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      Oceanographic drivers of sablefish recruitment in the California Current
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31 ABSTRACT

32 Oceanographic processes and ecological interactions can strongly influence recruitment success 33 in marine fishes. Here, we develop an environmental index of sablefish recruitment with the 34 goal of elucidating recruitment-environment relationships and informing stock assessment. We 35 start with a conceptual life-history model for sablefish Anoplopoma fimbria on the US west coast 36 to generate stage- and spatio-temporally-specific hypotheses regarding the oceanographic and 37 biological variables likely influencing sablefish recruitment. Our model includes seven stages 38 from pre-spawn female condition through benthic recruitment (age-0 fish) for the northern 39 portion of the west coast U.S. sablefish stock (40-50 °N). We then fit linear models and use 40 model comparison to select predictors. We use residuals from the stock-recruitment relationship 41 in the 2015 sablefish assessment as the dependent variable (thus removing the effect of spawning 42 stock biomass). Predictor variables were drawn primarily from ROMS model outputs for the 43 California Current Ecosystem. We also include indices of prey and predator abundance and 44 freshwater input. Five variables explained 57% of the variation in recruitment not accounted for 45 by the stock-recruitment relationship in the sablefish assessment. Recruitment deviations were 46 positively correlated with (1) colder conditions during the spawner preconditioning period, (2) 47 warmer water temperatures during the egg stage, (3) stronger cross shelf transport to near-shore 48 nursery habitats during the egg stage, (4) stronger long-shore transport to the north during the 49 yolk-sack stage, and (5) cold surface water temperatures during the larval stage. This result 50 suggests that multiple mechanisms likely affect sablefish recruitment at different points in their 51 life history.

52

53 Key Words: Sablefish, Anoplopoma fimbria, recruitment, oceanic drivers, California Current

54 INTRODUCTION

Climate plays an important role in coastal marine ecosystems, driving changes in horizontal and vertical transport that affect nutrient availability and primary production that, in turn, propagate through both lower and upper trophic levels (Di Lorenzo *et al.*, 2013; Chavez *et al.*, 2003; Hunt Jr and McKinnell, 2006). In coastal upwelling systems the bottom-up forcing hypothesis has been the prevailing paradigm (Di Lorenzo *et al.*, 2013). However, recent synthesis by the Global Ecosystems Dynamics Program (GLOBEC) of four regions (Gulf of Alaska, Northern California

61 Current, Northwest Atlantic and Southern Ocean) suggests that horizontal transport (cross-shelf, 62 long-shore) is also highly important, providing the foundation for a new horizontal-advection 63 bottom-up-forcing paradigm (Di Lorenzo et al., 2013). This new paradigm suggests that-in 64 addition to the indirect effects of climate manifested through variability in primary production 65 and the timing of the availability of food resources-horizontal transport directly affects the 66 reproductive success of taxa like fish by influencing their transport to suitable settlement habitat 67 (reviewed by Di Lorenzo et al., 2013). At the same time, other oceanographic parameters like 68 temperature can affect growth, which can mediate an individual's vulnerability to predation and 69 the susceptibility of larvae to starvation (Litvak and Leggett, 1992; Houde, 1987; Leggett and 70 DeBlois, 1994. Chezik et al., 2014). Likewise, ecological interactions such as prey availability, 71 predator abundance and density dependence may also be important in determining reproductive 72 success for marine fishes and other species (Tolimieri, 2015; Hunt Jr and McKinnell, 2006; 73 Frank et al., 2007; Field et al., 2006). Recruitment, in turn, directly affects age structure and 74 population size in marine fishes (Bailey, 1981; Hjort, 1914; Myers, 1998; Maunder and Watters, 75 2003). Therefore, it is important to understand the role of oceanographic processes and 76 ecological interactions in determining recruitment success in marine fishes—both to gain a better 77 understanding of their population dynamics and to better manage these species.

78 Sablefish (Anoplopoma fimbria) inhabit waters along the west coast of North America 79 from the Baja California through Alaska and extend west (and south) to Japan (Allen and Smith, 80 1988; Hart, 1973; Johnson et al., 2016). Traditionally, two stocks have been recognized in the 81 northeastern Pacific: (1) an Alaskan/British Columbian stock extending from the Bering Sea to 82 Vancouver Island, and (2) a U.S. west coast population extending from southwest Vancouver 83 Island to Baja California (Johnson *et al.*, 2016; Schirripa and Colbert, 2006), with each stock 84 being subject to independent management. The spawning stock biomass (SSB) of the U.S. west 85 coast sablefish stock (the focus of this work) has declined steadily since the 1980's (Fig. 1a), 86 concurrent with high landings during 1976-1990 (Johnson et al., 2016) and highly variable, but 87 declining recruitment (Fig. 1b). The stock-recruitment relationship appears weak (Fig. 1c) 88 suggesting that environmental factors are likely important.

Sablefish recruitment-environment investigations have generally focused on large-scale
climate or oceanographic variables (Schirripa and Colbert, 2006; Schirripa *et al.*, 2009; Schirripa
and Methot, 2001; Coffin and Mueter, 2015; Shotwell *et al.*, 2014; Sogard, 2011). For example,

92 in the California Current Ecosystem along the U.S. west coast, sablefish recruitment has been 93 correlated with changes in sea surface height (SSH) and both northward and eastward Ekman 94 transport (Schirripa and Colbert, 2006). SSH has been used as an index of recruitment success in 95 recent sablefish stock assessments (Schirripa et al., 2009; Stewart and Forrest, 2011; Johnson et 96 al., 2016), in part as a proxy for abundance of copepods (Schirripa and Colbert, 2006; Schirripa 97 et al., 2009; Schirripa and Methot, 2001), which are an important food source for larvae and 98 juveniles (McFarlane and Beamish, 1992; Grover and Olla, 1987). Changes in SSH serve as a 99 proxy for large-scale climate forcing that drives regional changes in alongshore and cross-shelf ocean transport. These changes directly impact the transport of water masses, nutrients, and 100 101 organisms. SSH tracks these changes on a gross scale, explaining the weak but significant 102 relationship with sablefish recruitment (Di Lorenzo et al., 2013; Schirripa and Colbert, 2006).

103 The relationship between sablefish recruitment and SSH is compelling from an ecological 104 standpoint. However, the use of the index has not had a large effect on stock-assessment results 105 because good data on year-class strength from fishery and fishery-independent survey data 106 already inform the stock assessment regarding variability in recruitment (Stewart *et al.*, 2011). While a significant predictor ($r^2 = \sim 0.3$), SSH provides too little additional information to lower 107 108 uncertainty enough in the stock assessment to improve the precision of future recruitment 109 predictions. To do so, an environmental index would need to explain more than 50% of the 110 variability in recruitment unrelated to the stock-recruitment relationship and age structure 111 (Basson, 1999; Johnson et al., 2016). Ideally, such an index should derive from environmental 112 variables at scales relevant to the sablefish life history and be able to forecast into the future, 113 potentially allowing managers and fishers to better respond to potential shifts in sablefish 114 recruitment. Given a robust environmental index with greater explanatory power, hind-casting 115 recruitment strength on the basis of past environmental data can better inform recruitment 116 estimates during time periods in which there is no information on fishery or survey length and 117 age compositions.

