Barn Owl Productivity Response to Variability of Vole Populations

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Abstract

We studied the response of the barn owl annual productivity to the common vole population numbers and variability to test the effects of environmental stochasticity on their life histories. Current theory predicts that temporal environmental variability can affect long-term nonlinear responses (e.g., production of young) both positively and negatively, depending on the shape of the relationship between the response and environmental variables. At the level of the Czech Republic, we examined the shape of the relationship between the annual sum of fledglings (annual productivity) and vole numbers in both non-detrended and detrended data. At the districts’ level, we explored whether the degree of synchrony (measured by the correlation coefficient) and the strength of the productivity response increase (measured by the regression coefficient) in areas with higher vole population variability. We found that the owls’ annual productivity increased linearly with vole numbers in the Czech Republic. Furthermore, we also found that synchrony between dynamics in owls’ reproductive output and vole numbers increased with vole population variability. However, the strength of the response was not affected by the vole population variability. Additionally, we have shown that detrending remarkably increases the Taylor’s exponent relating variance to mean in vole time series, thereby reversing the relationship between the coefficient of variation and the mean. Instead, we suggest that higher synchrony could result from high food specialization of owls on the common vole in areas with highly fluctuating vole populations.

Introduction

In recent years, ecologists have increasingly recognised the importance of environmental variability on population growth rate, behavioural strategies and individual fitness [1, 2]. The nonlinear responses of organisms to environmental variability are prevalent in biology. In a fluctuating environment, their long-term response measured as the average in a long run is inevitably either higher or lower than that measured in a less variable or constant environment. This is a
direct result of Jensen’s inequality, a mathematical property stating that for a non-linear function \( f(x) \) the average function \( \bar{f}(x) \) does not equal to \( f(\bar{x}) \) (see [3] or [4] for detailed graphical explanation). Based on this rationale, life history theory predicts that temporal environmental variability in food resource availability can affect the long-term organism’s response (e.g., production of young) positively if the function is accelerating (convex up, the second derivative is positive), i.e., organisms respond to environmental stochasticity by slightly decreasing productivity in periods of low food abundance but greatly increasing productivity in good years. Strategies that increase this convexity are favoured by natural selection as they lead to higher fitness [3, 4]. On the other hand, when productivity response is a concave function of a stochastically fluctuating resource, strategies that decrease the concavity are favoured by avoiding the detrimental effects of bad years. Organisms, thus respond evolutionarily to low food abundance by substantially decreasing their breeding capacity which in years of high food abundance becomes saturated and allows only for slightly increased productivity. This saturating response can be expected in some avian predators nesting once a year with a limited capacity to increase the clutch size in years of high food availability [4, 5]. There is no effect of environmental variability on a long-term productivity average if the function is linear, or the effect can further complicate if the underlying function is sigmoid [6].

Birds and mammals feeding on small herbivore voles whose population numbers fluctuate largely in time represent a useful model system to test this theory. Vole dynamics typically consist of long intervals of low abundance and short intervals of superabundance [7]. This is why the voles are commonly considered a good example of environmental stochasticity referred to as pulsed resources [8–10]. Annual productivity of vole-hunting animals has been documented to decrease in years of low vole abundances in both avian and mammalian predators. Such reports on their breeding responses include the significant decrease in clutch sizes in the hen harrier [11], the tawny owl [12], and the Ural owl [13, 14], the reduced number of fledglings in the barn owl [15], or reduced egg size and hatching success in the Eurasian kestrel [16]. In mammals, arctic foxes decrease ovulation rate [17] resulting in smaller litter sizes and productivity [18, 19]. Weasels have also been observed to lower embryo and offspring survival rates [20].

Unlike evidence on how these predators respond to changes in resource availability, we are much less informed on how they respond to changing variability in resource availability. Hušek et al. [5] examined the white stork–common vole system and concluded that breeding responses of storks, measured as cross-correlation between annual stork productivity and vole abundances, were stronger in areas with higher variability of vole numbers. Interestingly enough, the relationship between the numbers of fledglings and vole abundances was observed to be concave, thereby predicting the opposite. One way of reconciling the theory with this odd observation is that the strength of the breeding response was erroneously estimated using cross-correlations. Cross-correlations can give us an estimate of closeness of points to a linear relationship or temporal synchrony between productivity and vole numbers but not the rate of change in productivity as a function of changes in vole numbers. Regression coefficients do that, hence these measure the strength of the stork breeding response to vole variability more appropriately and accurately. Another potential explanation for the positive relationship between the breeding response and population variability has been suggested by Barraquand and Hušek [4] who found an exceptionally high exponent \( b \) from Taylor’s law describing the dependence of variance on the population mean in the vole population index. Their best-fit estimate of \( b \) for the Czech and Polish populations was found to be somewhere between 4 and 6, thereby deviating greatly from the interval of 1 to 2 observed for most organisms [21]. As a result, voles’ population variability increases with the population mean. The positive effect of
variation in vole numbers on the strength of response then arises from its covariation with mean vole density. To assess the generality of previous studies, analyses of further empirical cases are therefore warranted.

