

Figure 1: Life cycle diagram and census points for pre-reproduction census of gizzard shad.

Graphical Abstract

- An integral projection model for gizzard shad (Dorosoma cepedianum) utilizing
- density-dependent age-0 survival
- ⁴ J. Peirce^{1,4}, G. Sandland^{3,4}, B. Bennie¹, R.A. Erickson²

5

- 6 Highlights
- An integral projection model for gizzard shad (Dorosoma cepedianum) utilizing
- $_{8}$ density-dependent age-0 survival
- ⁹ J. Peirce^{1,4}, G. Sandland^{3,4}, B. Bennie¹, R.A. Erickson²
- Research highlight 1
- Research highlight 2

An integral projection model for gizzard shad (*Dorosoma cepedianum*) utilizing density-dependent age-0 survival

J. Peirce^{1,4}, G. Sandland^{3,4}, B. Bennie¹, R.A. Erickson²

1 University of Wisconsin - La Crosse, Mathematics & Statistics Department
2 U.S.G.S. Upper Mississippi Environmental Science Center
3 University of Wisconsin - La Crosse, Biology Department
4 River Studies Center

$_{19}$ Abstract

12

13

14

15

16

17

18

Gizzard shad (*Dorosoma cepedianum*) are a common freshwater fish throughout the central and eastern portions of North America. Within these areas, gizzard shad play a number of critical roles in the freshwater community. Because of this, it is important that we understand how gizzard shad populations respond to environmental changes and what these changes may mean for aquatic communities in general and fish assemblages in particular. Here we introduce an integral projection model for gizzard shad based on empirical data and include density-dependent survival in age-0 fish. Integral projection models (IPM) are a generalization of stage-based, matrix population models that have been used to describe a wide range of organisms. IPMs are a natural choice for gizzard shad since many aspects of their life cycle have been studied. In this paper, we compared model outcomes to empirical patterns reported for this fish species at a key location along the Illinois River. Results of our work suggest that this model could serve as an important tool for predicting gizzard shad population responses to changing environmental conditions, including those mediated through species invasions.

- 20 Keywords: population dynamics, fisheries, Mississippi River basin, population ecology,
- 21 invasive species impact

2 1. Introduction

- Gizzard shad (*Dorosoma cepedianum*) is a laterally compressed, deep-bodied fish species
- that occupies numerous aquatic systems throughout central, southern and eastern regions
- of the United States (Pierce et al., 1981; Vanni et al., 2005). In more eutrophic habitats,

such as reservoirs, gizzard shad can reach high abunadances and can come to dominate fish assemblages. Because of this, gizzard shad have the potential to influence freshwater systems in a number of ways. 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). Thus, this species can serve as an important trophic link within aquatic food webs. Second, because detritus can serve as a primary food source throughout much of gizzard shad development (i.e. from the age-0 stage onward), these fish can transport 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 (Aday et al., 2003; Schaus and Vanni, 2000). Finally, the fact that 37 detritus can comprise a substantial portion of gizzard shad diet also makes this species an important connector between terrestrial inputs and aquatic processes (Schaus and Vanni, 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.

Interactions within and between species in interconnected environments can have important consequences for fish populations across space and time (Thorp et al., 2006). For gizzard shad, previous work has suggested that fish densities can play an important role in the length-distribution patterns observed in populations. For example, (Buynak et al., 1992) and (Welker et al., 1994) reported inverse relationships between densities and lengths in age-0 gizzard shad under both field and semi-natural conditions. These patterns were attributed to intraspecific competition among young shad for prey (zooplankton) resources. Although intraspecific competition may be occurring in subsequent stages of gizzard shad development, little work has actually been conducted to address this (DiCenzo et al., 1996). There is also evidence that the densities of other co-occurring fish species (such as invasive carps) can negatively influence aspects of gizzard shad biology, such as body condition (Irons et al., 2007; Love et al., 2018).

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 gizzard shad. Work by Catalano and Allen (2010, 2011) used empirically-based simulations of gizzard shad and focused on population-level responses. This work did examine population-structure by examining length and did not explore different sources of density feedback on the population dynamics. Here, we introduce an integral projection model for gizzard shad based on empirical data with density-dependent survival in age-0 fish. We then 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).