Here, we (1) develop a literature-based, conceptual life-history model for sablefish that includes seven stages from female conditioning through benthic recruitment (age-0 fish) for the northern portion of the west coast U.S. sablefish stock (40-50 $^{\circ}$ N). We then (2) use this conceptual model to generate stage-specific and spatio-temporally-specific hypotheses regarding the physical and biological variables likely to influence sablefish recruitment. Next we (3) use

123 linear models and model comparison to develop predictive models of sablefish recruitment using 124 oceanographic drivers taken from a Regional Ocean Modeling System (ROMS) model of the 125 California Current System (Neveu et al., 2016). We also evaluate support for biological indices 126 like predator and prey abundance, temperature impacts on growth rates (e.g., Sogard and Olla, 127 2001), and freshwater input from the Columbia River as a proxy for nutrient input or effects on ocean currents. Our goal is to develop an ecologically meaningful, robust environment-128 129 recruitment relationship that has higher explanatory power than those developed in the past, with 130 the end goal of enhancing the sablefish stock assessment and improving short term forecasts of 131 sablefish recruitment required by fishery managers.

132

133 METHODS

134 We investigate predictors of recruitment for the northern portion of the U.S. west coast stock 135 from 40-50 N. We focus on the northern portion of the stock because recruitment estimates 136 from the coast-wide stock assessment model are most strongly informed by age and length 137 composition data from the north (Johnson et al., 2016). Although the two-stock definition for 138 eastern Pacific sablefish (Alaskan/British Columbian and U.S. west coast) is commonly 139 accepted, there is evidence that the dynamics of the U.S. west coast stock differs north and south of Cape Mendocino (~40.4 °N). Head *et al.* (2014) concluded that the sablefish population south 140 141 of Cape Mendocino may be a separate subpopulation based on differentiation in ages and lengths 142 at 50% maturity, as well as potential differences in reproductive success. Maximum body size is 143 larger and growth rates are slower north of Cape Mendocino. North of Cape Mendocino, the 144 highest concentration of age-0 fishes observed in the West Coast Groundfish Bottom Trawl Survey (WCGBTS) is between 44-45° N, just south of the Columbia River (Fig. S1). 145

146

147 Sablefish life history: female preconditioning to age-0 recruits

We began our conceptual life-history model by first identifying each life-history stage that could potentially contribute to determining the size of each sablefish year class, beginning with female condition prior to the start of the spawning season (Table 1). The energetic status of females may influence their propensity to spawn, and the quality and number of eggs produced (Sogard *et al.*, 2008; Rodgveller *et al.*, 2016). Thus, the summer and fall prior to spawning (June-Dec) may be important for female preconditioning. Spawning occurs from December to March with a

154 peak in February. Most spawning takes place at the edge of the continental shelf at depths 155 greater than 300 m with eggs initially found from 200 m to greater than 825 m (Hunter et al., 156 1989; Kendall and Matarese, 1987; Moser et al., 1994; Mason et al., 1983). Eggs are buoyant, 157 rising to 200-300 m in the water column but are most common between 240 and 480 m, where 158 they remain for approximately 12-17 days until hatching (Moser et al., 1994; Mason et al., 1983; 159 Kendall and Matarese, 1987; Boehlert and Yoklavich, 1985; McFarlane and Beamish, 1992). 160 Post hatch, larvae sink to 1000-1200 m where they can be found between February and May as 161 yolk-sack larvae. By 14-17 days post-hatch larvae have consume about 50% of their yolk sack 162 and may show initial attempts at feeding approximately a week later. By 40-days post hatch 163 larvae are in surface waters from the 500-m isobath out to 150 nautical miles (277 km) from 164 shore where they are found between February and May (Moser et al., 1994; Brock, 1940; 165 McFarlane and Beamish, 1992). Pelagic juveniles are also found in these surface waters and are 166 present from April through November (Mitchell and Hunter, 1970; Kendall and Matarese, 1987). 167 Sablefish settle to the benthos as age-0 recruits between August and November with most fish 168 likely settling to habitats 250 m or shallower.

169

170 Regional Ocean Modeling System (ROMS)

The majority of our predictors were physical oceanographic parameters including temperature, 171 172 long-shore and cross-shelf currents, and mixed layer depth. We derived these variables from a 173 California Current System (CCS) configuration of the Regional Ocean Modeling System 174 (ROMS) with 4-Dimensional Variational (4D-Var) data assimilation (Neveu et al., 2016). The 175 ROMS model domain covers the region 30-48°N and from the coast to 134W at 0.1° (\sim 10 km) 176 horizontal resolution, with 42 terrain-following vertical levels. In the 1980-2010 CCS reanalysis 177 used for this study, satellite observations (SST, SSH) and in situ data (temperature and salinity 178 from ships, floats, moorings) are assimilated into the model to improve its representation of the 179 true ocean state. This reanalysis has been used extensively to study spatial upwelling variability 180 (Jacox et al., 2014), basin-scale climate influences (Jacox et al., 2015a; Jacox et al., 2015b), and 181 bottom-up controls on primary production (Jacox et al., 2016) in the CCS. All ROMS output 182 was averaged in 4-day increments and then either averaged or summed over the appropriate 183 period (as defined in Table 1) for each of the 30 years (n = 30 for each time series in the 184 analysis). Additional predictors are described in relation to their specific hypotheses below.

185 Oceanographic data are scarcer at depth than at the surface. As a result, the ROMS 186 output, which assimilates available data, is not as strongly constrained by observations at the 187 subsurface as it is at the surface. It is therefore likely that the ROMS output deviates from nature 188 more at the subsurface than it does at the surface, though the scarcity of subsurface 189 measurements (or transport metrics at any depth) precludes a definitive model evaluation in this 190 regard. The paucity of subsurface data, however, is the motivation for using model output in this 191 study, and despite our inability to validate the subsurface transport in the model, there is reason 192 to believe it provides useful information. First, the purpose of data assimilation is to further 193 improve a model that already captures the dynamics of the California Current faithfully without 194 data assimilation. The fidelity of this model to nature has been documented extensively for 195 applications with and without data assimilation (e.g., Jacox et al., 2015a; Veneziani et al., 2009). 196 The model is forced by realistic winds, surface heat fluxes, and lateral boundary conditions, 197 which drive realistic physical variability even in the absence of data assimilation. Second, the 198 data that are assimilated, even if only available at the ocean surface, can impact the subsurface 199 dynamics. For example, assimilating sea surface height measurements constrains the geostrophic 200 flow and therefore transport at the surface and in the subsurface. The impact of assimilated data 201 on different metrics of the CCS circulation is discussed in detail in Moore et al. (2017). For a transport example, see Fig. 4e in that paper. 202

203

204 *Hypotheses and additional data sources*

205 We developed 21 a priori, life-stage-specific and space-specific (considering time, depth, and 206 longitude) hypotheses for ecological and environmental covariates that may drive variation in 207 sablefish recruitment (Table 1). Two hypotheses were represented by two predictors, and we 208 included SSH for historical reasons (see below). Thus, we tested 24 total predictors. While the 209 current sablefish stock assessment estimates coast-wide spawning biomass and recruitment time 210 series, we limited the predictor time series to 40-50 °N because a majority of the length and age 211 composition data come from the northern California Current (Johnson et al., 2016). For each 212 hypothesis, we specified the time period, depth and longitudinal extent of the potential 213 predictor-for example, net cross-shelf transport between January and April at 300-825 m depth, 214 between 40-50 N and from the 500-m isobath to 170 nautical miles off shore may affect 215 transport and distribution of sablefish eggs (Table 1).