Here we analyse the barn owl–common vole system to study the responses of the barn owl annual productivity to common vole population numbers and variability in central Europe. By applying two different methodological approaches, one with non-detrended original data and the other with detrended data prior to analysis, we then described the functional relationship between the annual sum of fledglings and vole numbers at the level of the whole Czech Republic. Next we focus on the variation in annual productivity relative to the vole population variability using data from ten Czech districts. Addressing the main question: when responding to the same change in vole numbers, do the owls increase or decrease the number of fledglings equally in all areas irrespective of vole population variability? To describe the strength of the responses in annual productivity, we use the slope estimate from a linear regression of annual productivity on the vole numbers in the 10 districts. Finally, we demonstrate how detrending the time series can shift Taylor’s exponent and potentially divert ecological inference.

**Methods**

**Barn owl data**

The barn owl (*Tyto alba*) is a long-lived medium-sized owl with high breeding site fidelity [22, 23]. They typically prefer farm buildings and church towers as nesting sites and their home range covers an area of 5 to 10 km² [24]. This species nests from April to July and occasionally nests twice or three times a year [25]. In open farmlands small rodents are crucial prey constituting up to 90% of their diet [24]. The common vole (*Microtus arvalis*) is the most common food item out of all small rodents [26]. Other prey like birds, amphibians and insects are of reduced importance. Obvious prey switching from mammals to other prey classes does not take place even when small mammals decrease to 7% of the available food, suggesting that barn owls are highly specialized predators of small rodents [27].

The fledglings remain dependent on their parents for one to two months. In Western Europe, mainly juvenile barn owls were able to migrate over large areas and follow places of high prey abundance [28]. The loss of suitable nesting sites has been suggested as one of the main causes of population decline in Western Europe [24, 29]. To support the barn owl population in the Czech Republic, 1470 artificial nest-boxes specifically designed to protect young against predators were installed in farm buildings between 1996 and 2001. Within few years of the national monitoring and ringing programme, these nest-boxes have become widely used by owls. Between 1998 and 2013, we monitored the barn owls’ nesting population by checking all nest-boxes. Additionally, we also searched for natural nests in suitable habitats, such as sacral buildings, castles and farm buildings. All known nesting sites were checked at least once a year in late May to July. The occupied nests were then repeatedly visited for ringing of young and in autumn to check the second nesting attempt. On rare occasions, nesting was delayed until August. These cases were identified during the following year control by signs such as pellets or abandoned eggs left by. In total, 662 potential nesting sites were examined regularly every year during 1998–2013. Each occupied nesting site was visited repeatedly over the whole nesting period between May and November and the number of eggs, hatchlings and fledglings were recorded, these data include direct observations of eggs, hatchlings or fledglings but also back calculations from nestlings and unhatched eggs (about 40%). In total, we collected 1667 observations on the number of eggs, 1580 observations on the number of successful hatchlings, and 1579 observations on the number of successful fledglings from 681 different nesting sites situated in 17 Czech districts (Fig 1). In our sample, the second and third nesting attempts were
recorded in 31.7% and 0.3% nesting sites, respectively, similar to reported data from Germany [30]. We assume that the second and third clutches were produced by the same female due to evidence of very high site fidelity in adult barn owls [22, 23].

**Ethics statement**

The ringing data on barn owls were collected in accordance with ethical standards following the Act No. 418 246/1992 Coll. on the protection of animals against cruelty by KP and IM who are official bird ringers in the Czech Republic. They have authorisation by the Ministry of the Environment of the Czech Republic to research and protect the Western Barn Owl (No. 6826/01–OOP/2110/01–V847) and after revision of the jurisdiction the permission was issued by the Nature Conservation Agency of the Czech Republic No. 426/PA/2007. They also hold the permission issued by Prague Ringing Centre No. 804 to ring birds such as the Western Barn Owl. The areas in the open agricultural land are freely accessible in the Czech Republic and no permission is required. We co-operated with hundreds of agricultural building’s owners (mostly farmers). And we have written agreements from all of the owners allowing us to enter the agricultural buildings set up nest-boxes and carry out the research.

