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6	Article type : Original Article
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9	Oceanographic drivers of petrale sole recruitment in the California Current Ecosystem
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This is the author manuscript accepted for publication and has undergone full peer review but has not been through the copyediting, typesetting, pagination and proofreading process, which may lead to differences between this version and the <u>Version of Record</u>. Please cite this article as <u>doi:</u> <u>10.1111/FOG.12459</u>

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41 Abstract

42 This paper investigates environmental drivers of US west coast petrale sole (Eopsetta 43 *jordani*) recruitment as an initial step towards developing an environmental recruitment index 44 that can inform the stock assessment in the absence of survey observations of age-0 and age 1-45 fish. First, a conceptual life-history approach is used to generate life stage- and spatio-46 temporally-specific mechanistic hypotheses regarding oceanographic variables that likely 47 influence survival at each life stage. Seven life history stages are considered, from female-48 spawner condition through benthic recruitment as observed in the Northwest Fisheries Science 49 Center West Coast Groundfish Bottom Trawl Survey (age-2 fish). The study area encompasses 50 the region from 40-48 °N in the California Current Ecosystem. Hypotheses are tested using 51 output from a regional ocean reanalysis model outputs and model selection techniques. Four 52 oceanographic variables explained 73% of the variation in recruitment not accounted for by 53 estimates based exclusively on the spawning stock size. Recruitment deviations were (1) 54 positively correlated with degree days during the female precondition period, (2) positively 55 correlated with mixed-layer depth during the egg stage, (3) negatively correlated with cross-shelf 56 transport during the larval stage, and (4) negatively correlated with cross-shelf transport during 57 the benthic juvenile stage. While multiple mechanisms likely affect petrale sole recruitment at 58 different points during their life history, the strength of the relationship is promising for stock 59 assessment and integrated ecosystem assessment applications.

60 Key Words: Petrale sole, *Eopsetta jordani*, recruitment, oceanic drivers, California Current

61

# 62 Introduction

63 Petrale sole (Family: Pleuronectidae, Eopsetta jordani) has consistently been the most 64 commercially valuable flatfish targeted in the California Current Ecosystem. During the 1980s through the 2000s, the petrale sole spawning biomass was around or below 10% of the 65 unexploited stock size (Haltuch et al. 2013). More recently the fishery has depended on 66 67 infrequent above average recruitments, generally followed by several years of low recruitments, 68 that drive fluctuations in the spawning biomass (Fig. 1) (Haltuch et al. 2013). This dependence of 69 the fishery on relatively few strong recruitment events makes understanding the drivers of petrale 70 sole recruitment a priority.

71 The horizontal-advection bottom-up forcing paradigm provided a mechanistic framework 72 for testing the hypothesis that climate-driven changes in cross-shelf and alongshore advection 73 drive petrale sole recruitment strength (Parrish et al. 1981, Di Lorenzo et al. 2013). This 74 paradigm suggested that large-scale climate forcing drives regional changes in alongshore and 75 cross-shelf ocean transport that directly impact ecosystem functions such as productivity, species 76 composition, and spatial connectivity. Oceanographic studies illustrated the influence of large-77 scale climate drivers on mesoscale processes in the California Current Ecosystem that drive 78 cross-shelf and alongshore transport at local scales (Hickey 1979, Chelton et al. 1982, Keister et 79 al. 2011, Combes et al. 2013, Di Lorenzo et al. 2013). Variability in cross-shelf transport in the 80 California Current Ecosystem was important for understanding population and ecosystem 81 dynamics because of variable offshore and northward transport of nutrients, mass, and organisms 82 (Combes et al. 2013, Di Lorenzo et al. 2013, Jacox et al. 2016). Additionally, fish life cycles and 83 population connectivity patterns can be impacted via the transport of larvae by either coastal 84 upwelling or offshore advection of water from the poleward flowing California Undercurrent 85 (Combes et al. 2013, Di Lorenzo et al. 2013). Research has suggested that horizontal advection 86 affects density-independent survival during early life stages, specifically that offshore transport 87 of eggs and larvae results in poor recruitment (Castillo et al. 1994, Castillo 1995). Furthermore, 88 decadal-scale variability in wind-driven cross-shelf transport of early life stages of other flatfish

species towards favorable nursery areas influenced recruitment strength (Wilderbuer et al. 2002,
Hollowed et al. 2009, Wilderbuer et al. 2013). Concurrently, temperature can influence growth
rates of early life history stages, mediating larval vulnerability to predation and the likelihood of
starvation (Houde 1987, Litvak & Leggett 1992, Leggett & DeBlois 1994, Chezik et al. 2014).

93 This study leveraged recruitment estimates from an information-rich stock assessment for 94 the California Current Ecosystem petrale sole stock (Stawitz et al. 2016) to build a mechanistic 95 framework for investigating oceanographic drivers of survival using a conceptual life-history 96 approach to frame hypotheses regarding factors that affect survival at each early life history 97 stage. Petrale sole were an excellent candidate for investigations into oceanographic drivers of 98 recruitment strength for multiple reasons. First, the petrale sole data set were likely the most 99 information-rich groundfish data set due to the large amount of age and length data that spans 100 multiple generations (compared to <1 generation for most rockfishes). Next, multiple stock 101 assessment models and model sensitivity runs, and thus historical recruitment estimates, 102 produced highly similar results. Finally, previous fisheries and oceanographic research suggested 103 potential mechanisms for oceanographic drivers of petrale sole recruitment. However, prior to 104 using oceanographic recruitment drivers to forecast recruitment for either near-term tactical 105 management or long-term strategic decision-making, statistical models must be built and 106 evaluated. Our goal was to develop a statistically robust environment-recruitment relationship to 107 explain variability in recruitment (Basson 1999) for use in both near-term and long-term petrale 108 sole recruitment forecasting. We developed and tested a series of generalized linear models 109 (GLMs) to explore the influence of oceanographic drivers on petrale sole survival at different life 110 stages, from female spawner condition through benthic recruitment.

