

Using Cell Cycle Measurements with Flow Cytometry to Predict the Growth Rate of Walleye  
Pollock *Gadus chalcogrammus* Larvae

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## 14    Abstract

15        Growth rate during early life stages can be an important factor in the recruitment process of  
16    marine fishes, and in this study we develop a methodology based on cell cycle measurements to  
17    predict the growth rate of Walleye Pollock (*Gadus chalcogrammus*) larvae. Results from  
18    cell-cycle analysis of muscle cell nuclei of laboratory-reared Walleye Pollock larvae measured  
19    with flow cytometry were used as covariates in a generalized additive model to predict the  
20    growth rate (growth in length per day,  $\text{mm d}^{-1}$ ) of individual larvae from the day exogenous  
21    feeding began (first feeding) to the time of capture ( $r^2 = 0.79$ ). Additional covariates used were  
22    temperature and standard length. A generalized additive model to classify a larva as fast- or  
23    slow-growing was also formulated using the same covariates. Validation testing with an  
24    independent set of 45 laboratory-reared larvae showed that 33% of the laboratory growth rates  
25    (i.e., growth rate based on age and size) fell within the 95% confidence interval of the predicted  
26    growth rates, and predicted growth rates were significantly less than otolith-derived growth rates.  
27    The growth classification model was more accurate than the growth rate model, correctly  
28    classifying the growth rate type (fast or slow growing) of 71% of the same set of larvae showing  
29    that flow cytometric cell cycle analysis may be better suited for classifying larvae as fast- or  
30    slow-growing rather than for predicting absolute growth rate. Predicted growth rates for larvae  $\leq$   
31    11 mm in length collected from the Gulf of Alaska were within the range of published values for  
32    that area, and were not significantly different than corresponding otolith-derived growth rates.  
33    The growth rates of Gulf of Alaska larvae  $> 11$  mm were overestimated when compared to both  
34    otolith-derived growth rates and published values, and that may be due in part because those  
35    larvae were outside of the size range used to formulate the model. Predicted growth rate and  
36    growth classification of late-stage, field-collected Walleye Pollock larvae would be improved by

37 adding larvae as large as 15 mm to the model to encompass the size range of larvae typically  
38 collected during spring ichthyoplankton surveys conducted in the Bering Sea or Gulf of Alaska.  
39 Flow cytometric cell cycle analysis offers promise as an alternative method for determining  
40 growth rate of fish larvae when otolith daily increments cannot be reliably counted.

41

42 Keywords

43 Flow cytometry

44 Cell cycle

45 Growth rate

46 Fish larvae

47 Walleye Pollock

48 *Gadus chalcogrammus*

49

## Introduction

Growth rate plays a crucial role in determining survival during the larval stage of marine fishes, and small changes in growth or mortality rates during early life may affect year-class strength (Houde, 1987). The ‘growth-mortality’ hypothesis is used to explain survival of marine fish larvae and is based on fast growth being advantageous (Anderson, 1988). Components of that hypothesis are the ‘bigger is better’, ‘growth rate’, and ‘stage duration’ hypotheses. The ‘bigger is better’ hypothesis states that for individuals of the same age, larger larvae are more likely to survive than smaller ones due to size-dependent predation mortality (Leggett and DeBlois, 1994). A decrease in the growth rate of fish larvae due to suboptimal prey abundance or temperature or other unfavorable environmental conditions can increase the duration that high mortality rates operate over (Houde, 1987). The ‘growth rate’ hypothesis relates fast growth to reducing the amount of time spent at sizes where mortality rates are highest (Ware, 1975). The ‘stage duration’ hypothesis is based on fast growth reducing the amount of time an individual spends in the high mortality larval stage; that is, increased survival by transitioning to the juvenile stage sooner (Chambers and Leggett, 1987). Thus, an individual’s survival is affected by both its growth rate and size-specific mortality rate. Model simulations have shown selectivity for fast-growing larvae (Rice et al., 1993; Akimova et al., 2016). For example, 80% of Bloater *Coregonus hoyi* larvae surviving to day 60 came from the upper 25% of the initial growth rate distribution (Rice et al., 1993), most likely because of reduced mortality with increasing size. Atlantic Cod *Gadus morhua* larvae surviving through settlement in the North Sea had growth rates approximately equal to fish fed *ad libitum* in the laboratory (Akimova et al., 2016). Field studies also show selection for faster growing fish larvae. Slow-growing Japanese Anchovy *Engraulis japonicus* larvae were more likely to be preyed upon than faster growing individuals

(Takasuka et al., 2003, 2004). Otolith microstructure analysis showed that fast-growing Atlantic Cod larvae, and Bluefish *Pomatomus saltatrix* larvae had higher survival (Meekan and Fortier, 1996; Hare and Cowen, 1997). Walleye Pollock larvae in the Bering Sea grew faster and had higher survival into summer during relatively warm years compared to cold years (Hunt et al., 2011). A laboratory study of yellowtail kingfish, *Seriola lalandi*, larvae reared under projected climate change conditions (increasing ocean temperature and  $p\text{CO}_2$ ) showed that larvae at 25°C grew and morphologically developed faster than larvae at 21°C, but larvae at the higher temperature had lower survival (Watson et al., 2018) indicating that future climate change could negate the advantage of growing fast because of higher mortality. Growth during the larval stage may also affect over-winter survival of young-of-the-year, larger young-of-the-year Walleye Pollock *Gadus chalcogrammus* are more likely to survive than smaller individuals (Heintz and Vollenweider, 2010).

Otolith microstructure analysis, the RNA-DNA ratio, and flow cytometric cell-cycle analysis have been used to estimate growth rates of fish larvae (Dougherty, 2008; Caldarone et al., 2003; Caldarone, 2005; Theilacker and Shen, 2001; Domingos et al., 2012). The number of otolith increments and/or increment width is widely used to estimate the growth rate of fish larvae and typically involves counting the daily increments formed on the otolith and then modeling growth based on size at age (Dougherty, 2008). Problems with that method arise from increment compression due such conditions as cold temperature or lack of prey that slow growth and hamper the ability to reliably age larvae. The RNA-DNA ratio and temperature have been used to model protein-specific, short-term growth rate of Atlantic Cod and Haddock *Melanogrammus aeglefinus* larvae (Caldarone et al., 2003; Caldarone, 2005), and weight-specific growth rate for different species of fish larvae (Buckley et al., 2008). The amount of variance in growth rate

96 explained by those models ranged from 39% to 82%. Using flow cytometric cell-cycle analysis  
97 to estimate larval growth is based on the premise that cell proliferation is related to somatic  
98 growth rate. In flow cytometry, individual cells or nuclei are stained with target-specific dyes,  
99 passed through laser beams, and their fluorescence is measured. Flow cytometric cell-cycle  
100 analysis is a technique that measures the DNA content of individual cells in a population based  
101 on their fluorescence to determine the proportion of cells in different stages of the cell cycle, and  
102 can be used on specific tissues of individual fish larvae or whole larva homogenates (Theilacker  
103 and Shen, 2001; Domingos et al., 2012). The cell cycle is divided into two parts: interphase and  
104 mitosis (M). Interphase consists of three stages: gap 1 (G1), DNA synthesis (S), and gap 2 (G2).  
105 Cell growth occurs during the G1 stage before cell division begins, and cells may enter a G0  
106 resting state after this phase in response to starvation or other unfavorable environmental  
107 conditions (Murray and Hunt, 1993). For cells to divide they must first replicate their DNA (S),  
108 and then grow and produce the structures necessary for mitosis (G2). The DNA content of cells  
109 in the S-stage increases to up to twice the amount of DNA as G1 stage cells. G2 and mitosis  
110 (before cell division) stage cells have twice the amount of DNA as G1 stage cells. The fraction of  
111 cells in S, and G2 and mitosis (hereafter referred to as G2 and M) are indicative of cells that may  
112 divide. For Walleye Pollock larvae the fraction of cells in S, and G2 and M-stages are responsive  
113 to recent feeding history (Theilacker and Shen, 2001; Porter and Bailey, 2011) and can indicate  
114 changes in cell division. The fraction of S-stage nuclei of muscle cells of Walleye Pollock larvae  
115 significantly decreased after a few days of prey being withheld ( $\leq 3d$ , Porter and Bailey, 2011)  
116 suggesting reduced cell division. The fraction of G2 and M-stage nuclei increased during  
117 starvation (Porter and Bailey, 2011) and may indicate cells staying in the G2-stage due to a  
118 checkpoint between this stage and mitosis that slows or halts cell division if energy reserves are

low (Murray and Hunt, 1993). Cell-cycle analysis of muscle tissue of Walleye Pollock larvae (Theilacker and Shen, 2001), and whole Barramundi *Lates calcarifer* larvae (Domingos et al., 2012) has shown that the fraction of cells dividing (S + G2 and M stages) is positively correlated with both larval size and growth in length. For both studies slightly more than 50% of the variance was explained in models using cell cycle analysis to estimate growth of those larvae (Theilacker and Shen, 2001; Domingos et al., 2012).