216 We focused on hypotheses for which we had data (or model output) available for testing. 217 We excluded some potential hypotheses for which data were not available rendering these 218 hypotheses untestable. These hypotheses are not included in Table 1. The most obvious 219 omission is food availability for larvae and pelagic juveniles (McFarlane and Beamish, 1992). 220 Unfortunately, continuous data for northern copepod biomass anomaly begin in 1996¹, long after 221 recruitment estimates are available for sablefish. We also did not include potential mechanisms 222 with indirect effects such as upwelling, which might affect larval survival by influencing food 223 availability. Note, however, that upwelling effects will likely be caught in many of the transport 224 and temperature indices we did include. Finally, in some cases, the literature suggested multiple 225 potential depth ranges over which environmental or biological variables might influence 226 recruitment: one broad and one more restricted. For example, sablefish eggs are found between 227 240-825 m (Hunter et al., 1989; Kendall and Matarese, 1987; Mason et al., 1983; Moser et al., 228 1994) with the highest occurrence between 240-480 m (Moser et al., 1994). When selecting 229 environmental predictors for inclusion in model selection, we initially evaluated variables (e.g., 230 total cross shelf transport) for each depth range. However, preliminary analyses showed these 231 paired predictors (i.e., net cross shelf transport between 300-825 m and 240-480 m) to be highly 232 correlated (r > 0.88), and we chose to include only the broader depth range version of each in our 233 analyses to reduce the number of predictors. The resulting testable hypotheses fall into six 234 general categories, which may overlap life-history stages (Table 1): temperature, transport, 235 mixing, prey, predators, and nutrient fertilization or effects on oceanic currents from freshwater 236 input from rivers.

237

238 <u>Temperature</u>

Ambient temperature may affect the production of recruits through multiple mechanisms. During the preconditioning period for spawning females (Table 1, hypothesis 2, hereafter, H2) warmer temperatures may increase energy demands (H2), forcing female sablefish to allocate less energy to reproduction, reducing the production of eggs. Temperature may also act as a spawning cue (H4). For eggs and larvae, temperature may also affect growth, development,

¹ http://www.noaa.gov/iea/regions/california-current-region/indicators/climate-and-oceandrivers.html

244 survival, and susceptibility to predation through multiple mechanisms, among which we cannot 245 distinguish (H7, H10, H13, H17 and H19). For example, warmer temperatures allow for faster 246 growth allowing larvae to outgrow potential predators (e.g., stage duration or 'bigger-is-better' 247 hypotheses; Houde, 1987; 1997), but warm waters may also increase energy demands making 248 larvae more susceptible to starvation, especially if warmer waters bring poor feeding conditions. 249 Temperature data were obtained from ROMS outputs. In most cases, we include temperature as 250 degree days (cumulative temperature above a threshold value, Chezik et al., 2014), setting a slightly conservative threshold temperature of 3.5 °C (Alderdice et al., 1988; Sogard and 251 252 Spencer, 2004).

- 253
- 254 <u>Transport</u>

Transport to or advection away from appropriate settlement habitat can strongly affect larval supply and settlement of marine species with pelagic eggs and larvae (Schirripa and Colbert, 2006; Montgomery *et al.*, 2001). Transport to settlement habitat was characterized by ROMS estimates of net long-shore transport (LST: H8, H11, H14 & H17) and net cross-shelf transport (CST: H6, H9, H12, H15 & H18) at specific depths and time periods for each sablefish lifehistory stage.

Sablefish spawn at depth but eggs are initially buoyant, rising in the water column. The mixed-layer depth (MLD) may influence how high eggs rise in the water column, thereby affecting transport (H5). We include the ROMS estimates of MLD (mean depth and standard deviation) from Jan-Apr when eggs are in the water column.

265

266 Prey availability and female condition

267 Prey availability (H1) during the months prior to spawning (June-Dec) may affect female 268 condition and in turn, egg quality, egg production or even the probability of spawning in a given 269 year. For example, given poor prey availability, individual sablefish in Alaskan waters may skip 270 spawning in some years (Rodgveller et al., 2016). We included the abundance of age-0 hake 271 *Merluccius productus* (Prey_{Hake}) and age-0 longspine thornyhead *Sebastolobus altivelis* (Prey_{Lsp}) 272 from their most recent stock assessments (Taylor et al., 2015; Stephens and Taylor, 2014) as 273 indices of prey abundance for female sablefish. Both species make up substantial proportions of 274 the sablefish diet (Laidig *et al.*, 1997). We included them as separate indices because hake were

275 much more abundant than thornyheads and a combined index was highly correlated with 276 $\operatorname{Prey}_{\operatorname{Hake}}(r = 0.99).$

- 277
- 278 Predation on recruits

279 Predation (H20) in the period immediately following settlement can strongly affect population 280 size and structure for benthic fishes (Hixon and Jones, 2005; Tolimieri, 2015). Therefore, an 281 index of predator abundance was developed based on known sablefish predators including: lingcod Ophiodon elongatus, arrowtooth flounder Atheresthes stomias, Pacific halibut 282 Hippoglossus stenolepis, widow rockfish Sebastes entomelas, yelloweye rockfish S. ruberrimus 283 284 and black rockfish S. melanops. Data were drawn from two sources: (1) stock assessments for 285 each species (Hamel et al., 2009; Taylor and Wetzel, 2011; He et al., 2011; Stewart et al., 2015; Wallace et al., 2008; Kaplan and Helser, 2007), and (2) West Coast Groundfish Bottom Trawl 286 287 Survey (Keller et al., 2008). Stock assessments for some of these species did not cover the 288 period under investigation (1980-2010). Therefore, we used multivariate autoregressive state-289 space (MARSS) models to combine the stock assessment and trawl time series for each species 290 into one population trend. We then summed the trends for each species and year to create a 291 single index of sablefish predator abundance (Holmes et al., 2014; Holmes et al., 2012; Tolimieri 292 et al., 2017; see Supplementary Material for more detail).

293

294 Freshwater input effects on nutrient fertilization and circulation

295 Freshwater outflow (H21) from major rivers influence the nearby marine environment. For 296 example, freshwater discharge from the Fraser River appears to play a role in sablefish 297 recruitment in the Gulf of Alaska, possibly through its influence on nearshore currents (Coffin 298 and Mueter, 2015). While total catch was low for age-0 fishes, preliminary analysis of the West 299 Coast Groundfish Bottom Trawl Survey data showed a preponderance individuals in the vicinity of the mouth of the Columbia River suggesting a similar effect and the possibility of nutrient 300 fertilization (Fig. S1). Annual discharge from the Columbia River measured at The Dalles². 301 302 Oregon is included as an index of potential fertilization or influence on currents.

303

² www.cbr.washington.edu/dart/query/streamflow_daily

304 <u>Sea-surface height</u>

305 Finally, for historical purposes we include an index of north coast spring sea surface height 306 (SSH, H22), which has been evaluated in recent stock assessments (Johnson et al., 2016; Taylor 307 et al., 2015). While not spatio-temporally specific like the other predictor variables, including 308 the SSH index aids in testing the consistency of mechanisms hypothesized to impact sablefish 309 recruitment and whether or not the more specific mechanistic variables identified in this study 310 allow for better recruitment prediction. For 1980-1992 the index is derived from tide-gauge data 311 and is the average of monthly SSH at Neah Bay and Toke Point, Washington and Astoria and Newport, Oregon from April-June. From 1993 onwards, the data come from JASON and 312 313 TOPEX satellites and represents a north coast spring index of SSH for 44-50° N (Schirripa, 314 2007).

315 Since, in this analysis, the 1980 recruitment deviation depends upon data on the 316 preconditioning period in 1979, our analysis covers recruitment residuals from 1981-2010.