111

#### 112 Methods

This study applied the methodology implemented by Tolimieri et al. (2018) for sablefish to the U.S. west coast petrale sole stock using estimates of log recruitment deviations from the petrale stock assessment (Stawitz et al. 2016) from 1981 - 2010 and model output from a California Current Ecosystem configuration of the Regional Ocean Modeling System (ROMS) with data assimilation (Neveu et al. 2016). While reliable recruitment estimates from the petrale sole stock assessment ranged from 1945, due to earlier collections of length and age composition data, through 2013, the ROMS model outputs did not span the full time series of recruitment 120 estimates. ROMS model outputs were available from 1981 through 2010. We focused on the 121 northern portion of the stock (40-48°N) because a majority of the spawning biomass and 122 spawning grounds were distributed in the north and recruitment estimates from the coast-wide 123 stock assessment\_model were most strongly informed by age and length composition data from 124 the north (Stawitz et al. 2016). Petrale sole stock assessments prior to 2009 were conducted 125 separately for northern and southern regions of the California Current Ecosystem (Lai et al. 126 2005). However, due to strong similarities in both spawning biomass and recruitment trends the 127 regional stock assessments were combined into a single coast-wide stock assessment during 2009 128 (Haltuch & Hicks 2009). We considered the time from female pre-spawning condition through 129 benthic recruitment (age-0 fish). This conceptual approach allowed for the generation of life-130 stage- and spatio-temporally-specific hypotheses regarding the physical variables likely to 131 influence petrale sole survival at each life stage leading to recruitment. Hypotheses were tested 132 using linear modeling, model selection, and model validation.

#### 133

## 134 *Petrale sole life history: female preconditioning to age-0 recruits*

135 The conceptual life-history approach tracked each life stage beginning with female 136 condition prior to the spawning season through recruitment at age-0 (Table 1). Research on 137 rockfishes and sablefish has shown that female condition may influence whether or not an 138 individual spawns, and the quality and number of eggs produced by female fishes (Sogard et al. 139 2008, Rodgveller et al. 2016). While petrale sole may have the potential for skip spawning, 140 currently available data are inconclusive (Pers. Comm. M. Head). Petrale sole spawn during the 141 winter, thus the summer and fall prior to spawning (May to October) were important for female 142 preconditioning.

143 Spawning occurred from November to March with a peak during December to January 144 (Cleaver 1949, Harry 1959, DiDonato & Pasquale 1970, Pedersen 1975). Petrale sole form deep 145 water spawning aggregations (270-460 m) at localized sites along the shelf break, with most 146 spawning taking place at depths ranging from 250 m to 500 m (Harry 1959, Best 1960, Gregory 147 & Jow 1976, Castillo et al. 1994, Castillo 1995, Love 1996, Moser 1996, Casillas et al. 1998). 148 Females spawn once each year and fecundity varies allometrically with fish size, with one large 149 female laying as many as 1.5 million eggs (Porter, 1964). Eggs rise to the surface between 150 November and April, remaining in surface waters for 6 to 14 days (Alderdice & Forrester 1971,

151 Hart 1973, Love 1996, Casillas et al. 1998), then sink to approximately 400 m depth before 152 hatching (Mason et al. 1983, Kendall & Matarese 1987, Hunter et al. 1989, Moser et al. 1994) as 153 yolk-sac larvae. After hatching, the larvae rise to the upper 50 m of the water column and remain 154 there for approximately five months, through the feeding larval stage (Alderdice & Forrester 155 1971, Hart 1973, Pearcy et al. 1977, Love 1996, Casillas et al. 1998). Planktonic petrale sole 156 larvae range in size from approximately 3 to 20 mm, and were found up to 150 km offshore 157 foraging upon copepod eggs and nauplii (Hart 1973, MBC Applied Environmental Sciences 158 1987, Moser 1996, Casillas et al. 1998). At about 2.2 cm in length petrale sole larvae settle to the benthos as age-0 recruits on the inner continental shelf (Pearcy et al. 1977). Benthic juveniles 159 160 inhabit sandy or sand-mud bottoms (Eschmeyer et al. 1983, MBC Applied Environmental 161 Sciences 1987)-, ranging in size from approximately 2.2 cm to the size at maturity. Fifty percent 162 of the population was mature at approximately 38 cm and 41 cm for males and females, 163 respectively (Casillas et al. 1998). Specific nursery areas for juvenile petrale sole have not been 164 identified. Due to a lack of sampling of newly settled petrale sole, this study evaluated two 165 alternative hypotheses regarding settlement depths, from 50 m to 150 m depth and from 150 m to 166 500 m depth.

In northern and central California petrale sole were found primarily on the middle and
outer continental shelf (Allen et al. 2006). Pacific Fisheries Information Network (PacFIN)
fishery logbook data showed that a majority of the adult petrale sole catch off the U.S. West
Coast takes place between 70 and 220 m during March through October, and between 290 and
440 m from November through February.

