1	
2	
3	
4	
5	
6	
7	
8	
9	
10	
11	
12	
13	
14 15	
15	A Life Cycle Simulation Model for Exploring Causes of Population Change in Alewife (Alosa
10	R Ene Cycle simulation Model for Exploring Causes of Fopulation Change in Alewite (Alosa pseudobarengus)
18	pseudonarengus)
19	
20	Gary A Nelson ^{1*} Benjamin I Gahagan ¹ Michael P Armstrong ¹ Adrian Jordaan ² Alison
21	Bowden ³
22	
23	
24	
25	
26	
27	
28	
29	
30	
31	
32	
33	¹ Massachusetts Division of Marine Fisheries, 30 Emerson Avenue, Gloucester, MA 01930
34	² Department of Environmental Conservation, University of Massachusetts, Amherst, MA 01003
35	³ The Nature Conservancy, 99 Bedford Street, Suite 500, Boston, Massachusetts 02111
36	
37	
38	*Corresponding author
39	Email address: gary.nelson@mass.gov
40	

41 Abstract

Over the last two decades, major changes in abundance and population characteristics of Alewife 42 43 (Alosa pseudoharengus), an anadromous herring species, have been observed along the US 44 Atlantic coast. Loss of spawning habitat, bycatch mortality in the directed pelagic fisheries, 45 increased predation mortality by rebounding predators such as Striped Bass, changes in water 46 flow and temperature affecting recruitment success, changes in ocean thermal habitat and direct 47 and indirect effects of changes in zooplankton community have been expounded by different 48 researchers as putative hypotheses for population changes in Alewife. Unfortunately, long-term, 49 concurrently-measured time series of regional factors and direct measures of biological processes 50 needed to elucidate underlying causes are severely lacking for Alewife. Therefore, we 51 developed, calibrated and validated a mechanistic, spatially-explicit, full life-cycle simulation 52 model that can be used to explore population responses of Alewife to various exogeneous 53 drivers. Daily processes such as spawning, recruitment, mortality, exploitation, predation and 54 movements are generated by using empirically-derived deterministic and stochastic relationships 55 and time-series of environmental data linked to specific life stages. We demonstrate the use of 56 the model as an investigative tool by simulating three hypotheses and comparing model results to 57 observed trends in Alewife populations from southern New England.

59 **1. Introduction**

60 Alewife (Alosa pseudoharengus) is an ecologically-important, migratory, anadromous 61 herring species that ranges along the Atlantic coast from Labrador to South Carolina (Loesch 62 1987). Alewife spend most of their life at sea but enter freshwater in spring to spawn (primarily in lakes and ponds). After spawning, Alewife adults exit the freshwater system and are presumed 63 to move offshore and northward to summer feeding grounds. The large-scale movement makes 64 65 them important forage for many marine and freshwater fish predators such as striped bass 66 (Morone saxatilis), cod (Gadus morhua), and yellow perch (Perca flavescens)(Loesch, 1987) as 67 well as birds (Dalton et al., 2009). In addition, they are a key link in the transfer of nutrients 68 between freshwater to marine systems (Mullen et al., 1986; Walters et al., 2009; Dias et al., 69 2019).

70 Over the last two decades, major changes in abundance and population characteristics of 71 Alewife have been observed along the US Atlantic coast (Schimdt et al., 2003; ASMFC 2017). 72 In southern New England, dramatic increases in run abundances of Alewife occurred during the 73 1980s and 1990s, but precipitous declines occurred after 2000 (Figure 1A)(Nelson et al., 2011; 74 Davis and Schultz, 2009). In addition, declines in mean body size (Figure 1B) and mean age 75 (Figure 1C) of Alewife have been observed in Massachusetts rivers since the 1980s and currently 76 are about 20-25 mm smaller than in the past (Nelson et al., 2011; ASMFC, 2017). Concurrently, 77 increases in total mortality and decreases in recruiting age classes have been noted (Figure 78 1D)(Nelson et al., 2011; ASMFC, 2017).

Researchers have expounded many hypotheses to explain the observed changes in
population characteristics. Hall et al. (2012) suggested that historical loss of spawning habitat
and productivity due to damming of rivers has decreased the resiliency of Alewife populations to

82 increases in mortality sources such as harvesting. Direct in-river harvest has varied over time, 83 but currently occurs only in Maine and a handful of other locations with approved management 84 plans (ASMFC 2017). Moratoriums limit direct harvest throughout much of the species range. 85 However, the occurrences of large incidental catches (bycatch) of Alewife in Atlantic Herring 86 (Clupea harengus) and Atlantic Mackerel (Scomber scombrus) fisheries off Southern New 87 England during fall and winter (Cournane et al. 2013) have been identified by many (ASMFC 88 2012; Hasselman et al. 2016; Palkovacs et al. 2013) as a potential cause of changes in 89 abundances. Increased predation mortality by rebounding predator populations like Striped 90 Bass (Morone saxatilis) and Double-Crested Cormorants (Phalacrocorax auritus) have been 91 suggested as a major cause as well (Dalton et al., 2009; Davis et al., 2012; Savoy and Crecco, 92 1995; Schmidt et al., 2003). Other putative causes include variability in water temperature and 93 flow affecting recruitment success (Tommasi et al., 2014), changes in ocean thermal habitat 94 (Lynch et al. 2015; Nye et al. 2009) and direct and indirect effects of changes in zooplankton and 95 phytoplankton composition (Kane 2011; Pershing et al., 2005), as a result of climate change. 96 With many factors potentially affecting growth and survival, the complex nature of interactions 97 between biotic and environmental factors, and regional differences in some of those factors, 98 such as bycatch (Hasselman et al., 2016), it seems unlikely that a single factor is responsible for 99 the observed changes in Alewife populations. Unfortunately, long-term, concurrently-measured 100 time series of regional factors and direct measures of biological processes (e.g., egg survival, 101 predation rates, etc.) needed to elucidate underlying causes are severely lacking for Alewife. 102 In the absence of long-term datasets, one method commonly used to understand complex 103 biological and environmental relationships in natural systems is simulation modeling. Through 104 simulation, population responses to changes in hypothesized causal mechanisms can be explored

105 and evaluated with models of population dynamics (Archambault et al., 2018; Watermeyer et al., 106 2018; Zeug et al., 2012). Such models allow researchers to conceptualize, describe and analyze 107 population behavior and ask "what-if" questions about the real system. Because Alewife 108 populations are under the influence of multiple environmental, predator and anthropogenic 109 pressures at all life stages, exploration of causal factors requires the development of a life cycle 110 model that incorporates relationships between the stressors and population dynamics. In this 111 paper, we present a comprehensive, mechanistic, spatially-explicit, full life-cycle simulation 112 model for predicting population responses of Alewife. The model is the first of its kind as 113 processes such as spawning, recruitment, mortality, exploitation, predation and movements are 114 generated by using empirically-derived deterministic and stochastic relationships and time-series 115 of environmental data are linked to specific life stages. In addition, we validate the model by 116 comparing predictions to data from published and unpublished studies not used in model 117 calibration, and show a detailed analysis of model sensitivities. Further, we demonstrate how the 118 model may be used as an investigative tool for exploring hypotheses related to population 119 changes in Alewife.

120

- 121 **2.** Methodology
- 122 2.1. General Model Description

123 The simulation model, developed in *R* (R Development Core Team, 2016), is 124 empirically-based and incorporates egg, yolk-sac larval, post-larval/young-of-the-year, juvenile 125 (immature fish of age 1 and greater) and adult (mature) stages of Alewife, although the amount 126 of detail varies among stages. The population structure is sex-specific, length-platoon (Goodyear 127 1989; 2002) and age-based meaning groups of individuals of a particular sex, length platoon, and 128 age are followed through time over a daily time step. The platoon-based structure is an 129 intermediate stance between all individuals of the same age progressing identically (age-based 130 model) and every individual have its own life path (individual-based model). The model is 131 specified to represent the regional (ocean, estuary, river and lake), sex (female and male), 132 maturity and length differences in population dynamics (Figure 2A). The model was developed 133 to differentiate among habitats such that different causes of mortality and changes in productivity 134 can be examined. Specifically, in the ocean, juveniles and adults grow and die due to length-135 related natural mortality and can experience size-selective, seasonal bycatch mortality from the 136 bottom trawl and mid-water trawl Atlantic herring fisheries. Mature adults migrate to lake 137 spawning habitat through estuary and river systems (where they can experience in-river harvest) 138 and undergo temperature-dependent batch spawning in the lake. Once spawning is completed, 139 adults return to the ocean depending on the duration spent in each system (Figure 2A). 140 The sub-model for age-0 dynamics is more detailed (Figure 2B). Daily batches of eggs 141 laid by all spawning individuals on the same day are followed through time. Temperature 142 determines development of egg and yolk-sac stages. First-feeding post-larvae experience initial 143 carrying capacity mortality and length-based natural mortality occurs thereafter. In the post-144 larval/young-of-the-year (YOY) stage, a bioenergetics model is used to grow length platoons 145 using area temperatures as input and platoons emigrate based on size, temperature and rainfall. 146 Duration of time spent in river and estuary is dependent on swimming speed and system length. 147 All YOY are moved to the ocean region by January 1 of the following year.

148

149 2.2 Juvenile/Adult Sub-Model

In this section we provide full details of the model processes for the juvenile/adult submodel. All functional relationships are empirically-based and were taken from literature or estimated from unpublished field data. Descriptions of indexes are listed in Table 1, definitions of mathematical symbols are given in Table 2, and mathematical equations are listed in Table 3. Parameter values are listed in Appendix Table A. The general order of daily processes for the juvenile/adult model is:

Movement among habitats (when applicable)→ Harvest/bycatch (fishing) mortality+Natural
 mortality→Growth

158

159 2.2.1 Population Structure

160 The female and male components of juvenile and adult subgroups were defined as having 161 nine ages (last age is a plus-group) and one-hundred length platoons in each age. The number of 162 length platoons was selected to obtain similar length frequency statistics (mean, standard 163 deviation and percentiles) as an individual-based model that used identical growth rates.

164

165 2.2.2 Population Dynamics

166 The number of individuals (N) of juvenile and adult fish is modeled using an exponential decay equation specific to each region. While in the ocean, the number of fish surviving to the 167 168 start of day d is dependent on by catch mortality (F) and natural mortality (M) during day d-1169 (Eq. JA.1). Length-dependent daily bycatch mortality is calculated from annual instantaneous 170 bycatch mortality, user-specified proportions representing monthly fractions of the fishing 171 pattern, and the number of days in a month. The daily F is further sub-divided in two fleet-172 specific daily bycatch mortalities based on landings ratios derived from Massachusetts Division 173 of Marine Fisheries (MADMF) port sampling of the Atlantic herring fisheries (Eq. JA.2).

Length-dependent, fleet-specific daily *F* is calculated by multiplying fleet-specific daily *F* and
fleet-specific selectivity-at-length values developed by using an equilibrium model and observed
bycatch lengths (Supplementary Document Section 1).