**Common vole data**

The common vole (*Microtus arvalis*) is the most abundant microtine rodent in central European farmlands primarily inhabiting open grassy habitats and agricultural fields with forage crops, such as alfalfa and clover. Their abundances fluctuate strongly with peaks occurring at intervals of three to four years [7]. The peculiar feature of this species is that autumn local population densities can attain more than two thousand individuals per hectare in peak years [31], whereas during population lows the numbers may decrease to virtually zero. The common vole population densities in this study were assessed at the district level, administrative units of on average 1000 km² in size, using a population index based on counts of active burrow entrances per hectare. These are indicated by the presence of signs, such as smooth margins of the entrances, fresh plants placed inside the burrow openings, fresh heaps of soil and/or fresh droppings. Since 2000, the State Phytosanitary Administration has monitored its numbers in

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**Fig 1.** (a) Map of the districts in the Czech Republic showing the distribution of barn owl nesting sites and (b) the dynamics of barn owl productivity (solid lines) and the common vole numbers (dashed line) in autumn during the period 1998–2013. The shaded areas in (a) indicate the 10 districts used in the analysis of vole population variability effects on the strength of the responses in the barn owl productivity parameters. Barn owl productivity was measured as the annual number of successfully produced fledglings (solid line). Vole numbers were measured by a vole index based on the number of active burrow entrances per hectare.

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various crops in the Czech Republic twice a year during spring (March–April) and autumn (October–November). Here we use data collected exclusively in fodder, such as alfalfa, clover, or cultivated meadows known as the common vole preferred habitats. In each district, 10 sites were surveyed for the number of active burrow entrances by walking along four 100-m strips, each 2.5 m wide. The counts collected on a total area of 1000 m² were then multiplied by 10 to obtain the numbers per hectare.

**Statistical analysis**

We analysed the relationship between the barn owl reproduction and vole abundances at the level of the Czech Republic. The data on barn owl productive output and vole numbers are time series. Temporal trends in non-stationary series may present a problem not only when measuring population variability [32] but also when measuring the relationships between the two time series by causing spurious correlations due to correlated trends [33]. Since ecologists are most often interested in measuring correlations for fluctuations around the trend rather than those for trends, these are routinely removed prior to analysis [34]. This approach is justified if the time series are long and trends are proven to be real at a reasonable level of certainty. However, if the time series are rather short, as with our data (14 years), and there is no external evidence of environmental change, the data are best analysed both with and without detrending in order to document the effects of detrending on the obtained results [34:176]. We, therefore, adopted two analytical approaches, one is based on the original data for barn owls and voles and another on the data with linear trends removed. Detrending of vole time series was done on a log scale (S1 Fig). The annual barn owl productivity was measured as the mean number of the successful fledglings produced per site.

We started by performing a correlation analysis to compare abilities of spring and autumn vole indices to predict the annual number of fledglings. The seasonal abundances of voles for the Czech Republic were derived as the means calculated from the district vole abundances weighted by the number of barn owl data from that district, i.e., we used the district abundances of voles in frequencies that corresponded to the numbers of barn owl measurements from that district. The annual number of clutches and annual fledgling productivity were then regressed on the annual vole abundances by fitting a weighted linear model. Reciprocals of variance for averaged annual number of clutches and annual numbers of fledglings were used as weights. Next, we analysed the relationship between the numbers of fledglings and vole numbers at the district level. We have chosen 10 districts in which the data on both owls and voles were available for more than eight years. To characterize the closeness or temporal synchrony between the dynamics of fledgling and vole numbers, we calculated cross-correlation coefficients. To quantify the strength of owls’ responses to variation in vole numbers, we used regression slopes. Standard errors for the cross-correlation coefficients were obtained by bootstrapping using 10000 randomly drawn samples. Lastly, the cross-correlation coefficients and regression slopes were modelled as a function of vole variability measured as the $s$-index, i.e., the standard deviation of log$_{10}$-transformed vole densities. We fitted a weighted linear regression to consider the different uncertainties in parameter estimates. Reciprocals of variances were used as weights for the cross-correlation and slope coefficients. For the detrended data approach, we calculated detrended $s$-index, i.e., $s$-index based on detrended vole time series of abundances. The best-fitting model was selected according to the lowest AICc [35] from two models: intercept-only model and model containing the effect. The difference in AICc ($\Delta$AICc) between these two models, when greater than two was considered as a strong evidence for the best model. To describe autumn vole population dynamics, specifically the cycle period, we fitted autoregressive log-linear models of order 2 using the function arima in R