Measuring the growth rate of marine fish larvae is valuable for understanding recruitment dynamics because fast growth increases the probability of surviving this life stage. In this study we formulate and test a generalized additive model to predict the growth rate (growth in length per day,  $\text{mm d}^{-1}$ ) of individual Walleye Pollock larvae using covariates from cell-cycle analysis with flow cytometry, temperature, and standard length. This approach may offer an alternative for growth rate estimates when otolith microstructure analysis is unreliable. The model was developed and validated with laboratory-reared larvae and then tested on larvae collected from the Gulf of Alaska. Growth rates predicted from the model and those calculated from otolith microstructure analysis were compared.

## Material and methods

### *1.1 Laboratory rearing and growth studies*

Adult Walleye Pollock were collected by trawl in Shelikof Strait, Gulf of Alaska, during the spawning season in March 2014, 2015, and 2017. Eggs from single fish pairings (one female and one male) were fertilized and maintained aboard ship in the dark at 3°C before being transported to the Alaska Fisheries Science Center, Seattle, WA. Larvae were reared at 3.2° and 6.1°C in 2014, 1.6° and 3.0°C in 2015, and 8.3°C in 2017 (Table 1). All rearing experiments used one-

micron filtered seawater (33 PSU) and a 16-hour daylight cycle with a  $2.5 \mu\text{mol photon m}^{-2} \text{s}^{-1}$  light level at the water surface from overhead full spectrum fluorescent lights. Two replicate 20 L black, circular tanks each containing approximately 500 larvae were used for each temperature and feeding treatment following procedures described in Porter and Theilacker, 1996. Feeding treatments for the 3.2°, 6.1°, and 8.3°C experiments were high density of prey (10 rotifer  $\text{ml}^{-1}$  and 3 natural zooplankton  $\text{ml}^{-1}$ ), low prey density (1.0 natural zooplankton  $\text{ml}^{-1}$ ), and an unfed treatment (Table 1). Natural zooplankton, which included primarily copepod nauplii (*Acartia* spp.) and gastropod veligers, were collected from a local lagoon and screened through a 202- $\mu\text{m}$  mesh. The high prey density treatment at 1.6°C did not use rotifer prey because they become inactive and settle out of the water column at that temperature. Instead a zooplankton density of 10  $\text{ml}^{-1}$  was used and the other treatments remained unchanged. Removing the rotifers and increasing the density of zooplankton to 10  $\text{ml}^{-1}$  should not negatively affect larval growth rate because Walleye Pollock larvae fed natural zooplankton grow faster than those fed rotifers (Porter and Bailey, 2007). Thus the growth rate of larvae fed 10  $\text{ml}^{-1}$  zooplankton would be expected to be at least equal to that of larvae fed the high density rotifer and zooplankton diet. First feeding was defined as the first day when 50% of the larvae were observed feeding. Mean standard length (SL) of 10 larvae at first feeding for each feeding treatment and temperature were used as the size at that stage when calculating laboratory growth rate. Here laboratory growth rate is defined as growth beginning at first feeding determined from known age and size in the laboratory. For each larva from a specific feeding treatment and temperature, laboratory growth rate was calculated as the difference between size at capture and mean SL at first feeding, divided by the number of days after first feeding. Treatments were sampled three or four times at intervals based on equivalent degree-days among rearing temperatures, ranging from 3 to 12



days between sampling depending on temperature. Larvae in the 1.6°C treatment were sampled on the same days as the 3°C treatments. This ensured that an adequate sample number was taken due to high mortality that was expected to occur at this temperature. The 2015 experiment was designed to take into account the effect of intermittent feeding on growth. This experiment was conducted at 3.0°C using larvae starved for 6 days after first feeding and then fed at the high prey density treatment, and another treatment that fed at high prey density for 6 days after first feeding followed by starvation. An always-fed high prey density treatment was used as a control. For all experiments the number of larvae taken from a tank on a given sampling day ranged from 5 to 36 depending on whether larvae would also be used for model validation testing. Larvae taken from rearing tanks were euthanized in a 1% solution of tricaine methanesulfonate (MS-222), and then a digital photograph was taken of each larva before it was frozen at -80°C. Calibrated digital photographs were used to measure SL and body depth at anus (BDA) of each larva using Image-Pro Plus vers. 4.5 (Media Cybernetics, Inc., Rockville, MD) image analysis software.

## *1.2 Flow cytometry*

DAPI (4',6-diamidino-2-phenylindole), a fluorescent DNA stain, was used at a concentration of 10 µg mL<sup>-1</sup> for cell-cycle analysis with flow cytometry. Nuclear RNA was stained with Invitrogen Syto RNASelect green fluorescent cell stain (S32703, Life Technologies Corp., Carlsbad, CA; hereafter referred to as Syto RNASelect stain) and was used at a concentration of 1000 nM. Tissue preparation, DNA and RNA staining protocols followed Porter and Bailey (2011 and 2013), modified from Theilacker and Shen (2001). A frozen larva was placed into an approximately 100 µL mixture of DAPI and Syto RNASelect stains on a glass depression slide. The head and gut were dissected away from the trunk musculature, and the tissue was sliced into

4–5 pieces with 2 scalpels. The pieces of the trunk musculature were transferred into a microcentrifuge tube that contained a 230- $\mu$ L mixture of DAPI and Syto RNASelect stains and then the mixture was triturated 6 times using a 1 mL syringe with a 25-gauge needle to release the nuclei from the cells. The solution was filtered through a 48- $\mu$ m filter into another microcentrifuge tube to separate the stained nuclei from large cellular debris. Prepared samples were kept on ice until they were analyzed with a BD Biosciences LSRII flow cytometer (BD Biosciences, San Jose, CA), typically within 6 hours of preparation. DAPI was excited with a 405 nm, 100 mW violet laser, and Syto RNASelect stain was excited with a 488 nm, 100 mW blue laser. The DAPI/DNA detector filter was 450/50, and filters used for Syto RNASelect stain/RNA were 505 long pass and 530/30.

Chicken erythrocyte nuclei (Biosure, Inc., Grass Valley, CA) stained with the same mixture of DAPI and Syto RNASelect dyes were used as flow cytometry controls. At the beginning of each flow cytometry session, the control was run and necessary adjustments were made to the laser voltages to keep control fluorescence values similar to previous sessions. Cell cycle analysis is sensitive to tissue quality. High-quality tissue preparation (i.e., quickly freezing larvae and minimizing tissue processing time) is necessary to keep the amount of tissue debris in prepared samples low, and this in turn results in small coefficients of variation for G0 and G1. Samples that had < 5000 nuclei analyzed or a coefficient of variation for G0 and G1 > 9.00 were not used in further analyses and this criteria resulted in rejection of about a third of all samples.

For each larva, the fraction of nuclei in G0 and G1, S, and G2 and M stages of the cell cycle were calculated using MultiCycle AV software, vers. 4.0 (Phoenix Flow Systems, San Diego). FCS Express flow cytometry analysis software, vers. 3.0 (De Novo Software, Los Angeles) was used to calculate the ratio of the number of S-stage nuclei to the number of G1-

stage nuclei with high RNA content (Porter and Bailey 2013; hereafter, this ratio will be referred to as RSG1). RSG1 is a measure of potential cell division based on progression of nuclei from the G1 to S stage. Slower growing larvae may have fewer dividing cells and thus potentially less S-stage nuclei, so the ratio of S to G1 nuclei may vary with growth rate (Porter and Bailey, 2013).

### *1.3 Otolith microstructure analysis*

Heads from frozen larvae were preserved in 100% ethanol after thawing. Otoliths were prepared for microstructure analysis as described in Dougherty (2008). For each larva, otoliths were dissected from the head and mounted to glass microscope slides using clear acrylic finger nail polish. Walleye Pollock larvae deposit increments on their otoliths daily (Dougherty, 2008) and their otoliths typically have an optically distinct increment representing the day of first feeding (Bailey and Stehr, 1988). Otolith increment width is a relative measure of somatic growth between days that the increments were deposited (Campana, 1996). Increments from first-feeding to the otolith edge of both sagitta otoliths were counted, and that distance was measured to nearest micron with an ocular micrometer at 1000X magnification using a compound microscope. Both measurements correspond with growth from first feeding to capture. Otoliths were read and measured along the maximum diameter axis, and the largest value for increment number and distance was recorded.