67 2. Model Development

68 2.1. Gizzard shad life history

Mating within this species can be temperature-dependent, but tends to occur between
May and June. Males and females aggregate and then broadcast gametes into the surrounding water; fertilized eggs then settle and adhere to the bottom substrates. After a period of
days, eggs hatch and fish develop from the larval stage to juveniles and eventually to adults.
In many habitats, individuals can reach sexual maturity within a year. As gizzard shad
mature, their diet preferences typically shift from phytoplankton and zooplankton early in
development to detritus and zooplankton as adults. Given the large number of eggs produced by shad females (> 300,000/year), there is evidence that intraspecific competition
can be intense during early developmental stages in this species. However, the strength
of competition can subside as fish transition to different food types during latter stages of
development.

80 2.2. Equations

We used an integral projection model to describe the life history of gizzard shad in the Upper Mississippi River (UMR) system. Integral projection models (IPM) were introduced by Easterling (Easterling et al., 2000), as a generalization of stage-based, matrix population

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

Parameter	Meaning (units)	Mean	Source
Logistic survival probability func-			
tion, $s(z)$			
$s_{ m min}$	minimum survival	0.002	(Bodola, 1955)
$s_{ m max}$	maximum survival	$1 - 8.871K^{0.73}L_{\infty}^{-0.33}$	THEN 2015
α_s	inflection point	104.34	Estimated from LTRM dataset
eta_s	slope	-637.93	Estimated from LTRM dataset
Growth function, $G(z, z')$			
L_{∞}	maximum length (in mm)	394.30	(Catalano and Allen, 2010)
K_g	growth rate	0.26	(Michaletz, 2017)
σ_g	growth standard deviation	25	(Michaletz, 2017)
Normal distribution of length of age-			
$1, C_1(z')$			
μ_r	mean length of recruitment (in	105	(Michaletz, 2017)
	mm)		
σ_r	standard deviation of length	25	(Michaletz, 2017)
Eggs produced, $egg(z)$			
$\mathrm{egg}_{\mathrm{max}}$	maximum number of eggs pro-	742,094	Estimated from (Jons and Mi-
	duced		randa, 1997)
α_e	inflection point	314.44	Estimated from (Jons and Mi-
			randa, 1997)
eta_e	slope	-7.12	Estimated from (Jons and Mi-
			randa, 1997)
Survival of age-0, $s_0(d(t))$			
a_0	intercept	0.27	Estimated from (Michaletz, 2010)
b_0	decay rate	0.003	Estimated from (Michaletz, 2010)
Spawning			
ν	probability that egg becomes vi-	0.002	(Bodola, 1955)
	able		
p_b	probability that female spawns	0.90	

models. IPMs have been used to describe a wide range of organisms (Ellner et al., 2016;
Merow et al., 2014; Rees et al., 2014) and have only recently be used to model fish populations
(Erickson et al., 2017; Liao et al., 2019; White et al., 2016; Pollesch et al., 2022). The
availability of empirical size observation to aid with model parameterization makes IPM
a natural choice for gizzard shad since many aspects of their life cycle have been studied.
Specifically, functions used in our model incorporate data from studies on egg production
and adult size, survival of the age-0 stage and density, and nearly thirty years of length
measurements in the main channel of the Upper Mississippi River system.

We assumed that variations among individual gizzard shad can be summarized by its length z (in mm) 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). Specifically, 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 $\int_a^b n(z,t) \, dz$.

Between times t and t+1, individual gizzard shad may grow, die, and produce offspring 97 that vary in length depending on the individuals current length (Figure ??). At time t+1 the population will have a length distribution defined by n(z, t+1). For our model, we partition 99 the life cycle of gizzard shad into two stages: 1) survival and growth, and 2) reproduction. 100 For an individual of length z at time t, $P(z',z)\Delta z$ is the probability that the individual 101 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 102 approximation that is valid for small Δz , and the exact probability is given by an integral 103 like the one above). Similarly, $F(z',z)\Delta z$ is the number of new offspring in the interval 104 $[z', z' + \Delta z]$ present at time t + 1, per length-z individual at time t. 105

2.2.1. Growth and survival

We define P(z'z) = s(z,T)G(z',z) where s(z) is the adult annual survival probability and G(z',z) describes the annual length transitions. We assumed that the survival function is a logistic function,

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

with four parameters: the minimum survival rate s_{\min} ; a maximum survival rate, s_{\max} ; and intercept parameter, α_s ; and a slope parameter, β_s (Bolker, 2008).

