Revisiting the von Bertalanffy Seasonal Cessational Growth Function of Pauly et al. (1992)

Derek H. Ogle

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# Version notes

1. Likely will not include simulated data results at the bottom. They are a weak test at best.
2. Would like to add 1-3 more examples of real data with varying seasonal effects (clear cessation (like Bonito), clear cessation with decline in mean length-at-age, and no clear cessation).
3. The citation styles are not correct in this automated version.

# Introduction

The mean length-at-age for many fish (Haddon 2011) and other aquatic animals (e.g., (**???**); Harwood et al. (2014)) is often modeled with the von Bertalanffy growth function (VBGF; von Bertalanffy (1938)). The parameterization of the VBGF attributable to Beverton and Holt (1957) is most common and may be expressed as

with

where is the expected or average length at time (or age) , is the asymptotic mean length, is a measure of the exponential rate of approach to the asymptote (Schnute and Fournier 1980), and is the theoretical time or age (generally negative) at which the mean length 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)). Equation 2 of the traditional VBGF has been modified, usually with a sine function, to model these seasonal oscillations in growth. The most popular of these modifications, from Hoenig and Choudaray Hanumara (1982) and Somers (1988), is

where modulates the amplitude of the growth oscillations and corresponds to the proportional decrease in growth at the depth of the oscillation (i.e., "winter"), and is the time between time 0 and the start of the convex portion of the first sinusoidal growth oscillation (i.e., the inflection point). If =0, then there is no seasonal oscillation and Equation 3 reduces to Equation 2 and the typical VBGF (Figure 1). If =1, then growth completely stops once a year at the "winter-point" (), whereas values of 0<<1 result in reduced, but not stopped, growth during the winter (Figure 1). Note that because the sine function in Equation 3 has a period (i.e., the growth period) of one year. Some confusion has surrounded the use of Equation 3, although Garcia-Berthou et al. (2012) carefully clarified its form.

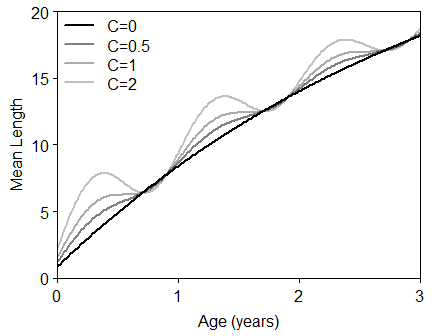


Figure 1: Example VBGF using Equation 3 with =30, =0.3, =-0.1, =0.05 (with =0.55) and four different values of .

Values of >1 (or <0) in Equation 3 allow seasonal decreases in mean length-at-age (Figure 1). A decrease in mean length is unlikely for organisms whose skeletons largely preclude shrinkage (Pauly et al. 1992), although a seasonal decrease in mean length-at-age is possible if size-dependent overwinter mortality occurs (Garcia-Berthou et al. 2012). Pauly et al. (1992) modified Equation 3 to include a true seasonal no-growth period where mean length was not allowed to decrease and smoothly transitioned into and out of the no-growth period. Specifically, their modification is

where 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, Pauly et al. (1992) noted that the units of changed from in Equation 3 to in Equation 4. To eliminate confusion, they suggested using in Equation 4, as we do here.

Pauly et al. (1992) devised Equation 4 by assuming =1 and replacing in Equation 3 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 3 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).

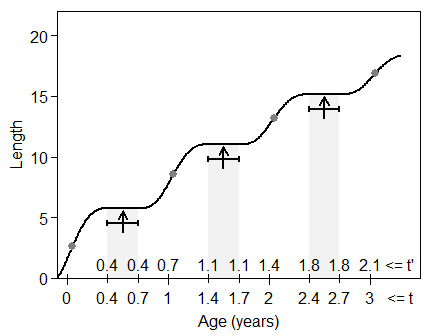


Figure 2: Example VBGF using Equation 4 with =30, =0.35, =-0.1, =0.3, and =0.05 (with =0.55). Each is shown by a gray point, by a vertical arrow, and no-growth period by the horizontal interval centered on the arrow and the gray region that extends to the x-axis. The ages adjusted for the (i.e., ) are shown above the x-axis.

Pauly et al. (1992) provided a "diskette" that contained a computer program to estimate the parameters of Equation 4. The diskette is difficult (at best) to obtain 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 description of how was operationalized. This is an important step in using Equation 4 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 (open-source) algorithm for the calculation of and Equation 4 for use in model fitting, and (iii) demonstrate the use of this algorithm.

# 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 algorithm may be used to convert from observed ages () to ages adjusted for cumulative prior to age ().