### *1.4 Growth model formulation and validation*

Generalized additive models (GAM; Wood, 2006) were used to predict the growth rate (growth in length per day,  $\text{mm d}^{-1}$ ) of a larva from the start of exogenous feeding (first feeding) to the time of capture. GAMs are non-parametric regression techniques, in which the effect of a covariate is estimated with a smooth function, typically a natural cubic spline, and therefore may

not necessarily be linear (Wood, 2006). Beginning at first feeding, larvae are dependent upon  
 their surrounding environment for energy, and measuring growth during at this stage gives an  
 overall indication of how well a larva has been feeding and growing. Laboratory growth rate was  
 used as the response variable, and the independent variables (covariates) were SL, BDA,  
 temperature, fraction of S-stage nuclei, fraction of G2-and-M-stage nuclei, and RSG1. The arcsin  
 $\sqrt{x}$  transformation was used to normalize the fraction of S nuclei, fraction of G2 and M nuclei,  
 and RSG1. Four model formulations were tested; these models are not comprehensive of all  
 possible models that could be tested but represent formulations 1) similar to previous linear  
 models that used only cell cycle parameters to model growth rate (Theilacker and Shen, 2001;  
 Domingos et al., 2012), and 2) take into account that the relationship between cell cycle  
 covariates and growth rate may vary with either size or temperature. Two additive GAMs,  
 representative of previous linear models, included all potential cell cycle covariates as well as  
 either SL or the product of SL and BDA for the size covariate (BDA was used to take into  
 account growth in body depth, and the product of SL and BDA is correlated with larval weight,  
 van der Meeren, 1991; models 1 and 2, Table 2). SL and BDA could not be used as separate  
 covariates in the same model because they were highly autocorrelated ( $r=0.96$ ) resulting in  
 multicollinearity in the model. Two variable coefficient GAMs (Bachelier et al., 2009) allowed  
 the relationship between cell cycle covariates and growth rate to vary with either SL or  
 temperature (models 3 and 4, Table 2). The two variable coefficient GAM formulations  
 represented that the relationship between the cell cycle parameter and growth rate was linear  
 (fraction of S-stage nuclei, fraction of G2-and-M-stage nuclei, and RSG1), but could vary with  
 either SL (model 3, Table 2) or temperature (model 4, Table 2). Cell cycle parameters may vary  
 with standard length for such reasons as larvae becoming better predators (i.e., improved

feeding, Porter et al., 2005), and increased digestive tract nutrient assimilation efficiency (Porter and Theilacker, 1999). For Walleye Pollock larvae the fraction of S-stage nuclei increases with temperature (Porter and Bailey, 2011), suggesting that temperature also affects the cell cycle. Backward stepwise variable selection based on Akaike Information Criterion (AIC) was used to select the covariates included in models. A covariate with a p-value  $> 0.05$  was removed from the model and the model was rerun. The covariate was not included in the final model formulation if the AIC decreased upon its exclusion, but it was included if AIC increased. Residuals were visually assessed to check for normality and independence. R version 3.4.0 (R Core Team, 2017) and mgcv package (version 1.8-17; Wood, 2006) were used.

To determine the model that best predicted growth rates, the model formulations were compared using independent validation testing based on the following criteria: 1) how accurately the model predicted laboratory growth rates. This test compared predicted growth rate and corresponding laboratory growth rate for each larva to determine the model with the highest accuracy; 2) how well predicted growth rates followed somatic growth. The distance from the increment at first feeding to the otolith edge represents growth over the time period from first feeding to when the larva was captured. The Pearson correlation coefficient was calculated for the relationship between predicted growth rate and the distance from the first feeding increment to the otolith edge, and the model with the highest correlation would indicate that predictions tracked somatic growth better than the other models; and 3) similarity between predicted growth rates and otolith-derived growth rates. Otolith-derived growth rate is the most common method used to calculate growth rates of fish larvae in the field, and this test gives an indication of how comparable results from the two methods are. Model predicted and otolith-derived growth rate for the same larva can differ because otolith-derived growth rate uses the number of increments

from the first feeding mark to the otolith edge for the number of days from first feeding to capture, and the growth rate models use the first day when 50% of larvae were observed to be feeding as day of first feeding for determining number of days after first feeding a larva was at capture. Some larvae start feeding before and after day of 50% feeding and that will affect the calculation of growth rate. The testing data set consisted of 45 laboratory reared larvae that were not used in the formulation of the models ranging in SL from 6.42 to 8.63 mm from high and low prey density treatments from 1.6°, 3.2°, 6.1°, and 8.3°C (Table 3). A laboratory growth rate and a predicted growth rate from each model was calculated for each larva. Differences between laboratory growth rate and predicted growth rates among the models were compared using repeated measures ANOVA. Otolith-derived growth rate was calculated as the difference in length between the mean SL at first feeding for the corresponding temperature and feeding treatment and size of a larva at the time of capture, divided by the number increments from first feeding to the otolith edge. Differences between otolith-derived growth rate and predicted growth rates among the models were compared using repeated measures ANOVA. SYSTAT vers. 13 (SYSTAT software, inc., San Jose, CA) was used for all statistical testing.

The utility of using flow cytometry to classify larvae as fast- or slow-growing was investigated by formulating binomial GAMs with logit link function similar to the growth rate models. The rational was to compare accuracy of growth rate prediction (absolute growth rate) to growth rate classification to determine which method would be most useful for use in the field. For the response variable, each larva was determined to be fast or slow growing based on predicted SL from a linear regression of age and SL from the high prey density feeding treatment for each temperature. Only the high prey density treatment was used for that procedure because larvae in that treatment were most likely to have growth rates similar to healthy larvae in the

field. For a given sampling day, a larva less than predicted size for that day was identified as slow-growing, and a larva equal to or greater than that size was fast-growing. Growth classification models predicted the probability of fast growth, and were formulated using the same stepwise procedure described for the growth rate models. The validation testing set was used to determine the model that most accurately classified larvae as fast- or slow-growing.

### *1.5 Walleye Pollock larvae growth rate in the western Gulf of Alaska*

Walleye Pollock larvae were opportunistically sampled from the Gulf of Alaska near Kodiak Island, and from the area between the Shumagin Islands and the southern end of Kodiak Island from 18-30 May 2013 during an ichthyoplankton survey conducted by the Alaska Fisheries Science Center (Fig. 1). Collections were made using a 60 cm bongo frame fitted with 505- $\mu$ m mesh nets towed obliquely to a depth of 100 m or 10 m off bottom, whichever was shallower. Water temperature was measured during each tow using a Sea-Bird SBE 19 Plus SeaCat attached to the towing wire. At completion of a tow, Walleye Pollock larvae were immediately removed from the net codend, placed onto a microscope slide with chilled seawater, imaged with a digital camera and individually frozen at -80°C. Mean temperature between the surface and 50 m depth was used as the value for temperature in the growth model because most Walleye Pollock larvae in Shelikof Strait, Gulf of Alaska are located at this depth interval (Kendall et al., 1994). Temperature, SL, transformed fractions of S-stage nuclei, G2-and-M stage-nuclei, and RSG1 were used to predict a growth rate for each larva at a station, and to classify them as fast- or slow-growing. The mean growth rate and mean probability of fast growth of larvae at a station were used as the value for that station. A mean otolith-derived growth rate was also calculated for each station. Size at first feeding was estimated from a linear regression of age and size of Walleye Pollock larvae from Shelikof Strait for 1983-2001 from hatching to 35 days after

hatching ( $n = 4159$ ; A. Dougherty unpublished data). Mean predicted SL of larvae 5, 6, and 7 days after hatching was used as the size at first feeding (5.49 mm) because those ages represent the time when larvae typically begin feeding at the temperatures measured in the study area. Mean model predicted growth rates and otolith-derived growth rates for each station were compared using the paired t-test.

## Results

### *2.1 Growth model formulation*

Linear regressions of SL and days after first feeding for the high prey density treatments at 1.6°, 3.0°, 3.2°, 6.1°, and 8.3°C showed that growth rates were 0.08, 0.09, 0.09, 0.18, and 0.18 mm d<sup>-1</sup>, respectively. Those growth rates indicate that larvae grew well when compared to results from previous rearing experiments at approximately the same temperatures (0.04, 0.06, 0.14, 0.16 mm d<sup>-1</sup> for 1.4°, 3.2°, 5.9°, 9.0°C, respectively; S. Porter unpublished data). Additionally, 9 larvae age-22 days after first feeding, reared at 2.8°C and fed high prey density were included to increase the number of older larvae for temperatures near 3°C. A total of 318 larvae ranging in SL from 5.32 to 11.17 mm were used to formulate the models (Table 3). The 1.6°, 3.0°, 3.2°, and 8.3°C treatments had no larvae remaining after three sampling dates (Table 3).

Models 1, 3, and 4 produced nearly identical  $r^2$  values and similar AIC values, indicating that these models fit the data similarly (Table 2). Model 2 had the largest AIC and smallest  $r^2$  value for the four models tested and was not considered further (Table 2). Models 1 and 3 included all covariates, and model 4 covariates were SL, and the effects of G2-and-M, and S varying with temperature (Table 2). Other covariates (rearing temperature alone, and RSG1) were not included in this model because they were not significant and did not decrease the AIC score.



Predicted growth rates from models 1, 3, and 4 resulting from the independent validation data set were compared to the corresponding laboratory growth rate, correlation with the distance between the first feeding increment and the otolith edge, and to otolith-derived growth rates to determine the model that most accurately predicted growth rates. Overall, predictions from all models were correlated with laboratory growth rates (Fig. 2a, b, c; Table 4). The slopes of linear regressions for the relationship between laboratory and predicted growth rates among models were not significantly different showing that all models predicted growth rate similarly (ANCOVA,  $p = 0.35$ ; Fig. 2a, b, c; Table 4). Mean predicted growth rate for each model was not significantly different from laboratory growth rate for rearing temperatures 1.6°, and 8.3°C (repeated measures ANOVA,  $p > 0.05$ ; Table 5). For 3.2° and 6.1°C, mean predicted growth rates were significantly different from mean laboratory growth rates (repeated measures ANOVA,  $p < 0.01$ ; Table 5). Growth rate was significantly overestimated by each model for 3.2°C, and for 6.1°C all models significantly underestimated growth rate (Table 5). Overall mean percent error between predicted and laboratory growth was similar among models, 23% for models 1 and 4, and 24% for model 3. Model 1 had the highest percentage of accurate predictions (i.e., predicted growth rate of a larva was equal to the corresponding laboratory growth rate; 16%, Table 4), and this model also had the highest percentage of laboratory growth rates that fell within the 95% confidence interval of the predictions (33%; Table 4).