177 Natural mortality experienced by fishes is often related to body size or weight (Pepin 178 1991; Lorenzen, 1996). For Alewife, we developed an equation to predict daily M from length 179 that is used for all juvenile and adult fish regardless of sex or region. The Lorenzen (1996) 180 weight-based M equation was altered to predict annual mortality from length by substituting a 181 weight-length equation for Alewife sampled as bycatch (Eq. JA.3). Length-dependent daily M is 182 calculated by dividing the estimated annual M from length by the days in a year (365). Migrating 183 adult fish experience the same natural mortality rate while in estuary, river and lake systems, but 184 additional harvest and predation mortality can occur in the river. Regional multipliers (δ) are 185 available in Eqs. JA.1, JA.4 and JA.5 to adjust rates up or down if required. In the simulation, 186 regional multipliers of 1.0 are used to obtain a baseline annual M of about 0.6 for ages 5-8 187 similar to Gibson and Myers (2003).

188

189 2.2.3 Growth

The body length of platoon p on day d is calculated from body length at the start of day d-1 and the daily growth increment (Δ L) (Eq. JA.6). Δ L is calculated from the annual growth increment formulation of the von Bertalanffy equation (Quinn and Deriso, 1999)(Eq. JA.7) and a sine function that generates the fraction of annual growth that occurs on a given day of the year (Eq. JA.8). Because Alewife exhibit sexually-dimorphic growth rates, separate growth equations for female and male Alewife were derived (Supplementary Document Section 2).

197 2.2.4 Movement of Mature Fish to Estuary

Based on the University of Rhode Island Graduate School of Oceanography Fish Trawl
Survey (https://web.uri.edu/fishtrawl/), mature Alewife appear to enter an estuary during late
February-early March. To move mature fish from the ocean to the estuary region, the day of
movement (*ES*) is selected from a uniform distribution (Appendix A). All mature fish move
together at the beginning of the selected day.

203

204 2.2.5 Movement of Mature Fish to River

205 Mature fish move through the freshwater system (into river and lake and return) using 206 prior knowledge of temperature, swimming speed (v) and system length, and estimated spawning 207 duration is determined when a platoon enters the lake. The number of fish that move into the 208 river, and those that remain in the estuary, at the beginning of day d are calculated according to 209 Eqs. JA.9 and JA.10, respectively, where θ_d is the fraction migrating on day d. The mechanisms 210 controlling when and how many alewife enter a river are unknown; therefore, θ_d is simulated 211 over time. For a given year, the start day of a run is the day on which the average of the current 212 daily temperature and temperatures from the prior four days is $\geq 9.5^{\circ}$ C. This method was 213 developed to obtain start days within the ranges observed in the Monument River, MA (e.g., 214 Sheppard and Bednarski 2015). The peak day and length of a run are selected randomly from 215 uniform distributions (run peak: day 114-137; run length: 63-87 days). The start, peak and end 216 days are input parameters to a triangle distribution (Eq. JA.11) and auto-correlated errors are 217 added to the probabilities to mimic the fluctuating run trends observed in the Monument River (Eq. JA.12). The initial probabilities $(p^*_d s)$ are standardized to 1 (Eq. JA.13). To ensure that all 218 fish move into the river by the end of the run, θ_d is determined sequentially following Eq. JA.14. 219

220 The duration spent in the river (*rdur*) is determined by the river length, swimming speed and

221 average body length (Eq. JA.15).

222 2.2.6 Movement of Mature Adults from River to Lake

The day of lake entry for each migrant group is calculated as the river entry day plus the time spent by the group in the river, and the entire group of fish moves together on that day.

226 2.2.7 Duration Spent in Lake

The duration that a group spends in the lake (*sdur*) depends on when it enters during the run (Kissel 1974). The duration is generated from an exponential equation fitted to Kissel (1974) data (Eq. JA.16) and random deviates (Eq. JA.17) are added to mimic the distributions of durations.

231

232 2.2.8 Reproduction

233 For a group of fish entering the lake, spawning activity commences on the day when the 234 lake temperature reaches a minimum temperature threshold (10.5 °C; Fay et al., 1983). If a 235 group of fish arrives before the minimum threshold is reached, the first day of spawning is 236 registered as the day when the minimum threshold is reached. If a group of fish arrives after the 237 minimum threshold is reached, the first day of spawning (spday) is selected randomly from 4 to 238 8 days after lake entry as long as lake duration is >8 days; otherwise, the first spawning day is 239 the day of lake entry. The number of spawning events is determined by dividing the days spent 240 in the lake by the average number of days between spawned egg batches (*Bdur*). The next 241 spawning event is determined by $B_{dur}+1 + d$ (day of spawning). Spawning occurs at the

242	beginning of the day and the total number of eggs spawned by a group on a given day (E_d) is
243	calculated by Eq. JA.18 using literature values and a weight-length equation (Eq. JA.19).
244	
245	2.2.9 Lake Emigration
246	The day on which adult moves out of the lake is calculated as lake entry day plus the
247	number of days spent in the lake.
248	
249	2.2.10 River and Estuary Emigration
250	The day on which movement to the estuary occurs is calculated as the river entry day plus
251	<i>rdur</i> . The day on which movement to the ocean occurs is calculated as the river exit day plus the
252	the time spent in the estuary (edur). edur is calculated by using the same equation for rdur except
253	estuary length (a user-specified parameter that can be adjusted to retain fish longer) is used (Eq.
254	JA.15). Any fish remaining in the estuary are automatically moved to the ocean on day 1 of the
255	following year.
256	
257	2.2.11 Juvenile Maturation
258	On day 1 of each year, a fraction of juveniles in each platoon is matured according to sex-
259	specific proportion-mature-at-length relationships (Eq. JA.20). The maturity-at-length
260	relationships were estimated following Maki et al. (2001) using scale age and spawning check
261	data collected from Monument River Alewife (full details in Supplementary Document Section
262	3).
263	
264	2.3 Egg, Yolk-sac Larval, Post-larval/YOY Sub-model

In this section, we provide full details of the model processes. Definitions of mathematical symbols and mathematical equations for the egg-YOY sub-model are given in Tables 4 and 5, respectively. Parameter values are listed in Appendix B. The general order of daily processes for the egg-YOY model is the same as the juvenile/adult model. The sub-model consists of egg, yolk-sac, and post-larval/YOY stages.

270

271 2.3.1 Population Dynamics

272 A batch (b) of eggs (all eggs laid by spawning adults on a given day) is followed through 273 time and survival is modeled by using an exponential decay equation (Eq. EY.1). Daily M^{eggs} is 274 drawn randomly from a uniform distribution with range 0.14-0.30 (Appendix B) developed from 275 the literature (Supplemental Document Section 4). The day on which a batch of eggs hatch is determined using a hatch time (h) and temperature (T) relationship (Eq. EY.2) developed from 276 277 Edsall (1970)(full details in Supplementary Document Section 5). The rate of development to 278 hatching is dependent on daily temperature and the fraction of the development that occurs on a 279 given day is calculated as 1/h. Hatching occurs on day d according to Eq. EY.3. The number of 280 yolk-sac larvae that hatch is derived from the number of surviving eggs reduced for hatching 281 success (Eqs. EY.4 and EY.5). The hatching success versus temperature relationship was 282 developed from Edsall (1970)(full details in Supplementary Document Section 6). Survival of 283 yolk-sac larvae also follows the exponential decay equation and a multiplier is available to 284 rescale M if desired (Eq. EY.5). The day on which complete yolk-sac absorption occurs (d(y)) is 285 derived from the inverse of a days-to-yolk absorption and temperature relationship created from 286 literature values (Eqs. EY.6 and EY.7)(full details in Supplementary Document Section 7). 287 Multipliers are available to rescale *M* in the egg and yolksac stages (Eqs. EY.1 and EY.5).

288

- 289 2.4 Post-Larvae/ YOY Population Dynamics
- 290 2.4.1 Initial Density-Dependent Mortality

Population growth is limited by a simple density-dependent function. On the first day that yolk-sac larvae become post-larvae (Eq. EY.8), the number is further reduced by a batch carrying capacity (k_b) derived from a user-specified number of post-larvae per km² that a lake can hold, the lake area (km²) and the total number of batches (Eqs. EY.9 and EY.10)(full details in Supplementary Document Section 8).

296

297 2.4.2 Length Platoons

The number of post-larvae in each batch is distributed into length platoons (L_p) on the first day. The length bins are first created by using the minimum and maximum observed lengths of post-larval Alewife (Eq. EY.11), density frequencies from a log-normal distribution are generated using the mean of log-transformed length and standard deviation of log values (Eq. EY.12), and then the number of post-larvae is assigned across length bins by randomly sampling from a multinomial distribution parameterized with probability π (Eqs EY.13 and EY.14))(full details in Supplementary Document Section 9).

305

306 2.4.3 Population Dynamics

307 A batch of post-larvae is followed through time and the decline in numbers is modeled 308 using the exponential decay model with a stage multiplier (Eq. EY.15) where M^{Post} is length-309 dependent daily instantaneous natural mortality (Eq. EY.16). Natural mortality for fish \leq 22 mm 310 total length (TL) is dependent on length and water temperature, whereas it is only length-

- dependent for fish > 22 mm TL. These relationships were developed from and calibrated with
 published literature values (full details in Supplementary Document Section 10).
- 313

314 2.4.4 Growth in Lake

315 The body length of platoon p on day d is calculated from body length at the start of day 316 *d-1* and the daily growth increment derived by using the Alewife bioenergetics model of Stewart 317 and Binkowski (1986) with improvements by Klumb et al. (2003). This model is based on an 318 energy balance equation, $\Delta B = C \cdot (R + S + ER + U)$, where ΔB is the daily increase in body weight, C 319 is daily consumption, R is energy lost through metabolism, S is energy lost to specific dynamic 320 action, ER is energy lost in egestion, and U is energy lost in excretion. The definitions of 321 symbols and the equations governing the functional relationships are given in Tables 6 and 7, 322 respectively. Parameter values are listed in Appendix C. Daily consumption and standard 323 metabolism are the only temperature-dependent processes in the model.

324 In the growth sub-model, length of platoon p is first converted to weight using lengthweight equations depending on body length (Eq. BE.1). Daily consumption (joules• $g^{-1} \cdot d^{-1}$) by 325 326 platoon p is then calculated following Eq. BE.2 where C_{max} is the maximum specific feeding rate $(g \bullet g^{-1} \bullet d^{-1})$ dependent on size (Eq. BE.3), *EDP* is the prevenergy density (joules g⁻¹; calculated 327 328 from prey specific energy densities for the average stomach contents of Alewife post-larvae in 329 Cohen (1976), pr is the proportion of maximum consumption dependent on body length and 330 abundance (Eqs. BE.4 and BE.5), and $f(T_d)$ is the temperature-dependence function for cold-331 water species (Thornton and Lessem, 1978)(Eqs. BE.6- BE.12). The pr and length L_p relationships (Eq. BE.4) were developed and calibrated to obtain length ranges described in 332 333 several published studies and unpublished field data from Massachusetts and Rhode Island

sampling programs that occurred in rivers, lakes, estuaries and ocean waters (full details in Supplementary Document Section11). A relationship between slope of the *pr* relationship and number of young on a given day was further developed to simulate density-dependent growth (Eq. BE.5)(full details in Supplementary Document Section12). Regional growth multipliers are included and the estuary region multiplier (Ω_2) is set to 1.40 to match size data of young-of-theyear Alewife from Narragansett Bay, Rhode Island.

