# AN INVESTIGATION OF INTEGRAL PROJECTION MODELS UTILIZING DENSITY-DEPENDENT SURVIVAL AND GROWTH IN GIZZARD SHAD

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Abstract. XXX

#### 1. Introduction

Gizzard shad are a common freshwater fish throughout the central and eastern portions of North America, occupying both lotic and lentic habitats (Pierce et al., 1981; Vanni et al., 2005). Within these areas, gizzard shad play a number of important roles in the freshwater community. First, young shad often serve as a critical food source for many fish species, including those of commercial and recreational importance (such as walleye and largemouth bass) (Jester et al., 1972). Second, because detritus serves as an important food source throughout much of gizzard shad development (i.e. from the young-of-the-year stage onward), these fish translocate nutrients from benthic regions into pelagic habitats (Mather et al., 1995; Schaus and Vanni, 2000; Vanni et al., 2005). This process can result in an increase in the nutrients available to organisms within the water column leading to increases in phytoplankton biomass, algal blooms, and, due to these conditions, shifts in freshwater community structure (Vanni et al., 2000; Aday et al., 2003). Finally, the fact that detritus comprises gizzard shad diets makes this species an important link between aquatic and terrestrial ecosystems through nutrient inputs (Vanni et al., 2000). Given its potentially important role in aquatic ecosystems, interest has intensified in understanding how gizzard shad populations respond to environmental changes (both natural and anthropogenic) and what these changes may mean for freshwater communities in general and fish assemblages in particular.

Although substantial empirical work on gizzard shad biology has accumulated over the decades, few if any studies have attempted to use these data to model the population dynamics of the particular species. Here, we introduce an integral projection model for gizzard shad based on empirical data, with density-dependent survival in age-0 fish, and compare model outcomes to dynamics reported for this fish species in the La Grange Station along the Illinois River. The model itself could be an important tool for predicting gizzard shad population responses to changing environmental conditions, including those mediated through species invasions (i.e. silver and bighead carp).

#### 2. Model Development

## 2.1. Gizzard shad life history.

2.2. **Equations.** We use an integral projection model to describe the life history of gizzard shad. Integral projection models have been used to describe a wide range of organisms. Ellner and Rees references. We assume that variations among individual gizzard shad can be completely summarized by its length z (in cm) ranging from the minimum possible length L to the maximum value U. The state of the population at time t (in years) is described by the length distribution n(z,t). Technically, for each time t, n(z,t) is a smooth function of z such that the number of individuals

of length z in the interval [a, b] at time t is  $\int_{a}^{b} n(z, t) dt$ .

Between times t and t+1, individual gizzard shad may grow, die, and produce offspring that vary in length all depending on their current state. In many ecological settings one can partition the life cycle into two stages: growth and maturation and fecundity/reproduction. We define two

functions P(z',z), representing survival and growth, and F(z',z) representing the production of new recruits. In both cases, z is the length at time t and z' is the length at time t+1. Note, For an individual of length z at time t,  $P(z',z)\Delta z$  is the probability that the individual is alive at time t+1, and its size is in the interval  $[z',z'+\Delta z]$  (as with n(z,t) this is an approximation that is valid for small Deltaz, and the exact probability is given by an integral like the one above). Similarly,  $F(z',z)\Delta z$  is the number of new offspring in the interval  $[z',z'+\Delta z]$  present at time t+1, per length-z individual at time t.

2.2.1. Growth and survival. We assume that P(z'z) = s(z)G(z',z) where s(z) is the survival probability and G(z',z) describes the length transitions. The survival function is a logistic function with four parameters: the minimum survival rate  $s_{\min}$ ; a maximum survival rate,  $s_{\max}$ ; and intercept parameter,  $\alpha_s$ ; and a slope parameter,  $\beta_s$  (Bolker 2008)

(1) 
$$s(z) = s_{\min} + \frac{s_{\max} - s_{\min}}{1 + e^{\beta_s(\log(z) - \log(\alpha_s))}}.$$