A subset of larvae from the independent validation data set were used for model assessment involving otolith microstructure analysis ( $n = 19$ ). Larvae reared at 1.6° and 3.2°C were excluded because of either increment compression (1.6°C) or non-daily increment deposition (3.2°C) that made determination of the number of daily increments and measurement of the distance from the first feeding increment to otolith edge unreliable. Model 1 predictions had the highest correlation

with the distance between the first feeding increment and the otolith edge indicating that this model had better fit with somatic growth than the other models (Table 6). Mean otolith-derived growth rate was significantly greater and model 1 and 3 predicted growth rates were significantly less than mean laboratory growth rate (repeated measures ANOVA,  $p < 0.01$ ; Table 6). The percentage of otolith-derived growth rates that were within the 95% confidence interval of the predictions of each model was 11% (2/19), 5% (1/19), and 16% (3/19) for models 1, 3, and 4, respectively (Table 6).

Of the models tested, model 1 was chosen as the best model for predicting Walleye Pollock larvae laboratory growth rate (i.e., growth from first feeding). This model showed that increasing temperature had a positive effect on growth over the range of temperatures tested (Fig. 3a), and also indicated that larvae smaller than approximately 6.5 mm SL grew slower than larger larvae, most likely because the smaller size range included starving larvae and larvae that were just beginning to feed (Fig. 3b). Higher growth rates were associated with increasing values of the fraction of G2-and-M-stage nuclei, and fraction of S-stage nuclei (Fig. 4a, b). The effect of RSG1 on growth rate was weak for large ratios as evidenced by the wide 95% confidence interval for those ratios (Fig. 4c). This model had both the highest percentage of correct predictions and the highest percentage of laboratory growth rates that fell within the 95% confidence interval of the predictions for the independent validation testing set (Table 4). Additionally, correlation between growth rate predictions and the distance between the first feeding increment and the otolith edge was highest with model 1 (Table 6). The percentage of otolith-derived growth rates that were within the 95% confidence interval of the predicted values of model 1 was nearly equal to that of model 4 (Table 6).

## 2.2 Classification of growth

Three binomial GAMs for classifying larvae as slow- or fast-growing, similar in formulation to growth models 1, 3, and 4 (growth classification models 1, 2, and 3, respectively; Table 7) were tested using the independent validation data set. The models predicted the probability of fast growth, and a larva with a probability  $\geq 0.50$  was classified as fast-growing. Growth classification models 1 and 3 had identical results with an accuracy of 69% (Table 7), and both more accurately classified fast-growing (82%) versus slow-growing larvae (57%). Growth classification model 2 was slightly more accurate overall (71%; Table 7), correctly classifying one additional slow-growing larva than the other two models and was chosen as the best growth classification model based upon highest overall accuracy. Model 2 indicated that larvae larger than approximately 9.5 mm were more likely to be fast growing than smaller sizes (Fig. 5), possibly because by the time larvae reach that size most individuals in poor condition (slow growing) have died. In general, cell cycle covariates had a strong influence on the probability of small larvae being fast growing and had a lesser effect on large larvae (Fig. 6a, b).

### *2.3 Comparing predicted and otolith-derived growth rates of western Gulf of Alaska Walleye Pollock larvae*

Thirty-one larvae ranging in size from 5.91 to 14.91 mm SL were analyzed from 14 stations: 6 stations were located in the vicinity of Kodiak Island and 8 stations were in the area between Kodiak Island and the Shumagin Islands (Table 8, Fig. 1). One to four larvae were analyzed at each station, and otolith microstructure analysis was completed on all larvae except for the larva from station 187. Poor increment definition of those otoliths prevented the increments from being reliably counted. Predicted growth rates were highly correlated with the distance from the first-feeding increment to the otolith edge ( $r = 0.98$ ), indicating that predicted rates were related to somatic growth. Overall, predicted and otolith-derived growth rates were not significantly

different from each other ( $0.21 \pm 0.08 \text{ mm d}^{-1}$  and  $0.18 \pm 0.01 \text{ mm d}^{-1}$  respectively; paired t-test,  $p = 0.31$ ). However, the greatest differences between predicted and otolith-derived growth rates were consistently at stations where mean larvae size was  $> 11 \text{ mm SL}$  (stations 161, 162, 164, and 191; Table 8). For those stations, predicted growth rates were significantly higher than otolith-derived growth rates ( $0.31 \pm 0.04 \text{ mm d}^{-1}$  and  $0.17 \pm 0.005 \text{ mm d}^{-1}$  respectively; paired t-test,  $p = 0.009$ ; Table 8), and were also higher (range  $0.26$  to  $0.36 \text{ mm d}^{-1}$ ) than the range of published values for Shelikof Strait larvae calculated using otolith microstructure analysis ( $0.14$  to  $0.24 \text{ mm d}^{-1}$ ; Bailey et al., 1996). A contributing factor to the large discrepancy in growth rates between the two methods for larvae  $> 11 \text{ mm SL}$  may be that the growth model was formulated using larvae between  $5.32$  and  $11.17 \text{ mm SL}$  (only one larvae larger than  $11 \text{ mm}$  was used), so growth rates were predicted outside of the range of sizes used in the model causing growth to be overestimated. Additionally, no transforming larvae were included in the growth model, and Walleye Pollock larvae growth rate changes when they begin juvenile transformation about  $12 \text{ mm SL}$  (Brown et al., 2001). Predicted growth rates for stations where mean larvae size was  $\leq 11 \text{ mm}$  were not significantly different from otolith-derived growth rates ( $0.16 \pm 0.04 \text{ mm d}^{-1}$  and  $0.19 \pm 0.01 \text{ mm d}^{-1}$  respectively; paired t-test,  $p = 0.15$ ), and nearly all of the predicted rates were within the range of published values for Shelikof Strait larvae (Table 8). For this smaller size range, at stations where more than one larva was analyzed, mean otolith-derived growth rates were within the 95% confidence interval of the mean predicted growth rates for 6 out of 7 stations, and the mean percent error between the predicted and otolith-derived growth was 20%, much lower than 80% for larvae larger than  $11 \text{ mm}$  in size. Based on these results, the model appears to be most appropriate for larvae  $\leq 11 \text{ mm}$ , and this equates to about the first 4 weeks of feeding based on temperatures that typically occur at depths where Walleye Pollock larvae

reside. Larvae at all stations except for station 78 were classified as fast-growing (i.e., probability of fast growth  $\geq 0.50$ ; Table 8). The predicted growth rate at station 78 was  $0.14 \text{ mm d}^{-1}$ , and that corresponds with the lower end of the range of published growth rates (Bailey et al., 1996) supporting that those larvae were growing slowly. There was a discrepancy between predicted growth rate and growth classification for stations 18 and 187 because both stations were classified as fast-growing but predicted growth rates were the slowest of all the stations (Table 8).

## Discussion

The growth model developed in this study represents an improvement over other models that have used flow cytometric cell cycle analysis to predict growth rates of fish larvae (Theilacker and Shen, 2001; Domingos et al., 2012). Cell cycle covariates, larval length, and temperature accounted for 79% of the variance used to predict growth rate of Walleye Pollock larvae in our study. Previous models used only cell cycle covariates and explained about 55% of the variance in growth rate and were based on only one temperature, so they may be better suited for application in the laboratory rather than to the field. Growth rate of larvae reared in the high prey density treatment at  $6.1^{\circ}\text{C}$  ( $0.18 \text{ mm d}^{-1}$ ) was within the range of growth rates for Gulf of Alaska Walleye Pollock larvae at similar temperature (Bailey et al., 1996) supporting that predicted growth rates should reflect growth in the field. A laboratory study that reared Walleye Pollock larvae at  $0^{\circ}$ ,  $2^{\circ}$ ,  $5^{\circ}$ , and  $12^{\circ}\text{C}$  reported that larvae grew fastest (% mass  $\text{day}^{-1}$ ) at the highest temperature tested, and survival was highest at  $2^{\circ}\text{C}$  and lowest at  $12^{\circ}\text{C}$  (Koenker et al., 2018). Survival at  $5^{\circ}\text{C}$  was nearly equal to survival at  $2^{\circ}\text{C}$  (Koenker et al., 2018) suggesting that the optimal temperature range for survival in the field may be  $2^{\circ}$  to  $5^{\circ}\text{C}$ . The optimal temperature for

Walleye Pollock larvae growth is not consistent among studies. In our study growth rate at 6.1° and 8.3°C was alike indicating that a temperature threshold may have been reached, and other studies have found no significant difference in growth between 7.4° and 9.9°C (Porter and Bailey 2007) or that growth rate continued to increase up to 12°C (Koenker et al., 2018). The growth model produced reasonable results when applied to Gulf of Alaska larvae  $\leq 11$  mm SL as shown by predicted growth rates within the range of published values for larvae in that area. Confirmation of the growth model's utility is shown for the larva at station 187 (Table 8). The otoliths of that larva were unreadable due to poor increment definition so the otolith aging method could not be used to calculate growth rate. In this circumstance, the predicted growth rate was 0.05 mm d<sup>-1</sup> (Table 8) indicating slow growth. Poor increment definition can be an indication of weak growth (A. Dougherty, pers. comm.), supporting the model prediction.