172Juveniles show little coast-wide or bathymetric movement, while studies suggested that173adults generally move inshore, and northward to the continental shelf feeding grounds during the174spring and summer and offshore and southward to deep water spawning grounds during the fall175and winter (Hart 1973, MBC Applied Environmental Sciences 1987, Horton 1989, Love 1996).

176

# 177 Regional Ocean Modeling System (ROMS)

Predictors considered here were physical oceanographic parameters including
temperature, long-shore and cross-shelf currents and mixed layer depth (Table 1). Variables
were obtained from a California Current Ecosystem configuration of the Regional Ocean
Modeling System (ROMS) of the California Current Ecosystem with 4-Dimensional Variational

182 (4D-Var) data assimilation (Neveu et al. 2016). The ROMS model domain covered the region 183 30-48°N and from the coast to 134°W at 0.1° (~10 km) horizontal resolution, with 42 terrain-184 following vertical levels. For the 1980-2010 California Current Ecosystem reanalysis used for 185 this study, satellite observations (SST, SSH) and in situ data (temperature and salinity 186 measurements from ships, floats, moorings) are assimilated into the model to improve its 187 representation of the true ocean state. This reanalysis has been used extensively to study physical 188 and biogeochemical dynamics in the California Current Ecosystem (Jacox et al. 2014, Jacox et 189 al. 2015a, Jacox et al. 2015b), and to develop models of marine species recruitment (Tolimieri et 190 al. 2018) and distribution (e.g., Brodie et al. 2018). All ROMS outputs for the study region were 191 obtained as 4-day averages, and then either averaged or summed over the appropriate life-history 192 time period (as defined in Error! Reference source not found.) within each of the 30 years 193 (1980 to 2010) for each time series in the analysis. Specific hypotheses for each predictor below 194 are described below.

195 Oceanographic data were scarcer at depth than at the surface. As a result, the ROMS 196 output that assimilates available data was less constrained by observations at the subsurface than 197 at the surface. Therefore, the ROMS output likely deviates from nature more at the subsurface 198 than at the surface. However, the paucity of subsurface data was the motivation for using model 199 output in this study; the model output was useful despite the inability to validate the subsurface 200 model transport. Data assimilation was used to improve models that already capture the 201 California Current Ecosystem dynamics without data assimilation. The fidelity of this model to 202 nature has been documented extensively for applications with and without data assimilation (e.g., 203 Veneziani et al. 2009, Jacox et al. 2015a). The model was forced by realistic winds, surface heat 204 fluxes, and lateral boundary conditions, which drive realistic physical variability even in the 205 absence of data assimilation. Assimilated data, even if only available at the surface, can 206 influence the subsurface dynamics. For example, assimilating sea surface height measurements 207 constrains the geostrophic flow and therefore transport at the both the surface and subsurface. 208 The impact of assimilated data on different metrics of the California Current Ecosystem 209 circulation is discussed in detail in Moore et al. (2017).

210

#### 211 Hypotheses and additional data sources

We developed 18 *a priori*, life-stage-specific and space-specific (considering time, depth, and longitude) hypotheses for environmental covariates that may drive variation in petrale sole recruitment (**Error! Reference source not found.**). Hypotheses for age-0 benthic juveniles were tested using two depth ranges given the uncertainty and lack of literature regarding the depth range of newly settled to age-2 juvenile petrale sole. Given the 18 hypotheses, three of which had two covariates, 21 total predictors were tested. Each hypothesis tested had specific temporal, longitudinal, and depth ranges (**Error! Reference source not found.**).

Only hypotheses that had ROMS model outputs available were tested (Table 1), thus excluding potential hypotheses such as food availability for larvae and pelagic juveniles (McFarlane & Beamish 1992), which remain untestable due to insufficient data. This study did not test mechanisms with potential indirect effects on survival at each life stage such as drivers of food availability (e.g. upwelling). However, some of these effects are implicit in the transport and temperature indices that are included (e.g., upwelling cools surface waters and drivers offshore transport).

226

## 227 Temperature

228 Temperature may affect recruitment through multiple mechanisms. During spawner 229 preconditioning (Table 1, hypothesis 1, hereafter, H1) warmer temperatures may increase energy 230 demands (H1), forcing female petrale sole to allocate less energy to reproduction by reducing 231 egg production or aborting spawning. Temperature may also act as a spawning cue (H2 and H3), 232 or affect growth, development, survival, and susceptibility to predation of eggs and pelagic 233 larvae through multiple mechanisms, among which we cannot distinguish (H7, H10, H13, and 234 H16). For example, warmer temperatures allow for faster growth and may help larvae to 235 transition through developmental stages (e.g., egg stage) or outgrow potential predators (e.g., 236 stage duration or 'bigger-is-better' hypotheses; Houde 1987, 1997), but warm waters may also 237 increase energy demands making larvae more susceptible to starvation, especially if warmer 238 waters bring poor feeding conditions. In most cases, temperature data was included as degree 239 days (cumulative temperature above a threshold value, Chezik et al. 2014). There is limited 240 information on the effects of temperature on the development of petrale eggs and larvae.

Alderdice and Forrester (1971) saw slight curvature and an increase in the rate of development from  $4 - 8.5^{\circ}$ C. Therefore, we set the cumulative threshold to  $3.5^{\circ}$ C in this study.

