An Algorithm for the von Bertalanffy Seasonal Cessation in Growth Function of Pauly et al. (1992)

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Abstract

Pauly et al. (1992; Australian Journal of Marine and Freshwater Research 43:1151–1156) introduced a modified von Bertalanffy seasonal growth function that allowed for a period of no growth. Pauly et al. (1992) provided special purpose software to fit the model to length-at-age data but this software is no longer available and specific details to implement a critical aspect of the new growth function were not clear. I provide details for this critical aspect of the function, implement the function in the open-source R environment, and briefly demonstrate the use of this function with four data sets. With this, the growth function of Pauly et al. (1992) is now readily available to all scientists with access to software that can fit nonlinear models to data. Thus, this growth function may be implemented in more situations and its fit rigorously compared to the results from other models of seasonal fish growth.

Keywords: Growth, Seasonal, Cessation, Nonlinear Modeling

# Introduction

The mean length-at-age for many fish (Haddon, 2011) and other aquatic animals (e.g., Hota, 1994; Harwood et al.*,* 2014) is often modeled with the von Bertalanffy growth function (VBGF; von Bertalanffy, 1938). A common foundation for several parameterizations of the VBGF is

where is the expected or mean length at time (or age) , is the asymptotic mean length, and is at least a function of . For example, the most common parameterization of the VBGF attributable to Beverton and Holt (1957) uses

(1)

where is a measure of the exponential rate at which approaches (Schnute and Fournier 1980) and is the theoretical time or age at which would be zero.

Many fish exhibit seasonal oscillations in growth as a response to seasonal changes in environmental factors such as temperature, light, and food supply (e.g., Bayley, 1988; Pauly et al., 1992; Bacon et al.2005; Garcia-Berthou et al., 2012; Carmona-Catot et al., 2014). Various modifications of Equation 1 have been used to model these seasonal oscillations in growth. The most popular of these modifications, from Hoenig and Choudaray Hanumara (1982) and Somers (1988) with a clarification by Garcia-Berthou et al. (2012), uses

(2)

with . In Equation 2, is the amount of time between time 0 and the start of the convex portion of the first sinusoidal growth oscillation (i.e., the inflection point) and is the proportional decrease in growth at the depth of the growth oscillation (i.e., "winter"). Equation 2 may represent no seasonal oscillation in mean length (=0), a reduced but not stopped increase in mean length (for 0<<1), a complete stop in the increase in mean length (=1), or a decrease in mean length (>1) during the “winter” (Figure 1). The point where the increase in mean length is smallest is called the "winter-point" () and is at because the sine function in Equation 2 has a period (i.e., the growth period) of one year.

Pauly et al. (1992) argued that a decrease in mean length with increasing age is unlikely for organisms whose skeletons largely preclude shrinkage and, thus, values of >1 from Equation 2 were unrealistic for length (but not weight) data (however, see Nickelson and Larson (1974), Huusko et al (2011) and Garcia-Berthou et al. (2012)). Pauly et al. (1992) then proposed a modification to Equation 2 that included a no-growth period where mean length was not allowed to decrease. Specifically, their modification is

(3)

with . In Equation 3, is the “no-growth time” or the length of the no growth period (as a fraction of a year) and is found by “subtracting from the real age () the total no-growth time occurring up to age ” (Pauly et al. 1992). Furthermore, because the units of changed from in Equation 2 to in Equation 3, Pauly et al. (1992) suggested using in Equation 3 to minimize confusion with from Equation 2.

Pauly et al. (1992) devised Equation 3 by assuming =1 and replacing in Equation 2 with (i.e., restricting the seasonal oscillation to the growth period and noting that only operates during the growth period). Their modification may be described geometrically (though not algorithmically) in two steps. First, Equation 2 with (fixed) =1 is fit to the observed lengths and ages that have had the cumulative subtracted (i.e., using ). This growth trajectory is then separated at each and horizontal segments that are units long are inserted at these points. This forms a growth trajectory that smoothly transitions into and out of the no-growth periods (Figure 2).