The model for classifying larvae as fast- or slow-growing had a somewhat lower overall classification accuracy (71%) than a logistic regression model developed for Walleye Pollock larvae using the sum of the fractions of nuclei in S-stage and G2-and M-stages, and standard length as covariates (79%; Theilacker and Shen, 2001). The difference in classification accuracy between these models may be due, in part, because the Theilacker and Shen model was tested with the same data used to formulate it. In that case, classification accuracy would likely be higher compared to using an independent data set as was done here. Flow cytometric cell cycle analysis may be better suited for classifying larvae as fast- or slow-growing rather than predicting absolute growth rates because the accuracy of the growth classification model was higher than the growth rate model. Thirty-three percent of laboratory growth rates of the test set larvae were within the 95% confidence interval of the predicted growth rates compared to 71% of larvae being correctly classified as fast- or slow-growing. Growth rate predictions and growth

classification for late-stage, field-collected Walleye Pollock larvae would be improved by adding larvae as large as 15 mm to the models to encompass the range of sizes of larvae collected during spring ichthyoplankton surveys conducted in the Bering Sea or Gulf of Alaska.

Differences among predicted, laboratory, and otolith-derived growth rates for the same larva may be partially due to how age and size at day of the start of exogenous feeding (first feeding) was determined. Day of first feeding was defined for this study as the day when 50% of the larvae were observed to have started feeding, and the mean SL for each treatment on this day was used as size at first feeding. The initiation of first feeding for Walleye Pollock larvae in the laboratory generally occurs over a 3-day period (Porter and Theilacker, 1999). Larvae that start feeding at the beginning of this period would be younger and most likely smaller than the mean length at 50% feeding; larvae that start feeding near the end of the first feeding period would be older and possibly larger than the mean. This variation in the onset of feeding can affect calculation of growth rates. For example, larvae from the 6°C independent validation test set were classified as age 13 days after first feeding based on the day when 50% of the larvae started feeding. Otolith increment counts of that group of larvae ranged from 10 to 15 days after first feeding, indicating that some larvae began feeding five days earlier than others. This range in age when feeding actually began will cause differences between predicted growth rates (that are based on age beginning when 50% of the larvae started feeding), and otolith-derived growth rates that use the number of increments from first feeding to the otolith edge for age. Additionally, flow cytometric cell cycle analysis can detect differences between starved and fed fish larvae after 1 to 3 days of starvation (Theilacker et al., 1996; Catalán et al., 2007; Porter and Bailey, 2011), indicating that cell cycle stages may be more related to recent growth rate (i.e.,

growth within the last 1 to 3 days before capture) rather than growth rate from first feeding to capture that the otolith-derived growth rates represented in our study.

The recruitment process of Walleye Pollock is initiated at the egg and larval stage, with larval survival being a critical first step. Variability in growth rate during early life can affect the survival of larvae due to size-dependent mortality (Leggett and DeBlois, 1994), and that can potentially affect year class strength. The abundance of Walleye Pollock larvae reaching 15 mm SL was positively related (although not statistically significant) to recruitment for years 1979 to 1991 (Bailey et al., 2012). In our study, the models worked best for larvae during the period of about the first 4 weeks of feeding, and studies indicate that growth early in the larval period may have a more significant effect on survival than growth later in this period. For example, field studies have shown that for a shallow-water reef fish the Rainbow Wrasse, *Coris julis*, size at age 30 days was positively correlated with recruitment (Fontes et al., 2011). For King George Whiting, *Sillaginodes punctate*, larvae, growth from hatching to 25 - 30 days was the most critical period for survival (Jenkins and King, 2006). Flow cytometric cell cycle analysis offers promise as an alternative method for determining growth rate of fish larvae when otolith daily increments cannot be reliably counted.

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- 532 National Marine Fisheries Service, NOAA. Reference to trade names does not imply
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## Figure Captions

Figure 1. Sampling stations where Walleye Pollock larvae (*Gadus chalcogrammus*) were collected in the Gulf of Alaska in May 2013.

Figure 2. Relationship between laboratory and predicted growth rate for models 1 (a), 3 (b), and 4 (c).

Figure 3. Partial effects of temperature (°C; a) and standard length (mm; b) on growth rate (mm d<sup>-1</sup>) of Walleye Pollock larvae (*Gadus chalcogrammus*) for growth model 1. Tick marks on X-axis show either rearing temperatures or individual standard length measurements. The 95% confidence interval is shown by dashed lines.

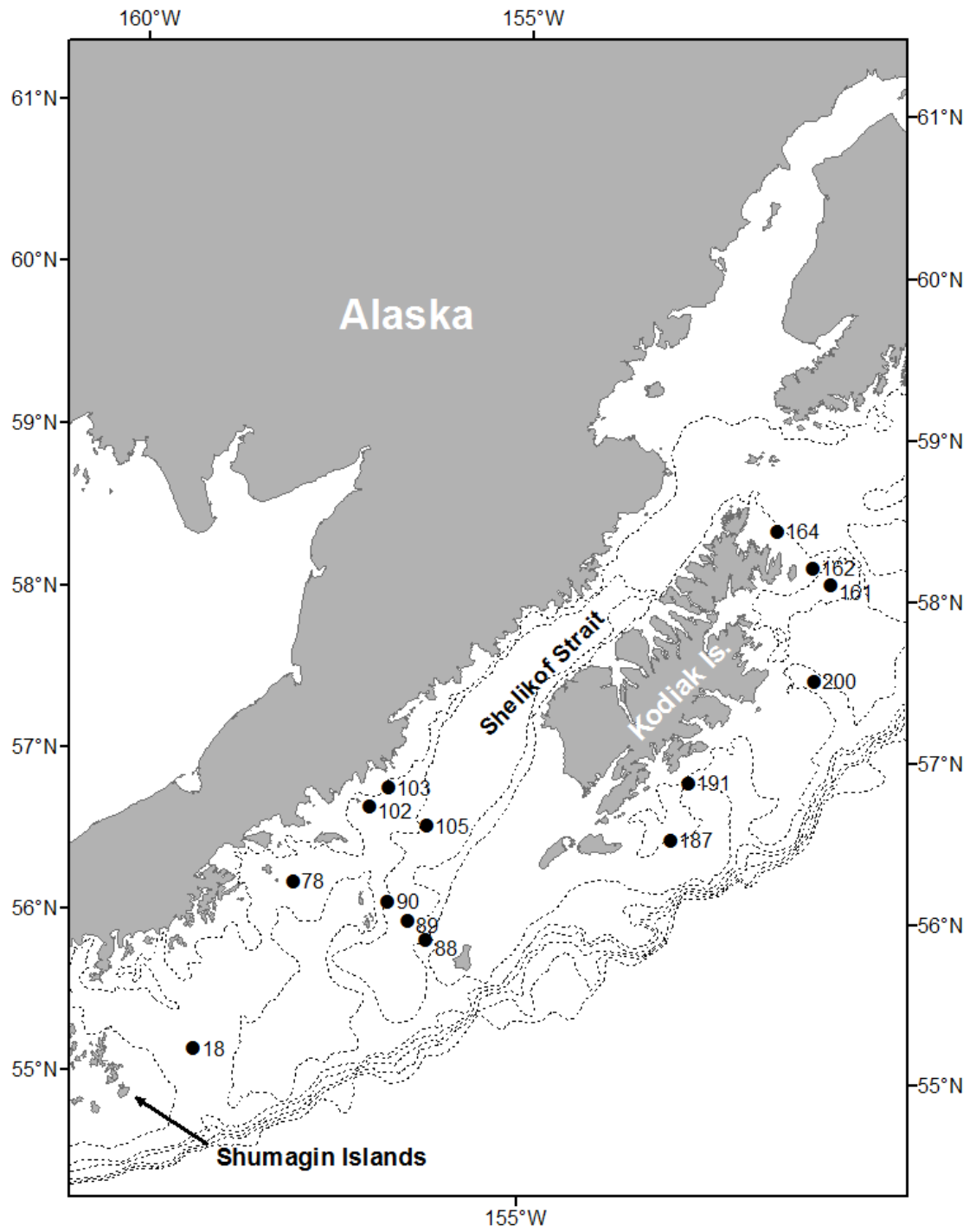
Figure 4. Partial effects of fraction of G2-and-M-stage nuclei (a), fraction of S-stage nuclei (b), and ratio of the number of S-stage nuclei to the number of G1-stage nuclei with high RNA content (c) on growth rate (mm d<sup>-1</sup>) of Walleye Pollock (*Gadus chalcogrammus*) larvae for growth model 1. Cell cycle data were normalized using the arcsin  $\sqrt{x}$  transformation. Tick marks on X-axis show individual measurements. The 95% confidence interval is shown by dashed lines.