The growth function is a two-variable normal distribution centered around a modified von Bertalanffy function of the length at time t. The von Bertalanffy equation, commonly used to describe the length of a fish over time, is given by  $z(t) = L_{\infty} \left(1 - e^{-r(t-t_0)}\right)$  where  $L_{\infty}$  is maximum asymptotic length, r is the growth rate, and  $t_0$  is the initial time. The expected length in the next year

$$z' = z(t+1) = L_{\infty} \left( 1 - e^{-r(t+1-t_0)} \right) = L_{\infty} - L_{\infty} e^{-r(t-t_0)} e^{-r} = L_{\infty} \left( 1 - e^{-r} \right) + z(t)e^{-r}.$$

We assume that

(2) 
$$G(z',z) = \text{Prob}(z'|z, L_{\infty}, r, \sigma_g) = \text{NormalPDF}(\mu_g, \sigma_g)$$

where  $\mu_g = L_{\infty} (1 - e^{-r}) + z(t) e^{-r}$  and  $\sigma_g$  is the standard deviation.

2.3. **Fecundity.** The fecundity/recruitment kernel,

(3) 
$$F(z',z) = p_b(z) \operatorname{egg}(z) \nu s_0(n(z,t)) C_1(z',z)$$

where  $p_b(z)$  is the probability of reproducing, egg(z) is the mean number of eggs produced,  $\nu$  is the probability that an egg is viable,  $s_0(n(z,t))$  is the density-dependent probability of surviving to age-1, and  $C_1(z',z)$  is the length distribution of new recruits at age-1 (when they are first censused).

In traditional matrix population models, the state vector is multiplied on the left by a matrix to project the current measurement to its future value. In an integral projection model, the projection matrix is replaces by an integral kernel K(z',z) that projects the population forward in time. The population at time t+1 is the sum of the contributions from each individual alive at time t,

(4) 
$$n(z',t+1) = \int_{L}^{U} K(z',z)n(z,t) dz,$$

where K(z',z) = G(z',z) + F(z',z) and [U,L] is the range of all possible lengths.

#### 3. Methods and Analysis

3.1. **Determination and Parameterization of Function.** The size-dependent probability of females spawning,  $p_b(z)$ , is estimated from a logistic regression with a logit link function scaled to maximum of 90% (Figure 1a). We apply the inverse of the logit transformation to write this probability as  $p_b(z) = (1 + e^{-\nu_f})^{-1}$ , where the estimated linear predictor  $\nu_f = \alpha_s + \beta_s z$ . The mean eggs produced per female is gleaned from a linear approximation of the data from Jons & Miranda (1997). Specifically we assume that  $\text{egg}(z) = \beta_e z$  for z > 140 and zero otherwise(Figure ??).

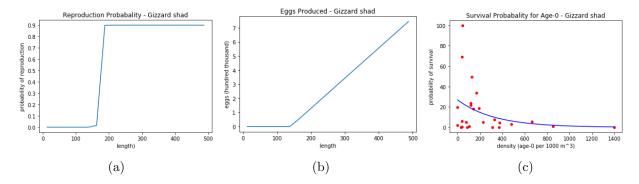


FIGURE 1. (a) Graph of  $p_b(z)$  (b) Graph of egg(z), (c) Density-dependent survival of age-0 gizzard shard.

Table 1. A summary of parameters, their biological meaning, and source for mean values.

Parameter	Meaning (units)	Mean Value	Source
$\alpha_s$	intercept for linear regression for probability	-73.954	Bodola(1955)
	of spawning		
$\beta_s$	slope for linear regression for probability of	0.430	Bodola(1955)
	spawning		
$\beta_e$	eggs per female	15000/7	Jons & Mi-
			randa (1997)
$\nu$	viability of eggs	0.002	
$A_s$	intercept for exponential model for age-0 sur-	0.289	Michaletz(1997)
	vival		
$r_s$	decay rate for exponential model for age-0 sur-	0.003	Michaletz(1997)
	vival		
$\mu_c$	mean length of age-0	112	
$\sigma_c$	standard deviation of age-0	40	

The survival to age-0 is assumed to be dependent on the density of age-0 gizzard shad (Figure 1c). We assume that the probability of survival in age-0 fish is approximated by the exponentially decaying function

(5) 
$$s_0(n(z,t)) = A_s e^{-r_s d(n,t)}$$
 where the density  $d(n,t) = 10^{-3} \int_L^U \nu \, \text{egg}(z) p_b(z) n(z,t) \, dz$ . To complete the recruitment process we assign a length to t

To complete the recruitment process we assign a length to the recruited individuals by simulating a Gaussian random variable with mean  $\mu_c$  and standard deviation  $\sigma_c$ .