340 Respiration is calculated through a standard metabolism-body weight relationship, 341 activity multiplier and a temperature dependence function (Eq. BE.13) where R_p is specific rate 342 of respiration (joules• $g^{-1} \cdot d^{-1}$) for platoon p, ACT is a length-dependent activity multiplier to 343 obtain active metabolism (Eqs. BE.14 and BE.15), $f(T_d)$ is the temperature dependence function 344 (Eq. BE 16), and OC is the oxycalorific coefficient used to convert grams to joule equivalents. Egestion rate (ER; joules• $g^{-1} \cdot d^{-1}$) is calculated assuming a constant fraction of consumption (Eq. 345 BE.17). Excretion (U; joules• $g^{-1} \cdot d^{-1}$) is derived assuming a constant fraction of assimilated 346 347 energy (Eq. BE.18), and specific dynamic action (SDA) is calculated assuming a constant 348 fraction of assimilated energy lost (Eq. BE.19).

349 Daily body weight increase is derived via Eqs BE.20- BE.22. The energy density (ED;
350 joules• g⁻¹) of Alewife to convert joules to grams is calculated from seasonal relationships
351 between ED and body length (Eq. BE.22). To obtain the daily change in length, weight is
352 converted to length by using length-weight equations (Eq. BE.23).

Because the observed daily lake temperature may not match the temperature actually selected by the young Alewives, an optimal temperature-length relationship is used to replace the observed lake temperature (usually measured at the surface) when it exceeds the optimal temperature (Eq. BE.24). This assumes YOY are actively seeking optimal temperatures. 357

358 2.4.5 Emigration from Lake

359 Based on results of Gahagan et al. (2010), movement of YOY from the lake is dependent 360 on body size, temperature and rainfall. The first day of emigration can begin when the body 361 length of any platoon exceeds the user-specified migration size. The end day of emigration is 362 randomly picked to occur between day 304 and day 334. The number of fish that emigrate at the 363 beginning of day d is calculated according to Eq. EY.17 in Table 5 and the fraction emigrating is 364 determined using temperature and rainfall relationships developed from raw data of Gahagan et 365 al. (2010) (Eqs. EY.18-EY.20)(full details in Supplementary Document Section 13). Number of 366 fish remaining in the lake is determined by Eq. EY.21.

367

368 2.4.6 Movement from River to Ocean

The amount of time spent in the river (*rdur*) and estuary (*edur*) is determined from swimming speed (V; cm s⁻¹) of YOY and system length (Eqs. EY.23- EY.24). Any remaining YOY in the estuary are moved to the ocean region on January 1 of the following year and the YOY become immature age 1 fish. Numbers are split between sexes using a constant proportion of females ($p \stackrel{\frown}{\rightarrow}$).

- 374
- 375 2.4.7 Mortality in River, Estuary and Ocean
- 376 Mortality is calculated following Eq. EY.16.
- 377
- 378 2.4.8 Growth in River, Estuary and Ocean

379 Daily growth in the river, estuary and ocean is calculated by using the bioenergetics sub380 model, the daily temperature from each region and the optimal temperature-body size
381 relationship.

382

383 2.5. Global Sensitivity Analysis

384 The sensitivities of model outputs to input parameters were explored by using the method 385 of Morris (1991). This global sensitivity method was used to identify parameters with a range of 386 negligible to strong impacts. In addition, the Morris method identifies linear and nonlinear 387 effects. The Morris method consists of several randomized one-at-a-time experiments in which 388 the effect of a parameter on the model output is assessed while keeping the other parameters 389 constant (van Houwelingen et al., 2011). The first step is to randomly draw a set of starting 390 values within defined ranges of possible values for all input parameters, running the model using 391 these initial starting values, and saving the model output. The second step changes the value of 392 one parameter by a random interval (all other parameter values are those from the last run) and 393 the model output is then compared the previous run. For comparison, the elementary effect of the 394 ith (*EE*) input parameter is calculated by

395

$$EE_i(x^t) = \left(y(x^t) - y(x^{t-1})\right) / \pm \Delta_i$$

397

where *y* is the model output variable of interest, *t* is the current run and Δ_i is the random change interval based on p=4 (See van Houwelingen et al., 2011 for more details). The process is repeated for the remaining variables. The entire procedure is repeated *r* times (*r*=10 in this paper), each time with a different set of initial starting values. The total number of runs needed 402 for the analysis are r(n+1) where *n* is the number of parameters. The mean (μ^*), absolute mean 403 (μ) and standard deviation (σ) are then calculated for the *i*th input parameter by 404

405

$$\mu_{i}^{*} = \sum_{j=1}^{r} EE_{i,j}/r$$

$$\mu_{i} = \sum_{j=1}^{r} |EE_{i,j}|/r$$

$$\sigma_{i} = \sqrt{\sum_{j=1}^{r} (EE_{i,j} - \mu_{i}^{*})^{2}/r}$$

406

407

408 The interpretation of a parameter's effect is based on the coupled μ and σ values. Low μ 409 and σ values imply a parameter has a low impact on the model output, high μ and low σ imply a 410 parameter has high linear impacts on the model, low μ and high σ values imply a parameter has 411 high nonlinear impacts on the model, and high μ and high σ values imply a parameter has high 412 nonlinear impacts on the model and/or strong interactions with other parameters (Loubiere et al., 413 2016). We explored the sensitivity of only seventy-five parameters because valid minimum-414 maximum ranges could not be derived for some given lack of literature (e.g., the temperature 415 function parameters used in the bioenergetics model) or changes in individual parameters could 416 not be examined independently (e.g., the polynomial relationship parameters between hatching 417 success and temperature) without producing non-sensical values. The list of parameters with 418 corresponding ranges for the uniform distributions are provided in Appendix Table D.1. Ranges 419 were derived mainly from literature, but when lacking, ranges were also derived from standard 420 error estimates (+2 SE) of parameters (e.g., male and female von Bertalanffy growth) or from 421 within a biologically-realistic range based on our best judgement (e.g., p_{\pm}^{\bigcirc}).

422 For each run, the model was initialized with an equilibrium population of juveniles and423 adults (full details in Supplementary Document Section 14). The model was then run for 30

424 years (duration population reaches equilibrium with new parameters) using regional temperature 425 and rainfall data from 1962-1992 (obtained from various data sources or developed from air 426 temperature for the region of Narragansett Bay through Southern Massachusetts (full details in 427 Supplementary Document Sections 15-17)). The last year's derived values of run size, mean 428 length of adults in the river, sex ratio on the run, egg abundance, YOY cumulative river 429 abundance, and age-1 mean length on January 1 were recorded and used to examine input 430 parameter sensitivities.

431

432 2.6. Model Validation

Confidence in the performance of the model may be evaluated in terms of its ability as a predictive tool (Balci, 1998). To validate model results, we compared predicted dates of run starts, sex-specific mean ages on the run, ranges of days of first hatching and mean total length of age1 fish in spring to published and unpublished field observations not used in the calibration of the model. The model was run 50 times with only baseline natural mortality, and temperature and rainfall data from 1962-2016, and the mean and 95th percentiles of the outputs were used for comparison.

440

441 2.7. Exploration of Hypotheses of Population Changes

To demonstrate the model's potential to explore factors that may have caused historical changes in Alewife population characteristics, we investigated three spatial mortality hypotheses: historical responses were caused by 1) in-river harvest only, 2) in-river striped bass predation and harvest, and 3) ocean by-catch in the Atlantic herring fishery and in-river harvest. We expect population responses to differ among these hypotheses because most of the population is 447 vulnerable to incidental capture in the ocean region, whereas only immigrating, mature adults are 448 vulnerable to predation by Striped Bass in the river region. Time series of mortality rates 449 attributed to each component were created from existing data. In-river exploitation rates from 450 the Monument River (Nelson et al., 2011) were used to represent the fraction of adults harvested 451 in the river system. No harvest was assumed after 2005 because a moratorium in Massachusetts 452 has been in place. Since Striped Bass prey on migrating river herring (e.g., Davis et al., 2012), 453 predation mortality rates were associated with the river region and represented the daily fraction 454 of the adults eaten in the river system. Values were derived from data on Striped Bass 455 abundance and estimates of instantaneous total mortality from a statistical escapement-at-age 456 model. Bycatch mortality rates, representing the fraction of ocean population harvested by 457 incidental capture, were similarly derived by using New England Atlantic herring fishery 458 landings (see section 18 of Supplementary Document for full details). For each scenario, the 459 model was run 75 times (runtime: ~ 65 hours with Intel i7-6700 CPU @ 3.40 GHz) for 55 years 460 with water surface temperature and rainfall for years 1962-2016 (trends related to climate change 461 are evident in the time series; Supplemental Document Section 15-17). The model was initialized at the start of each run with the equilibrium population of juveniles and adults described above. 462 463

464 **3. Results**

465 3.1 Global Sensitivity

Results of the sensitivity analysis indicated that the model is highly non-linear and there
are strong interactions among parameters (Figure 3). The top twenty parameters based on μ*
that play a significant role in the magnitude of fluctuations in model output are shown in Table 8.
The run size and mean length of adults on the run are sensitive primarily to changes in

470 parameters from the bioenergetics sub-model, juvenile/adult mortality equations, post-471 larvae/YOY mortality and adult maturity (Table 8). The run sex ratio is most sensitive to changes 472 in the $p^{\bigcirc}_{\downarrow}$ (proportion used to split YOY into sexes), and maturity and bioenergetics model 473 parameters (Table 8). Egg abundance is most sensitive to parameter changes in the bioenergetics model, juvenile/adult mortality equation and post-larvae/YOY mortality equation. Abundance of 474 475 YOY is very sensitive to changes in lake carrying capacity, post-larvae/YOY mortality and 476 bioenergetics model parameters (Table 8). The mean length of age-1 fish on January 1 is 477 sensitive primarily to changes in the bioenergetics model parameters (Table 8).

478

479 *3.2 Model Validation*

480 Figure 4 shows the model output and data from published and unpublished field data. 481 Model predictions of run starts matched well ranges of run starts observed in the Parker River, 482 Massachusetts during 1972-1978 (Cole et al., 1976; Cole et al., 1978), in the Annaquatucket 483 River, Rhode Island during 1971 and 1972 (Richkus, 1974) and in multiple Massachusetts 484 systems monitored in 2014 (Rosset et al., 2017)(Figure 4A). The predicted mean age of female 485 and male Alewife during 1972-1978 agreed well with observed mean ages in the Parker River 486 (Cole et al., 1976; Cole et al., 1978) (Figure 4B). The range of model predictions of the first day 487 of hatching was similar to those observed in field studies (Yako, 1998; Iafrate and Oliviera, 488 2008; Devine, 2018) (Figure 4C). The model predicted well the mean lengths and 95% length 489 percentiles of age-1 Alewife compared to length data recorded in the MA DMF trawl survey and 490 Atlantic herring bycatch sampling during 2012-2016 (Figure 4D). Overall, the model produced 491 realistic trends and ranges of population characteristics observed historically.