Figure 5. Partial effect of standard length (mm) on the probability of fast growth of Walleye Pollock (*Gadus chalcogrammus*) larvae for growth classification model 2. Tick marks on X-axis show individual measurements. Y-axis represents values that larval size contributes to predicted probability of fast growth. The 95% confidence interval is shown by dashed lines.

716 Figure 6. Partial effects of standard length (SL, mm) and fraction of G2-and-M-stage nuclei (a),  
717 and SL and fraction of S-stage nuclei (b) on the probability of fast growth of Walleye Pollock  
718 (*Gadus chalcogrammus*) larvae for growth classification model 2. Cell cycle data were  
719 normalized using the  $\arcsin \sqrt{x}$  transformation. Grey scale shows probability of fast growth  
720 increasing from dark to light colors. Contour lines indicate probability of fast growth.

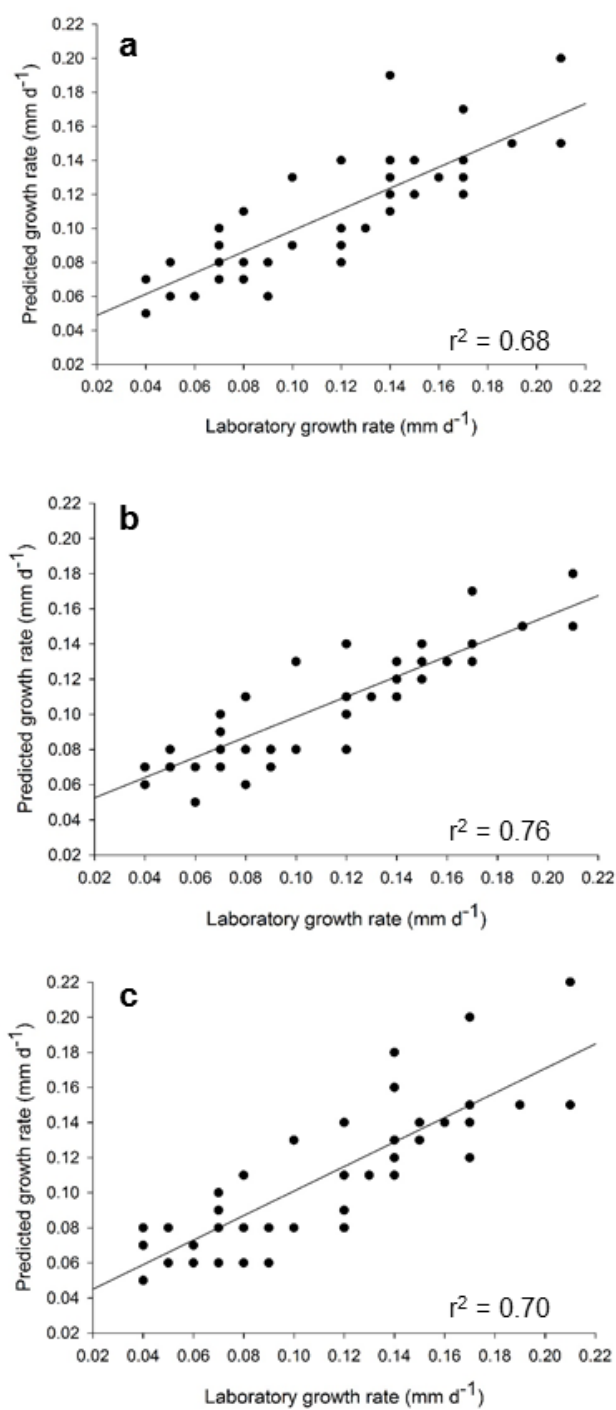
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Fig.1

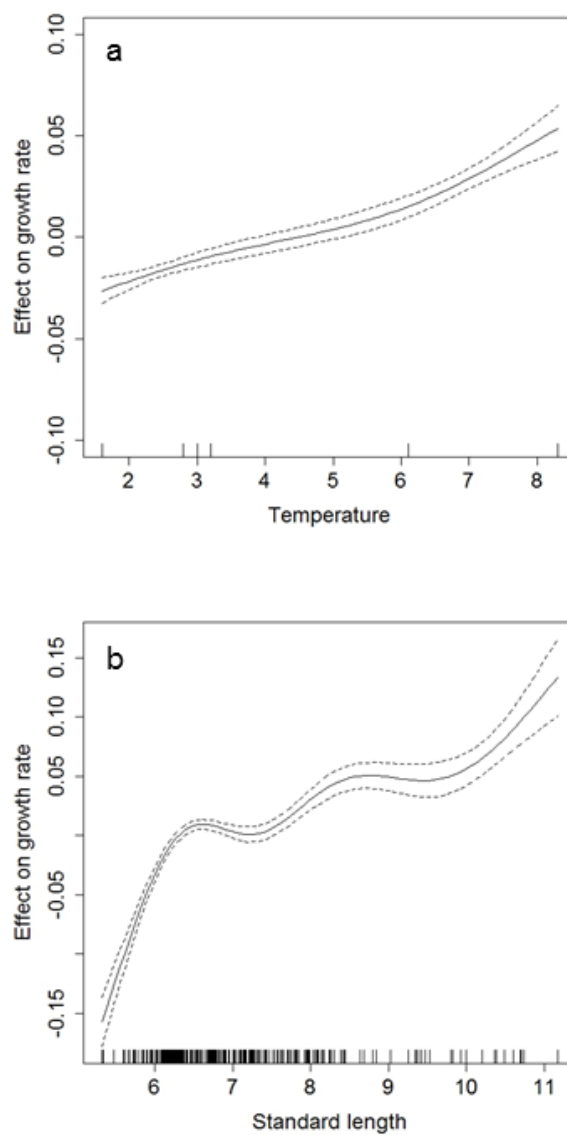


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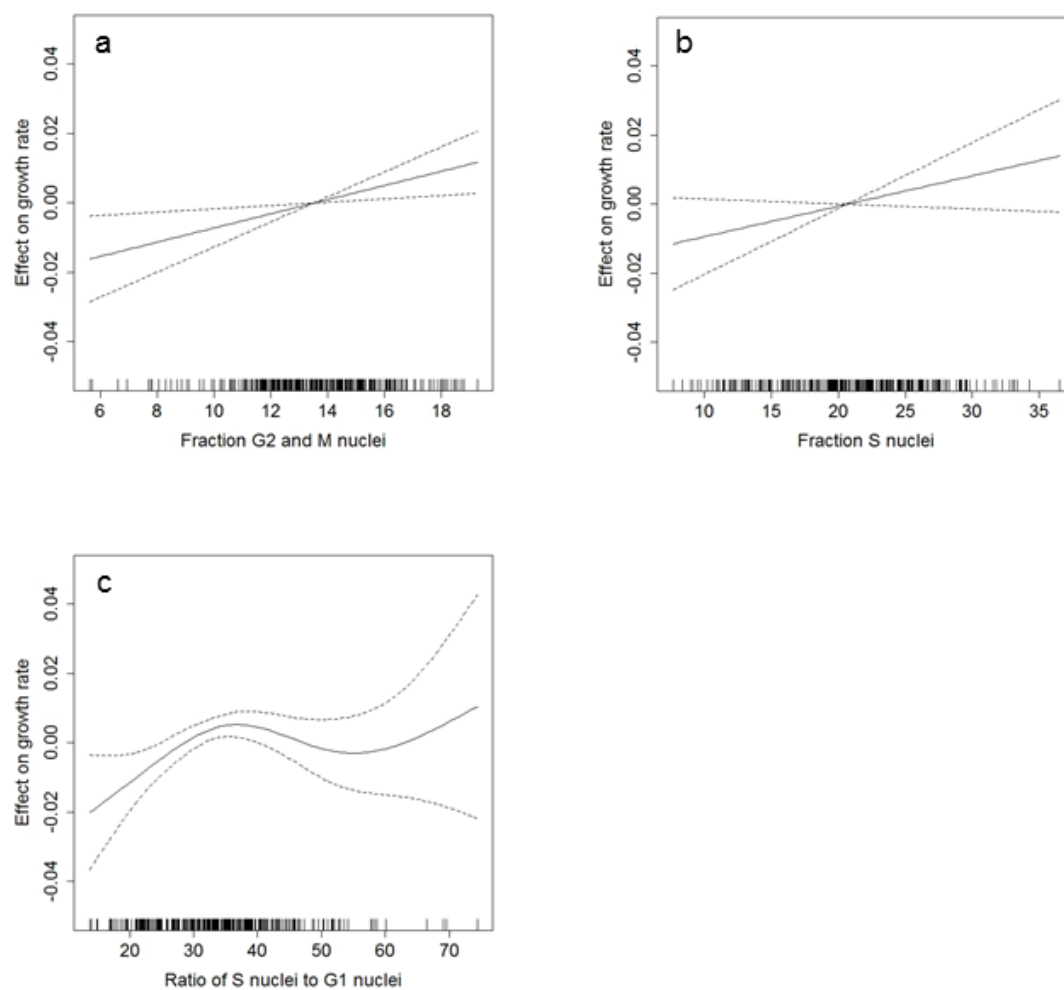
723 Fig. 2