The growth function in Pauly et al. (1992) does not appear to have been widely used. Pauly et al. (1992) has been cited at least 70 times (from Google Scholar and ResearchGate searches on 31-May-16); though it appears that only two of 43 English journal citations (excludes book, dissertation, report, other non-journal citations, and journals not published in English) actually fit Equation 3 to data. Of these, Chatzinikolaou and Richardson (2008) used the special purpose LFDA software (www.mrag.co.uk/resources/lfda-version-50) to fit Equation 3 to length frequency data, whereas it is not clear how Beguer et al. (2011) fit the function, though they did fix to a constant value.

Perhaps the growth function of Pauly et al. (1992) has not been widely adopted because it is not clear how to actually fit the function to length-at-age data. Pauly et al. (1992) provided a then ubiquitous but now obsolete 3.5-in “diskette” with a computer program to estimate the parameters of Equation 3; however, the last diskette has been lost and the source code is no longer available (D. Pauly, pers. comm.). Pauly et al. (1992) did describe the operations performed by their program, but there is no detailed description of how should be operationalized. This is an important step in using Equation 3 because is a function of , but it is also a function of and , which are parameters to be estimated during the model-fitting process. Thus, the values for change with each iteration of the non-linear model-fitting algorithm.

Therefore, the objectives of this note are to (i) operationalize the calculation of , (ii) provide an algorithm for the calculation of to be used when fitting Equation 3 to observed data, and (iii) illustrate the use of this algorithm with real data. With this description, Equation 3 can now be implemented in more situations and rigorously compared with other seasonal growth models (e.g., Equation 2).

# Methods

## 2.1 Calculating

As noted by Pauly et al. (1992) the calculation of depends on the observed age () and the cumulative no-growth time prior to . In practice, the calculation of also depends on the position of the no-growth period within a year. Here, the position of the no-growth period is defined relative to and , such that the following six-step algorithm may be used to compute ages adjusted for cumulative prior to age (i.e., ), from observed ages (i.e., ).

1. Shift the age () by subtracting the start of the no-growth () period (i.e., ; Chatzinikolaou and Richardson 2008) from, such that a whole number will represent the start of a no-growth period. For example, if =0.4, then =2.4 will become 2.0 and =2.9 will become 2.5.
2. Subtract the whole number age (i.e., fully completed growth years) from the shifted age from Step 1 such that the remaining decimal represents the fraction of a shifted year. For example, a 0 will result if the shifted age is 2.0 and a 0.5 will result if the shifted age is 2.5.
3. Substract the from the value from the previous step.
4. If the value from the previous step is negative, then the age is within the no-growth period and the negative value should be replaced with a zero. Otherwise, the positive value represents the amount of time into a growth period.
5. Add the value from the previous step to the total growth time completed (i.e., the product of the number of growth periods completed and the length of the growth period ()).
6. Compute by adding back the that was subtracted in Step 1.

Further examples of values relative to values are shown in Figure 2.

The values that result from this algorithm are then input values, along with observed lengths, to a function for fitting Equation 3 with any nonlinear model fitting software. For convenience, an R (R Development Core Team 2016) function to represent Equation 3, including use of the algorithm to compute , is included in the vbFuns() function of the FSA package (Ogle 2016b). Use of this function is demonstrated in the Supplementary information.

