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**Larger but younger fish when growth compensates for higher mortality in warmed ecosystem**

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Abstract

Ectotherms are often predicted to “shrink” with global warming, in line with the temperature-size rule (TSR), which states warming leads to smaller adult body sizes. But the TSR also predicts faster growth rates. As such, whether the average size of a population declines with warming depends on the mortality rates. We used data from an artificially heated (+8C) lagoon to analyze how warming has affected body growth, mortality rates and population size-structure of Eurasian perch (*Perca fluviatilis*). Compared to a reference area, we found that the size was larger for all ages and growth was faster for all sizes, to the extent that the size-spectrum slope was larger (indicating a higher proportion of large fish in) – despite the increased mortality rates. General size-temperature rules derived from experimental studies may not always translate to natural populations, and thus predictions of climate change impacts should incorporate a wider range of demographic parameters.

Introduction

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Results

Results

Discussion

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Materials and Methods

*Study system*

The Biotest basin is a 1 km2 artificial enclosure of Swedish Baltic Sea archipelago, built in conjunction with the construction of the nuclear power plant in Forsmark in 1977. Since 1980, the lake has received cooling water from the plant, after which the water temperature has been approximately 8 higher than in surrounding reference area (Adill *et al.* 2013; Huss *et al.* 2019). All types of fishing (apart from the regular surveys) have been banned since the construction (Huss et al., 2019). The Biotest Lake was also provided with a grid at its outlet which, together with the strong current, prevented fish larger than 10 cm from migrating in and out of the area (Adill et al., 2013; Huss et al., 2019). Genetic studies confirm the reproductive isolation between the two populations (Björklund et al., 2015) between 1980-2003 (the grid was removed in the spring 2004 and the Biotest Lake has since then been an open system (Adill et al., 2013); hence all analyses in this report are based on data collected before 2004). The study speces is Eurasian perch (*Perca fluviatilis*), henceforth only perch, which is a widely distributed and common predatory freshwater native to Europe and northern Asia.

*Data*

The Biotest basin and the reference bay outside (henceforth the warm and cool area) have been scientifically sampled using survey-gillnets since the construction of the enclosure. Fishing took place in October in the Biotest lake and in August in the reference area when temperatures are most comparable between the two areas (Huss et al., 2019).

The catch was recorded by 2.5 cm length classes during 1987-2000, and into 1 cm length groups between 2001-2003. To express lengths in a common length standard, 1 cm intervals were converted into 2.5 cm intervals. The unit of catch data is hence number of fish caught per 2.5 cm size class per night. All data from fishing events with disturbance affecting the catch (e.g., seal damage, strong algal growth on the gears, clogging by drifting algae, boat traffic or other human inference) were removed.

Age and length-at-age was reconstructed for a semi-random length-stratified subset of individuals each year. This was done using annuli rings on the operculum bones (with control counts done on otoliths), and an established power-law relationship between the distance of annual rings and fish length: , where is the length of the fish, the operculum radius, *k* the intercept of the line, and *b* the slope of the line for the regression of log-fish length on log-operculum radius (Thoresson, 1996). Back-calculated length-at-age were obtained from the relationship , where is the back-calculated body length at length , is the final body length (body length at catch) and is the distance from the centre to the annual ring corresponding to age . Since perch exhibit sexual size-dimorphism, and age-determination and back calculation was only done for males in select years, we only used females for our analyses.

To focus on the long term differentiation between the two areas, and to ensure that we had enough back-calculated length-at-age observation within individual fish born after the onset of heating, we avoided the initial years of warming (when mortality rates increased only to decline again (Sandström *et al.* 1995), and used data from 1987-2003. 2003 corresponds to the year before the populations were no longer reproductively separated, and when the back-calculation parameters potentially changed (Magnus Appelberg, pers. comm.)

*Statistical Analysis*

The differences in growth, mortality and size structure between the two areas were quantified using hierarchical linear and non-linear regression models fitted in a Bayesian framework. First we describe each statistical model and then we provide details of model fitting, model diagnostics and comparison.

We fit two models to illustrate differences in growth. First, the von Bertalanffy growth equation (VBGE) describing length as a function of age: , where is the length at age (, years), is the asymptotic size, is the Brody coefficient (). Secondly, an allometric model describing how specific growth rate scales with length: , where, the specific growth, is defined as: and is the geometric mean weight: .