243

244 Transport

245 Transport to or away from appropriate settlement habitat can strongly affect larval supply 246 and settlement of marine species with pelagic eggs and larvae (Montgomery et al. 2001, 247 Schirripa & Colbert 2006). Transport to settlement habitat was characterized by estimates of 248 mean long-shore (LST: H5, H9, H11, H14 and H17) and cross-shelf transport (CST: H6, H8, H12, H15 & H18) at specific depths and times for each petrale sole life-history stage. 249 250 Petrale sole spawn at depth but eggs rise in the water column. The mixed-layer depth 251 (MLD) may influence how high eggs rise in the water column, thereby affecting transport (H4) 252 (Sundby 1991, Hinckley et al. 1996) and available food resources post-egg hatch (Cushing 253 1982). Therefore, the mean MLD from November to April, when eggs are in the water column, 254 was included.

255

# 256 Preliminary Analyses

257 The correlations among predictor variables and individual linear and quadratic 258 regressions for each predictor against recruitment deviations were calculated (Tables S1 and S2) 259 and evaluated prior to model fitting and selection. Strongly correlated predictors (Table S1, r >260 0.70), were not permitted in the same model (Dormann et al. 2013). Quadratic terms were 261 included in the main modeling exercise, in addition to linear terms, if the individual quadratic 262 model fit better than the individual linear model (Akiake's Information Criterion, AIC, was <2.0 263 that of the linear model, see Table S1, Burnham & Anderson 1998). Based on preliminary 264 analyses, a quadratic term for degree-days during the egg stage (DD<sub>egg</sub>) and degree-days during 265 the larval stage (DD<sub>larv</sub>) were included as potential predictors in the final model selection.

266

# 267 *Recruitment deviations*

Model estimates of petrale sole recruitment were taken from the most recent stockassessment (Fig. 1c here, Fig. C in Johnson et al. 2016, Stawitz et al. 2016). Specifically, the model-estimated log deviations where the deviations were the annual deviations from log-mean recruitment were used. Mean recruitment was a function of the assumed Beverton-Holt stock272 recruitment relationship, aging error assumptions, and the data included in the stock assessment273 model.

- 274
- 275 Model selection

A series of generalized linear models (GLMs) were run including all possible permutations of the 21 covariates (n = 11611 models) excluding highly correlated terms ( $|\mathbf{r}| > 0.7$ ) from the same model. Each covariate matched a specific hypothesis in Table 1. The number of predictors in a candidate model was limited to five (one covariate per six data points in the time series) to prevent over-fitting. The potentially quadratic predictors were allowed to enter the models as linear, or linear + quadratic forms. The best-fit models were selected using  $\Delta AICc$ values, retaining models with  $\Delta AICc < 2.0$  (Burnham & Anderson 1998).

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285

284 Model validation and testing

Performance of the best fit model was evaluated using

(1) resampling with replacement of recruitment deviations to estimate  $r^2$  values using 286 287 100 randomized data sets; (2) bootstrapping whole years with replacement to estimate bias and 288 calculate standard error of the parameter estimates; (3) annual jackknife resampling to determine the effect of any single year on the  $r^2$ ; (4) resampling annual recruitment deviations from a log-289 290 normal distribution using the annual mean and standard deviation estimated from the assessment 291 (Fig. 1c here, Stawitz et al. 2016, Fig c.), then recalculating recruitment residuals and refitting 292 the model 1000 times; (5) refitting the model using data for 1981-2005 and predicting 293 recruitments deviations for 2006-2010; (6) jackknife resampling to re-run the entire model fitting 294 and comparison exercise, to determine if removal of any individual year would change the 295 selected oceanographic variables; and (7) the entire model fitting exercise was re-run 100 times 296 using the re-sampled petrale recruitment deviations with error (from Step 4 above), comparing 297 AICc selected models from each run. Finally, (8) we used jackknife resampling but fit only the vears 1981-2005, compared the resulting models to the best-fit model above, and used the 1981-298 299 2005 model to predict recruitment deviations for 2006-2010.

300

#### 301 **Results**

302 There were six candidate models with a  $\Delta AICc < 2.0$ . All of these models had the same 303 three core terms (Error! Reference source not found.): degree days during the female 304 preconditioning stage (DDpre), mixed-layer depth during the egg stage (MLDegg), and cross-305 shelf transport during the larval stage (CSTlarv) (Fig. 2). Two of these models (Model 1 and 306 Model 2) included four total predictors, while the remaining models included five. The fourth 307 terms for Model 1 and Model 2 were highly correlated and essentially indistinguishable: cross-308 shelf transport during the benthic juvenile stage (CSTbjuv.a) and long-shore transport during the 309 same stage (LSTbjuv.a)(Fig. 2). We chose Model 1 as the best-fit model because it had the lowest AICc, highest weight, and fewest parameters. However, we note that CSTbjuv.a and 310 311 LSTbjuv.a are highly correlated with several other variables and distinguishing between them 312 was not possible.