## 2.2 Demonstrating the Algorithm

The algorithm developed to fit Equation 3 is demonstrated with four data sets. The first data set is the fork lengths (mm) and decimal ages (the number of opaque zones observed on otolith thin sections plus the proportion of the year after the designated birthdate) from 215 Australian bonito (*Sarda australis*) sampled from commercial landings as detailed in Stewart et al. (2013). Stewart et al. (2013) fit Equation 2 to these data but constrained to not exceed 1. These data were chosen to illustrate how Equation 3 may provide a better and more appropriate fit then Equation 2 with the boundary condition of . The remaining three data sets are for invasive Eastern mosquitofish (*Gambusia holbrooki*) from southern France to southern Spain detailed by Carmona-Catot et al. (2014). Standard lengths (mm) were measured for each fish and annual ages were estimated from length frequencies and analysis of scales, with decimal ages determined from capture date and estimated birth dates for a cohort. Carmona-Catot et al. (2014) fit Equation 2, without constraining , to fish from ten locations. Data from three locations were chosen to examine here to demonstrate how Equation 3 fits relative to Equation 2 with varying estimates of (i.e., much greater than 1 for Site 2, only slightly greater than 1 for Site 4, and much less than 1 for Site 9).

We used the “port” algorithm in the nls() function in R to estimate the parameters for both Equations 2 and 3 for all four data sets. Starting values for , , and were obtained from the vbStarts() function in the FSA package v0.8.8 (Ogle 2016b) as described in Ogle (2016a). Starting values for , and were obtained by visual examination of the length versus age plot. Starting values for were the starting value for divided by 1 minus the starting value for . Values of , , and were constrained to be positive, and were constrained to be between 0 and 1, and was constrained to be between 0 and 1 for the Australian bonito data and positive for the mosquitofish data. The growth function with the lowest Akaike Information Criterion (AIC) value was chosen as the better fit for each data set. Confidence intervals for each parameter were the 2.5% and 97.5% percentile values of non-parametric bootstrap parameter estimates computed with the nlsBoot() function from the nlstools package v1.0-2 (Baty et al. 2015) in R. All code used in these analyses is in the Supplementary information.

# Results

Equation 3 fit the Australian bonito data slightly better (a lower AIC value; Table 1) than Equation 2. The length of the no-growth period was estimated to be 0.13 or 13% of the year. The parameters were equal and the parameters were similar, but the parameters differed somewhat between the two functions (Table 1). Graphically, there was little perceptual difference in the fits of the two growth functions (Figure 3A).

Equation 3 did not fit the mosquitofish data better in situations where there was some evidence for a decrease in mean length with increasing age (i.e., C>>1 in Equation 2; e.g., Site 2; Table 1; Figure 3B) or no evidence for a cessation in growth (i.e., C<1 in Equation 2; e.g. Site 9; Table 1; Figure 3D). However, Equation 2 appeared to respond too dramatically to one sample of ages (approx. 0.4) at Site 2, and Equation 3 likely provides more realistic estimates of mean length throughout the seasonal cessation in growth period in this example (Figure 3B). Equation 3 fit better than Equation 2 when a cessation in growth was evident without an apparent decline in mean length with age for mosquitofish (i.e., Site 4; Table 1; Figure 3C).

# Conclusion

The carefully described algorithm provided here for computing , which allows for Equation 3 to be statistically fit to seasonal age data, appears to provide reasonable parameter estimates for the four examples provided. However, Equation 3 is likely not the globally best seasonal growth model, as demonstrated here with mosquitofish. Perhaps a better understanding of the utility of the Pauly et al. (1992) growth function for modeling the seasonal growth of fishes will be forthcoming now that this function is readily available to all scientists with access to software (e.g., R) that can fit nonlinear models to data.

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### Supplementary Information

R code for all figures and analyses.

### References

Bacon, P., W. Gurney, W. Jones, I. McLaren, and A. Youngson. 2005. Seasonal growth patterns of wild juvenile fish: Partitioning variation among explanatory variables, based on individual growth trajectories of Atlantic salmon (*Salmo salar*) parr. Journal of Animal Ecology 74:1–11. DOI: 10.1111/j.1365-2656.2004.00875.x

Baty, F., C. Ritz, S. Charles, M. Brutsche, J.-P. Flandrois, M.-L. Delignette-Muller. 2015. A toolbox for nonlinear regression in R: The package nlstools. J. Stat. Software 66(5):1-21. DOI: [10.18637/jss.v066.i05](http://dx.doi.org/10.18637/jss.v066.i05)