725 Fig. 3

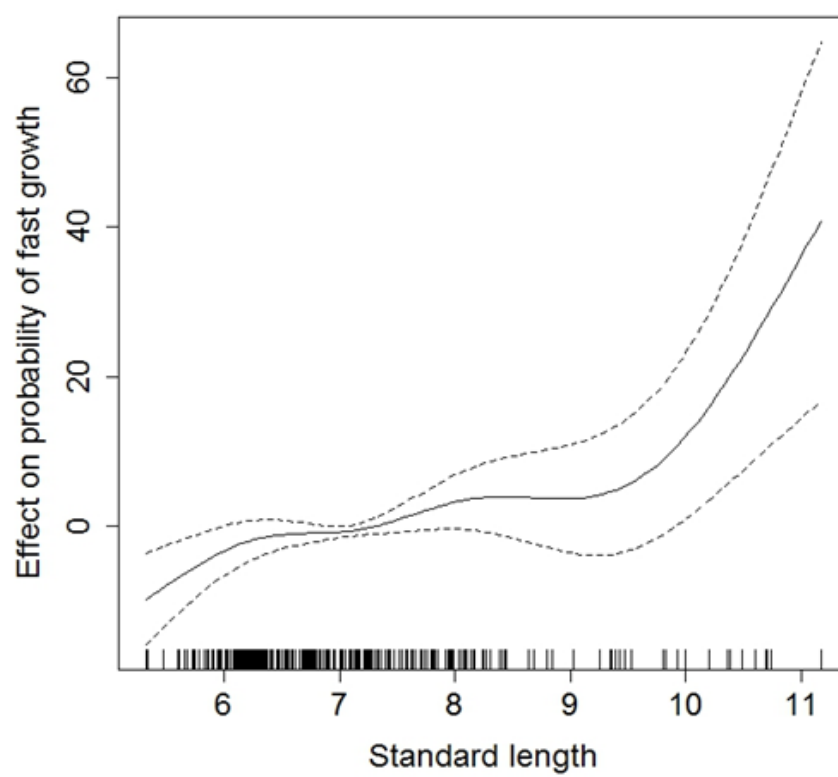


727 Fig. 4



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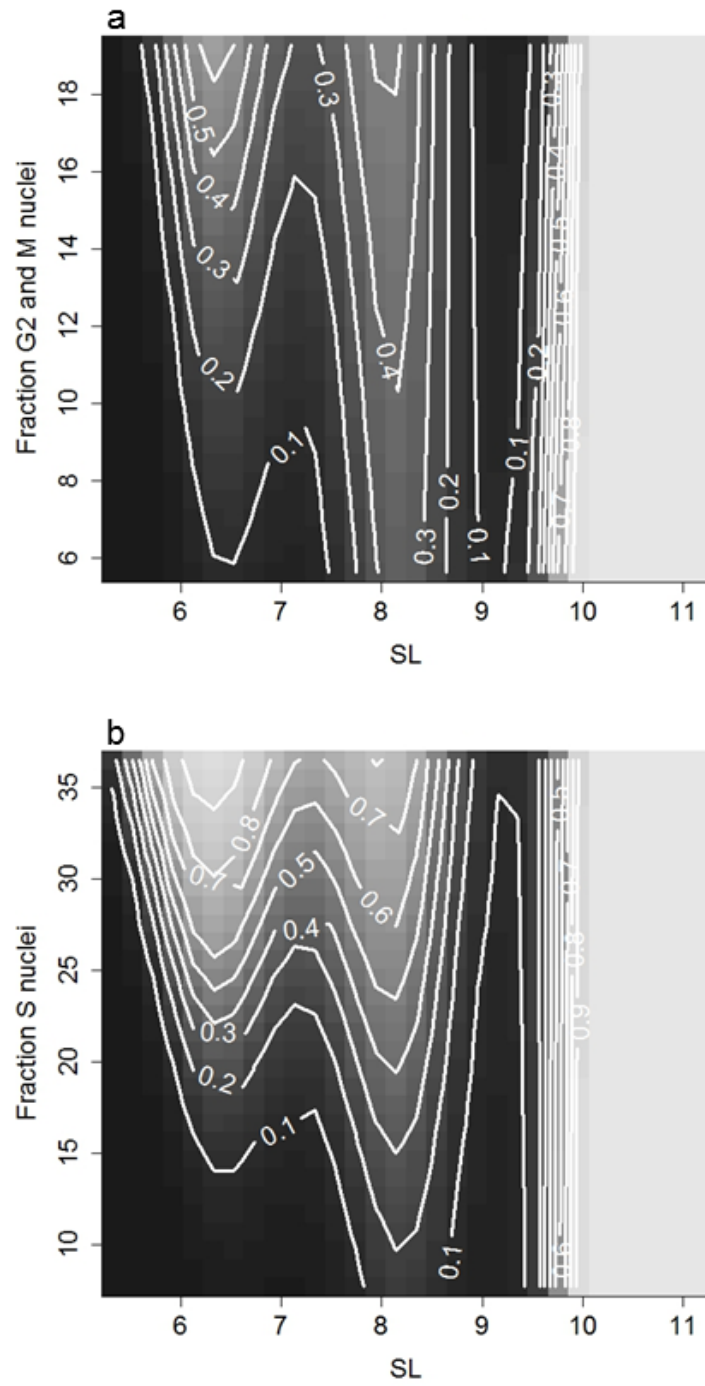
Fig. 5



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731

Fig. 6



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Table 1. Experimental year, rearing temperature, and feeding treatments for Walleye Pollock (*Gadus chalcogrammus*) larvae laboratory studies used to formulate the growth model.

Year	Temperature $\pm$ SD <sup>a</sup> (°C)	Feeding treatments <sup>b</sup>
2014	3.2 $\pm$ 0.1	HP, LP, UF
2014	6.1 $\pm$ 0.4	HP, LP, UF
2015	1.6 $\pm$ 0.1	HP, LP, UF
2015	3.0 $\pm$ 0.1	S6F, F6S, HP
2017	8.3 $\pm$ 0.5	HP, LP, UF

<sup>a</sup>standard deviation

<sup>b</sup>HP = high prey density (10 rotifer ml<sup>-1</sup> and 3 natural zooplankton ml<sup>-1</sup>); LP = low prey density (1.0 natural zooplankton ml<sup>-1</sup>); UF = unfed; S6F = starved for 6 days then fed; F6S = Fed for 6 days then starved

Table 2. Additive and variable coefficient generalized additive models formulated to predict Walleye Pollock (*Gadus chalcogrammus*) larvae growth rate. AIC is Akaike Information Criterion.

Model Type	Model	AIC	r <sup>2</sup>
Model 1 : Additive, standard length used as size covariate	$GR^a = a^b + s^c(t^d) + s(sl^e) + s(tg2^f) + s(ts^g) + s(trsg1^h) + \mathcal{E}$	-1375.68	0.79
Model 2: Additive, product of standard length and body depth at anus used as size covariate	$GR = a + s(t) + s(slbda^j) + s(tg2) + s(ts) + s(trsg1) + \mathcal{E}$	-1247.86	0.68
Model 3: Variable coefficient, cell cycle parameters vary with SL	$GR = a + s(t) + s(sl) + s(sl) \times tg2 + s(sl) \times ts + s(sl) \times trsg1 + \mathcal{E}$	-1389.55	0.80
Model 4: Variable coefficient, cell cycle parameters vary with temperature	$GR = a + s(sl) + s(t) \times tg2 + s(t) \times ts + \mathcal{E}$	-1378.39	0.79

<sup>a</sup>growth rate (growth in length, mm d<sup>-1</sup>)

<sup>b</sup>model intercept

<sup>c</sup>smooth function

<sup>d</sup>rearing temperature

<sup>e</sup>standard length

<sup>f</sup>arcsin  $\sqrt{x}$  transformed fraction of nuclei in the G2-and-M-stage of the cell cycle

<sup>g</sup>arcsin  $\sqrt{x}$  transformed of nuclei in the S-stage of the cell cycle

<sup>h</sup> arcsin  $\sqrt{x}$  transformed ratio of the number of cells the S-stage to number of cells in the G1-stage with high nuclear RNA content

<sup>i</sup>model error

<sup>j</sup>product of standard length and body depth at anus, and is correlated with larval weight (van der Meeren, 1991).

758 Table 3. Rearing temperature, days after first feeding sampled, size range (standard length, mm),  
 759 and number of larvae (n) used to formulate and validation test the growth models.