317

318 Preliminary Analyses

We conducted two sets of preliminary analyses prior to model fitting. First, we evaluated correlations among predictor variables (Tables S1 and S2). As noted above, paired predictors that were the same variable but with different spatial extents were highly correlated (Table S1), and we eliminated the smaller-scale version from the model fitting to reduce the number of potential predictors. Some of the remaining predictors showed strong correlation (Table S2, r > 0.75), and we required both correlated variables not appear together in any model under consideration.

326 Next, we ran individual linear and quadratic regressions for each predictor against 327 recruitment deviations to determine if the data supported non-linear relationships. Quadratic terms were included in the main modeling exercise for those terms where, in these preliminary 328 329 analyses, the individual quadratic model fit better than the individual linear model (Akiake's 330 Information Criterion, AIC, was <2.0 that of the linear model, see Table S1, Burnham and 331 Anderson, 1998). Based on preliminary analyses, a quadratic term for long-shore transport 332 during the benthic juvenile stage (LST_{bjuv}) was included as a potential predictor in the final 333 model selection. Additionally, SSH was included as a potential quadratic term based on the 334 observed relationship in Schirripa and Colbert (2006).

335

336 Recruitment residuals

337 Estimates of sablefish recruitment and spawning stock biomass (SSB) were taken from the most 338 recent stock-assessment (Fig. 1c, Johnson et al., 2016). The recruitment time-series from the 339 stock assessment is generated, in part, by an asserted stock-recruitment relationship (asserted 340 because the form is assumed to be Beverton-Holt; steepness, unfished spawning biomass, and 341 recruitment at unfished biomass are fixed; and the data are derived data from the stock 342 assessment process, see Eqs. 1-3 below). If the environmental indices are to be useful for 343 prediction within the stock assessment, it is necessary to remove this relationship to avoid double 344 counting the effect of SSB on recruitment. Moreover, the stock-recruitment relationship is 345 known, and we want to predict the variability in recruitment unaccounted for by the SSB. In 346 order to remove the effect of the asserted stock-recruitment relationship, recruitment residuals for this study were calculated as the difference between the estimated recruitments from the 347 348 stock assessment and the assumed stock-recruitment relationship, and are given by:

- 349
- 350

Recruitment residuals = $R_{assessment} - R_{sr.curve}$ (1)

351

where $R_{assessment}$ are the estimated recruitments from the stock assessment and $R_{sr,curve}$ are the predicted recruitments from the stock-recruitment relationship. Note that the recruitment residuals calculated for this study are not the same as the estimated recruitment deviations from the stock assessment, which are constrained to sum to zero over the time period during which recruitments are estimated in the stock assessment model. $R_{sr,curve}$ is calculated as:

357

358
$$R_{sr.curve} = R_0 * S/(b+S)$$
 (2)

359

360 where R_0 is recruitment at unfished biomass (equilibrium recruitment), S is the spawning 361 biomass, and b is:

(3)

362

363
$$b = (R_0 * 0.2 * S_0)/(h - R_0) - (0.2 * S_0)$$

364

where, S_0 is unfished biomass and *h* is steepness (the ratio of recruitment at 20% vs. 100% of unfished biomass). From the 2015 assessment, $R_0 = 115,622$ thousand age-0 fish, $S_0 = 17,198$ metric tons, and h = 0.6.

368

369 Model selection

370 We fit a series of generalized linear models (GLMs) including all possible permutations of the 24 371 covariates (n = 27,191 models) with the above restrictions for total number of parameters and 372 exclusion of highly correlated terms (|r| > 0.75) from the same model. Each covariate matched a 373 specific hypothesis in Table 1 with the exceptions of Prey_{Hake} and Prey_{LSP}, which both mapped 374 to H1, prey availability. The number of predictors in a candidate model was limited to five (one 375 covariate per six data points in the time series) to prevent over-fitting. The two potentially quadratic predictors (LST_{bjuv}, and SSH) were allowed to enter the models as linear, quadratic 376 377 only, or linear + quadratic forms. The best-fit model(s) was selected using $\Delta AICc$ values, 378 retaining models with $\Delta AICc < 2.0$ (Burnham and Anderson, 1998).

379

380 Model validation and testing

381 We conducted additional analyses to evaluate the performance of the best-fit models. First, recruitment deviations were resampled with replacement to estimate r^2 values for randomized 382 383 data (1000 permutations) for the final, best-fit model. Second, standard bootstrapping 384 (resampled whole years with replacement) was used to estimate bias and calculate standard error 385 of the parameter estimates. Third, jackknife resampling was used to determine the effect of any single year on the r^2 of the best-fit model. Fourth, since the dependent variable was based on 386 387 estimated recruitments from a stock-assessment, there is error for each recruitment residual that 388 is not accounted for in the best-fit model. Therefore, we resampled the recruitment values for 389 each year from a log-normal distribution where the mean was the value for that year and 390 recruitment standard deviation for each year was taken from the sablefish stock assessment (Johnson et al., 2016, Table 15). We then recalculated the recruitment residuals and refit the 391 392 best-fit model repeating the process 1000 times. Fifth, we refit the best-supported model using 393 data for 1981-2005 and used this model to predict recruitments for 2006-2010. Sixth, jackknife 394 resampling was used to re-run the entire model fitting and comparison exercise, rather than re-395 fitting only the best fit model, to determine if removal of any individual year would change the

396 oceanographic/biological variables in the final model. Finally, we re-ran the entire model fitting
397 exercise 100 times using the re-sampled sablefish recruitments with error (from Step 4 above)
398 and compared top models from each run.

399

400 *Northern copepod biomass anomaly*

Previous work has suggested that recruitment success in sablefish is influenced by the abundance 401 402 of northern copepods (McFarlane and Beamish, 1992), which are larger and fattier than southern 403 species. Sablefish stock assessments have included zooplankton indices as environmental 404 predictors (Schirripa, 2002; Schirripa, 2007). We could not test the northern copepod biomass 405 anomaly directly in our main analyses because the continuous time series begins in 1996 406 (although there are some earlier data). However, after selecting a best-fit model from the main 407 analysis, we refit this model for years 1996-2010 both with and without this index. Data were the northern copepod biomass anomaly, $\log_{10}(\text{mg C/m}^3)$, taken from the California Integrated 408 Ecosystem Assessment³. We averaged monthly values over the summer months (July-Sept) in 409 410 each year.

411

412 **RESULTS**

Model fitting produced a clear best-fit model, with no other candidate models with a $\Delta AICc <$ 413 414 2.0 (Table 2, Table 3). In fact, the next best model had $\triangle AICc > 3.9$. The best-fit model 415 included five covariates (Fig. 2) related to temperature and transport, which explained 57% of 416 the variation in recruitment residuals (from the stock-recruitment relationship) during 1981-417 2010. Model predictions closely followed the estimated recruitments from the stock assessment 418 with the exception of 2006-2009 when the model over or under-predicted recruitment (Fig. 3). 419 Recruitment residuals were negatively correlated with degree days during the period of female 420 preconditioning (DD_{pre}) and larval stage (DD_{larv}) . Conversely, cross-shelf transport during the egg stage (CST_{egg}), degree days during the egg stage (DD_{egg}), and long-shore transport during 421 422 the yolk-sack stage (LST_{volk}) were positively correlated with recruitment residual (Fig. 4, Table

³ https://www.integratedecosystemassessment.noaa.gov/regions/california-current-region/indicators/ecological-integrity.html

423 3). Standardized coefficients suggest that DD_{egg} had the strongest effect on recruitment 424 deviations, while the other predictors had similar impact (Table 3).