Bayley, P. 1988. Factors affecting growth rates of young tropical floodplain fishes: Seasonality and density-dependence. Environmental Biology of Fishes 21:127–142. DOI: 10.1007/BF00004848

Beguer, M., S. Rochette, M. Giardin, and P. Boet. 2011. Growth modeling and spatio-temporal variability in the body condition of the estuarine shrimp *Palaemon longirostris* in the Gironde (Sw France). Journal of Crustacean Biology. 31:606-612. DOI: [10.1651/10-3376.1](http://dx.doi.org/10.1651/10-3376.1)

Bertalanffy, L. von. 1938. A quantitative theory of organic growth (inquiries on growth laws II). Human Biology 10:181–213.

Beverton, R. J. H., and S. J. Holt. 1957. On the dynamics of exploited fish populations. United Kingdom Ministry of Agriculture; Fisheries, 533 p. DOI: 10.1007/978-94-011-2106-4

Carmona-Catot, G., A. Santos, P. Tedesco, and E. Garcia-Berthou. 2014. Quantifying seasonality along a latitudinal gradient: From stream temperature to growth of invasive mosquitofish. Ecosphere 5:1–23. DOI: 10.1890/ES14-00163.1

Chatzinikolaou, E. and C.A. Richardson. 2008. Population dynamics and growth of *Nassarius reticulatus* (Gastropoda: Nassariidae) in Rhosneigr (Anglesey, UK). Marine Biology 153:605-619. DOI: 10.1007/s00227-007-0835-5

Garcia-Berthou, E., G. Carmona-Catot, R. Merciai, and D. H. Ogle. 2012. A technical note on seasonal growth models. Reviews in Fish Biology and Fisheries 22:635–640. DOI: 10.1007/s11160-012-9262-x

Haddon, M. J. 2011. Modelling and quantitative methods in fisheries. Second edition. Chapman & Hall/CRC, Boca Raton, FL, 449 p.

Harwood, L., M. Kingsley, and T. Smith. 2014. An emerging pattern of declining growth rates in belugas of the Beaufort Sea: 1989-2008. Arctic 67:483–492. DOI: [10.14430/arctic4423](http://dx.doi.org/10.14430/arctic4423)

Hoenig, N., and R. Choudaray Hanumara. 1982. A statistical study of a seasonal growth model for fishes. Technical Report, Department of Computer Sciences; Statistics, University of Rhode Island.

Huusko, A., A. Maki-Petays, M. Stickler, and H. Mykra. Fish can shrink under harsh living conditions. Functional Ecology 25:628-633.

Nickelson, T. E., and G. L. Larson. 1974. Effect of weight loss on the decrease of length of coastal cutthroat trout. The Progressive Fish-Culturist 36:90-91.

Ogle, D.H., 2016a. Introductory Fisheries Analysis with R. Chapman & Hall/CRC Press, Boca Raton, FL.

Ogle, D.H., 2016b. FSA: Fisheries stock analysis. Available from: https://cran.r-project.org/web/packages/FSA/.

Pauly, D., M. Soriano-Bartz, J. Moreau, and A. Jarre-Teichmann. 1992. A new model accounting for seasonal cessation of growth in fishes. Australian Journal of Marine and Freshwater Research 43:1151–1156. DOI: 10.1071/MF9921151

R Development Core Team. 2016. R: A Language and Environment for Statistical Computing, v3.3.0. R Foundation for Statistical Computing, Vienna, Austria.

Schnute, J., and D. Fournier. 1980. A new approach to length-frequency analysis: Growth structure. Canadian Journal of Fisheries and Aquatic Sciences 37:1337–1351. DOI: 10.1139/f80-172

Somers, I. F. 1988. On a seasonally oscillating growth function. Fishbyte - Newsletter of the Network of Tropical Fisheries Scientists 6(1):8–11.