492

493 *3.3 Exploration of Hypotheses of Population Changes*

494 Generally, harvest mortality has changed in an inverse relationship to recently dominant 495 striped bass predation and bycatch mortality (Figure 5). All hypotheses produced similar trends 496 in run size, mean length and mean age through 1990, but changes in population characteristics 497 were larger under hypotheses 2 (in-river harvest plus Striped Bass mortality) and 3 (in-river 498 harvest plus by catch mortality) because of higher combined mortalities (Figure 6). The largest 499 decline in run size occurred when bycatch mortality was present, but the simulated decline began 500 much earlier (1991) than was observed in Monument River (Figure 6). Under hypotheses 2 and 501 3, the model simulated declines in mean total length starting in 1991, which was close to 502 observed start years (1989-1990), but similar magnitudes in body size reduction (20-30 mm TL) 503 as observed in the Monument River were not reproduced (Figure 6). The model under 504 hypotheses 2 and 3 predicted declines in mean age of females, and the trends and magnitudes 505 under each hypothesis partially matched the trends and magnitudes observed in the Monument 506 River (Figure 6).

507

508 **4. Discussion**

We have developed, calibrated and validated a full life-cycle model for Alewife for evaluating hypotheses of potential causal factors affecting population dynamics. Key features of the model are that it (1) is structured sufficiently to provide realistic dynamics without being individually-based; (2) includes the full life cycle; (3) includes a realistic growth model for YOY Alewife; (4) includes environmental drivers that are known to influence growth, survival and migration processes during the first year of life; (5) has a generalized spatial structure and (6) incorporates exploitation and predation. This model should improve Alewife restoration efforts 516 by providing a tool to better understand factors influencing demographic trends and the517 consequences of potential management actions or environmental change.

518 The model is the first of its kind for Alewife and is an advancement towards 519 understanding the impact of exogeneous factors on the population dynamics of Alewife; however 520 it is generic and is not a complete representation of all processes that may affect Alewife 521 population dynamics. For example, there are currently no links between prey abundance and 522 growth of juvenile/adults and YOY Alewife or coupling between adult ocean migration and 523 environmental variables. In some aspects, we had to develop methods that would simulate trends 524 in population characteristics similar to those observed in the field without understanding 525 underlying processes. In other aspects, processes had to be simplified or left out, and many 526 parameters were assumed constant because of lack of data. As data become available, additional 527 details of underlying dynamics may be easily added given the flexible sub-model structures and 528 coding in the R language.

529 We have demonstrated the power of the model to explore population responses by 530 simulating three hypotheses concerning spatial exploitation and predation. Under these simple 531 hypotheses, the model did predict similar trends in population characteristics as those observed 532 historically in southern New England runs, but the timing and/or magnitudes of change were not 533 always replicated. This is not surprising given the limited hypotheses examined. These results 534 stress the fact that the reasons for the dramatic changes observed in the field are likely more 535 complex. To that end, the model has the ability to evaluate complex hypotheses if sufficient data 536 are available representing the specific aspects of growth and survival in a system. The population 537 consequences of management actions pertaining to juvenile/adult Alewife and its predators can 538 be simulated through the link with mortality. The effects of climate change, expected to be a

539 major stressor on Alewife populations (Hare et al., 2016; Lynch et al., 2015), on population 540 responses could be examined because water temperature is directly coupled to the start of adult 541 migration, spawning initiation, egg hatching rate, yolk-sac absorption, growth, mortality and 542 emigration of post-larval/YOY stages, while rainfall is linked to emigration of YOY. 543 For instance, if forecasts of water temperatures are available, the consequences of temperaturerelated changes in reproductive processes (e.g., earlier spawning times and/or hatching times) on 544 545 dynamics that can impact future recruitment (e.g., Lowerre-Barbieri et al., 1998) could be 546 explored. Similarly, if forecasts of rainfall are available, the consequences of potential changes 547 in YOY emigration timing as rainfall changes on similar dynamics and resulting processes could 548 be examined.

549 The sensitivity analysis showed that the most influential parameters in the model are 550 those associated with the post-larvae/YOY bioenergetics growth model and the 551 juvenile/adult/YOY mortality relationships. It is important that those parameters have a low 552 degree of uncertainty associated with them to allow for accurate prediction. Unfortunately, 553 measures of uncertainty are not available for most parameters used in the model because many 554 authors did not include statistics associated with model parameters (e.g., Stewart et al., 1983), 555 although some did include information on model fit (e.g., R²; Pepin, 1991) (Appendix D). For 556 parameters with measures of uncertainty (e.g., coefficients of variation), precision appears 557 reasonably high (CVs <20%) in most cases. The lack of measures of uncertainty for many 558 parameters does not mean the model predictions will be inaccurate. Stewart et al. (1983) showed 559 the bioenergetics model produces realistic growth predictions, and will likely do so under a wide 560 range of scenarios, because most parameters were determined through laboratory studies. In 561 addition, the pr relationships and Ω_2 were calibrated to field observations of post-larvae and

562 YOY sizes, and predictions were validated against data not used in the calibration (Figure 4D).

563 Similarly, we are confident that the mortality relationships will also produce realistic values

564 because they were calibrated to literature-based observations of survival.

565 There are many biotic and environmental drivers affecting productivity of young Alewife while in freshwater system (Kosa and Mather 2001; Yako et al., 2002). Based on the sensitivity 566 567 analysis results, the model outputs reflecting system productivity (e.g., YOY abundance) were 568 most affected by changes in parameters of the growth and mortality components (including the 569 carrying capacity parameter). Although we developed components of the relationship from 570 limited data, in reality, they are simple and generic. It is unlikely that more detailed lake-specific 571 predictions of production could be made unless data on important abiotic and environmental 572 drivers are known for each system. Ultimately, lake-specific differences in drivers will result in 573 varying levels of production that will determine individual population resilience to 574 anthropogenic and environmental stressors.

575 A benefit of creating an empirically-based life cycle model is learning where gaps in our 576 current knowledge of biological processes and population dynamics are. More detailed 577 information on Alewife population dynamics are needed at all life stages. Some areas of 578 necessity include Alewife-specific egg and yolk-sac mortality rates, relationships between YOY 579 growth and prey, and environmental and physiological influences that affect adult immigration 580 and YOY emigration. Some of the biggest gaps in our knowledge pertain to the times YOY and 581 adults spend in an estuary and ocean. For YOY, there is some general information on growth and 582 broad habitat use between freshwater and estuarine environments (Turner and Limburg 2012; 583 Turner and Limburg 2016). However, no information exists on the duration spent in an estuary, 584 prey species eaten in estuaries and the ocean, spatial distribution, mortality and movement

585 patterns that would help improve our understanding of YOY dynamics and allow us to enter 586 more detail into the model. For adults, general migration patterns are assumed (north into the 587 Gulf of Maine after spawning, south to waters south of Cape Cod in the winter (Munroe, 2002)), 588 but details on movements and spatial distribution for individual stocks or regional groups are 589 needed to help resolve a multitude of fishery and climate change impacts (ASMFC 2012; 590 Hasselman et al. 2016; Hare et al. 2016; Lynch et al. 2015; Palkovacs et al. 2013). This is 591 particularly important since migration routes will influence the exposure of Alewife to fisheries 592 impacts over space and time (Bethoney et al., 2017). 593 Over the past ten years, Alewife have been the focus of two Endangered Species Act 594 Listing Determinations, a Benchmark Stock Assessment, and Stock Assessment update. The 595 resounding conclusion of all of these examinations is that status determination, management and 596 recovery of this species is hindered by a lack of data and tools. This model represents a 597 significant step forward in our ability to understand Alewife population dynamics and should 598 improve restoration efforts. Still, until deeper understanding of underlying processes and long-599 term time series of environmental and biological measures are available, it may be impossible to 600 identify primary factors responsible for the historical changes in Alewife populations. New 601 technologies, including bio-logging (e.g., Dean et al., 2017), and additional data collected across

the life cycle will be key to closing the data gaps and making Alewife assessment and

603 management a more data-rich effort.

604

605 Acknowledgements

606

607 Funding for this project was provided, in part, by the Atlantic States Marine Fisheries

608 Commission and by Sportfish Restoration Grant F-57R. We thank the following individuals and

609 institutions for providing Alewife data: Matthew Devine of the University of Massachusetts-

610 Amherst, John Sheppard, Bradley Schondelmeier and the Age and Growth staff of Massachusetts

611 Division of Marine Fisheries, Phil Edwards of the Rhode Island Department of Fish and

612 Wildlife, Kurt Gottschall of the Connecticut Department of Environmental Protection and the

613 University of Rhode Island Graduate School of Oceanography. All members of the NOAA

614 funded Linking Lifestages project provided helpful comments during the initial stages of model

615 creation. We also thank two anonymous reviewers for helpful comments that improved the

616 manuscript. The entire model code and data used in this study are available upon request from

617 the senior author.

References

620	Archambault, B., Rivot, E., Savina, M., Le Pape, O., 2018. Using a spatially structured life cycle
621	model to assess the influence of multiple stressors on an exploited coastal-nursery-
622	dependent population. Estuar. Coast. Shelf Sci. 201: 95-104.
623	https://doi.org/10.1016/j.ecss.2015.12.009
624	
625	ASMFC (Atlantic States. Marine Fisheries Commission), 2012. River herring benchmark stock
626	assessment Volume I. Stock Assessment Report 12-02 of the Atlantic States Marine
627	Fisheries Commission, Washington, DC.
628	
629	ASMFC (Atlantic States. Marine Fisheries Commission), 2017. River Herring Stock Assessment
630	Update, Volume I: Coastwide Summary. 172 p. Arlington, VA.
631	
632	Balci, O., 1998. Verification, Validation and Testing. In: Handbook of Simulation: Principles,
633	Methodology, Advances, Applications and Practice, J. Banks (ed.), John Wiley and Sons,
634	Inc. New York, New York, p.335-393.
635	
636	Bethoney, N. D., Schondelmeier, B. P., Kneebone, J., Hoffman, W. S., 2017. Bridges to best
637	management: effects of a voluntary bycatch avoidance program in a mid-water trawl
638	fishery. Mar. Pol. 83: 172-178. https://doi.org/10.1016/j.marpol.2017.06.003.
639	
640	Burbidge, R. G., 1974. Distribution, growth selective feeding, and energy transformations of
641	young-of-the-year blueback herring, Alosa aestivalis (Mitchill), in the James River,
642	Virginia. Trans. Am. Fish. Soc.103: 297-311. https://doi.org/10.1577/1548-
643	8659(1974)103<297:DGSFAE>2.0.CO;2
644	
645	Cole, C. F., Libey, G. S., Huber, M. E., Jimenez, D., 1976. Optimal alewife run management,
646	Parker River, Massachusetts. Final Report to U.S. Department of Commerce, National
647	Oceanic and Atmospheric Administration, National Marine Fisheries Service. Project
648	AFC-12. 37 p.
649	
650	Cole, C. F., Essig, R., Sarnelle, O., 1978. Biological investigation of the alewife population,
651	Parker River, MA. Anadromous Fish Act Yearly Report to U.S. Department of
652	Commerce, National Oceanic and Atmospheric Administration, National Marine
653	Fisheries Service. Project AFC-16-2. 16.p.
654	
655	Cournane, J. M., Kritzer, J. P., Correia, S. J., 2013. Spatial and temporal patterns of anadromous
656	alosine bycatch. Fisheries Research 141: 88-94.
657	https://doi.org/10.1016/j.fishres.2012.08.001
658	
659	Dalton, C. M., Ellis, D., Post, D. M. 2009. The impact of double-creasted cormorant
660	(Phalacrocorax auritus) predation on anadromous alewife (Alosa pseudoharengus) in
661	south-central Connecticut, USA. Can. J. Fish. Aquat. Sci. 66: 177-186.
662	https://doi.org/10.1139/F08-198
663	