425 Only three additional models had $\Delta AICc < 4.0$ and all had $\Delta AICc > 3.9$. These models 426 were generally consistent in structure with the best-fit model (Table 2). The terms DD_{egg} and 427 LST_{yolk} occurred in all three models, and DD_{pre} , and DD_{egg} in two. Model 4 was a subset of 428 Model 1 minus the DD_{larv} term (r² = 0.43). The other two models included alternate terms that 429 had less support from the data including prey abundance, long-shore transport during the pelagic 430 juvenile stage and SSH.

431 There was some moderate correlation between the covariates with the highest being for DD_{pre} and DD_{ecc} (r = 0.62, Table 4). Generalized variance inflation factor values (VIF; which 432 measure how much the variance of the estimated regression coefficients are inflated as compared 433 434 to when the predictor variables are not linearly related) were low to moderate with the exception of DD_{egg} , which was moderate (Table 4). Because there was moderate correlation between 435 DD_{egg} and the other two DD terms, we refit the best-fit model adding interactions between 436 DD_{egg} and the other two DD terms. These additional models included all nested possibilities 437 from an individual interaction term to all interaction terms including a three-way interaction. 438 439 Adding interactions increased AICc values by 2.99 – 14.78 points for all models suggesting that 440 non-interaction model best fit the data. Residuals from the best-fit model did not show signs of 441 autocorrelation (Fig. S2).

442

443 *Model testing and validation: best-fit model*

Randomly resampling the recruitment deviations (with replacement) gave a median expected $r^2 = 0.17$ (95% confidence limits of 0.03-0.40) for a five-parameter model suggesting that the observed value of $r^2 = 0.57$ was unlikely to be observed at random.

Removing individual years and refitting the best-fit model (jackknifing) had little impact on the model fit (Figs. 3 and. 5, median $r^2 = 0.57$). Predicting the missing year from any iteration produced estimates very similar to those for the full model (Fig. 3). The three years that showed the highest impact on the model's ability to explain the data were 1999, 2000 and 2006. Removing 1999 reduced the explained variance the most ($r^2 = 0.51$), while removing 2000 or 2006 increased the r^2 to 0.63. Recruitment was higher than predicted in 1999 and 2000 but lower than predicted in 2006. Predicting recruitments for 2006-2010 based on a 1981-2005 model
produced estimates very close to those for the 1981-2010 model (Fig. 3).

455 Resampling individual recruitments with error had little effect on the model's predictive 456 power with median $r^2 = 0.55$ (95% C.I. = 0.59-0.70) across all trials. This result suggests that 457 uncertainty in the recruitment time series (given the current assessment parameters) did not 458 greatly affect the results.

459 Using the jackknife resampling and re-running the entire model fitting process produced 460 results that were largely consistent with the primary analysis. However, removing 1984, 1987, 461 1989 and 1991 did result in different best-fit models. Three of these models included Prey_{Hake} as 462 a predictor. Hake recruitment was above the 95% confidence intervals for estimated recruitment 463 at unfished biomass in 1984 and 1987 but below it in 1989 and 1991 (see Figure e and Table c in 464 Taylor et al., 2015). However, there were other extreme years as well, which did not alter the 465 terms included in the model here. SSH was also selected in three models. Removal of any other 466 year produced the same set of predictors as in the primary analysis (Table 5). Model predictions 467 for the best-fit model closely followed observed model data for 1984, 1987 and 1989 but 468 recruitment in 1991 was lower than predicted. Interestingly, there was no overlap between the 469 three years that had the largest impact on the explanatory value of the best-fit model and those 470 years that affected the re-fitting process.

Finally, the results from resampling the recruitment values (with error) and re-running the entire model fitting exercise one hundred times were also generally consistent with the best-fit model from the primary analysis. The predictors from the best-fit model also occurred in the majority of best-fit models from each iteration: DD_{pre} (72% of models), CST_{egg} (73%), DD_{egg} (89%), LST_{yolk} (83%), and DD_{larv} (70%). Additional predictors included $Prey_{Hake}$ (18%), LST_{pjuv} (18%) and SSH^2 (21%).

477

478 Northern copepod biomass anomaly

When refit to 1996-2010 data, the best-fit model explained 67% of the variation around the stock recruitment-relationship from the stock assessment. Alone the copepod index explained 24% of the variation in recruitment over the 1996-2010 period. However, adding the northern copepod biomass anomaly did not increase the variation explained (also 67%). The anomaly was moderately correlated with most other predictors in the best-fit model with the exception of 484 CST_{egg} : DD_{pre} (r = -0.46), CST_{egg} (r = -0.01), DD_{egg} (r = 0.56), LST_{yolk} (r = 0.48), and DD_{larv} (r 485 = -0.57). Thus, while food supply is likely important, the other physical parameters may act as 486 sufficient proxies in the absence of a full time series of copepod data.

487

488 **DISCUSSION**

Our results suggest that multiple mechanisms acting at different stages the life history of sablefish drive recruitment (Fig. 6). The best-fit model explained approximately 57% of the variation around the stock-recruitment curve and fit the recruitment data well (specifically residuals from the stock-recruitment relationship in the assessment in Fig. 3). Nevertheless, it is not unreasonable to consider this work as an effort to formalize potential hypotheses that should be investigated further.

495 Colder conditions during the spawner pre-conditioning period led to higher recruitment. 496 Given sufficient food, warmer temperatures allow for faster growth and larger female size, which 497 results in higher fecundity (Harvey, 2009). However, colder water temperatures may lower 498 metabolic costs allowing female sablefish to divert more energy to egg production. However, 499 the variability in the ROMS output for this parameter is likely driven primarily by temperature 500 variation at shallower depths, and my not represent female exposure in deeper regions. More 501 likely, colder conditions may be linked to stronger upwelling, which leads to productivity and 502 food availability (Harvey, 2005; Chavez et al., 2003; Peterson, 2009; Sydeman et al., 2011). While we are unaware of data connecting female condition to egg production in sablefish, 503 504 individuals may skip spawn (Rodgveller et al., 2016) given limited energy reserves and degraded 505 body condition as seen in other species like Atlantic cod Gadus morhua (Skjæraasen et al., 2009; 506 Skjæraasen et al., 2012) and Pacific ocean perch (Hannah and Parker, 2007). The results from 507 the jackknife refitting of the entire selection process support the hypothesis that food resources 508 are likely important for females during the period leading up to spawning. For three of the four 509 years that when omitted produced different best-fit models, Prey_{Hake} (positive correlation with 510 recruitment) replaced DD_{pre} as a predictor suggesting that a large hake prey base may also lead 511 to better female condition and recruitment.

512 Recruitment was positively correlated with water temperature during the egg stage 513 (DD_{egg}) . In marine fishes, increased temperature results in faster development, shorter stage 514 duration, earlier hatch date and increased hatching rates, up to a species-specific temperature

515 maxima after which developmental abnormalities occur (Llopiz et al., 2014; Pepin, 1991; Peck et 516 al., 2012; Tsoukali et al., 2016). Growth in sablefish is strongly related to temperature (Sogard 517 and Olla, 2001; Sogard, 2011) suggesting that a similar physiological response would be likely 518 for development. Moving more quickly through the egg state may reduce susceptibility to egg 519 predators and reduce dispersal, the latter potentially important for match-mismatch dynamics 520 with prey resources (Peck et al., 2012). Offshore transport due to upwelling occurs primarily in 521 surface waters. Since eggs are buoyant but yolk-sack larvae found at depth (1000-1200 m), 522 entering deep water sooner may help to avoid advection offshore.