3.2. Simulations. We numerically solved our integral model using the mid-point rule with large approximating matrices (Burden and Faires, 2005). The mid-point rule has been commonly used for integral projection models because of its simplicity and effectiveness (Ellner and Rees, 2006; Ramula et al., 2009; Merow et al., 2014). During the course of model development, we explored different step sizes for the mid-point rule point rule and found that about XXXX points provided numerically stable results. We integrated over lengths from 0.01 cm to 500 cm. The upper limit was chosen based upon numerical stability and consistency of the system (e.g., avoiding ?eviction? or the loss of individuals due to numerical errors; Williams et al., 2012).

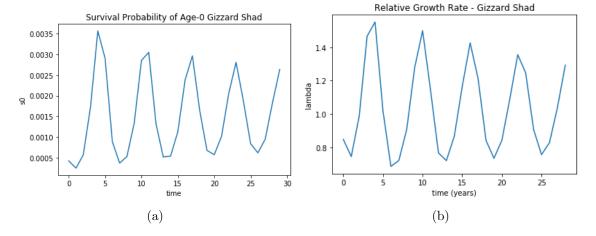


FIGURE 2. (a) survival of age-0, (b) Relative growth rate of gizzard shad density

We coded our model as python and published scripts on JP bithub page XXXXXX.

3.2.1. Relative Growth. The dependence on survival of next generation of age-0 fish on the present density of age-0 fish strongly influences the density, at all ages, of gizzard shad within the population. When the fish density is large, there may be more fish at longer lengths and consequently a greater number of eggs produced. More eggs leads to a higher density of age-0 fish and reflectively a reduction in the survival to age-1. If this reduced survival continues for a few years, the overall density of fish may decline and there may not be as many larger fish reproducing. If fewer eggs are spawned, there are less age-0 fish and the reduced density leads to a better survival probability. For these years we would expect the overall density to increase. This cycle of oscillation continues, reflected in our model by the time-dependent survival probability of age-0 (Figure 2a) and the relative growth rate

$$\lambda(t) = \int_{L}^{U} n(z, t+1) dz / \int_{L}^{U} n(z, t) dz$$

(Figure 2b).

## 4. Results

4.1. Length Distribution Over Time. We compared simulated length distributions with USGS LTRM data from La Grange reach. Selecting years with sample sizes greater than 200 observations limited the LTMR dataset to the years 2002 and 2005 (see Figure 3a). In both years, the length of the sampled gizzard shad was organized into 11 bins of equal width  $\Delta z = 30$ cm starting with 60-90cm. We used the 2002 length distribution to define the initial distribution  $n(z_i, 0)$ , i = 1, ..., 11, as the observed values per  $\Delta z$ . Solving equation (4) with parameter values summarized in Table for 3 years resulted in a relative distribution that can be compared with the observed values in 2005 (Figure 3b).

# 4.2. **Total Population Size vs. Time.** Graphs of n total n vs t and compare with Pool 26

#### 5. Discussion

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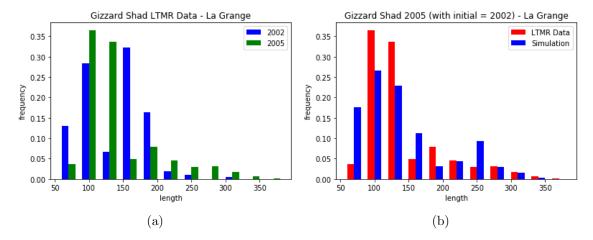


FIGURE 3. (a) Relative frequency of length from LTMR data in the years 2002 and 2005, (b) A comparison length frequency in 2005 from LTRM data and simulated frequency from the model with initial condition given by LTRM data in 2002.

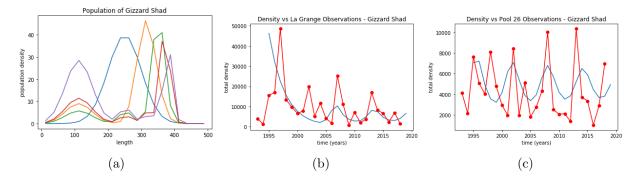


FIGURE 4. (a) length distribution over time, (b-c) Density of gizzard shad vs La Grange reach and Pool 26 observations