The model is defined by the equation (e.g., [37]):

\[ X_t = a_0 + (1 + a_1)X_{t-1} + a_2X_{t-2} + \epsilon_t \]  

(1)

in which \( X_t \) is the logarithm of autumn vole index in a given year \( t \), \( \epsilon_t \) is a Gaussian noise term and \( a_1 \) and \( a_2 \) are the estimates of annual direct and delayed density dependence determining the length of the cycle period [38]. In particular, the more negative the parameter \( 1 + a_1 \), the shorter the cycle period. Through relationships to density dependence parameters, vole population variability can be related to the variation in a cycle period, i.e., the time interval between the two population peaks. We then examined the relationship between the degree of synchrony and strength of productivity responses on density dependence parameters by fitting weighted linear regression models with direct or delayed density dependence parameters as predictors.

Lastly, to demonstrate how the detrending procedure can affect the parameter value of a Taylor’s exponent \( b \) and the relationship between the coefficient of variation (CV), which is another proposed measure of population variability [39], and mean population density, we computed the exponent \( b \) from Taylor’s power law for the variance–mean relationship \( \text{Var} = a \text{Mean}^b \). By taking logs, a linear functional form can be obtained

\[ \log(\text{Var}) = \log(a) + b \log(\text{Mean}) \]  

(2)

in which, the parameter \( b \) is estimated by the ordinary least square method. In a vast majority of organisms, the exponent \( b \) falls between 1 and 2 [32]. The Taylor’s law also determines the relationship between the CV and the mean population density:

\[ CV = a^{0.5} \text{Mean}^{0.4b-1} \]  

(3)

where, \( CV = (1+1/n)s/\text{Mean} \), \( s \) is the standard deviation of vole population numbers \( N \), and \( n \) is the sample size. If \( b < 2 \), then the CV is negatively related to the population mean.

Results

The overall mean number of clutches produced each year per nesting site over the study period 1998–2013 was 1.31 (SE 0.001, \( n = 2071 \), max = 3). The overall mean number of fledglings produced per nesting site was 4.82 (SE 0.004, \( n = 1369 \), min = 0, max = 17). Based on original data, the number of fledglings per year \( t \) co-varied less with vole spring abundances in year \( t \) (correlation coefficient 0.438, 95% CI –0.121 to 0.786) than autumn abundances in year \( t \) (correlation coefficient 0.656, 95% CI 0.192 to 0.880; Fig 1). At the country level, there was no time lag in synchrony of owl annual productivity with vole abundances. Since the annual productivity of owls was better predicted by autumn vole abundances, only autumn indices of vole numbers were used in further analyses. By applying linear regression analysis, we found evidence for the mean number of clutches and fledglings to be positively related to the number of voles in autumn for non-detrended (clutches: \( \Delta AICc = 2.45 \); fledglings: \( \Delta AICc = 10.68 \)). For detrended data, evidence for clutches was weak (\( \Delta AICc = 1.68 \)) but strong for fledglings (\( \Delta AICc = 16.12 \); Fig 2). There was no indication of nonlinearity in the responses of owls.

At the districts’ level, we fitted weighted linear regression models to examine the relationship between temporal synchrony for barn owl productivity and vole index and the strength of the response of fledgling number to vole population variability. We obtained evidence of increased synchrony with vole population variability with both non-detrended (\( \Delta AICc = 5.99 \)) and detrended data (\( \Delta AICc = 4.16 \); Fig 3A and 3B). However, there was no evidence for the strength of the response in the numbers of fledglings to increase with population variability,
with the intercept-only models giving a slightly better fit to the data (non-detrended: ΔAICc = 1.06; detrended: ΔAICc = 0.31; Fig 3C and 3D). We also did the same analysis for the effects of vole population means in 10 districts. The only effect we have found was for detrended vole index which increased synchrony between barn owl productivity and vole index (S2 Fig).