Stewart, J., W. Robbins, K. Rowling, A. Hegarty, and A. Gould. 2013. A multifaceted approach to modelling growth of the Australian bonito, *Sarda australis* (Family Scombridae), with some observations on its reproductive biology. Marine and Freshwater Research 64:671–678. DOI: 10.1071/MF12249

Table 1. Parameter estimates and Akaike Information Criterion (AIC) values from the fits of Equation 2 and Equation 3 to the Australian bonito and three sites of Eastern mosquitofish data. The lower AIC between the two equations for the same dataset is boldfaced.

|  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |  |
| --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- | --- |
| Equation 2 (Somers (1988) function) | | | | | | |  | | | Equation 3 (Pauly et al. (1992) function) | | | | | | | | |
|  | Australian | Mosquitofish | | |  | | |  | | | | Australian | | Mosquitofish | | |
|  | Bonito | Site 2 | Site 4 | Site 9 | |  | | |  | | bonito | | Site 2 | | Site 4 | Site 9 | |
|  | 71.9  (59.6,125.8) | 35.9  (34.5,37.6) | 46.0  (40.1,56.2) | 41.6  (39.1,44.9) | |  | | |  | | 71.7  (58.5,124.7) | | 35.1  (33.8,36.8) | | 44.0  (38.9,57.6) | 47.0  (42.4,57.0) | |
|  | 0.27  (0.09,0.46) | 2.01  (1.68,2.35) | 1.05  (0.63,1.57) | 1.31  (1.00,1.71) | |  | | |  | | 0.31  (0.10,0.76) | | 4.64  (3.25,6.70) | | 1.60  (0.85,2.58) | 0.77  (0.51,1.12) | |
|  | -1.9  (-3.0,-1.2) | -0.02  (-0.04,-0.01) | -0.20  (-0.28,-0.14) | -0.21  (-0.30,-0.15) | |  | | |  | | -1.6  (-2.8,-0.7) | | 0.43  (0.35,0.50) | | 0.07  (-0.04,0.18) | -0.41  (-0.49,-0.19) | |
|  | 0.09  (0.00,0.19) | 0.88  (0.87,0.89) | 0.75  (0.72,0.78) | 0.72  (0.66,0.76) | |  | | |  | | 0.09  (0.00,0.17) | | 0.92  (0.91,0.93) | | 0.76  (0.69,0.79) | 0.61  (0.55,0.65) | |
|  | 1.00a  (0.44,1.00) | 1.95  (1.82,2.05) | 1.28  (1.13,1.44) | 0.62  (0.46,0.80) | |  | | |  | | 0.13  (0.00,0.49) | | 0.43  (0.37,0.48) | | 0.26  (0.16,0.45) | 0.00  (0.00,0.26) | |
|  |  |  |  |  | |  | | |  | |  | |  | |  |  | |
| AIC | 1435.9 | **4159.4** | 4070.6 | **4995.8** | |  | | | AIC | | **1435.4** | | 4175.4 | | **4059.9** | 5018.4 | |

a was constrained to be less than or equal to 1 during model fitting.

### Figure Labels

Figure 1. Example VBGF using Equation 2 with =30, =0.3, =-0.1, =0.05 (with =0.55) and four different values of .

Figure 2. Example VBGF using Equation 3 with =30, =0.35, =-0.1, =0.3, and =0.05 (with =0.55). Each is shown by a gray point, by a vertical arrow, and no-growth period by the horizontal interval centered on the arrow and the gray region that extends to the x-axis. The ages adjusted for the (i.e., ) are shown above the x-axis.

Figure 3. Fork lengths at age for Australian Bonito (A) and standard lengths at age for Eastern mosquitofish at Sites 2 (B), 4 (C), and 9 (D) with the best-fits of Equation 3 (black line) and Equation 2 (gray line) superimposed. Parameter estimates (and 95% confidence intervals) from the model fits are shown in Table 1.