523 Cross-shelf transport was important from January to April corresponding to life-history 524 stages for which sablefish are non- or only minimally motile: eggs and yolk-sack larvae. The 525 positive correlation with CST_{egg} is fairly straightforward: being transported to near-shore nursery environments leads to higher recruitment. At first glance this result may seem at odds with 526 527 Schirripa and Colbert (2006) who found a negative correlation between recruitment strength and 528 on-shore cross-shelf transport during June. However, the time periods (Jan to Apr versus June) 529 and the depths (300-825 m vs. approximately 0-150 m) differ suggesting that the two predictors 530 relate to different processes: (1) transport of eggs onshore following spawning (our analysis) and 531 (2) and upwelling leading to higher productivity and food resources later in the season.

532 Recruitment was also positively correlated with transport to the north during the yolk-533 sack stage (LST_{volk}: Feb to May at 1000-1200 m). Again this result may initially appear to 534 contradict Schirripa and Colbert (2006) who found that stronger southerly transport of surface 535 waters (50-100 m) in February correlated with higher recruitment. However, the depths differ, 536 and these two results may tell different parts of the same story. Southerly transport of surface 537 waters brings boreal copepods into the Northern California Current Ecosystem. These copepods 538 are larger, higher in fatty acids, and a better food source than southern copepods (McFarlane and 539 Beamish, 1992; Peterson, 2009; Peterson and Keister, 2002). Yolk-sack larvae, however, are 540 found at much deeper depths (1000-1200 m). Northern transport at these depths likely brings 541 these larvae to the north where they are more likely encounter these energy-rich copepods once 542 the larvae rise to the surface, start feeding, and eventually grow large enough to feed on the 543 copepods. Access to this food resource might then result in high recruitment through any 544 number of mechanisms including faster growth rates (size-related predation avoidance) and 545 reduced starvation risk.

546 Temperature during the larval stage (DD_{larv}, Feb-May, surface waters) was negatively 547 correlated with recruitment. Sogard (2011) found increased growth at warmer temperatures and 548 higher recruitment in years with good growth, the latter after adjusting for fish size and 549 temperature. However, Schirripa and Colbert (2006) found higher recruitment with offshore 550 transport of surface waters, which coincides with colder upwelled water. Combined, these 551 results suggest a trade-off between better food resources under colder conditions but faster 552 growth for metabolic reasons in warm water. Colder temperatures may index two different 553 mechanisms that result in higher prey availability or quality: enhanced upwelling and higher 554 primary production, and the southerly transport of northern copepods. At the same time, larvae 555 may be more susceptible to starvation under warm conditions due to increased metabolism and 556 lack of prey. Both daily growth and daily mortality are positively correlated with temperature in 557 marine fish larvae in general, although stage-duration decreases (Houde, 2008). Growth of early 558 juvenile sablefish increases with increasing water temperature, but this effect requires sufficient 559 food resources (Sogard, 2011; Sogard and Olla, 2001). Since sablefish have limited capacity or 560 compensatory growth (Sogard and Olla, 2002), they appear to have a risk-prone strategy growing 561 as quickly as possible (Sogard and Olla, 2002; Sogard, 2011) and may be susceptible to 562 starvation under warm conditions. In the lab larvae will move to colder water as rations are 563 decreased, indicating an energy-conservation strategy when food is limited (Sogard and Olla, 564 2001; 1998). Adjusting for water temperature and fish size, Sogard (2011) found higher than 565 expected growth under cold conditions in some years suggesting better food resources under cold 566 conditions.

567 Rapid growth may help individuals avoid predation through a number of mechanisms 568 (e.g., 'bigger-is-better', stage-duration) that cannot be untangled in this study. However, Sogard 569 (2011) did not find evidence for size-selective mortality, possibly because the higher temperatures also lead to increased predator activity and consumption (e.g., Akimova et al., 570 571 2016). One might also expect the offshore transport associated with cold, upwelled waters to 572 advect sablefish larvae away from suitable settlement habitat. Sablefish larvae are found in these 573 surface waters where most of this offshore transport occurs. However, sablefish larvae develop 574 large pectoral fins suggesting that they may have evolved to control their position in the water 575 column (Shotwell et al., 2014; Kendall and Matarese, 1987), and many fish larvae are strong 576 swimmers, especially prior to settlement (Montgomery et al., 2001). Thus, rapid development earlier in the season (warm waters during DD_{egg}) may allow sablefish larvae to develop the swimming capacity to utilize high quality food resources later in the year (DD_{larv} , indexed by cold water) while maintaining position in the water column and staying closer to appropriate settlement habitat.

581 Many of the years that either affected the model fit in the jackknife exercises or in which 582 observed recruitment fell outside the 95% confidence intervals for the best-fit model appear to be 583 associated with El Niño or La Niña events in some way. For example, recruitment was higher than predicted in 2000 and 2008 coincident with La Niña conditions as measured by the ONI⁴ 584 from the summer of 1998 through spring of 2001 and late summer 2007 through late spring of 585 586 2008. Observed recruitment was lower than predicted 2005-2007, with variable El Niño 587 conditions between late summer of 2005 and early 2007. However, in other El Niño or La Niña 588 years the observed recruitment was within the 95% confidence limits for the model predictions. We did try adding both the tri-monthly ONI and bi-monthly MEI⁵ indices to the best-fit model to 589 590 determine if El Niño/La Niña information would improve the model fit, but all such models had 591 higher AICc values than the best-fit model. This result does not necessarily mean that El 592 Niño/La Niña events do not affect sablefish recruitment as the effects on food availability, 593 temperature, and transport are likely already caught in the ROMS parameters.

594 In summary, multiple mechanisms likely drive recruitment of sablefish (Fig. 6). Cold 595 conditions during the pre-conditioning period for spawners may allow females to divert more 596 energy to egg production due to lower metabolic costs, or these colder temperatures may be 597 indicative of upwelling or southerly transport of boreal copepods and better food resources also 598 allowing for more egg production. Warmer temperatures during the egg stage likely allow for 599 higher growth rates and predation avoidance through bigger-is-better mechanisms. On shore 600 transport during the egg and yolk-sack stages retains individuals on shore near settlement habitat 601 and reduces losses due to advection. Northern long-shore transport appears to bring larvae to 602 regions where they are more likely to encounter food resources as they rise to surface waters to 603 start feeding. The mechanism behind the negative correlation between recruitment and water 604 temperature (degree days) during the larval stage is less clear. However, colder water may index

⁴Oceanic Niño Index (ONI) http://www.cpc.ncep.noaa.gov/products/analysis_monitoring/ensostuff/ensoyears.shtml

⁵Multivariate ENSO Index (MEI) https://www.esrl.noaa.gov/psd/enso/mei/

better or alternate food resources due to upwelling or the southerly transport of boreal copepods,and may also reduce basal metabolism and reduce the risk of starvation.

607

608 Implications for stock assessment

609 This study provides the underlying process-orientated hypotheses to address three issues in the 610 stock assessment and management of U.S. west coast sablefish: 1) the reconstruction of what 611 historical recruitment could have been in the absence of length- and age-composition data to 612 inform recruitment during the period when the fishery was operating, 2) sub-annual to one year ahead forecasting of recruitment based on both observed oceanographic conditions and 613 614 potentially sub-annual forecasts of environmental conditions, and 3) long-term projections of 615 potential future stock productivity and the evaluation of the robustness of harvest control rules 616 (HCRs) to changes in stock productivity evaluation via Management Strategy Evaluation (MSE).