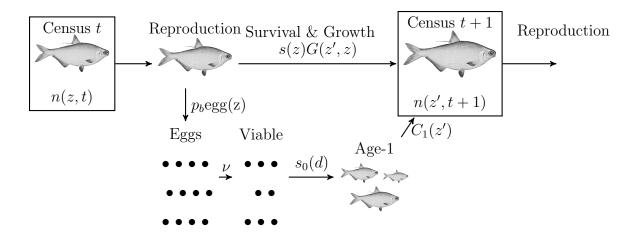


Figure 2: Life cycle diagram and census points for pre-reproduction census of gizzard shad.

We assumed that 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 $z(t) = L_{\infty} \left(1 - e^{-K(t-t_0)}\right)$ where L_{∞} is maximum asymptotic length, K is the growth rate, and t_0 is the initial time. The expected length in the next year

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

Consequently, we assumed that $G(z',z) = \text{Prob}(z' | z, L_{\infty}, K_g) = \text{NormalPDF}(\mu_g, \sigma_g)$ where K_g is the individual growth rate, $K_g = L_{\infty} \left(1 - e^{-K_g}\right) + z(t)e^{-K_g}$, and K_g is the standard deviation.

115 2.3. Fecundity

We define the fecundity kernel,

$$F(z',z) = p_b \exp(z) \nu s_0(n(z,t)) C_1(z')$$
(2)

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

We assumed that the mean number of eggs produced by females of a certain length is a three-parameter logistic function,

$$\operatorname{egg}(z) = \frac{\operatorname{egg}_{\max}}{1 + e^{\beta_e(\ln(z) - \ln(\alpha_e))}}.$$
 (3)

The probability of survival of gizzard shad during their first year may depend on many factors (Michaletz, 2010) such as mean temperature, mean total length, the present density of age-0 fish. In this study, we focus only on the density factor and define the probability of survival of age-0 fish as the exponential function

$$s_0(d(t)) = a_0 e^{-b_0 d(t)} (4)$$

where a_0 is the intercept, b_0 the decay rate, and d(t) is the density at time t of age-0 gizzard shad per 1000 m³,

$$d(t) = 10^{-3} \int_{L}^{U} p_b \operatorname{egg}(z) \nu n(z, t) dz.$$

Finally, after computing the total number of viable eggs produced and survive to age-1 fish, we multiplied this number with a normal distribution of length, $C_1(z') = \text{NormalPDF}(\mu_r, \sigma_r)$ where μ_r is the mean length of age-1 gizzard shad and σ_r is the standard deviation.

132 2.4. Dynamical Model

The population at time t+1 is the sum of the contributions from each individual alive at time t,

$$n(z', t+1) = \int_{L}^{U} K(z', z) n(z, t) dz,$$
 (5)

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

3. Methods

3.1. Study Area and LTRMP Data Collection

The 129 km long La Grange Reach is located between La Grange Lock and Dam (L&D) and Peoria L&D on the Illinois River, U.S., and is about midway between the Mississippi River, and Lake Michigan. The Illinois River is a major tributary of the Mississippi River, draining nearly two-thirds of the state of Illinois. Along with the main channel of the UMR,

the fish community of La Grange Reach has been monitored by the Long-term Resource
Monitoring Program (LTRMP) from 1990 to the present, with approximately 500 random
collections each year from 15 June to 31 October. The LTRMP fish collection methodology
included a multiple gear approach (netting and electrofishing) to monitor the general fish
community of the UMR system through time (Gutreuter et al., 1995). The total lengths were
recorded for all fishes captured. Methodology, protocols and modifications to the LTRMP
can be found in Gutreuter et al. (1995), and Ickes and Burkhardt (2002).