In the VBGE model, we used only age-and-size at catch as the response variables (i.e., not back-calculated length-at-age) to only have on observation per individual fish. To evaluate the support for shared versus unique VBGE parameters for the two areas, we fit models with a dummy-coded area effect and eight combinations of shared and unique parameters. For example, the full model was:

where we assume distribution for length-at-age to account for extreme observations, with , and representing the degrees of freed, mean and the scale parameter. and are dummy variables such that and if it is the cold area, and vice versa, and and subscripts indicate area-specific parameters. The multivariate normal distribution in Eq. 3 is the prior for the cohort-varying parameters , ,and (for cohorts …,1997), with hyper-parameters , , , describing the non-varying population means and a covariance matrix with the between-cohort variation along the diagonal (note we did not model a correlation between the parameters hence off-diagonals are 0). The remaining seven models include some or all parameters as common parameters, e.g., substituting and with . To aid convergence of this non-linear model, we used informative priors that we chose after visualizing draws from prior predictive distributions (Wesner & Pomeranz 2021) using probable parameter values (see *Supporting Information)*. We used the same prior for each parameter for both areas to not give introduce any other sources of differences in parameter estimates between areas. The following priors for the VBGE model: , (uniform to facilitate convergence), , , (where is one of the following parameters: *,* , , , and .

We estimated growth…

Growth, geometric, allometric

Mortality, Z, instant

Size spectrum, weight, Edwards method…

All analyses were done using *R* (R Core Team 2020) version 4.0.2. The packages within the *tidyverse* (Wickham *et al.* 2019) collection were used to processes and visualize data. Models where fit using the R-package `*brms`* (Bürkner 2017)*.* We used 3 chains and 4000 iterations in total per chain. Sub-models were compared by evaluating their expected predictive accuracy (expected log pointwise predictive density) using leave-one-out cross-validation (LOO-CV)(Vehtari *et al.* 2017) while ensuring pareto values < 0.7, in the R-package ´*loo´* (Vehtari *et al.* 2020)*. bayesplot* (Gabry *et al.* 2019) and *tidybayes* (Kay 2019) were used to process and visualize model diagnostics and posteriors. Model convergence and fit was assessed by ensuring potential scale reduction factors () where <1.1 (suggesting all three chains converged to a common distribution) (Gelman *et al.* 2003), and by visually inspecting trace plots, residuals QQ-plots and with posterior predictive checks (*Supporting Information*).

Code and Data Availability

All data and *R* code can be downloaded from a GitHub repository (<https://github.com/maxlindmark/warm_life_history>) and will be archived on Zenodo upon publication.

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Author Contributions

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Additional Information

Additional Information

Tables

Figures

Fig. 1. The average length-at-age is larger for all ages in the warm (red) compared to the cold (blue) area. Points in panel (A) depicts individual-level length-at-age and lines show the global posterior prediction without group-level effects (i.e., cohort) from the von Bertalanffy model with area-specific coefficients (MX). The shaded areas correspond to 50% and 90% credible intervals. Panel B shows the posterior distributions for parameters and , and panel C shows the distribution of differences. Panel D shows the posterior distributions for parameters and , and panel E shows the distribution of differences.

Diagram

Description automatically generated with medium confidence

Fig. 2. The faster growth rates in the warm area are maintained as fish grow. The points illustrate specific growth estimated from back-calculated length-at-age (within individuals) as a function of length (expressed as the geometric mean of the length at the start and end of the time interval). Lines show the global posterior prediction without group-level effects (i.e., individual within cohort) from the allometric growth model with area-specific coefficients (MX). The shaded areas correspond to the 90% credible interval. The equation uses mean parameter estimates. Panel B shows the posterior distributions for parameters and , and panel C shows the distribution of differences. Panel D shows the posterior distributions for parameters and , and panel E shows the distribution of differences.

Chart

Description automatically generated

Fig. 3. The instantaneous mortality rate is higher in the warm area. Panel (A) shows the as a function of , where the slope corresponds to the global . Lines show the posterior prediction without group-level effects (i.e., cohort) and the shaded areas correspond to the 50% and 90% credible intervals. The equation uses mean parameter estimates. Panel B shows the posterior distributions for parameters and , and panel C shows the distribution of differences.

Chart

Description automatically generated

Fig. 4. The warm area has a larger proportion of large fish, illustrated both as histograms of proportions and the generally smaller size-spectrum slopes (which also increased over time in both areas). Panel (A) shows mean annual size-spectrum slopes as points and their 95% confidence interval, estimated using the ANDREWS method. Lines and shaded areas correspond to the median and 90% credible interval. The equation uses mean parameter estimates. Panel B shows the posterior distributions for parameters and (i.e., the intercepts in the model shown in panel A of size-spectrum slopes as a function of year), and panel C shows the distribution of differences. Panel (D) illustrates histograms of length groups as proportions (for all years pooled).