617 The five oceanographic indicators identified as important drivers of sablefish recruitment 618 in this study can inform recruitments for historical periods that lack fishery age- and length-619 composition data and for near-term forecasts. These historical and near-term forecasts generally 620 rely upon average recruitment from the stock recruitment curve. However, recruitment is often 621 far above or below the average, with large annual deviations around the stock-recruitment curve. 622 Furthermore, the most uncertain periods in the stock assessment model are those that either lack 623 age and length composition data, or have sparse data. Hind-casting recruitments during periods 624 of high fishery removals without good age and length composition data should result in a 625 reduction in uncertainty during early model years. The oceanographic indices identified during 626 this study could be used as a survey index of recruitment in the stock assessment during the 627 historical time period. Using the five oceanographic indicators from this study to estimate what 628 recruitment deviations could have been, or may be into the near-term future, can provide both 629 better estimates of historical recruitment, and therefore the un-fished spawning biomass 630 reference point used for managing the fishery, as well as the incoming recruitment one year 631 ahead, providing better catch advice based on what recruitment is likely to be entering the fishery 632 in the near term (Kaplan et al., 2016; Siedlecki et al., 2016). Finally, if long-term forecasts of 633 these five oceanographic drivers of sablefish recruitment are available from either Global or 634 downscaled regional Climate Models (GCMs), they could be used to provide long-term strategic 635 projections for fishery management planning that could include not only the fishing industry and

636 managers (PFMC) but those groups interested in the identification of potential climate-637 adaptation strategies given the long-term outlook for the sablefish fishery. This could include 638 societal adaptations that build community resilience to climate driven changes in the sablefish 639 resource such as: 1) market development for a higher quality and priced product, 2) evaluation of 640 the resilience of the sablefish supply chain to projected changes in catch, and 3) aquaculture 641 research and development (Norman-López *et al.*, 2013; Norman-LATA *et al.*, 2014)

642 A number of issues with respect to the oceanographic modeling and data availability will 643 need continuing work. First, the easily available ROMS outputs that were used for this study 644 span the period from 1980-2010. However, the current recruitment estimates from the sablefish 645 stock assessment model continue through 2014, and will continue into the future. There is 646 currently a discontinuity in the ROMS model outputs between 2010 and 2011 because drivers for 647 the surface forcing (heat flux, wind) come from different products for 2011-2015, as does the 648 input for the ocean boundary conditions. This discontinuity prevents the use of the most recent 649 recruitment estimates in evaluating the hypotheses posed in this study. Continuous ROMS model 650 outputs for both the pre-1980 and the post-2010 periods are necessary for fully using the results 651 of this study in sablefish stock assessment and management. Future oceanographic modeling 652 work should investigate the ability of ROMS models to provide informative, short-term seasonal 653 to annual forecasts of relevant oceanographic covariates. Skillful ROMS forecasts could better 654 inform recruitments entering the fishery prior to the availability of survey data.

655 It is important that the oceanographic drivers of sablefish recruitment identified in this 656 study continue to be periodically re-evaluated as the fundamental relationships could change 657 through time, particularly if the impacts of global climate change cause variability in the 658 California Current to exceed recent historical levels (Cheung et al., 2016). This continued re-659 evaluation depends upon both maintaining current oceanographic and biological monitoring 660 programs that will allow for tracking of potential non-stationarity. Finally, best methods for 661 integrating these relationships into the stock assessment model should be investigated. The 662 current stock assessment (Johnson et al., 2016) uses the environmental data as a survey index of 663 recruitment, but new tools to incorporate environmental drivers in the population dynamics with 664 uncertainty are becoming available (Methot Jr and Wetzel, 2013).

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678 LITERATURE CITED.

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936 Figure Legends

937 Figure 1. Sablefish biological parameters from the 2015 stock assessment for 1981-2010. (a)

- 938 spawning stock biomass (SSB) in metric tons, (b) thousands of age-0 recruits, and (c) stock-
- 939 recruitment relationship asserted in the assessment (line) versus observed data (points).
- 940

941 Figure 2. Time-series of independent variables in the best-fit model. (a) degree days during 942 female preconditioning, (b) net cross-shelf transport during the egg stage, (c) degree days during 943 the yolk-sack stage, (d) long-shore transport during the yolk-sack stage, and (e) degree days 944 during the larval stage.

945

Figure 3. Fit of the best-supported model ($r^2 = 0.57$) to the estimated recruitment residuals from the stock assessment. Solid line is the predicted recruitment residuals from the full time series. Dotted lines= 95% confidence limits. Black points are recruitment residuals from the Beverton-Holt stock recruitment relationship from the 2015 sablefish assessment. Stars are predicted values from jackknife analysis removing individual years one at a time. Grey points are predicted values for the retrospective analysis of the last five years. Note, all three points overlap in 2010.

953

Figure 4. Partial residual plots for (a) degree days during female pre-conditioning , (b) net crossshelf transport during the egg stage, (c) degree days during the yolk-sack stage, (d) long-shore
transport during the yolk-sack stage, and (e) degree days during the larval stage.

957

958 Figure 5. Results of jackknife resampling showing distribution of r^2 values. (a) Frequency 959 distribution of r^2 values, (b) r^2 for when the indicated year was removed from the model.

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961 Figure 6. Conceptual model for sablefish showing the environmental drivers at specific life962 history stages that lead to higher recruitment. Signs in parentheses indicate the partial
963 correlation of each term with residuals from the sablefish stock-recruitment relationship. See
964 Figure 4 for plots of these relationships.

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Life history	Time	Donth	Sablefish	Urmathaaia	Stage	Coveriates	Donth ortont	Longitudinal autont	Data source
Life-mstor	y stage period	Deptii	location	nypomesis	Stage	Covariates	Deptil extent	Longitudinai extent	Data source
Precondition	<i>ing</i> Jun - Dec	50-1200m	Bottom	(H1) Food availability affects energy	Prey _{Hake}	Index of age-0 hake	50-1200 m	50-1200 m isobath	Stock
	(Yr 0)			allocation to reproduction with higher	$\operatorname{Prey}_{\operatorname{Lsp}}$	and age-0 longspine			Assessments
				recruitment when more prey are		thornyhead			
	()			available during the preconditioning		abundance			
				period					
	()			(H2) Higher temperature increases	$\mathrm{DD}_{\mathrm{pre}}$	Degree days	50-1200 m	50-1200 m isobath	ROMS
				food demand resulting in lower egg			(150-400 m)	(150-400 m isobath)	
				production, egg quality, or					
				probability of spawning resulting in					
				lower recruitment					
Spawning	Dec - Mar	300-500 m	Bottom	(H3) Temperature acts as a spawning	$Temp_{spawn} \\$	Mean temperature	300-500 m	300-500 m isobath	ROMS
	\mathbf{O}			cue with fish less likely to spawn at					
				high temperature resulting in lower					
				recruitment					
				(H4) As H3 but degree days not mean	$\mathrm{DD}_{\mathrm{spawn}}$	Degree days	300-500 m	300-500 m isobath	ROMS
				temperature					
Eggs	Jan-Apr	300-825 m	Open water	(H5) Eggs are buoyant so Mixed	$MLD_{egg;mn}$	Location of mixed	300-825 m	Inshore limit: 500 m bottom	ROMS
				Layer Depth may limit how far they	$MLD_{egg.sd} \\$	layer depth (m); mean	(240-480 m)	depth,	
	\bigcirc			rise in the water column affecting		and standard		Seaward to 170 nmi	
				later transport		deviation			
				(H6) Transport to settlement habitat	$\mathrm{CST}_{\mathrm{egg}}$	Net cross-shelf	300-825 m	Inshore limit: 500 m bottom	ROMS
				affects recruitment		transport	(240-480 m)	depth,	
								Seaward to 170 nmi	
				(H7) Growth/Predation hypothesis:	$\mathrm{DD}_{\mathrm{egg}}$	Degree days	300-825 m	Inshore limit: 500 m bottom	ROMS
				growth rate is faster in warm water			(240-480 m)	depth,	
				leading to reduced time vulnerable to				Seaward to 170 nmi	
				predators					
Yolk-sack la	arvae Feb-May	1000-1200 m	Open water	(H8) Longshore transport to	LST_{yolk}	Net long-shore	1000-1200 m	1000 m to 170 nmi	ROMS
				settlement habitat affects recruitment		transport			

Table 1 Life history model. See the text for data sources.