664 665 666	Davis, J. P., Schultz, E. T. 2009., Temporal shifts in demography and life history of an anadromous alewife population In Connecticut. Mar. Coast. Fish. Dyn. Manage and Ecosystem Science. 1: 90:106. https://doi.org/10.1577/C08-003.1
667 668	Davis, J. P., Schultz, E. T., Vokoun, J. C., 2012. Striped Bass consumption of Blueback herring during vernal riverine migrations: does relaxing harvest restrictions on a predator help
009	Management and Example Science 4, 220, 251
670 671	https://doi.org/10.1080/19425120.2012.675972.
672	Dean, M. J., Hoffmans, W. S., Zemeckis, D. R., Armstrong, M. P., 2017, Fine-scale diel and
673	gender-based patterns in behaviour of Atlantic cod (Gadus morhua) on a spawning
674	ground in the Western Gulf of Maine, ICES J. Mar. Sci. 71: 1474-1489.
675	https://doi.org/10.1093/icesims/fsu040
676	
677	Diaz B S Frisk M G Jordaan A 2019 Opening the tap: increasing riverine connectivity
678	strengthens marine food web pathways. PLoS ONE 14(5):e0217008.
679	https://doi.org/10.1371/journal. pone.0217008.
680	
681	Edsall, T. A., 1970. The effect of temperature on the rate of development and survival of alewife
682	eggs and larvae. Trans. Am. Fish. Soc. 99: 376-380. https://doi.org/10.1577/1548-
683	8659(1970)99<376:TEOTOT>2.0.CO:2
684	
685	Ellis, D., Vokoun, J. C., 2009. Earlier spring warming of coastal streams and implications for
686	alewife migration timing. N. Am. J. Fish. Manage. 29: 1584-1589.
687	https://doi.org/10.1577/M08-181.1
688	
689	Fay, C.W., Neves, R. J., Pardue, G. B., 1983. Species profiles: life histories and environmental
690	requirements of coastal fishes and invertebrates (Mid-Atlantic) - alewife/blueback
691	herring. U. S. Fish and Wildlife Service, Division of Biological Services,
692	FWS/OBS.82/11.9. U. S. Army Corps of Engineers TR EL-82-04. 25 p.
693	
694	Gahagan, B. I., Gherard, K. E., Schultz, E. T., 2010. Environmental and endogeneous factors
695	influencing emigration in juvenile anadromous alewives. Trans. Am. Fish. Soc. 139:
696	1069-1082. https://doi.org/10.1577/T09-128.1
697	
698	Ganias, K., Divino, J. N., Gherard, K. E., Davis, J. P., Mouchlianitis, F., Schultz, E. T., 2015. A
699	reappraisal of reproduction in anadromous alewives: determinate versus indeterminate
700	fecundity, batch size, and batch number. Trans. Am. Fish. Soc. 144: 1143-1158.
701	https://doi.org/10.1080/00028487.2015.1073620
702	
703	Gibson, A. J. F., Myers, R. A., 2003. A statistical, age-structured, life-history-based stock
704	assessment model for anadromous Alosa. Pages 275-283 in Limburg, K. E., Waldman, J.
705	R., editors. Biodiversity, status, and conservation of the world's shads. American
706	Fisheries Society Symposium 35, Bethesda, Maryland.
707	

Goodyear, C. P., 1989. LSIM – a length-based fish population simulation model. NOAA Tech. 708 709 Memo. NMFS-SEFC 219. 53 p. 710 711 Goodyear, C. P., 2002. Negative implications of large minimum size regulations on future mean 712 size at age: an evaluation using simulated striped bass data. Am. Fish. Soc. Sym. 30: 217-713 229. 714 715 Hall, C. J., Jordaan, A., M. G. Frisk., 2012. Centuries of anadromous forage fish loss: 716 consequences for ecosystem connectivity and productivity. Bioscience 62 (8): 723-731. 717 https://doi.org/10.1525/bio.2012.62.8.5 718 719 Hare, J.A., Morrison, W. E., Nelson M. W., Stachura, M. M., Teeters, E. J., Griffis, R. B., 720 Alexander, M. A., Scott, J. D., Alade, L., Bell, R. J., Chute, A. S., Curti, K. L., Curtis, 721 T.H., Kircheis, D., Kocik, J. F., Lucey, S. M., McCandless, C. T., Mike, L. M., 722 Richardson, D. E., Robillard, E., Walsh, H. J., McManus, M. C., Marancik, K. E., 723 Griswold, C. A., 2016. A vulnerability assessment of fish and invertebrates to climate change on the Northeast U. S. Continental Shelf. PLoS ONE 11(2): e0146756. 724 725 https://doi.org/10.1371/journal.pone.0146756. 726 727 Hasselman, D. J., Anderson, E. C., Argo, E. E., Bethoney, N. D., Gephard, S. R., Post, D. M., 728 Schondelmeier, Schultz, T. F., Willis, T. V., Palkovacs, E. P., 2016. Genetic stock 729 composition of marine bycatch reveals disproportional impacts on depleted river herring 730 genetic stocks. Can. J. Fish. Aquat. Sci. 73: 951-963. https://doi.org/10.1139/cjfas-2015-731 0402. 732 733 Iafrate, J., Oliveira, K., 2008. Factors affecting migration patterns of juvenile river herring in a 734 coastal Massachusetts stream. Environ. Biol. Fish. 81:101-110. 735 https://doi.org/10.1007/s10641-006-9178-1. 736 737 Kane, J., 2011. Multiyear variability of phytoplankton abundance in the Gulf of Maine. ICES J. 738 Mar. Sci. 68: 1833-1841. https://doi.org/10.1093/icesjms/fsr122. 739 740 Kissel, G. W., 1974. Spawning of the anadromous alewife, Alosa pseudoharengus, in Bride 741 Lake, Connecticut. Trans. Am. Fish. Soc. 103: 312-317. https://doi.org/10.1577/1548-742 8659(1974)103<312:SOTAAA>2.0.CO;2. 743 744 Kitchell, J. F. Stewart, D. J., Weininger, D., 1977. Applications of a bioenergetics model to 745 yellow perch (Perca flavescens) and walleye (Stizostedion vitreum vitreum). J. Fish. Res. 746 Board Can. 34: 1922-1935. https://doi.org/10.1139/f77-258. 747 748 Klumb, R. A., Rudstam, L. G., Mills, E. L., 2003. Comparison of alewife young-of-the-year and 749 adult respiration and swimming speed bioenergetics model parameters: implications of extrapolation. Trans. Am. Fish. Soc. 132: 1089-1103. https://doi.org/10.1577/T03-038. 750 751

752 Loesch, J. G., 1987. Overview of life history aspects of anadromous alewife and blueback 753 herring in freshwater habitats. In: M. J. Dadswell et al., eds., Common Strategies of 754 Anadromous and Catadromous Fishes. Am. Fish. Soc. Symp. 1: 89-103. 755 756 Lorenzen, K., 1996. The relationship between body weight and natural mortality in juvenile and 757 adult fish: a comparison of natural ecosystems and aquaculture. J. Fish Biol. 49: 627– 758 642. https://doi.org/10.1111/j.1095-8649.1996.tb00060.x. 759 760 Loubiere, P., Jourdan, A., Siarry, P., Chelouah, R., 2016. A sensitivity analysis method for 761 driving the Artificial Bee Colony algorithm's search process. Appl. Soft. Comput. 762 41:515-531. https://doi.org/10.1016/j.asoc.2015.12.044. 763 764 Lowerre-Barbieri, S. K., Lowerre, J. M., Barbieri, L. R. 1998. Multiple spawning and the 765 dynamics of fish populations from an individual-based simulation model. Can. J. Fish. 766 Aquat. Sci. 55: 2244-2254. 767 768 Lynch, P. D., Nye, J. A., Hare, J. A., Stock, C. A., Alexander, M. A., Scott, J. D., Curti, K. L., 769 Drew, K., 2015. Projected ocean warning creates a conservation challenge for river 770 herring populations. ICES J. Mar. Sci. 72: 374-387. 771 https://doi.org/10.1093/icesjms/fsu134. 772 773 Maki, K. L., Hoenig, J. M., Olney, J. E., 2001. Estimating proportion mature at age when 774 immature fish are unavailable for study, with applications to American shad in the York 775 River, Virginia. N. Am. J. Fish. Manage. 21: 703-716. https://doi.org/10.1577/1548-776 8675(2001)021<0703:EPMAAW>2.0.CO;2. 777 778 Morris, M. D., 1991. Factorial sampling plans for preliminary computational experiments. 779 Technometrics 33: 161-174. http://dx.doi.org/10.1080/00401706.1991.10484804. 780 781 Mullen, D.M., Fay, C. W., Moring, J. R., 1986. Alewife/Blueback Herring. Species Profiles: Life 782 Histories and Environmental Requirements of Coastal Fishes and Invertebrates (North 783 Atlantic series) USDI Fish and Wildlife Service. Biological Report 82(11.58). 22 pp. 784 785 Munroe, T. A., 2002. Herrings. Family Clupeidae: Alewife. In: Collette, B. B., Klein-MacPhee, G. (eds.), Bigelow and Schroeder's Fishes of the Gulf of Maine, 3rd edition. Smithsonian 786 787 Institution Press, Washington, DC. pp. 118-125. 788 789 Nelson, G. A., Brady, P. D., Sheppard, J. J., Armstrong, M. P., 2011. An assessment of river 790 herring stocks in Massachusetts. Massachusetts Division of Marine Fisheries Tech. Rep. 791 TR-46. 81 p. https://www.mass.gov/files/documents/2016/08/wi/tr-46.pdf. 792 793 Nye, J. A., Link, J. S., Hare, J. A., Overholtz, W. J., 2009. Changing spatial distribution of fish 794 stocks in relation to climate and population size on the Northeast United States continental shelf. Mar. Ecol. Prog. Ser. 393: 111-129. 795 796 https://doi.org/10.3354/meps08220. 797