313 Predictions of recruitment deviations from the best-fit model closely followed those estimated within the stock assessment ( $r^2 = 0.73$ , Fig. 3a). Analysis of partial residuals showed 314 315 that recruitment deviations were positively correlated with degree days during the female 316 preconditioning period and mixed-layer depth during the egg stages. Recruitment deviations 317 were negatively correlated with cross-shelf transport during both the larval stage and benthic 318 juvenile stages (Fig. 4, Error! Reference source not found.). Standardized coefficients 319 suggested roughly similar effects of the three core variables on recruitment deviations with 320 cross-shelf transport during the benthic juvenile stage having the smallest effect (Error! 321 Reference source not found.). Including only the three core terms in the model resulted in a moderate decrease in fit ( $r^2 = 0.64$ , Table S3). 322

There was weak correlation among the covariates in the best-fit model with the highest being -0.58 for the correlation between mixed-depth layer and cross-shelf transport during the benthic juvenile stage (**Error! Reference source not found.**). The variance inflation factor values (VIF), which measure how much the variance of the estimated regression coefficients are inflated as compared to when the predictor variables are not linearly related, were low (**Error! Reference source not found.**). The diagnostics for the best-fit model show good model fit (Figs. 4 and S1) with residuals that did not show signs of autocorrelation (Fig. S2).

#### 331 Model testing and validation: core-fit model

- Randomly resampling the recruitment deviations (bootstrap with replacement) gave a median expected  $r^2 = 0.14$  (95% C.I. = 0.02-0.33) for the core model suggesting that the observed value of  $r^2 = 0.73$  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 & 5, median  $r^2 = 0.73$  (95% C.I. = 0.70-0.76). Predicting the missing year from any iteration produced estimates very similar to those for the full model (Fig. 3a). The years that showed the highest impact on the model's ability to explain the data were 1991 (increased to  $r^2 = 0.77$ ) and 2008 (decreased to  $r^2 = 0.69$ ) (Figure 5).
- Resampling annual recruitment deviations with error produced a modest decline in model performance (median  $r^2 = 0.0.62$  (95% C.I. = 0.48-0.75. This suggests that uncertainty in the time-series of recruitment deviations (given the current assessment parameters) results in a somewhat lower ability to explain the variability in recruitment.
- Using the jackknife resampling and re-running the entire model fitting process produced
  results consistent with the primary analysis (Table 5). Degree days during preconditioning,
  mixed-layer depth, and cross-shelf transport during the larval stage appeared in all models, while
  cross-shelf transport and long-shore transport during the benthic juvenile stage alternated in
  appearance.
- Resampling the recruitment deviations (with error) and re-running the entire model fitting exercise 100 times was generally consistent with the core model from the primary analysis (Table 6). The three core terms appeared in the majority of models and years.
- 352 Finally, jackknife resampling and re-running the entire model fitting process for the 1981 353 - 2005 data supported the inclusion of the three core terms, which appeared in all best-fit models 354 given the exclusion of a given year (Table 7). The only exception was cross-shelf transport 355 during the larval stage, which appeared in 24 out of 25 years. Predicting recruitment deviations 356 for 1981-2005 generally did a good job of predicting recruitment deviations ( $r^2 = 0.61$ , Fig. 3b). 357 This three-term model did not predict the high recruitment deviations in 2006-2008 when used to 358 forecast recruitment (Fig. 3b) but did forecast increases in recruitment for these years and closely 359 forecast recruitment deviations for 2009 and 2010. Thus Model 1 appears to be an effective yet 360 conservative predictor.
- 361

#### 362 Discussion

363 The four oceanographic variables in the best-fit model explained 73% of the variation in 364 petrale sole recruitment not accounted for by estimates based exclusively on the spawning stock 365 size. Recruitment deviations were positively correlated with: (1) degree days at the female 366 preconditioning stage and (2) mixed-layer depth at the egg stage, but negatively correlated with 367 (3) cross-shelf transport during the larval stage and (4) cross-shelf transport during the benthic 368 juvenile stage. These results suggest that petrale sole recruitments were a product of multiple 369 mechanisms acting at different early life history life stages (Fig. 6). Notably, although the time 370 spans and depth ranges for each life stage were different, both the core model relationships for 371 petrale sole and the best fit model from a similar study for California Current sablefish 372 (Tolimieri et al. 2018) share some of the same linear covariates that suggested spawner condition 373 and horizontal transport impact recruitment success for both species.

374 The best-fit model (Model 1) fit the recruitment deviations from the stock-recruitment 375 relationship well. During the spawner preconditioning stage degree days from May through 376 October impacted subsequent recruitments, with better recruitment when females experienced 377 warmer conditions prior to spawning. Some species like Atlantic cod (Pedersen & Jobling 1989) 378 and bullhead (Abdoli et al. 2005) showed positive correlations between growth or egg production 379 and water temperature. Presumably warmer water temperatures allowed for higher metabolism 380 and more rapid conversion of energy into eggs resulting in higher growth and fecundity given 381 sufficient food resources. Earlier work by Castillo et al. 1994 also concluded that warmer 382 water temperatures were associated with increased egg and larval survival, which may be 383 influenced by increased egg production.

384 During the egg stage, mixed layer depth from November through April was an important 385 driver of recruitment. Specifically, water below the MLD was denser than that above, thus 386 petrale sole eggs that are positively buoyant at spawning depths below the MLD were neutrally 387 buoyant at or above the MLD (Alderdice and Forrester 1971). The vertical distribution of fish 388 eggs in the water column affected horizontal transport of eggs and larvae (Sundby 1991, 389 Hinckley et al. 1996). A shallower MLD may allow eggs to rise and be transported into more 390 productive waters, resulting in higher larval survival due to better feeding conditions (Cushing 391 1982).