We found no evidence for the models containing the effects of direct or delayed density dependence on synchrony or strength of the response (S3 and S4 Figs). Finally, we regressed log (Var) on log (Mean) for non-detrended and detrended vole data to check for the value of Taylor’s exponent (Fig 4). The exponents were 1.41 (95% CI 0.70 – 2.12) and 3.38 (95% CI 1.41 – 5.35) for non-detrended and detrended data, respectively. As a result of detrending, the relationship of CV to mean was reversed from a negative to a highly positive one, suggesting
that the loss of variation due to detrending introduced a positive dependence of the variance on the mean.

**Discussion**

While the breeding responses of most vole predators to changes in prey abundance are well documented [11–16], their responses to temporal variability in prey abundances are rare [5, 6] and poorly understood [4]. Here we examined the barn owls’ annual productivity (defined as the annual sum of fledglings produced) in relation to the common vole fluctuating populations in central Europe, specifically focusing on the effect of population variability on the strength of the productivity response. Using two analytical approaches, one with non-detrended and another with detrended data, we found that the annual number of fledglings increased linearly with autumn vole numbers. We obtained strong evidence that the degree of synchrony between the owls’ reproductive output and vole numbers increased with population variability. In contrast, the strength of productivity response did not. Moreover, we showed that detrending, substantially affected the value of the Taylor’s exponent and thus reversed the relationship.

![Fig 3. The relationship between barn owl productivity responses and vole population variability in ten districts of the Czech Republic.](image-url)
between population variability and population mean. Overall, these results suggest that unlike the mean, vole population variability does not influence the productivity response in barn owls. However, they do provide supportive evidence for the findings obtained by Hušek et al. [5] with white stork–vole system that the productivity responses of consumers to the fluctuating common vole abundances are more precise in areas with more variable vole populations.

As expected, we demonstrated that the barn owls’ productivity increased with the common vole numbers. The fact that autumn vole numbers are capable of better predicting the barn owls’ productivity than spring numbers can perhaps be best explained by the owls’ ability to adjust their reproductive investments steadily over the breeding season according to changing food availability. Alongside a change in brood survival, the responses in the number of clutches and clutch sizes can lead to a close fit between overall annual productivity and vole numbers in autumn. Hence, there is no need to invoke any anticipatory responses as proposed in seed predators such as rodents ([40] but see [41]).

The observed close relationship between the owls’ reproductive output and vole numbers once again confirms the importance of the common vole as the primary food resource for barn owls in central Europe and the fact that the response of owls is quick, with no time-lag [24, 26].

![The Taylor's power law relationships for the vole time series data.](https://doi.org/10.1371/journal.pone.0145851.g004)

The upper panels show the relationship between variance and mean for non-detrended (a) and detrended (b) data. The lower panels show the relationship between the coefficients of variation (CV) and mean for non-detrended (c) and detrended (d) vole data.
Although the close linkage between vole-eating birds and voles in Europe is well known, the quantitative descriptions of predators’ breeding responses to variation in prey abundance are rarely reported [30, 42, 43]. Some of them are concave, such as those for Montagu’s Harrier [42], white storks [5] or long-tailed skuas [6] but some are obviously linear, such as that reported for the barn owl in Scotland [22]. In particular, the latter is in agreement with our data. This linear relationship means that owls respond to increases and decreases in vole numbers equally, i.e., the rate of change in the number of fledglings produced per unit increment in prey numbers is constant over the whole range of vole densities. Consequently, the long-term mean in productivity is not affected by vole population variability. If there is any biological trait contributing to the linearity of the response without any upper bound, then it is the ability of barn owls to breed more than once as compared to Montagu’s harriers, storks or skuas whose reproductive capacity is constrained to one nesting event per year.

Like Hušek et al. [5], we observed a positive relationship between the degree of synchrony (correlation) and population variability of voles. However, because they interpreted the correlation as the strength of the breeding response, this observation is in conflict with Jensen’s inequality theory for concave curves, which predicts the opposite [4]. This discrepancy was then explained by unusually high Taylor’s exponent $b$ of about 4 to 6 inducing a strong positive relationship between voles’ temporal variability and population mean. Our results suggest that the high Taylor’s exponent $b$ is not a specific feature of the common vole population variability. Instead, it may be the result of data handling procedures which reduced the original amount of variation. In our data, it was due to detrending. In vole data used in Hušek et al. [5], it was the reconstruction procedure. Their continuous data came from Tkadlec et al. [44] and were derived from annual pest survey maps showing four abundance categories for the common vole. Through this process, a large portion of temporal variation in vole numbers was lost, particularly in districts with low vole’s population densities. These districts are characterized by high proportion of forested areas which are suboptimum habitats for the common vole. Incidentally, it is just these low-density districts that were analysed in the white stork study. Likewise, the elimination of the trend in our study reduced the variance in log population means. Because $b$ from Taylor’s power law is a standard regression coefficient computed as $\text{Cov (log (Var), log (Means))}/\text{Var (log (Mean))}$, any reduction in Var (log (Mean)) leads inevitably to an increased $b$. Hence, our study adds to a long-term search for situations in which Taylor’s exponent reaches values of $b > 2$ [45].