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			(H9) Cross-shore transport to	CST_{yolk}	Net cross-shelf	1000-1200 m	1000 m to 170 nmi	ROMS
			settlement nabitat affects recruitment		transport	1000 1000	1000 170	5010
			(H10) Growth/Predation hypothesis:	DD _{yolk}	Degree days	1000-1200 m	1000 m to 170 nmi	ROMS
			growth rate is faster in warm water					
	\mathbf{O}		leading to reduced time vulnerable to					
			predators					
Larvae	(start Feb-May ² Surface waters	Open water	(H11) Longshore transport to	LST_{larv}	Net long-shore	Surface waters	0-150 nautical nmi	ROMS
feeding)			settlement habitat affects recruitment		transport			
	\mathbf{C}		(H12) Cross-shore transport to	CST_{larv}	Net cross-shelf	Surface waters	0-150 nautical nmi	ROMS
			settlement habitat affects recruitment		transport			
	\mathbf{O}		(H13) Growth/Predation hypothesis:	$\mathrm{DD}_{\mathrm{larv}}$	Degree days	Surface waters	0-150 nautical nmi	ROMS
			growth rate is faster in warm water					
			leading to reduced time vulnerable to					
			predators					
Pelagic	juveniles Apr-Nov Surface waters	Open water	(H14) Longshore transport to	LST_{pjuv}	Net long-shore	Surface waters	0-150 nautical nmi	ROMS
(feeding p	elagics)		settlement habitat affects recruitment		transport			
			(H15) Cross-shore transport to	CST_{pjuv}	Net cross-shelf	Surface waters	0-150 nautical nmi	ROMS
			settlement habitat affects recruitment		transport			
			(H17) Growth/Predation hypothesis:	$\mathrm{DD}_{\mathrm{jpuv}}$	Degree days	Surface waters	0-150 nautical nmi	ROMS
			growth rate is faster in warm water					
			leading to reduced time vulnerable to					
			predators					
Benthic	Juvenile Aug-Nov 0 - 250 m	Bottom	(H17) Longshore transport to	LST_{bjuv}	Net long-shore	to 250 m	0-250 m isobath	ROMS
(Age-0)	0		settlement habitat affects recruitment		transport			
			(H18) Cross-shore transport to	CST_{bjuv}	Net cross-shelf	to 250 m	0-250 m isobath	ROMS
			settlement habitat affects recruitment		transport			
	+		(H19) Growth/Predation hypothesis:	$DD_{\textit{bjuv}}$	Degree days	to 250 m	0-250 m isobath	ROMS
			growth rate is faster in warm water					
			leading to reduced time vulnerable to					
			predators					
			(H20) Predation	$\operatorname{Pred}_{\operatorname{bjuv}}$	Some index of	to 250 m	0-250 m isobath	Trawl Survey
	7				predator abundance			& Stock

Assessments

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(H21) Nutrient input (fertilization)	FW_{bjuv}	Annual discharge	NA	NA	Columbia
affects productivity, food availability		from Columbia River			River DART
and therefore survival or river		measured at The			
discharge may affect currents.		Dalles.			
(H22) Sea surface height as an	SSH				Spring north
indicator of basin- scale processes.					coast SSH 44-
Used in previous assessments.					50° N

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Table 2. Results of model selection showing models with a $\triangle AICc < 6.0$. CST = cross-shelf transport, LST = long-shore transport, DD = degree days, Prey = prey index, pre = preconditioning, egg = egg stage, yolk = yolk-sack larvae, larv = larval stage, pjuv = pelagic juvenile, and bjuv = benthic juveniles. AICc = Akiake's Information Criterion adjusted for small sample size.

Model								R2	ΔΑΙϹ
Model 1	DD _{pre} CST _{egg}	DD _{egg}	LST _{yolk}	DD _{larv}				0.57	0
Model 2		$\mathrm{DD}_{\mathrm{egg}}$	LST_{yolk}		LST _{pjuv}		SSH ²	0.46	3.92
Model 3 Prey _{Hake}		$\mathrm{DD}_{\mathrm{egg}}$	LST_{yolk}		LST_{pjuv}		SSH ²	0.51	3.94
Model 4	DD _{Pre} CST _{egg}	$\mathrm{DD}_{\mathrm{egg}}$	LST_{yolk}			SSH		0.51	3.95
Σ									
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Table 3 Coefficients for the best-fit model (Model 1 from Table 2) showing both raw and standardized (beta) coefficients. Bias and standard error (s.e.) are from bootstrap resampling. DD = degree days, CST = cross-shelf transport, LST = long-shore transport, pre= preconditioning, egg = egg stage, yolk = yolk-sack larvae, larvae = larval stage.

	Coefficient	Bias	S.E.	Standardized Coefficient	Bias	S.E
Intercept	6973.8	-888.9	18216.9	-1534.3	38.1	1288.1
DD _{pre}	-523.2	8.2	147.4	-6149.0	103.1	1753.2
$\mathrm{CST}_{\mathrm{egg}}$	112489.8	3344.9	26593.2	6607.4	107.3	1641.2
$\mathrm{DD}_{\mathrm{egg}}$	1308.52	12.8	250.1	11126.9	-91.2	2100.1
LST_{yolk}	44813.1	209.9	15348.5	4920.7	-48.4	1669.3
DD _{larvae}	-231.3	-9.3	84.3	-5012.8	-134.7	1987.8

Table 4 Correlations among variables included in the best-fit model. VIF = generalized variance inflation factor. DD = degree days CST = cross-shelf transport, LST = long-shore transport, pre= preconditioning, egg = egg stage, yolk = yolk-sack larvae, larvae = larval stage.

	DDpre	CST _{egg}	DD _{egg}	LST _{yolk}	VIF
DD _{pre}	-				2.17
CST _{egg}	0.08				1.87
DD _{egg}	O _{0.62}	-0.43			3.38
LST_{yolk}	-0.17	0.23	-0.30		1.63
DD _{larvae}	0.38	-0.14	0.55	-0.54	1.92
	σ				
	5				
I					
	0				
	T				

Table 5. Best-fit models from jackknife refits of the entire model-fitting process. Removing the years 1984, 1987, 1989 and 1991 changed the best-fit model in those iterations. All other years produced models with the same covariates as in the main model.

0	Year remov	ved		
Main model	1984	1987	1989	1991
DD _{pre}	Prey _{Hake}	DD _{spawn}	Prey _{Hake}	Prey _{Hake}
CST _{egg}	CST _{larv}	LST_{yolk}	$\mathrm{DD}_{\mathrm{egg}}$	$\mathrm{DD}_{\mathrm{egg}}$
DD _{egg}	LST_{pjuv}	$\mathrm{CST}_{\mathrm{pjuv}}$	LST_{pjuv}	LST_{yolk}
LST _{yolk}	SSH^2	LST_{bjuv}	DD_{bjuv}	LST _{pjuv}
DD _{larv}		$\mathrm{FWI}_{\mathrm{bjuv}}$	SSH	SSH

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