The location of La Grange Reach has two important features motivation its choice for the 149 methods of our study. First, we parameterized the IPM using data from the main channel 150 of the UMR. As a part of Illinois River and UMR, La Grange Reach is upstream from the 151 main channel (should we include figure/map?). The proximity to, but relative independence 152 from, the main channel, made it a good choice. Secondly, La Grange Reach is a large pool 153 between the main channel of the Mississippi River and Lake Michigan. In recent years, 154 there have been concerns with the threatening introduction of invasive carp to the Great 155 Lakes. Consequently, the impact of invasive carp on the native fish populations in the pools leading to the Great Lakes have received an elevated level of attention. Understanding the 157 population dynamics of gizzard shad, may in the future make it easier to assess the impacts 158 carp has on native fish populations. 159

160 3.2. Parameterization

3.3. Growth and survival of adults

The parameters for the growth function were chosen as the mean values published on a study of gizzard shad located in large impoundments (Michaletz, 2017). The survival rate of adults gizzard shad dependent on their length is not well documented and required us to make some additional modeling assumptions. An investigation of gizzard shad in Lake Eerie (Bodola (1955)) provided the minimum and maximum survival rate of adults.

Based on the observed length distributions of gizzard shad sampled from the main channel of the Mississippi River (Figure 3), we assumed that solutions of our model will exhibit periodic behavior every 8-9 years. We used a least squares method to estimate the α_s and β_s parameters that minimized the total square-distance between the (observed) pre-carp

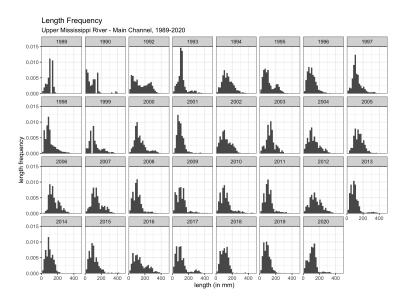


Figure 3: Length frequency of sampled gizzard shad in the main channel of the Upper Mississippi River system.

LTRMP length distribution in the main channels of the UMR (pools 4, 8, 13, 26, and the open river) and (predicted) model equilibrium, n(z,t) during a 8-year period 100 years after initialization. The slope parameter β_s was found to be large in magnitude resulting in a primarily two-valued survival probability. Gizzard shad less than α_s mm in length have a very low survival rate (s_{\min}) while lengths larger than α_s mm have the maximum survival rate s_{\max} . A two-valued survival probability model are common in fisheries model (GREG - CITATIONS)

3.3.1. Fecundity and recruitment

178

Maturity of female gizzard shad corresponds with lengths of approximately 140 mm and 179 the number of eggs produced increase as the females increase in size (Jons and Miranda, 180 1997). The logistic parameters for mean number of eggs produced by females of a certain 181 length were obtained by applying fitting a dose-response model to batch fecundity versus 182 length (Jons and Miranda, 1997) (Figure 4a). The survival to age-0 was assumed to be 183 dependent on the density of age-0 gizzard shad (Figure 4b). Parameters for the exponentially 184 decaying age-0 survival function were determined by fitting equation 4 to the survival means 185 for 2003-2007 cohorts of gizzard shard in five Missouri reservoirs (Michaletz, 2010). To 186 complete the recruitment process we assign a length to the recruited individuals by simulating 187

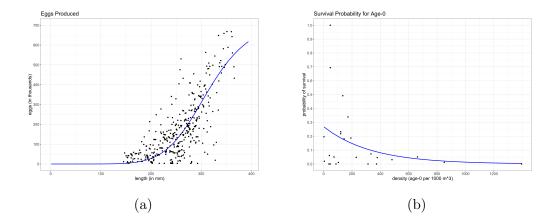


Figure 4: (a) Mean number of eggs produced by female gizzard shad egg(z). Data from (Jons and Miranda, 1997). (b) Age-0 density-dependent survival function $s_0(d)$. Data from (Michaletz, 2010).

a Gaussian random variable with mean μ_c and standard deviation σ_c . The parameters for the size distribution of age-1 fish were gleaned from a study of gizzard shad located in large impoundments (Michaletz, 2017) and the historic 1990-2020 LTRMP dataset of gizzard shad in La Grange Reach.