392 At both the larval (from December through May) and benthic juvenile (from April to 393 October) stages, cross shelf transport was an important driver of recruitment. Similarly, Castillo 394 et al (1994) found that winter horizontal transport played an important role in determing petrale 395 sole year class strength, describing over half recruitment variability estimated at that time. 396 Stronger cross-shelf transport from deep water spawning locations to shallower continental shelf 397 waters transported and retained individuals in near-shore settlement habitat, resulting in higher 398 recruitment and reduced advection of larvae and juveniles off-shore, potentially important for 399 match-mismatch dynamics with prey resources (Peck et al. 2012).

400 Hypotheses regarding recruitment drivers at other life stages including spawning, the 401 sinking egg stage, and pelagic juvenile stage were not included in the core model. However, the 402 AICc selected suite of models (Table 2) includes additional covariates that appear to be 403 occasional recruitment drivers including long-shore transport during the surface egg and benthic 404 juvenile stages, and degree days during the surface egg and larval stages. The cross-shore and 405 long-shore transport at the benthic juvenile stages, as well as degree days at the surface egg and 406 larval stages were highly correlated (Table S1). Including these additional covariates explained 407 up to about 75% of the variability in petrale sole recruitment deviations. However, model 408 diagnostics for these models were poor, indicating that they were likely overfitting the data. The 409 negative relationship with cross-shore transport and the positive relationship with long-shore 410 transport during the benthic juvenile stage may indicate that if benthic juveniles were transported 411 offshore and to the south then recruitment is lower, possibly due to relatively poor food resources 412 in this region. The positive relationship with warm conditions during the surface egg and larval 413 stages may indicate faster development in warmer waters, thus decreased predation pressure 414 (Chavez et al. 2003, Harvey 2005, Peterson 2009, Sydeman et al. 2011).

415 The multiple model validation methods applied to these data suggest that the core model 416 predictions were robust. Recruitment residuals from the Beverton-Holt stock recruitment 417 relationship, based on the 2015 petrale sole stock assessment fell within the core model 95% 418 confidence interval during 24 years, below during 3 years (1986, 2001, and 2005), and above 419 during 3 years (1989, 1990, and 2008) (Fig. 3a). The years in which recruitment residuals fell 420 outside of the predicted 95% confidence interval did not clearly align with large-scale climate 421 oscillations in the north Pacific, such as the El Niño – Southern Oscillation (ENSO). The core 422 model predictions from a leave-one-year-out jackknife analysis provided predictions within the

95% confidence limits of the fitted core model for all years, 1981 through 2010 (Fig. 3a). Core
model predictions that used the available data through 2005 then predicting 2006-2010 resulted
in under predicting three out of five years that were among the largest recruitment deviations in
the time series (Fig. 3b). These under predictions were likely due to increases in the spawning
biomass at the end of the time series that were removed from the 1981 through 2005 training
data set.

429 This study provides the basis for understanding and forecasting the implications of 430 climate change and climate variability on petrale sole recruitment within the stock assessment 431 process, ecosystem assessments, and MSEs (e.g., Hollowed et al. 2009). The stock assessment 432 uses the Northwest Fisheries Science Center West Coast Groundfish Bottom Trawl Survey data, 433 which initially samples petrale sole as age-2 fishes. However, petrale sole are not 50% selected 434 until approximately age-6 (Stawitz et al. 2016). The ability to predict recruitment more than one 435 year ahead would allow the stock assessment to provide better catch advice based on improved 436 estimates of the recruitments likely to be entering the fishery in the near term, prior these 437 recruitments being well sampled by the survey (Kaplan et al. 2016, Siedlecki et al. 2016). 438 Annually updating environmental predictors to inform recruitment forecasting based on both 439 observed oceanographic conditions and potentially sub-annual forecasts of environmental 440 conditions (Siedlecki et al. 2016, Jacox et al. 2017) could provide leading environmental 441 indicators of recruitment for managers and fishers. Additionally, long-term stock productivity 442 projections based on oceanographic output from global climate models or regional ocean models 443 could be useful for fishery stakeholders and for evaluating the robustness of harvest control rules 444 (HCRs) to changes in long-term stock productivity evaluation via Management Strategy 445 Evaluation (MSE). Of course, given concerns about nonstationarity in environment-recruitment 446 relationships (Myers 1998), periodic re-evaluations any environmental drivers of recruitment are 447 necessary, and are dependent upon the continuation of both oceanographic and biological 448 monitoring programs.

This work can also inform the selection of historical environmental data sets for hindcasting petrale sole recruitment in the absence of age and length composition data, rather than assuming some level of average recruitment into the past based on more recently observed recruitment variability. Using environmental indicators of what early recruitment might have been could reduce stock assessment uncertainty and provide better estimates of management 454 reference points. Stock assessments estimate age-0 recruitment conditioned on the available data,

- 455 a majority of which may come from only one environmental regime. Using stock-recruitment
- data from only one environmental regime may result in over- or underestimating past
- 457 recruitment, producing bias in the calculation of fishery reference points, particularly in the
- 458 California Current Ecosystem where many age and length time series are short relative to the
- timescales of low frequency environmental variability (e.g., the Pacific Decadal Oscillation,
- 460 Berger 2019).

Transitioning the environmental drivers of petrale sole recruitment from this research to an operational product for both stock assessments and ecosystem assessments for fishery managers ideally requires the extension of ocean reanalyses both forward and backwards in time, this study uses data from 1980 through 2010. Alternatively, more easily available observational data sets could serve as proxies for the selected covariates. Finally, if available, skillful, shortterm oceanographic model forecasts of relevant oceanographic covariates could be used to inform recruitment estimation prior to survey data availability.