Surprisingly enough, the shift in $b$ due to detrending does not seem to be the cause of increased synchrony of stork productivity with voles’ population variability. As demonstrated by our results, we identified high synchrony in both approaches, suggesting that it is not related to the detrending procedure. As an alternative explanation, we propose that owls in high vole variability areas are more specialized on the common vole which in turn reduces their diet breadth, especially in high-density years. As a result, a stronger ecological signal conveyed by more variable vole dynamics translates into consumer’s productivity dynamics more precisely, thereby resulting in higher correlations between both dynamics. Another feature contributing to the stronger correlations in more variable districts is that higher vole population densities are measured more precisely. For population densities above 1000 burrow entrances per hectare the relative sampling error falls below 10% [46] which, in general, is a desired level of precision for most population measurements [47].

We found insufficient evidence for an increased strength of the productivity responses to more variable vole populations. In fact, our evidence based on AICc did not allow us to discriminate between the models with and without effect. In particular, the approach with detrended data was a better fit than that without it. Surely, detrending non-stationary data does have the potential to shape the outcomes of time series analyses leading us to draw
different ecological inferences. Even if real, the effect of variability seems to be quite small and thus may perhaps require much larger sample sizes to actually prove it. However, even if so, it would be very difficult to interpret the effects of the common vole population variability in biological terms. In areas with a broader range of alternative prey and less dependence on voles, barn owls may respond not only less precisely to changes in vole numbers but also less strongly, thus eliciting no adaptive explanation.

By focussing on the barn owl–common vole system, we showed how the pulsed resource can influence the dynamics in the consumers’ reproductive output by bringing them into a close synchrony with the prey. From the life history perspective, it might be more insightful to examine the consumer responses in systems with a curvilinear relationship between productivity and pulsed resource variability which can facilitate testing more specific predictions (e.g., [5]). Furthermore, comparisons of responses in vole-eating consumers with different levels of vole specialization may also help us to better understand not only consumers’ life histories but also the dynamics of such interactive systems.

Supporting Information

S1 Fig. Linear trends (dashed line) in log-transformed autumnal vole numbers (solid line) for ten districts of the Czech Republic. With the exception of one district (ZN), the intercept-only models performed better than the models with time (the difference in AICc > 2).

S2 Fig. The relationship between barn owl productivity responses and vole population means in ten districts of the Czech Republic. The upper panels show the degree of synchrony between the barn owl productivity and vole population means using the non-detrending (a) and detrending approach (b). The lower panels show the strength of the productivity response to vole population means without detrending (c) and with detrending (d). The dashed lines indicate 95% confidence intervals for the regression line.

S3 Fig. The effects of direct density dependence on synchrony (a, b) and strength of productivity response of barn owls (c, d) using non-detrended (a, c) and detrended data (b, d). The intercept-only models performed better or equally well as the models containing direct density dependence.

S4 Fig. The effects of delayed density dependence on synchrony (a, b) and strength of productivity response of barn owls (c, d) using non-detrended (a, c) and detrended data (b, d). The intercept-only models performed better or equally well as the models containing delayed density dependence.

S5 Fig. Linear trends (dashed line) in log-transformed autumnal vole numbers (solid line) for ten districts of the Czech Republic. Except the district ZN, the intercept-only models performed better the model with time (the difference in AICc > 0).

S6 Fig. The relationship between barn owl productivity responses and vole population means in ten districts of the Czech Republic. The upper panels show the degree of synchrony between the barn owl productivity and vole population means using the non-detrending (a) and detrending approach (b). The lower panels show the strength of the productivity response to vole population means without detrending (c) and with detrending (d). The dashed lines
indicate 95% confidence intervals for the regression line.

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Author Contributions
Conceived and designed the experiments: PP IM ET. Analyzed the data: PP AG JL ET. Wrote the paper: PP AG JL ET. Collected data on barn owls: KP IM. Collected data on vole index: ET.