1. Shift the age () by subtracting the start of the no-growth () period (i.e., ) 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 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. This algorithm for computing is implemented in an R (R Development Core Team 2016) function as shown in Appendix 1. With this, Equation 4 is easly implemented as an R function as shown in Appendix 2.

# Fitting the Function

## Bonito Data

Stewart et al. (2013) examined the growth of 215 Bonito (*Sarda australis*) sampled from commercial landings. Fork lengths (mm) were measured for each fish and decimal ages were recorded as the number of opaque zones observed on otolith thin sections plus the proportion of the year after the designated birthdate (see Stewart et al. (2013) for more detailed methods). Stewart et al. (2013) fit Equation 3 to these data but constrained to not exceed 1. Their model fit resulted in the boundary condition of , which suggested that Bonito ceased to grow at *at least* one point. This result suggests that Equation 4 may fit these data better and allow the length of the no-growth period to be estimated (Pauly et al. 1992).

Equation 4 fit the Bonito data slightly better than Equation 3 with slightly lower residual sums-of-squares (RSS) and Akaike Information Criterion (AIC) values. The length of the no-growth period was estimated to be 0.133 or 13.3% of the year. The parameters were equal and the parameters were similar, but the parameters differed somewhat between the two models (Table 1). The from Equation 3 was equal to from Equation 4 multiplied by (Table 1). Graphically, there was little perceptual difference in the model fits (Figure 3).

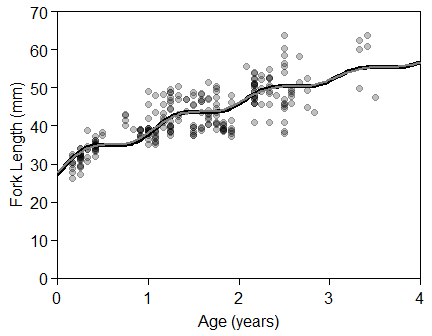


Figure 3: Fork length at age for Australian Bonito with the best-fit of Equation 4 (black line) and Equation 3, with fixed C=1, superimposed (gray line). The parameter estimates (and 95% confidence intervals) from the model fits are shown in Table 1.

Table 1: Parameter estimates, residual sums-of-squares (RSS), and Akaike Information Criterion (AIC) from the fits of Equation 3 and Equation 4 to the Bonito data.

Somers Model (Equation 3) fit

Est 2.5% 97.5%  
Linf 71.86 58.53 230.11  
K 0.27 0.04 0.51  
t0 -1.92 - -0.75  
C 1.00 - -  
ts 0.09 - 0.18  
RSS 4274.05 - -  
AIC 1435.86 - -

Modified Pauly Model (Equation 4) fit

Est 2.5% 97.5%  
Linf 71.68 58.46 223.03  
Kpr 0.31 0.05 0.66  
t0 -1.64 - -  
ts 0.09 - 0.18  
NGT 0.13 - 0.40  
RSS 4265.70 - -  
AIC 1435.37 - -

## Mosquitoefish Data

## Simulated Data

Simple simulated data are used initially to demonstrate the fitting of Equation 4. The data were simulated by randomly selecting 200 real numbers from a uniform distribution between 0 and 5 to serve as observed ages, plugging these ages into Equation 4 to compute a mean length for each age, and then adding a random deviate from a normal distribution with a mean of 0 and a standard deviation of to each mean length to simulate observed individual lengths. Eight data sets were simulated for all combinations of "early" () and "late" () maximum growth points, "short" () and "long" () no-growth periods, and "low" () and "high" () individual variabilities.

The nls() function from R was then used to estimate parameter values for the nonlinear Equation 4 fit to each simulated data set. The "port" algorithm was used so that and could be constrained to be positive and and could be constrained to be between 0 and 1. The estimated parameters for each data were consistently very close to the values used to create the data set, with the possible exception of . One example, for a "long" , "late" , and low individual variability, is shown in Figure 4. The results from fitting Equation 4 to these data sets are not surprising as one would expect the parameter estimates to be very close to the values used to create them, given that Equation 4 was used to both create and fit the data.

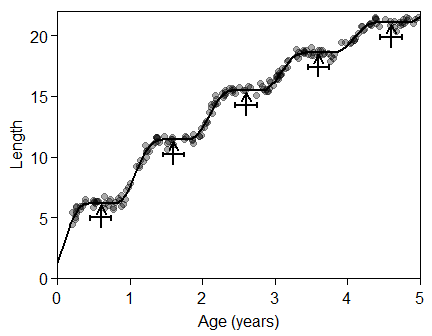


Figure 4: Example VBGF using Equation 4 with =30, =0.35, =-0.1, =0.2, =0.1 (with =0.55) and =0.3. Parameter estimates (and 95% confidence intervals) are shown in the first row of Table 2. Each is shown by the vertical arrow and each no-growth period is shown by the horizontal interval bar centered on the arrow.