798	Palkovacs, E. P., Hasselman, D. J., Argo, E. E., Gephaerd, S. R., Limburg, K. E., Post, M.m
799	Schultz, T. F., Willis, T. V., 2013. Combining genetic and demographic information to
800	prioritize conservation efforts for anadromous alewife and blueback herring. Evol. Appl.
801	7: 212-216. https://doi.org/10.1111/eva.12111.
802	
803	Pauly, D., 1984. Length-converted catch curves. A powerful tool for fisheries research in the
804	tropics (Part III). ICLARM Fishbyte 2(1): 17-19.
805	
806	Pepin, P., 1991. Effect of temperature and size on development, mortality, and survival rates of
807	the pelagic early life history stages of marine fish. Can. J. Fish. Aquat. Sci. 48: 503-518.
808	https://doi.org/10.1139/f91-065.
809	
810	Pershing, A. J., Greene, C. H., Jossi, J. W., O'Brien, L, Brodziak, J. K. T., Baily, B.A., 2005.
811	Interdecadal variability in the Gulf of Maine zooplankton community, with potential
812	impacts on fish recruitment. ICES J. Mar. SCi. 62:1511-1523.
813	https://doi.org/10.1016/j.icesjms.2005.04.025.
814	
815	Quinn, T. J., Deriso, R. B., 1999. Quantitative fish dynamics. Oxford University Press. 542
816	pages.
817	I B
818	R Development Core Team. 2016. R: A language and environment for statistical computing. R
819	Foundation for Statistical Computing, Vienna, Austria, ISBN 3-900051-07-0.
820	http://www.R-project.org/.
821	
822	Richkus, W. A., 1974. Factors influencing the seasonal and daily patterns of Alewife (Alosa
823	pseudoharengus) migration in a Rhode Island river. J. Fish. Res Boad Can. 31:1485-
824	1497. https://doi.org/10.1139/f74-178.
825	
826	Rosset J. Roy A. H. Gahagan, B. J. Whiteley, A. R. Armstrong, M. P. Sheppard, J. J.
827	Iordaan A 2017 Temporal patterns of migration and spawning of river herring in
828	coastal Massachusetts Trans Am Fish Soc 146: 1101-1114
829	https://doi.org/10.1080/00028487.2017.1341851
830	
831	Savoy, T. Crecco, V., 1995, Factors affecting the recent declined of blueback and American shad
832	in the Connecticut River. A report to the Atlantic States Marine Fisheries Commission
833	Washington DC
834	
835	Schmidt R F Jesson B M Hightower J F 2003 Status of river herring stocks in large
836	rivers In: Limburg K F Waldman I R editors Biodiversity Status and Conservation
837	of the World's Shads Am Fish Soc Symp 35:171-182
838	of the world's blads. All: 1151. 566. 5911p. 55.171 162.
830	Shennard I. I. Bednarski, M. S. 2015. Utility of single-channel electronic resistivity counters
840	for monitoring river herring populations N Am I Fish Manage 35: 11/1/ 1151
8/1	https://doi.org/10.1080/02755947.2015.1084407
041 8/17	$\pi up s. (uo 1.01g/10.1000/02755747.2015.1004407.$
042	

843 844 845	Stewart, D. J., Binkowski, F. P., 1986. Dynamics of consumption and food conversion by Lake Michigan alewives: an energetics-modeling synthesis. Trans. Am. Fish. Soc. 115: 643- 661. https://doi.org/10.1577/1548-8659(1986)115<643:DOCAFC>2.0.CO;2.
846 847	Stewart, D. J., Kitchell, J. F., Rottiers, D. V., Edsall, T. A., 1983. An energetics model for lake
848 849 850	Aquat. Sci. 40: 681-698. https://doi.org/10.1139/f83-091.
851 852 853 854 855	Tommasi, D., Nye, J., Stock, C., Hare, J. A., Alexander, M., Drew, K., 2015. Effect of environmental condition on juvenile recruitment of alewife (<i>Alosa pseudoharengus</i>) and blueback herring (<i>Alosa aestivalis</i>) in freshwater: a coastwide perspective. Can. J. Fish. Aquat Sci. 72:1037-1047. https://doi.org/10.1139/cjfas-2014-0259.
856 857 858 859	Turner, S. M., Limburg, K. E.,2012. Comparison of juvenile alewife growth and movement in a large and a small watershed. Mar. Coast. Fish: Dyn. Manage. Eco. Sci. 4: 337-345. https://doi.org/10.1080/19425120.2012.675974.
860 861 862 863	Turner, S. M., Limburg, K. E., 2016. Juvenile river herring habitat use and marine emigration trends: comparing populations. Oecologia 180: 77-89. https://doi.org/10.1007%2Fs00442-015-3443-y.
864 865 866 867	Van Houwelingen, H. C., Boshuizen, H.C, Capannesi, M., 2011. Sensitivity analysis of state- transition models: how to deal with a large number of inputs. Comput. Biol. Med. 41:838-842. https://doi.org/10.1016/j.compbiomed.2011.07.001.
868 869 870 871	 Walter s, A. W., Barnes, R. T., Post, D. M., 2009. Anadromous alewives (<i>Alosa pseudoharengus</i>) contribute marine-derived nutrients to coastal stream food webs. Can. J. Fish. Aquat. Sci.66: 439-448. https://doi.org/10.1139/F09-008.
872 873 874 875 876	Watermeyer, K. E., Jarre, A., Shannon, L. J., Mulumba, P. Botha, J., 2018. A frame-based modeling approach to understanding changes in the distribution and abundance of sardine and anchovy in the southern Benguela. Ecol. Model. 371: 1-17. https://doi.org/10.1016/j.ecolmodel.2017.12.017.
877 878 879	Yako, L. A., M. E. Mather, and F. Juanes., 2002. Mechanisms for migration of anadromous herring: an ecological basis for effective conversation. Ecol. Appl. 12: 521-534. https://doi.org/10.1890/1051-0761(2002)012[0521:MFMOAH]2.0.CO;2.
880 881 882 883 884 884	Zueg, S. C., Bergman, P. S., Cavallo, B. J., Jones, K. S., 2012. Application of a life cycle simulation model to evaluate impacts of water management and conservation actions on an endangered population of Chinook salmon. Environ. Model. Assess. 17: 455-467. https://doi.org/10.1007/s10666-012-9306-6.

886 Figure Captions

- Figure 1. Trends in Alewife A) run size for the Monument and Mattapoisett rivers in
 Massachusetts, B) mean total length (mm) for the Monument River and Stony Brook, C)
 mean age of female and male Alewives for the Monument River, and D) total mortality
 rates estimated from a statistical escapement-at-age model for the Monument River stock
 (Nelson et al., 2011) updated with data from 2011-2015 (ASMFC, 2017).
- 894Figure 2. Schematic of model processes for the A) juvenile/adult and B) egg, yolk-sac, post-895larval/young-of-the-year sub-models. M is natural mortality, CC is carrying-capacity896mortality, and ΔL is change in body length (growth).
- Figure 3. Plots of Morris σ versus Morris μ for six output variables: run size, adult mean lengths
 in the river, run sex ratio, total egg abundance, YOY abundance exiting lake, and age-1
 mean length on May 1.
- 900 Figure 4. Comparison of model predictions to observed literature and field data for A) first day 901 of run initiation in the Parker River (PR) reported in Cole et al. (1978), Annaquatucket 902 River (AR) reported by Richkus (1974), and multiple rivers observed by Rosset et al. 903 (2017), B) grand mean age (minimum and maximum of mean ages from observed (Cole 904 et al. 1978) and simulated years 1972-1978) of male and female Alewife in the Parker 905 River, C) first day of hatching and D) mean lengths (and 95% percentiles) of age-1 906 Alewife. Note that, although variable, the first day of hatching becomes earlier over time 907 as a result of increasing water temperatures (see Supplemental Document Section 15-17).
- Figure 5. Derived mortality rates used in the three simulation scenarios. Bycatch mortality
 represents the annual exploitation fraction, and those for in-river harvest and Striped Bass
 represent daily fractions.
- Figure 6. Simulated run size, run mean total length of adults and run mean age of females for the
 three hypotheses examined. Observed data from the Monument River only are shown.
 See Figure 1 from data from other rivers. Predictions are averages of 75 simulations.

















Index	Definitions	Range
d	Day of year	{1,,365}
p	Platoon	{1,,100}
a	Age	{1,,9+}
r	Region	{1=Ocean, 2=Estuary, 3=River, 4=Lake}
x	Sex	{1=Female, 2=Male}
т	Maturity	{1=Immature, 2=Mature}
f	Fleet	{1=Bottom Trawl, 2=Mid-Water}
b	Daily Egg Batch	$\{d(E_d),,d(E_{last})\}$

 Table 1. Description of Indexes used in the mathematical equations

Table 2. Definitions of symbols used in the mathematical description of the Alewife juvenile/adultsimulation sub-model.

D Number of days in the year (365)	
$N_{d,p,a,x,m}^{r}$ Numbers of platoon p , age a , sex x and maturity m individuals at t	the beginning
of day <i>d</i> in region <i>r</i>	
M_p^r, MA, MB, δ_r Daily instantaneous natural mortality rate (d^{-1}) for platoon p in reg	gion <i>r</i> ,
intercept and exponent of mortality equation, M multiplier for reg	gion <i>r</i>
μ , F_d Annual fraction of by-catch mortality, daily instantaneous fishing	mortality
$L_{d,x,p}$ Total length (mm) of platoon p of sex x on day d	
η_n, D_n Fraction of by-catch <i>F</i> occurring in month <i>n</i> , number of days in mo	onth <i>n</i>
P_f The annual fraction of bycatch attributed to fleet $f(P_2=1-P_1)$	
$s_{p,f}$ The selectivity at length platoon p in region 1 by fleet f (see Suppl	ementary
Document Section 1)	
$\Delta L_{d,x,p}, L \infty_x, K_x, t 0_x$ Growth increment (mm/day) of platoon p and sex x on day d, sex-	-specific von
Bertalanffy growth equation parameters	
$L_{x,p}^{o}, g_{d}, gA, gB, gC$ Length of sex x and platoon p on day 1, fraction of annual growth	increment,
sine function parameters	
ES Estuary entrance day	
<i>rT</i> , <i>ndaysT</i> Temperature for initiation of run start (°C), number of prior days u	used in
calculation	
<i>H</i> , <i>PRED</i> Fraction of Alewife run that is harvested, fraction of run eaten by	a predator
$\theta_d, a', b', crl, c'$ Daily fraction of platoon numbers immigrating to river, first day or	f
immigration, peak day of immigration, run length, last day of imn	nigration
(a'+crl)	
$\sigma_{\varepsilon}^{2}, \rho$ Normal variance for run error sampling, first-order autocorrelatio	n
$rdur, rl, edur, el, v, \bar{s}$ Days spent in river, river length, days spent in estuary, estuary len speed (body length/sec), mean body size	ngth, swimming
sdur, lA, lB, pdays, lC Spawning duration (days), intercept of spawning duration equation	on, slope of
spawning duration equation, day of year after initiation of emigra	tion expressed
as a proportion of run duration, value of <i>sdur</i> for <i>pdays</i> <0.45	
<i>CV_{sdur}</i> , <i>spdday</i> Coefficient of variation for <i>sdur</i> calculation, added delay before sp	pawning begins
<i>sT</i> , <i>Bdur</i> Lower spawning temperature limit (°C), days between batches	
W_p , B, WA, WB, \bar{e} , E_d Mean weight of females of platoon p (grams), mean number of eg	gg batches
produced per female, intercept of length-weight equation, slope of	of length-
weight equation, mean total number of eggs per gram produced p	per
females, total number of eggs produced by spawning females on	day <i>d</i>
$pm_{x,a,p}, pmA_x, pmB_x$ pm is the proportion of mature fish of sex x, age a and length plat	:oon <i>p,</i>
intercept of sex x maturity equation, slope of sex x maturity equal	tion