468

# 469 Acknowledgements

We thank Dr. Vlada Gertseva, Dr. Owen Hamel, and Dr. James Hastie, for discussion in
general and comments on the manuscript. Reviews from three anonymous reviewers greatly
improved this manuscript. ROMS model output was provided by the UC Santa Cruz oceanmodeling group and is available at <a href="http://oceanmodeling.ucsc.edu">http://oceanmodeling.ucsc.edu</a>. We acknowledge funding
from NOAA's Joint Polar Satellite System Proving Ground and Risk Reduction program.
Special thanks to S.M. Lieber and G. Stafford.

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# 477 Author Contributions

478 M.A.H. and N.T. conceived the study. L.Q., M.A.H, and N.T. developed the conceptual life

history model. N.T. and M.A.H. developed and undertook statistical analyses. M.G.J. provided
the oceanographic expertise and ROMs model outputs for the study. M.A.H. and N.T. drafted the
manuscript with input and revisions from all authors.

- 482
- 483 **Conflict of Interest**

The authors do not have any conflicts of interest, commercial or otherwise, that have influencedthe findings of this research.

486

# 487 Data Availability Statement

488 The data associated with this manuscript is available through the NOAA Northwest Fisheries

- 489 Science Center, Fishery Resource Analysis and Monitoring Division.
- 490
- 491

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Life Stage	Year	Stage Duration	Stage Depth	Ho Number	Hypothesis	ROMS Covariates (40° N – 48° N)
	Vear 0		Bottom		Higher bottom water temperatures increases food	Mean bottom water temperature
Draconditioning	May		dopths of EQ	1	demand resulting in lower egg production, egg	(°C, 4d)
Preconultioning	iviay-			T	quality, or probability of spawning and lowers	
	Uct		- 200 m		recruitment (likely a bell-shaped relationship)	
	Year 1,		Bottom		Bottom water temperature acts as a spawning cue	Mean bottom water temperature
Spawning	Nov -		depths of	2	with fish less likely to spawn at high temperature	(°C, 4d)
	Mar		250 - 500 m		resulting in lower recruitment	
			Bottom		Water column temperature acts as a spawning cue	Mean water column temperature
σ			depths of	3	with fish less likely to spawn at high temperature	(°C, 4d)
			250 - 500 m		resulting in lower recruitment	
	Voor 1		Water		Eggs are buoyant so Mixed Layer Depth limits how	Mean <b>mixed layer depth</b> (m)
Far outros	Neur		column from	4	far they rise in the water column affecting later	
Egg, surjace	NOV-	6-14 days	surface to	4	transport	
	Apr		MLD			
			Water		Transport in the water column above the MLD to	Mean long-shore transport
			column from	_	settlement habitat affects recruitment	between the surface and MLD (m/s,
<u> </u>			surface to	5		4d cumulative)
			MLD			
			Water		Transport in the water column above the MLD to	Mean cross-shelf transport
			column from	6	settlement habitat affects recruitment (aka	between the surface and MLD (m/s,
			surface to		Advection reduces recruitment while retention	4d cumulative)

MLD

			Water		Growth/Predation hypothesis: growth rate is faster	Degree days in the water column
			column from	7	in warm water leading to reduced time vulnerable	calculated from mean water column
			surface to	/	to predators	temperature between the surface
()			MLD			and MLD (days, 4d)
			Water		Transport in the water column to settlement	Mean long-shore transport from the
U)	Year 1,		column from		habitat affects recruitment (aka Advection reduces	MLD to 400m (m/s, 4d cumulative)
Egg, sinking	Nov-	6-14 days	MLD to 400	8	recruitment while retention enhances recruitment)	
	Apr		m			
			Water		Transport in the water column to settlement	Mean <b>cross-shelf transport</b> from the
<b>M</b>			column from		habitat affects recruitment	MLD to 400 m (m/s, 4d cumulative)
			MLD to 400	9		
			m			
			Water		Growth/Predation hypothesis, growth rate is faster	<b>Degree days</b> in the water column
			column from		in warm water leading to reduced time vulperable	calculated from mean water column
				10	to produces	targenerative between the MID and
0			MILD to 400		to predators	temperature between the MLD and
			m			400 m (days, 4d)
		~ 5			North to south transport in the water column	Mean long-shore transport in the
Larvae (both	Year 1,	months	Water		brings northern zooplankton and leads to higher	water column at 50-150 km offshore
yolk sac and	Dec-	for all	column from	11	survival and recruitment	(m/s, 4d cumulative)
feeding)	May	pelagic	0 - 50 m			
		stages				