Table 2: Parameter estimates from fitting Equation 4 to simulated data sets using =30, =0.35, =-0.1, and varying values of , , and .

tsset NGTset sigmaset tsest tslci tsuci NGTest NGTlci NGTuci Linfest  
[1,] 0.1 0.2 0.3 0.10 0.09 0.11 0.29 0.25 0.33 29.6  
[2,] 0.1 0.2 0.5 0.10 0.09 0.12 0.19 0.12 0.26 30.1  
[3,] 0.1 0.4 0.3 0.11 0.10 0.12 0.39 0.34 0.44 29.5  
[4,] 0.1 0.4 0.5 0.10 0.08 0.12 0.38 0.30 NA 31.0  
[5,] 0.5 0.2 0.3 0.51 0.50 0.52 0.15 0.10 0.19 30.0  
[6,] 0.5 0.2 0.5 0.50 0.48 0.52 0.14 0.06 0.23 30.0  
[7,] 0.5 0.4 0.3 0.49 0.48 0.50 0.41 0.36 0.46 31.2  
[8,] 0.5 0.4 0.5 0.51 0.49 0.52 0.44 0.36 0.52 30.5  
 Linflci Linfuci Kprest Kprlci Kpruci t0est t0lci t0uci  
[1,] 28.9 30.5 0.36 0.33 0.39 -0.09 -0.13 -0.06  
[2,] 29.1 31.3 0.35 0.31 0.40 -0.08 -0.13 -0.04  
[3,] 28.4 NA 0.35 0.32 0.40 -0.09 NA -0.05  
[4,] 29.0 33.6 0.32 0.26 0.39 -0.27 -0.39 -0.06  
[5,] 29.4 30.5 0.33 0.31 0.35 -0.13 -0.16 -0.10  
[6,] 29.1 31.1 0.33 0.29 0.37 -0.14 -0.20 -0.08  
[7,] 30.1 32.3 0.34 0.30 0.37 -0.10 -0.15 -0.05  
[8,] 29.1 32.3 0.37 0.31 0.43 -0.06 -0.14 0.02

# Discussion

* General
  + Fits perfectly simulated data well.
  + Parameters from real data seem reasonable
  + Other parameters by maths -- WP=ts+0.5, SNG=WP-NGT/2, K=Kpr\*(1-NGT)
  + Little practical difference between Equations 3 and 5 unless C>>1 and NGT>>0
* Model-Fitting
  + Fit Equation 3 first to see if C>=1
  + Problems due to 5 parameters
  + Bound parameters
* Assumptions
  + ts same time each year and age
  + NGT same length each year and age

# Acknowledgments

John Stewart (New South Wales Department of Primary Industries Fisheries) graciously provided the Bonito length-at-age data.

# Appendices

## Appendix 1

################################################################################  
## internal function to compute t-prime  
################################################################################  
iCalc\_tpr <- function(t,ts,NGT) {  
 ## Step 1  
 SNG <- ts+(1-NGT)/2  
 tmp.t <- t-SNG  
 ## Step 2 (in parentheses) and Step 3  
 tmp.t2 <- (tmp.t-floor(tmp.t))-NGT  
 ## Step 4  
 tmp.t2[tmp.t2<0] <- 0  
 ## Step 5 (in parentheses) and Step 6 (also returns value)  
 (floor(tmp.t)\*(1-NGT)+tmp.t2) + SNG  
}

## Appendix 2

################################################################################  
## Main Function  
## Linf, t0 as usual  
## Kpr = K-prime as defined in Pauly et al. (units are diff than usual K)  
## ts = start of sinusoidal growth (maximum growth rate)  
## NGT = "No Growth Time" = "fraction of a year where no growth occurs"  
## tpr = "t-prime" = actual age (t) minus cumulative NGT prior to t  
################################################################################  
  
VBSCGF <- function(t,Linf,Kpr=NULL,t0=NULL,ts=NULL,NGT=NULL) {  
 if (length(Linf)==5) { Kpr <- Linf[[2]]; t0 <- Linf[[3]]  
 ts <- Linf[[4]]; NGT <- Linf[[5]]  
 Linf <- Linf[[1]] }  
 tpr <- iCalc\_tpr(t,ts,NGT)  
 q <- Kpr\*(tpr-t0) +  
 (Kpr\*(1-NGT)/(2\*pi))\*sin((2\*pi)/(1-NGT)\*(tpr-ts)) -  
 (Kpr\*(1-NGT)/(2\*pi))\*sin((2\*pi)/(1-NGT)\*(t0-ts))  
 Linf\*(1-exp(-q))  
}

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