192 4. Analysis and Results

We numerically solved the integral model using the Midpoint Rule with large approxi-193 mating matrices (Burden and Faires, 2005). The Midpoint Rule has been commonly used 194 for integral projection models because of its simplicity and effectiveness (Ellner and Rees, 195 2006; Ramula et al., 2009; Merow et al., 2014). During the course of model development, we 196 explored different step sizes for the Midpoint Rule and found that about 50 points provided 197 numerically stable results. We integrated over lengths from 0 mm to 500 mm. The upper 198 limit was chosen based upon numerical stability and consistency of the system (e.g., avoiding 199 eviction or the loss of individuals due to numerical errors (Williams et al., 2012)). 200

4.1. Initial conditions

201

We assumed that the initial density of gizzard shad was $d_0 = 964.7$, the annual average density of gizzard shad observed in La Grange Reach from 1993-2019. The probability of an individual being length z at time t = 0 was assumed to be normally distributed with mean $0.5L_{\infty}$ and standard deviation $\sigma_0 = 30$ (similar to observations (1990-2020) in LTRMP fish

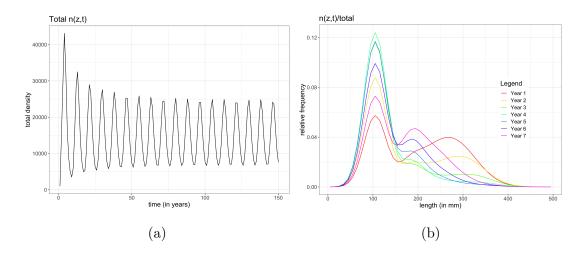


Figure 5: (a) The total number of gizzard shad in La Grange Reach predicted by IPM for first 150 years. (b) Simulated length distributions during an 8 year period of time (approximately 1 period of the total number function)

dataset). As a result, we initialized our model with length distribution

$$n(z,0) = d_0 \text{Norm}(0.5L_\infty, \sigma_0) = 964.7 \text{Norm}(166, 30).$$
 (6)

The model was coded in R (R Core Team, 2017) and the scripts are published on JP github page https://github.com/jppeirce.

209 4.1.1. Periodic orbit and comparison with the LTRMP La Grange dataset

The total number of gizzard shad in our simulation reached a stable periodic orbit (Figure 210 5a) within 50 years. The length distributions during the periodic orbit (Figure 5b) has 211 similarities to length observations from La Grange found in the LTRMP fish dataset (Figure 212 6a). In addition, the average simulated length distribution compares favorably to the length 213 frequencies of observations from La Grange. (Figure 6b). We notice that the peak frequencies 214 are near the same length with the model predicting slightly more adults lengths and fewer 215 juvenile lengths than the observations. This may be explained by gear bias. The methods 216 used to capture fish make it more likely to record longer lengths (> 200mm). Studies 217 [CITATION] suggest that due to the environmental stochasticity and other effects, smaller 218 recruitment fish densities can fluctuate annually and be difficult to measure accurately. While 219 our model uses age-0 density to effect age-0 survival, we assumed constant viability which 220 may be sensitive to the external factors mentioned above. In addition, the location of the

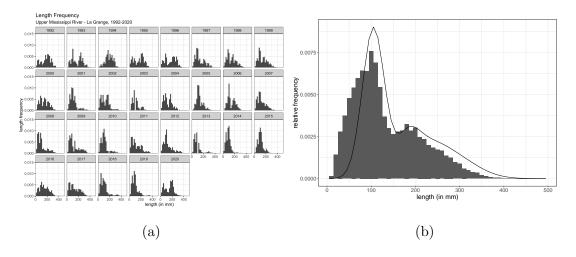


Figure 6: (a) Length frequencies of sampled gizzard shad in La Grange for each year, 1992-2020. (b) Length frequencies of sampled gizzard shad in La Grange from 1992-2020 compared with average (over 1 period) simulated length frequency.

maximum length and the variation in the of length of new recruits recorded in the LTRMP data, suggests that there may be smaller age-0 fish in La Grange Reach than in study location (Michaletz, 2017) used to parameterize the model.