Table 3. Equations describing dynamics and processes of the juvenile/adult sub-model.

$$N_{d,p,a,x,m}^{r} = N_{d-1,p,a,x,m}^{r} e^{-\sum s_{p,f}F_{d-1,f} - M_{p}^{r}\delta^{r}}, r = 1$$
 JA.1

$$F_{d,f} = \left(\left(-\log e(1-\mu)\eta_n \right) / D_n \right) P_f$$
 JA.2

$$M_p^r = (MA \cdot L_p^{MB})/D \qquad \qquad \text{JA.3}$$

$$N_{d,p,a,x,2}^{r} = N_{d-1,p,a,x,2}^{r} e^{-M_{p}^{r} \delta^{r}}, r = 2, 4$$
 JA.4

$$N_{d,p,a,x,2}^{r} = N_{d-1,p,a,x,2}^{r} (1-H)(1-PRED)e^{-M_{p}^{r}\delta^{r}}, r = 3$$
 JA.5

$$L_{d,x,p} = L_{d-1,x,p} + \Delta L_{d-1,x,p}$$
 JA.6

$$\Delta L_{d,x,p} = \left(L_{\infty,x} - L_{x,p}^{o}\right) \cdot \left(1 - e^{-Kx}\right) \cdot g_{d}$$
 JA.7

$$g_d = gA + gB * \sin\left(2\pi \cdot \frac{d - gC}{D}\right) / \sum_{d=1}^{d} gA + gB * \sin\left(2\pi \cdot \frac{d - gC}{D}\right)$$
 JA.8

$$N_{d,p,a,x,2}^3 = N_{d,p,a,x,2}^2 \cdot \theta_d$$
 JA.9

$$N_{d,p,a,x,2}^{2} = N_{d,xp,a,x,2}^{2} \cdot (1 - \theta_{d})$$
JA.10

$$p_{d} = \begin{cases} 2(d-a')/[(c'-a')(b'-a')], & d \le b'\\ 2(c'-d)/[(c'-a')(c'-b')] & , & d > b' \end{cases}$$
JA.11

$$p_{d}^{*} = \begin{cases} if \ d = a', p_{d} \exp^{\varepsilon_{d}} & where \ \varepsilon_{d}^{*} = N(0, \sigma_{\varepsilon}^{2}) \\ if \ d > a', p_{d} \exp^{\varepsilon_{d}^{*} - \sigma_{\varepsilon}^{2}/2} & where \ \varepsilon_{d}^{*} = \varepsilon_{d-1}^{*} \cdot \rho + \sqrt{1 - \rho^{2}} \cdot N(0, \sigma_{\varepsilon}^{2}) \end{cases}$$
 JA.12

$$p'_{d} = p^{*}_{d} / \sum_{d=a'}^{5} p^{*}_{d}$$
 JA.13

$$\theta_d = p'_d / \sum_{d}^{c'} p'_d$$
 JA.14

$$rdur = \begin{cases} 1, if \ rl/(v\bar{s} \cdot 0.0864) < 1\\ round(rl/(v\bar{s} \cdot 0.0864), 0), if \ rl/(v\bar{s} \cdot 0.0864) > 1 \end{cases}$$
 JA.15

$$sm = \begin{cases} e^{lA \cdot (pdays - lB)}, pdays \ge 0.45\\ lC , pdays < 0.45 \end{cases}$$
 JA.16

$$sdur = e^{LN(sm,(CV_{sdur},sm)^2)}$$
JA.17

$$E_d = \sum_a \sum_p (\overline{W}_p \overline{e} / \overline{B}) \cdot N^4_{d,p,a,1,2}$$
 JA.18

$$\overline{W}_p = 10^{WA + WB \log_{10} Lp}$$
 JA.19

$$pm_{x,a,p} = e^{pmA_x + pmB_x \cdot L_{x,a,p}} / (1 + e^{pmA_x + pmB_x \cdot L_{x,a,p}})$$
 JA.20

Table 4.	Definitions of symbols used in the mathematical description of the egg, yol	lk-sac larvae,	post-
larvae/Y	OY sub-model.		

Definitions
Number of daily egg batches
Number of eggs in batch <i>b</i> on day <i>d</i> , number of yolk-sac larvae from batch <i>b</i> on day <i>d</i> .
Daily instantaneous natural mortality rate for eggs, M multiplier for eggs
Egg natching time (days) at temperature on day <i>d</i> , day batch of eggs is
laid, day eggs natch, fraction of eggs that successfully natch at
temperature (°C) on day d
Hatching time-temperature equation parameters
Deily instantaneous natural mostality rate for yolkees large. M multiplier for
volksec larvae, wi multiplier for
Yolksac absorption time (days) at temperature on day d day yolksac is fully
absorbed
Yolksac absorption-temperature equation parameters
Water temperature (°C) on day d
Number of post-larvae in platoon <i>p</i> from batch <i>b</i> on day <i>d</i> in region <i>r</i>
Post-larval capacity (larvae per km ²), lake area (km ²), carrying capacity,
batch carrying capacity
Proportion of post-larvae from batch <i>b</i> in length platoon <i>p</i> , total length (mm) of
post-larvae in platoon <i>p</i> , the daily increment of length increase
variance of \overline{L}^{Post}
Minimum length, maximum length, mean length (natural log scale), number of
post-larvae length platoons
Daily instantaneous natural mortality rate for post-larvae/YOY of batch b and
platoon <i>p</i> on day <i>d</i> , regional M multipliers for post-larvae <22 mm TL and >22
mm TL
Post-larvae emigration length, day on which $L_p^{Post} \ge ml$, last day of emigration
Probability of emigration based on temperature at start of day d, proportion
emigrating on day d
Swimming speed (cm/sec), equation parameters
Mean weight (g) of platoon p derived from bioenergetics model
River and estuary duration multiplier

$N_{b,d}^{Eggs} = N_{b,d-1}^{Eggs} e^{-M^{Eggs} \zeta^{Eggs}}$	EY.1
$h_d = dhA \cdot (1.8T_d + 32)^{-dhB}$	EY.2
$d(h) = \sum_{j=d(b)}^{d-1} \frac{1}{h_j} \ge 1$	EY.3
$o_d = (oA + oB(1.8T_d + 32) - oC \cdot (1.8T_d + 32)^2)/100$	EY.4
$N_{b,d}^{Yolksac} = \begin{cases} N_{b,d-1}^{Eggs} e^{-M^{Eggs}\zeta^{Eggs}} o_{d-1}, & d = d(h) \\ N_{b,d-1}^{Yolksac} e^{-M^{Yolksac}\zeta^{Yolksac}}, & d > d(h) \end{cases}$	EY.5

$$y_d = dyA \cdot T_d^{dyB/T_d}$$
 EY.6

$$d(y) = \sum_{j=d(h)}^{d-1} \frac{1}{y_j} \ge 1$$
 EY.7

$$N_{b,d(y)}^{Post,4} = N_{b,d-1}^{Yolksac} e^{-M^{Yolksac}\zeta^{Yolksac}} , d = d(y)$$
 EY.8

$$k_b = k/nb$$
 EY.9

$$N_{b,d}^{Post,4} = \begin{cases} k_b, & N_{b,d}^{Post,4} \ge k_b \\ N_{b,d}^{Post,4}, & N_{b,d}^{Post,4} < k_b \end{cases}$$
EY.10

$$L_{p}^{Post} = \begin{cases} L_{\min}, & p = 1\\ L_{p-1}^{Post} + \frac{L_{\max} - L_{\min}}{np - 1}, p > 1\&p < np\\ L_{\max}, & p = np \end{cases}$$
EY.11

$$f(L_p^{Post}) = 1/L_p^{Post}\sigma\sqrt{2pi} \cdot \exp(-(\log eL_p^{Post} - \mu_{\log e})^2/2\sigma^2)$$
 EY.12

$$\pi_{b,p}^{Post} = f(L_p^{Post}) / \sum_{p=1}^{np} f(L_p^{Post})$$
EY.13

$$N_{b,d,p}^{Post,4} = multinomial(\pi_{b,p}^{Post}, N_{b,d}^{Post,4})$$
EY.14

$$N_{b,d,p}^{Post,r} = N_{b,d-1,p}^{Post,r} e^{-M_{d-1,p}^{Post} \zeta^{Post}}$$
EY.15

Table 5 cont.

$$M_{d,p}^{Post} = \begin{cases} MP1A \cdot \exp^{MP1B * T_d} (L_p^{Post})^{-MP1C} \zeta_{r,1}^{Post}, & L_p^{Post} \le 22 \ mm \ TL, \\ (MP2A \cdot (L_p^{Post})^{-MP2B}) / D \zeta_{r,2}^{Post}, & L_p^{Post} > 22 \ mm \ TL \end{cases}$$
EY.16

$$N_{b,d,p}^{Post,3} = \frac{\psi_d \tau_d}{\sum_a^{d(e)} \psi_d \tau_d} N_{b,d,p}^{Post,4}, \text{ where } L_p^{Post} > ml$$
EY.17

$$\psi_{d \ge d(s)} = \begin{cases} 1, & \frac{\exp^{uA - uB \cdot T_d}}{1 + \exp^{uA - uB \cdot T_d}} \ge 0.5 \\ 0, & \frac{\exp^{uA - uB \cdot T_d}}{1 + \exp^{uA - uB \cdot T_d}} < 0.5 \end{cases}$$
EY.18

$$n_{d} = \begin{cases} NB(1, mu = 10.71, size = 0.244), \text{rainfall} < 18mm \\ N(1, mean = 473, sd = 183.05), \text{ rainfall} \ge 18mm \\ NB(1, mu = 10.71, size = 0.244), \text{ } d > 285 \end{cases}$$
EY.19

$$\tau_d = n_{d-1} / \sum_{j=d-1}^{d(e)-1} n_{j-1}$$
 EY.20

$$N_{b,d,p}^{Post,4} = N_{b,d,p}^{Post,4} - N_{b,d,p}^{Post,3}$$
 EY.21

$$V = velp1 \cdot \overline{W}_p^{Post^{velp2}} e^{velp3 \cdot T_d}$$
 EY.22

$$rdur = \begin{cases} 1, & V/8.64e9/rl \cdot \varpi_3 < 1\\ round(V/8.64e9/rl \cdot \varpi_3, 0), & V/8.64e9/rl \cdot \varpi_3 \ge 1 \end{cases}$$
 EY.23

$$edur = \begin{cases} 1, & V/8.64e9/el \cdot \varpi_2 < 1\\ round(V/8.64e9/el \cdot \varpi_2, 0), & V/8.64e9/el \cdot \varpi_2 \ge 1 \end{cases}$$
 EY.24