References
1. Cohen D (1993) Fitness in random environments. In: Yoshimura J, Clark C, editors, Adaptation in Stochastic Environments, Lecture Notes in Biomathematics 98, Berlin: Springer. pp. 8–25.
2. Lande R, Engen S, Saether B-R (2003) Stochastic population dynamics in ecology and conservation. Oxford: Oxford University Press.
3. Ruel JJ, Ayres MP (1999) Jensen’s inequality predicts effects of environmental variation. Trends in Ecology and Evolution 14: 361–366. PMID: 10441312
4. Barraquand F, Hušek J (2014) Covariation between mean vole density and variability drives the numerical response of storks to vole prey. Population Ecology 56: 551–553.
5. Hušek J, Adamík P, Albrecht T, Cepák J, Kania W, Mikolášková E, et al. (2013) Cyclicity and variability in prey dynamics strengthens predator numerical response: the effects of vole fluctuations on white stork productivity. Population Ecology 55: 363–376.
6. Barraquand F, Høye TT, Henden J-A, Yoccoz NG, Schmidt NM, et al. (2014) Demographic responses of a site-faithful and territorial predator to its fluctuating prey: long-tailed skuas and arctic lemmings. Journal of Animal Ecology 83:375–387. doi: 10.1111/1365-2656.12140 PMID: 24128282
7. Tkadlec E, Stenseth NC (2001) A new geographical gradient in vole population dynamics. Proceedings of the Royal Society of London B 268: 1547–1552.
8. Ostfeld RS, Keesing F (2000) Pulsed resources and community dynamics of consumers in terrestrial ecosystems. Trends in Ecology and Evolution 15: 232–237. PMID: 10802548
9. Yang LH, Bastow JL, Spence KO, Wright AN (2008) What can we learn from resource pulses? Ecology 89: 621–634. PMID: 18459327
10. Yang LH, Edwards KF, Byrnes JE, Bastow JL, Wright AN, Spence KO (2010) A meta-analysis of resource pulse–consumer interactions. Ecological Monographs 80: 125–151.
11. Baines D, Richardson M (2013) Hen harriers on a Scottish grouse moor: multiple factors predict breeding density and productivity. Journal of Applied Ecology 50: 1397–1405.
12. Yoccoz NG, Bustnes JO, Bangjord G, Skaare JU (2009) Reproduction and survival of tawny owls in relation to persistent organic pollutants. Environment International 35: 107–112. doi: 10.1016/j.envint.2008.08.014 PMID: 18930319
13. Brommer JE, Pietiläinen H, Kolunen H (2002) Reproduction and survival in a variable environment: Ural Owls (Strix uralensis) and the three-year vole cycle. The Auk 119: 544–550.
14. Kontiainen P, Pietiläinen H, Huttunen K, Karell P, Kolunen H, Brommer JE (2009) Aggressive Ural owl mothers recruit more offspring. Behavioral Ecology 6: 789–796.
15. Charter M, Izhaki I, Leshem Y, Meyrom K, Roulin A (2015) Relationship between diet and reproductive success in the Israeli barn owl. Journal of Arid Environments 122: 59–63.
16. Valkama J, Korpimäki E, Wiehn J, Pakkanen T (2002) Inter-clutch egg size variation in kestrels Falco tinnunculus: seasonal decline under fluctuating food conditions. Journal of avian biology 33: 426–432.
17. Lindström E (1988) Reproductive effort in the red fox, Vulpes vulpes, and future supply of a fluctuating prey. Oikos 52: 115–119.
18. Angerbjörn A, Arvidson B, Norén E, Strömberg L (1991) The effect of winter food on reproduction in the arctic fox, Alopex lagopus: a field experiment. Journal of Animal Ecology 60: 705–714.
19. Angerbjörn A, Tannerfeldt M, Erlinge S (1999) Predator-prey relationships: arctic foxes and lemmings. Journal of Animal Ecology 68: 34–49.
20. Sundell J (2003) Reproduction of the least weasel in captivity: basic observations and the influence of food availability. Acta Theriologica 48: 59–72.
21. Linneberg M, Saether B-E, Grotan V, Engen S, Noble DG, Freckleton RP (2013) Interspecific differences in stochastic population dynamics explains variation in Taylor's temporal power law. Oikos 122: 1207–1216.
22. Taylor I (1994) Barn owls. Predator-prey relationships and conservation. Cambridge: Cambridge University Press.
23. Bond G, Burnside NG, Metcalfe DJ, Scott DM, Blamire J (2005) The effects of land-use and landscape structure on barn owl (Tyto alba) breeding success in southern England, UK. Landscape Ecology 20: 555–566.