5 4.1.2. Survival of Age-0 Cohort

The dependence on survival of next generation of age-0 fish on the present density of 226 age-0 fish strongly influences the density, at all ages, of gizzard shad within the population. 227 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 229 reflectively a reduction in the survival to age-1. If this reduced survival continues for a 230 few years, the overall density of fish may decline and there may not be as many larger fish 231 reproducing. If fewer eggs are spawned, there are less age-0 fish and the reduced density 232 leads to a better survival probability. For these years we would expect the overall density to 233 increase. This cycle of oscillation continues, reflected in our model by the time-dependent 234 survival probability of age-0 recruits (Figure 7). 235

236 4.2. Time Evolution of Initial State

Starting with n(z,0) defined by Equation 6, we computed the length distribution for 4 years. In year 1, there are two relative maximum frequencies corresponding to the recruit-

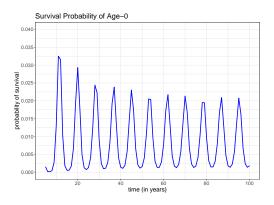


Figure 7

Figure 8: Survival probability of age-0 gizzard shad.

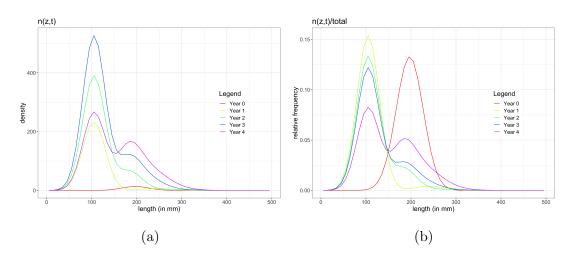


Figure 9: Simulated (a) density and (b) frequency of lengths of gizzard shad.

ment of age-1 fish and the survival of the adults. The greatest value of n(z,1) during year one is centered at the mean length of recruitment (Figure 9a). In the following years, the fluctuation in the maximum value is likely the result the density-dependent survival function for age-0 fish as explained in Section 4.1.2.

Figure 9a also demonstrates the length dynamics within the gizzard shad population. The peak length frequency of age-1 fish in n(z,1) near z=105 mm is seen as a peak in n(z,2) near 180 mm before becoming unindentifiable in the graph of n(z,3). This is consistent with the average lifespan of 4-6 years (CITATION).

5. Discussion

262

264

265

266

267

268

269

270

271

273

274

276

After fitting two adult survival parameters using LTRMP data from the main channel 248 if the Mississippi River, we compare a simulated length distribution with LTRMP data 249 collected from the La Grange Reach of the Illinois River in 1992-2020. Density dependence 250 in the age-0 cohort produced a range of simulated length distributions (Figure 5b) that 251 match well patterns in the LTRMP data from La Grange (Figure 6a). Specifically, the two peaks in Year 1 are similar to the observations made in 1997. In 1997, there is wider 253 range of mid-length and longer length gizzard shad and relative smaller amount of age-1 fish 254 recorded. In the next year, there was a higher frequency of age-1 fish in both the LTRMP 255 data and the simulations (Year 2). While not all observed length distributions are echoed in 256 the simulations, the similarity in the average length distributions (Figure 6b) suggest that our IPM may be a good prediction tool for future length distributions. (LAST SENTENCE 258 NEEDS A LOT MORE CLARITY) 259

- 260 1. Summary of findings and key discussion points
- (a) Comparison to empirical data [X]
 - (b) Deviations from empirical data
- (c) discussion about sameness and differences
 - (d) Sources of density within our model
 - 2. Comparison to existing literature
 - (a) Talk about Matt's work (Catalano and Allen, 2010, 2011)
 - (b) Broader need for models such as this
 - 3. Implications for management of species
 - (a) Invasive species
 - (b) Impact of size on harvest
 - (c) impact of size on movement
- 4. Future ideas to explore
 - (a) Multi-species model
 - (b) Spatial impacts
- (c) Climate on density
 - (d) Changing climate scenarios