			Water		Transport in the water column to settlement	Mean <b>cross-shelf transport</b> in the
			column from	12	habitat affects recruitment (aka Advection reduces	water column at 50-150 km offshore
			0 - 50 m		recruitment while retention enhances recruitment)	(m/s, 4d cumulative)
			Watar		Growth/Predation hypothesis: growth rate is faster	Degree days in the water column
			water	10	in warm water leading to reduced time vulnerable	calculated from mean water column
				15	to predators	temperature at 50 - 150 km offshore
$\mathbf{O}$			0 - 50 m			(days, 4d)
		~ 5			North to south transport brings northern	Mean long-shore transport in the
	Year 1,	months	Water		zooplankton and leads to higher survival and	water column at 80-120 km offshore
	Mar -	for all	column from	14	recruitment, Transport to settlement habitat affects	(m/s, 4d cumulative)
(Jeeding	June	pelagic	0- 150 m		recruitment	
pelugics)		stages				
$\mathbf{C}$			Water		Transport in the water column to settlement	Mean cross-shelf transport in the
$\sim$			column from	15	habitat affects recruitment (aka Advection reduces	water column at 80-120 km offshore
			0- 150 m		recruitment while retention enhances recruitment)	(m/s, 4d cumulative)
_			Wator		Growth/Predation hypothesis: growth rate is faster	Degree days in the water column
			vvalei	16	in warm water leading to reduced time vulnerable	calculated from mean water column
O				10	to predators	temperature at 80- 120 km offshore
			0- 150 11			(days, 4d)
			Pottom		Bottom water transport to settlement habitat	Mean long-shore transport at
T	Voor 1		dopths from		affects recruitment (Advection reduces recruitment	bottom depths of 50-150 m and 150
Benthic Juvenile	Apr			17	while retention enhances recruitment)	- 500 m (m/s, 4d cumulative); Two
(Age-0)	Api -			17		depth ranges are considered due to
	ULL		4NU 150 500 m			uncertainty regarding juvenile
			T20-200 III			distributions

Detters	Bottom water transport to settlement habitat	Mean <b>cross-shelf transport</b> at
Bottom	affects recruitment (aka Advection reduces	bottom depths of 50-150 m and 150
depths from	recruitment while retention enhances recruitment)	- 500 m (m/s, 4d cumulative); Two
50-150 m 18		depth ranges are considered due to
AND		uncertainty regarding juvenile
150-500 m		distributions

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Model	Intercept	DDpre	MLDegg	CSTlarv	CSTbjuv.a	LSTbjuv.a	DDegg1	DDlarv	LSTegg	$\mathbb{R}^2$	AICc	ΔAICc	Weight
Model 1	2 770	0.012	0.041	11 015	29 576					0.72	0.29	0	0.22
Model.1	-3.119	0.012	0.041	-41.013	-38.320					0.75	-0.58	0	0.52
Model.2	-4.043	0.012	0.040	-39.178		11.778				0.72	0.82	1.19	0.17
Model.3	-4.272	0.012	0.038	-47.960	-43.688		0.003			0.75	1.3	1.67	0.14
Model.4	-4.459	0.013	0.041	-47.763	-44.831			0.002		0.75	1.38	1.76	0.13
Model.5	-3.632	0.012	0.038	-39.673	-43.993				-1.640	0.74	1.49	1.87	0.12
Model.6	-4.737	0.012	0.036	-46.850		14.671	0.004			0.74	1.62	2.00	0.12

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				Standardized	Std	
Predictor	Coefficient	Bias	SE	Coefficient	Bias	Std SE
Intercept	-3.779	-0.010	0.499	0.006	0.002	0.038
DDpre	0.012	0.000	0.002	0.237	-0.005	0.040
MLDegg	0.041	0.001	0.008	0.212	0.006	0.047
CSTlarv	-41.815	-0.119	7.801	-0.292	0.001	0.055
CSTbjuv.a	-38.526	1.392	15.075	-0.153	0.008	0.062

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Parameter	DDpre	MLDegg	CSTlarv	VIF
DDpre				1.48
MLDegg	0.13			1.91
CSTlarv	0.31	0.54		1.65
CSTbjuv.a	0.28	-0.58	-0.38	2.02
+<	INK			

ROMS predictor	Times included
CSTbjuv.a	29
CSTbjuv.b	4
CSTegg1	0
CSTegg2	0
CSTlarv	66
CSTpjuv	0
DDegg1	3
DDegg1 <sup>2</sup>	0
DDegg2	0
DDlarv	0
DDlarv <sup>2</sup>	0
DDpjuv	0
DDpre	66
LSTbjuv.a	28
LSTbjuv.b	4
LSTegg	0
LSTegg2	0
LSTlarv	0
LSTpjuv	0
MLDegg	66
Tpre.a	0
Tpre.b	0
Total models	66

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ROMS	Number of jackknives	Number of models
CSTbjuv.a	56	88
CSTbjuv.b	21	24
CSTegg1	13	21
CSTegg2	12	17
CSTlarv	81	161
CSTpjuv	29	50
DDegg1	13	23
DDegg2	7	10
DDlarv	3	4
DDpjuv	0	0
DDpre	100	218
DDegg1 <sup>2</sup>	1	3
DDlarv <sup>2</sup>	0	0
LSTbjuv.a	35	42
LSTbjuv.b	31	36
LSTegg	11	19
LSTegg2	18	29
LSTlarv	5	5
LSTpjuv	11	20
MLDegg	81	164
Tpre.a	4	6
Tpre.b	4	6
Total	100	227

ROMS parameter	Time included
CSTbjuv.a	0
CSTbjuv.b	0
CSTegg1	0
CSTegg2	1
CSTlarv	24
CSTpjuv	1
DDegg1	0
DDegg1 <sup>2</sup>	0
DDegg2	0
DDlarv	0
DDlarv <sup>2</sup>	0
DDpjuv	0
DDpre	25
LSTbjuv.a	0
LSTbjuv.b	0
LSTegg	1
LSTegg2	0
LSTlarv	0
LSTpjuv	0
MLDegg	25
Tpre.a	0
Tpre.b	0

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