Symbol	Definitions
	For $I \leq 22 \ mm T I$ intercent and coefficient for length-weight equation
WAS,WDS WAI WRI	For $L_p \leq 22 \text{ mm TL}$, intercept and coefficient for length-weight equation
VV AL, VV DL	For $L_p > 22$ min 1 L, intercept and coefficient for length-weight equation Consumption rate of plateon $p(q/q/d_{2V})$, maximum feeding rate $(q/q/d_{2V})$
U_p, U_{max}	Dreparties of maximum consumption for plateen n bace properties
$pr_p, ps, \alpha, \alpha, \beta, \Omega_r$	Proportion of maximum consumption for platoon <i>p</i> , base proportion
	maximum consumption, intercept of <i>pr</i> versus length equation, intercept
	growth multiplier
f(TC), f(TR)	Temperature dependence function for consumption, respiration
KA, KB	Increasing portion of temperature dependence function, decreasing portion
	of temperature dependence function
EDP	Energy density of prey (joules/g)
CA, CB	Intercept and exponent of mass dependence function
<i>CQ,CK</i> 1	Temperature at which temperature dependence is a small fraction of
	the maximum rate, temperature dependence parameter
СТО, СК4	Water temperature corresponding to 0.98 of the maximum consumption
	rate, temperature dependence parameter
CTL, CTM	Temperature at which dependence is some reduced fraction of CK4 of the
D	maximum rate, temperature at which dependence is still 0.98 of the maximum
R_p	Specific rate of respiration (joules/g/day)
RAS, RBS, RQS	For $L_p \leq 49 \ mm \ TL$, intercept of mass function (g/g/day), slope of mass
	function, approximates the Q10 (1/°C)
RAL, RBL, RQL	For $L_p > 49 mm TL$, intercept of mass function (g/g/day), slope of mass
	function, approximates the Q10 (1/°C)
ACT, VEL	Activity multiplier as a function swimming speed, swimming speed (cm/sec)
RTO	Coefficient for swimming speed dependence on metabolism (sec/cm)
RK1, RK4, RTL	Intercept for swimming speed above cutoff temperature (cm/sec), mass
	dependence coefficient for swimming speed at all water temperatures,
	cutoff temperature at which the activity relationship changes (°C)
AM, BACT	Intercept for swimming speed versus mass relationship at
	temperature< <i>RTL</i> (cm/sec for a 1 gram fish at 0°C), water temperature
	RTL(1/°C)
<i>OC</i>	Oxycalorific coefficient used to convert grams to joule equivalents
ER _p , ERA	Egestion rate (g/g/day) of platoon <i>p</i> , constant proportion of <i>C</i>
U_p, UA, S_p, SDA	Excretion (joules/g/day), constant proportion of assimilated energy, proportion
	assimilated energy lost to specific dynamic action, specific dynamic action
ED, EDA_d, EDB_d	Energy density of Alewife (joules/g), intercept of length-ED relationship for day
	d, slope of length-ED relationship for day d
\tilde{T}_p, Ta, Tb, Tc	Optimal temperature (°C) for platoon <i>p</i> , asymptote parameters for optimal
	temperature versus platoon length equation

Table 6. Definitions of symbols used in the mathematical description of bioenergetics sub-model.

$$\overline{W}_{p} = \begin{cases} e^{WAS + WBS \cdot \log e(Lp)}, L_{p} \leq 22 \ mm \ TL, \\ e^{WAL + WBL \cdot \log e(Lp)}, L_{p} > 22 \ mm \ TL, \end{cases}$$
BE.1

$$C_p = C_{max,p} \cdot EDP \cdot pr_p \cdot f(T_d)$$
BE.2

$$C_{max,p} = \begin{cases} 0.8464, \ \bar{W}_p < 1 \ g \\ C_A \cdot \bar{W}^{CB} & \bar{W} > 1 \ g \end{cases}$$
BE.3

$$pr_p = \begin{cases} ps \cdot \Omega_r, L_p \le 22mmTL \\ (\alpha + \beta \cdot L_p) \cdot \Omega_r, L_p > 22mmTL \end{cases}$$
BE.4

$$\beta = \begin{cases} 0.0021, \quad \alpha' + \beta' / \sum_{b} N_{b,d}^{Post,4} < 0.0021 \\ \alpha' + \beta' / \sum_{b} N_{b,d}^{Post,4}, \quad 0.0021 \le \alpha' + \beta' / \sum_{b} N_{b,d}^{Post,4} \le 0.0026 \end{cases}$$
BE.5

$$\left(0.0026, \ \alpha' + \beta' / \sum_{b} N_{b,d}^{Post,4} \ge 0.0026\right)$$

$$f(T_d) = KA \cdot KB$$
BE.6

$$KA = \frac{CK1 \cdot L_1}{(1 + CK1(L_1 - 1))}$$
BE.7

$$L1 = \exp^{G_1 \cdot (T_d - CQ)}$$
BE.8

$$G_1 = \frac{1}{CTO - CQ} \ln\left(\frac{0.98(1 - CK1)}{CK1 \cdot 0.02}\right)$$
BE.9

$$KB = \frac{CK4 \cdot L_2}{1 + CK4(L_2 - 1)}$$
BE.10

$$L_2 = \exp^{G_2(CTL - T_d)}$$
BE.11

$$G_2 = \frac{1}{CTL - CTM} \ln\left(\frac{0.98(1 - CK4)}{CK4 \cdot 0.02}\right)$$
BE.12

$$R_{p} = \begin{cases} RAS(\overline{W}_{p}^{Post})^{RBS} \cdot ACT \cdot f(TR)_{d} \cdot OC, & Lp \leq 49 \ mm \ TL \\ RAL(\overline{W}_{p}^{Post})^{RBL} \cdot ACT \cdot f(TR)_{d} \cdot OC, & Lp > 49 \ mm \ TL \end{cases}$$
BE.13
(1 L₁ < 49 mm TL)

$$ACT = \begin{cases} 1, L_p \leq 49 \text{ mmTL} \\ \exp & \text{RTO-VEL}, L_p > 49 \text{ mmTL} \end{cases}$$
BE.14

$$VEL = \begin{cases} RK1(\overline{W}_p^{Post})^{RK4}, T_d > RTL \\ A_m \overline{W}_p^{Post}^{RK4} \exp^{BACT \cdot T_d}, T_d \le RTL \end{cases}$$
BE.15

$$f(TR)_d = \exp^{RQ \cdot T_d}$$
BE.16

$$ER_p = ERA \cdot C_p$$
 BE.17

$$U_p = UA(C_p - ER_p)$$
BE.18

$$S_p = SDA \cdot (C_p - ER_p)$$
BE.19

$$\overline{W}_{d,p}^{Post} = \overline{W}_{d-1,p}^{Post} + \Delta B_{d-1,p}$$
BE.20

$$\Delta B_{d-1,p} = \frac{(C_{d-1,p} - R_{d-1,p} - ER_p - U_p - S_p)}{ED_{d-1,p}} \overline{W}_{d-1,p}^{Post}$$
BE.21

$$ED_{d,p} = \begin{cases} EDA1 + EDB1 \cdot L_{d,p} , d \le 273 \\ EDA2 + EDB2 \cdot L_{d,p} , d > 273 \end{cases}$$
BE.22

$$L_{d,p} = \begin{cases} e^{(\log e(\bar{W}_{d,p}^{Post}) + WAS)/WBS}, L_{d-1,p} \le 22 \ mm \ TL \\ e^{(\log e(\bar{W}_{d,p}^{Post}) + WAL)/WBL}, L_{d-1,p} > 22 \ mm \ TL \\ \tilde{m} = \begin{cases} T_a/(1 + e^{T_b + T_c L_p}), & L_p < 42 \ mm \ TL \end{cases}$$
BE.23

$$\tilde{T}_{p} = \begin{cases} I_{a}/(1+e^{i_{b}}), & L_{p} < 42 \ mm \ IL \\ 18.2, & L_{p} \ge 42 \ mm \ TL \end{cases}$$
BE.24

	Run		Adult		Sex		Egg		YOY		Age-1
Parameter	Size	Parameter	Mean TL	Parameter	Ratio	Parameter	Ν	Parameter	Ν	Parameter	Mean TL
МВ	1060136	Ω2	23.14	p	0.448	Ω2	1.14E11	k	6129498	СА	68.99
СА	1032472	СВ	11.34	pmB1	0.317	MB	6.60E10	ξ_1	3262671	Ω_2	46.27
RA50	826973	MB	8.10	pmB2	0.293	СВ	6.29E10	СА	2637403	RTO	39.91
k	698598	СА	6.64	pmA2	0.267	СА	5.81E10	MP2B	2046166	RAL	39.89
Ω_2	639013	EDA273	6.53	pmA1	0.266	ξ_1	5.18E10	EDP	1848882	EDP	37.70
MP2B	555478	b'	5.56	MB	0.159	MP2B	4.05E10	EDA273	1414878	ERA	22.89
ξ_1	475550	pmA2	3.65	Ω_2	0.152	RAL	3.96E10	b'	1321743	СВ	19.85
α	457451	pmB2	3.24	СА	0.133	k	3.82E10	μ_{loge}	1244198	RAS	19.77
ERA	430475	EDB273	2.46	b'	0.112	ERA	3.21E10	ρ	1040644	α	16.95
RTO	428328	ERA	2.30	К2	0.109	\overline{B}	2.84E10	V	1032588	EDA273	16.83
СВ	417892	V	2.18	MP2B	0.107	α	2.55E10	MB	968857	RQS	16.58
RQS	348760	ξ_1	2.08	K1	0.086	EDB274	2.32E10	ERA	954126	UA	13.52

velp1

2.17E10

2.11E10

2.07E10

1.99E10

1.94E10

1.56E10

1.46E10

1.44E10 Ω_2

pmA1

ps

RAS

RQS

RAL

RTO

MP1C

0.077

0.067 RTO

0.063 *ē*

0.055 RQS

0.052 EDP

0.051 UA

0.050 RQL

0.065 EDA273

833169 *b*'

767496 μ_{loge}

750182 SDA

647212 RQL

615851 RBL

680269

650904

682251 pmA2

velp1

velp3

12.43

12.03

11.76

11.29

11.20

9.85

9.84

9.44

EDP

RQL

velp1

EDA273

EDB274

ndaysT

UΑ

ρ

333956 ndaysT

276194 pmA1

275004 MP2B

269654 pmB1

263594 EDB274

266816 dhB

248770 RQS

239476 RAL

1.78 EDP

1.61 *L*∞.1

1.60 *L*∞,2

1.49 RAS

1.51 ρ

1.59 *nyrsT*

1.50 EDA273

1.72 v

Table 8. Results of Morris sensitivity analysis showing the top-twenty μ s for six output variables. TL = total length (mm) and N = abundance. See Figure 3 for more details.