277 6. Acknowledgments

These data are a product of the U.S. Army Corps of Engineer's Upper Mississippi River
Restoration Program (UMRR) Long Term Resource Monitoring (LTRMP) element implemented by the U.S. Geological Survey in collaboration with the five Upper Mississippi River
System (UMRS) states of Illinois, Iowa, Minnesota, Missouri, and Wisconsin. The U.S.
Army Corps of Engineers (Corps) provides guidance and has overall program responsibility.
We thank the U.S. Geological Survey Biothreats program and Great Lakes Restoration
Initiative for funding. In addition, research was supported by NSF-DMS Grant #1852224,
"REU Site: Ecological Modeling of the Mississippi River Basin".

286 References

- R. J. Pierce, T. E. Wissing, B. A. Megrey, Aspects of the feeding ecology of gizzard shad in
 Acton Lake, Ohio, Transactions of the American Fisheries Society 110 (3) (1981) 391–395.
- M. J. Vanni, K. K. Arend, M. T. Bremigan, D. B. Bunnell, J. E. Garvey, M. J. Gonzalez,
 W. H. Renwick, P. A. Soranno, R. A. Stein, Linking landscapes and food webs: effects of
 omnivorous fish and watersheds on reservoir ecosystems, BioScience 55 (2) (2005) 155–167.
- D. B. Jester, B. L. Jensen, et al., Life history and ecology of the gizzard shad, *Dorosoma* cepedianum (Le Sueur) with reference to Elephant Butte Lake.
- M. E. Mather, M. J. Vanni, T. E. Wissing, S. A. Davis, M. H. Schaus, Regeneration of nitrogen and phosphorus by bluegill and gizzard shad: effect of feeding history, Canadian Journal of Fisheries and Aquatic Sciences 52 (11) (1995) 2327–2338.
- M. H. Schaus, M. J. Vanni, Effects of gizzard shad on phytoplankton and nutrient dynamics: role of sediment feeding and fish size, Ecology 81 (6) (2000) 1701–1719.
- D. D. Aday, R. J. H. Hoxmeier, D. H. Wahl, Direct and indirect effects of gizzard shad on bluegill growth and population size structure, Transactions of the American Fisheries

 Society 132 (1) (2003) 47–56.

- J. H. Thorp, M. C. Thoms, M. D. Delong, The riverine ecosystem synthesis: biocomplexity in river networks across space and time, River Research and Applications 22 (2) (2006) 123–147.
- G. L. Buynak, R. S. Hale, B. Mitchell, Differential growth of young-of-year gizzard shad
 in several Kentucky reservoirs, North American Journal of Fisheries Management 12 (3)
 (1992) 656–662.
- M. T. Welker, C. L. Pierce, D. H. Wahl, Growth and survival of larval fishes: roles of competition and zooplankton abundance, Transactions of the American Fisheries Society 123 (5) (1994) 703–717.
- V. J. DiCenzo, M. J. Maceina, M. R. Stimpert, Relations between reservoir trophic state and gizzard shad population characteristics in Alabama reservoirs, North American journal of fisheries management 16 (4) (1996) 888–895.
- K. S. Irons, G. Sass, M. McClelland, J. Stafford, Reduced condition factor of two native fish species coincident with invasion of non-native Asian carps in the Illinois River, USA Is this evidence for competition and reduced fitness?, Journal of Fish Biology 71 (2007) 258–273.
- S. A. Love, N. J. Lederman, R. L. Anderson, J. A. DeBoer, A. F. Casper, Does aquatic invasive species removal benefit native fish? The response of gizzard shad (Dorosoma cepedianum) to commercial harvest of bighead carp (Hypophthalmichthys nobilis) and silver carp (H. molitrix), Hydrobiologia 817 (1) (2018) 403–412.
- M. J. Catalano, M. S. Allen, A size-and age-structured model to estimate fish recruitment, growth, mortality, and gear selectivity, Fisheries Research 105 (1) (2010) 38–45.
- M. J. Catalano, M. S. Allen, A whole-lake density reduction to assess compensatory responses
 of gizzard shad Dorosoma cepedianum, Canadian Journal of Fisheries and Aquatic Sciences
 68 (6) (2011) 955–968.
- A. Bodola, The life history of the gizzard shad, Dorosoma cepedianum, in western Lake Erie,
 Ph.D. thesis, Doctoral dissertation. Ohio State University, Columbus, 1955.