24. de Bruin O (1994) Population ecology and conservation of the barn owl Tyto alba in farmland habitats in Liermers and Achterhoek (The Netherlands). Ardea 82: 1–109.
25. Marti CD (1994) Barn owl reproduction: patterns and variation near the limit of the species' distribution. The Condor 96: 468–484.
26. Bernard N, Michelat D, Raoul F, Quere JP, Delattre P, Giraudoux P (2010) Dietary response of Barn Owls (Tyto alba) to large variations in populations of common voles (Microtus arvalis) and European water voles (Arvicola terrestris). Canadian Journal of Zoology 88: 416–426.
27. Jaksic J, Jiménez JE, Castro SA, Feinsinger P (1992) Numerical and functional response of predators to a long-term decline in mammalian prey at a semi-arid Neotropical site. Oecologia 89: 90–101.
28. Honer MR (1963) Observations on the barn owl in the Netherlands in relation to its ecology and population fluctuations. Ardea 51: 158–195.
29. Toms MP (1994) Small mammals in agricultural landscapes. Raptor 21: 57–59.
30. Schönfeld M, Gribig G (1975) Beiträge zur Brutbiologie der Schleiereule, Tyto alba, unter besonderer Berücksichtigung der Abhängigkeit von der Feldmausschädigung. Hercynia N. F., Leipzig 12: 257–319.
31. Bryja J, Něsvadbová J, Heroldová M, Jánošová E, Losík J, Trebatická L, et al. (2005) Common vole (Microtus arvalis) population sex ratio: biases and process variation. Canadian Journal of Zoology 83: 1391–1399.
32. Gaston KJ, McArdril BH (1994) The temporal variability of animal abundances: measures, methods and patterns. Philosophical Transactions of the Royal Society of London B 345: 335–358.
33. Granger CWJ, Newbold P (1974) Spurious regressions in econometrics. Journal of Econometrics 2: 111–120.
34. Turchin P (2003) Complex population dynamics: a theoretical/empirical synthesis. Princeton: Princeton University Press.
35. Burnham KP, Anderson DR (2002) Model selection and multimodel inference: a practical information-theoretic approach. New York: Springer.
36. R Core Team (2014) R: A language and environment for statistical computing. Vienna (Austria): R Foundation for Statistical Computing.
37. Gouveia A, Bejček V, Flousek J, Sedláček F, Štastný K, Zima J, et al. (2015) Long-term pattern of population dynamics in the field vole from central Europe: cyclic pattern with amplitude dampening. Population Ecology 57: 581–589.
38. Royama T (1992) Analytical population dynamics. London: Chapman and Hall.
39. McArdril BH, Gaston KJ, Lawton JH (1990) Variation in the size of animal populations: patterns, problems and artefacts. Journal of Animal Ecology 59: 439–454.
40. Boutin S, Wauters LA, McAdam AG, Humphries MM, Tost G, Dhondt AA (2006) Anticipatory reproduction and population growth in seed predators. Science 314: 1928–1930. PMID: 17185600
41. White TCR (2013) ‘Anticipatory’ reproduction by small mammals cannot succeed without enhanced maternal access to protein food. New Zealand Journal of Zoology 40: 332–336.
42. Salamolard M, Butet A, Leroux A, Bretagnolle V (2000) Responses of an avian predator to variations in prey density at a temperate latitude. Ecology 81: 2428–2441.
43. Millon A, Bretagnolle V (2008) Predator population dynamics under a cyclic prey regime: numerical responses, demographic parameters and growth rates. Oikos 117: 1500–1510.
44. Tkadlec E, Zbořil J, Losík J, Gregor P, Lisická L (2006) Winter climate and plant productivity predict abundances of small herbivores in central Europe. Climate Research 32: 99–108.
45. Pertoldi C, Faury S, Reed DH, Nkape J, Bjorklund M, Lundberg P, et al. (2014) Scaling of the mean and variance of population dynamics under fluctuating regimes. Theory in Biosciences 133: 165–173.
46. Lisická L, Losik J, Zejda J, Heroldová M, Něsvadbová J, Tkadlec E (2007) Measurement error in a burrow index to monitor relative population size in the common vole. Folia Zoologica 56: 169–176.
47. White GC, Anderson DR, Burnham KP, Otis DL (1982) Capture-recapture and removal methods for sampling closed populations. Los Alamos: Los Alamos National Laboratory.