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Temperature (°C)	Days sampled	Model formulation		Validation testing	
		Size range	n	Size range	n
1.6	6, 12, 22	5.33 – 8.30	64	6.42 – 6.90	9
2.8	22	6.54 – 8.84	9		
3.0	6, 12, 25	5.60 – 8.41	72		
3.2	6, 12, 25	5.96 – 9.25	52	6.78 – 8.54	14
6.1	3, 6, 13, 25	5.32 – 11.17	88	7.40 – 8.63	15
8.3	3, 8, 17	5.66 – 8.25	33	7.14 – 8.61	7

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Table 4. Validation testing for growth models 1, 3, and 4 (see Table 2 for description of models). The Pearson correlation coefficient ( $r$ ) between predicted growth rates and laboratory growth rates, slope of linear regression for laboratory growth rate and predicted growth rate, the percentage of predictions that were correct (i.e., equal to laboratory growth), and percentage of laboratory growth rates within the 95% confidence interval of the predicted growth rates.

Model	$r$	Slope of linear regression	% correct predictions	% of laboratory growth rates within the 95% CI of the predicted growth rates
1	0.83	0.62	16	33
3	0.88	0.58	7	31
4	0.84	0.70	7	27

Table 5. Laboratory and model predicted growth rate ( $\text{mm d}^{-1}$ ) for each rearing temperature for the independent validation testing data set (see Table 2 for description of models).

	Temperature			
	1.6°C, n = 9 <sup>b</sup>	3.2°C, n = 14	6.1°C, n = 15	8.3°C, n = 7
Laboratory <sup>a</sup>	0.06 <sup>c</sup> ± 0.02	0.08 ± 0.02	0.15 ± 0.03	0.15 ± 0.03
Model 1	0.06 ± 0.01	*0.09 ± 0.02	*0.12 ± 0.02	0.16 ± 0.03
Model 3	0.07 ± 0.01	*0.09 ± 0.02	*0.12 ± 0.02	0.14 ± 0.03
Model 4	0.06 ± 0.01	*0.09 ± 0.02	*0.13 ± 0.02	0.16 ± 0.04
ANOVA <sup>d</sup>	F <sub>3,24</sub> = 0.44, p = 0.72	F <sub>3,39</sub> = 5.12, p = 0.004	F <sub>3,42</sub> = 66.58, p < 0.001	F <sub>3,18</sub> = 2.76, p = 0.07

<sup>a</sup>laboratory growth rate was calculated from size at age (see text)

<sup>b</sup>number of larvae tested

<sup>c</sup>mean growth rate ± standard deviation

<sup>d</sup>repeated measures ANOVA comparing laboratory and model predicted growth rates at each temperature

\*significantly different from laboratory growth rate ( $p < 0.05$ ) as indicated by Tukey Test

Table 6. Laboratory growth rate, growth rate derived from otolith microstructure analysis, and growth rate predicted from models 1, 3, and 4 for 6.1° and 8.3°C rearing temperatures of the validation testing data set (19 total larvae, see text for explanation). The Pearson correlation coefficient (r) for predicted growth rates and the distance between the first feeding increment and otolith edge. The percentage of otolith-derived growth rates within the 95% confidence interval of the predicted growth rates.

Growth rate type	Growth rate <sup>b</sup>	r	% of otolith-derived growth rates within the 95% CI of the predicted growth rates
Laboratory	0.16 ± 0.03		
Otolith derived <sup>a</sup>	0.17 ± 0.02*		
Model 1	0.13 ± 0.03*	0.91	11
Model 3	0.13 ± 0.02*	0.77	5
Model 4	0.14 ± 0.03	0.81	16

<sup>a</sup>otolith-derived growth rate was calculated from larval standard length and number of daily increments from first feeding to the otolith edge (see text)

<sup>b</sup>mm d<sup>-1</sup>; mean ± standard deviation

\*significantly different from laboratory growth rate; repeated measures ANOVA, Tukey test, p < 0.01

Table 7. Additive and variable coefficient binomial generalized additive models formulated to predict Walleye Pollock (*Gadus chalcogrammus*) larvae as fast- or slow-growing based on laboratory growth rate of larvae in the high prey density treatment at each temperature. AIC is Akaike Information Criterion.

Classification Model Type	Model	AIC	r <sup>2</sup>	% Correct <sup>j</sup>
Model 1: Additive, analogous to growth rate model 1.	$pGR^a = a^b + s^c(t^d) + s(sl^e) + s(tg2^f) + s(ts^g) + s(trsg1^h) + \mathcal{E}$	321.61	0.33	69 (31/45)
Model 2: Variable coefficient, cell cycle parameters vary with SL. Similar to growth rate model 3 <sup>k</sup> .	$pGR = a + s(t) + s(sl) + s(sl) \times tg2 + s(sl) \times ts + \mathcal{E}$	326.87	0.31	71 (32/45)
Model 3: Variable coefficient, cell cycle parameters vary with temperature. Analogous to growth rate model 4.	$pGR = a + s(sl) + s(t) \times tg2 + s(t) \times ts + \mathcal{E}$	325.46	0.31	69 (31/45)

<sup>a</sup>probability that a larva is fast growing

<sup>b</sup>model intercept

<sup>c</sup>smooth function

<sup>d</sup>rearing temperature

<sup>e</sup>standard length

<sup>f</sup> $\arcsin \sqrt{x}$  transformed fraction of nuclei in the G2-and-M-stage of the cell cycle

<sup>g</sup> $\arcsin \sqrt{x}$  transformed fraction of nuclei in the S-stage of the cell cycle

<sup>h</sup> $\arcsin \sqrt{x}$  transformed ratio of the number of cells the S-stage to number of cells in the G1-stage with high nuclear RNA content

<sup>i</sup>model error

<sup>j</sup>percent of larvae correctly classified as fast- or slow-growing, number of correct classifications shown in parenthesis

<sup>k</sup>based on AIC,  $\arcsin \sqrt{x}$  transformed ratio of the number of cells the S-stage to number of cells in the G1-stage with high nuclear RNA content was not included in the model

Table 8. Area sampled, station, water temperature (mean 0-50 m depth; °C), mean size of larvae, mean predicted growth rate, mean otolith-derived growth rate, and mean predicted probability of fast growth for Walleye Pollock (*Gadus chalcogrammus*) larvae collected from the western Gulf of Alaska 18 – 30 May 2013.

Area	Station	Date <sup>b</sup>	Temp.	n <sup>c</sup>	SL <sup>d</sup>	Pred. GR <sup>e</sup>	Oto. GR <sup>f</sup>	Prob. fast GR <sup>g</sup>
KI-SI <sup>a</sup>	18	18 May	4.9	1	6.70	0.10	0.20	0.77
KI-SI	78	21 May	4.0	1	9.47	0.14	0.19	0.10
KI-SI	88	22 May	4.6	3	10.36 ± 0.20	0.18 ± 0.01	0.19 ± 0.02	0.99 ± 0.01
KI-SI	89	22 May	4.2	2	8.93 ± 0.83	0.14 ± 0.02	0.18 ± 0.03	0.57 ± 0.16
KI-SI	90	23 May	4.5	3	10.21 ± 0.57	0.17 ± 0.03	0.18 ± 0.01	0.54 ± 0.24
KI-SI	102	23 May	4.5	3	9.90 ± 0.10	0.15 ± 0.003	0.20 ± 0.01	0.71 ± 0.19
KI-SI	103	23 May	4.1	2	10.07 ± 0.66	0.17 ± 0.03	0.19	0.54 ± 0.46
KI-SI	105	23 May	4.3	2	10.40 ± 0.60	0.18 ± 0.04	0.16 ± 0.01	0.79 ± 0.21
Kodiak Is.	161	27 May	5.1	3	12.73 ± 1.33	0.36 ± 0.10	0.17 ± 0.003	1.00
Kodiak Is.	162	27 May	5.1	1	12.10	0.30	0.17	1.00
Kodiak Is.	164	28 May	5.0	4	11.25 ± 0.99	0.26 ± 0.07	0.18 ± 0.01	0.75 ± 0.22
Kodiak Is.	187	29 May	5.5	1	5.91	0.05	- <sup>h</sup>	0.79
Kodiak Is.	191	29 May	6.0	1	12.10	0.32	0.17	1.00
Kodiak Is.	200	30 May	5.9	4	11.00 ± 0.60	0.24 ± 0.04	0.19 ± 0.02	0.93 ± 0.07

<sup>a</sup>area between Kodiak Island and Shumagin Islands (see Fig. 1)

<sup>b</sup>date larvae collected

<sup>c</sup>number of larvae analyzed

<sup>d</sup>mean standard length ± standard error



822 <sup>e</sup>model 1 predicted growth rate  $\pm$  standard error ( $\text{mm d}^{-1}$ ); mean growth rate for all larvae  
823 analyzed at a station  
824 <sup>f</sup>otolith-derived growth rate  $\pm$  standard error ( $\text{mm d}^{-1}$ ) calculated from age (number of  
825 increments from first feeding to otolith edge) and standard length  
826 <sup>g</sup>mean probability of fast growth  $\pm$  standard error; probability  $\geq 0.50$  indicates fast growth  
827 <sup>h</sup>otoliths from this larva were unreadable due to poor increment definition, so this larva was not  
828 used in further analyses