- P. Michaletz, Variation in characteristics among gizzard shad populations: The role of impoundment size and productivity, Fisheries Management and Ecology 24 (5) (2017) 361–371.
- G. Jons, L. Miranda, Ovarian weight as an index of fecundity, maturity, and spawning periodicity, Journal of Fish Biology 50 (1) (1997) 150–156.
- P. H. Michaletz, Overwinter survival of age-0 gizzard shad in Missouri reservoirs spanning a productivity gradient: roles of body size and winter severity, Transactions of the American Fisheries Society 139 (1) (2010) 241–256.
- M. R. Easterling, S. P. Ellner, P. M. Dixon, Size-specific sensitivity: applying a new structured population model, Ecology 81 (3) (2000) 694–708.
- S. P. Ellner, D. Z. Childs, M. Rees, et al., Data-driven modelling of structured populations,

 A practical guide to the Integral Projection Model. Cham: Springer.
- C. Merow, J. P. Dahlgren, C. J. E. Metcalf, D. Z. Childs, M. E. Evans, E. Jongejans, S. Record, M. Rees, R. Salguero-Gómez, S. M. McMahon, Advancing population ecology with integral projection models: a practical guide, Methods in Ecology and Evolution 5 (2) (2014) 99–110.
- M. Rees, D. Z. Childs, S. P. Ellner, Building integral projection models: a user's guide,

 Journal of Animal Ecology 83 (3) (2014) 528–545.
- R. A. Erickson, E. A. Eager, M. K. Brey, M. J. Hansen, P. M. Kocovsky, An integral projection model with YY-males and application to evaluating grass carp control, Ecological Modelling 361 (2017) 14–25.
- B. Liao, X. Shan, C. Zhou, Y. Han, Y. Chen, Q. Liu, A dynamic energy budget—integral projection model (DEB-IPM) to predict population-level dynamics based on individual data: a case study using the small and rapidly reproducing species Engraulis japonicus,

 Marine and Freshwater Research 71 (4) (2019) 461–468.

- J. W. White, K. J. Nickols, D. Malone, M. H. Carr, R. M. Starr, F. Cordoleani, M. L.
- Baskett, A. Hastings, L. W. Botsford, Fitting state-space integral projection models to
- size-structured time series data to estimate unknown parameters, Ecological Applications
- 26 (8) (2016) 2677–2694.
- N. Pollesch, K. Flynn, S. Kadlec, J. Swintek, S. Raimondo, M. Etterson, Developing integral projection models for ecotoxicology, Ecological Modelling 464 (2022) 109813.
- B. M. Bolker, Ecological models and data in R, Princeton University Press, 2008.
- S. Gutreuter, R. Burkhardt, K. S. Lubinski, Long Term Resource Monitoring Program procedures [microform] .
- B. S. Ickes, R. W. Burkhardt, Evaluation and proposed refinement of the sampling design for the Long Term Resource Monitoring Program's fish component, Tech. Rep., Geological Survey La Crosse WI Upper Midwest Environmental Sciences Center, 2002.
- R. Burden, J. D. Faires, Numerical analysis 8th ed, Thomson Brooks/Cole.
- S. P. Ellner, M. Rees, Integral projection models for species with complex demography, The
 American Naturalist 167 (3) (2006) 410–428.
- S. Ramula, M. Rees, Y. M. Buckley, Integral projection models perform better for small demographic data sets than matrix population models: a case study of two perennial herbs, Journal of Applied Ecology 46 (5) (2009) 1048–1053.
- J. L. Williams, T. E. Miller, S. P. Ellner, Avoiding unintentional eviction from integral projection models, Ecology 93 (9) (2012) 2008–2014.
- R Core Team, R: A Language and Environment for Statistical Computing, R Foundation for Statistical Computing, Vienna, Austria, URL https://www.R-project.org/, 2017.