

ENVIRONMENTAL
CONSERVATION



CAMBRIDGE
UNIVERSITY PRESS

**Using a qualitative model to explore impacts of ecosystem
and anthropogenic drivers upon declining marine survival in
Pacific salmon**

Journal:	<i>Environmental Conservation</i>
Manuscript ID	Draft
Manuscript Type:	Research Paper
Date Submitted by the Author:	n/a
Complete List of Authors:	Sobocinski, Kathryn; NOAA Northwest Fisheries Science Center, Fish Ecology; Long Live the Kings Greene, Correigh; NOAA Northwest Fisheries Science Center, Fish Ecology Schmidt, Michael; Long Live the Kings
Keywords:	Network model, Indicators, Ecosystem, Salmon, Qualitative

SCHOLARONE™
Manuscripts

Using a qualitative model to explore impacts of ecosystem and anthropogenic drivers upon declining marine survival in Pacific salmon

Kathryn L. Sobocinski^{*,1,2}, Correigh M. Greene¹, Michael W. Schmidt²

¹Northwest Fisheries Science Center
National Marine Fisheries Service
National Oceanic and Atmospheric Administration
2725 Montlake Blvd. E
Seattle, WA 98112

²Long Live the Kings
1326 5th Ave. #450
Seattle, WA 98101

*Corresponding Author:
kathryn.sobocinski@noaa.gov
(206) 302-2466

Word Count: 5,901

Running Head: Salish Sea salmon qualitative network model

Keywords: Network model, indicators, ecosystem, salmon, qualitative

Submitted to [*Environmental Conservation*](#)
May 12, 2017

Summary

Coho salmon (*Oncorhynchus kisutch*), Chinook salmon (*O. tshawytscha*), and steelhead (*O. mykiss*) in Puget Sound and the Strait of Georgia have exhibited declines in marine survival over the last 40 years. While the cause of these declines is unknown, multiple factors, acting cumulatively or synergistically, have likely contributed. To evaluate the potential contribution of a broad suite of drivers on salmon survival, we used qualitative network modeling (QNM). QNM is a conceptually based tool that uses networks with specified relationships among the variables. In a simulation framework, linkages are weighted and then the models are subjected to user-specified perturbations. Our network had 33 variables, including: environmental and oceanographic drivers (e.g., temperature, precipitation), primary production variables, food web components from zooplankton to predators, and anthropogenic impacts (e.g., habitat loss, hatcheries). We included salmon traits (survival, abundance, residence time, fitness, and size) as response variables. We invoked perturbations to each node and to suites of drivers and evaluated the responses of these variables. The model showed that anthropogenic impacts resulted in the strongest negative responses in salmon survival and abundance. Additionally, feedbacks through the food web were strong, beginning with primary production, suggesting that several food web variables may be important in mediating effects on salmon survival within the system. With this model, we were able to compare the relative influence of multiple drivers on salmon survival.

22 **Introduction**

23 Problems of complex interactions are common in many fields, including medicine,
24 economics, and ecology (Levins 1974). In ecology, much attention has been given to describing
25 food webs and interactions among species (Paine 1966, May 1974, Pimm et al. 1991, Dunne et
26 al. 2002a). But often, these food webs are nested within larger ecological or social-ecological
27 contexts where exogenous forces influence components of the food web system. External
28 forcings may include physical drivers, anthropogenic impacts, or ecosystem components that
29 are not characterized within the focal network. In social-environmental systems, tools that
30 incorporate ecological properties, abiotic variables, and management actions within the same
31 analytical framework are needed to accurately understand the dynamics of complex systems
32 and evaluate potential management actions (Liu et al. 2007). However, rarely are compatible
33 datasets available for this type of analysis. Here we use a qualitative network model, a
34 conceptually based modeling approach, and a suite of simulations to address questions about
35 the relative impacts of human and natural influences on early marine survival of juvenile
36 salmon.

37 In recent years, attention has turned to marine life-history stages of Pacific salmon
38 (*Oncorhynchus* spp.) in an effort to understand population declines and the subsequent failure
39 to rebound, given myriad conservation and restoration efforts in freshwater streams. In
40 Chinook salmon, Coho salmon, and steelhead (*O. tshawytscha*, *O. kisutch*, and *O. mykiss*,
41 respectively), declines in marine survival have been evidenced within the Salish Sea (Puget
42 Sound, WA, USA and the Strait of Georgia, BC, Canada) that have not been seen in coastal
43 populations (Beamish et al. 2010, Johannessen and McCarter 2010, Zimmerman et al. 2015,

Ruff et al. in press, Kendall et al. in press). These inland water bodies serve as habitats for juvenile salmon as they pass from natal streams to ocean waters during their outmigration. Yet, because of complex anthropogenic changes brought about by population increases and the associated human activity in these waters, it is likely that a number of factors and their cumulative—synergistic or additive—effects are contributing to early-marine mortality. Other salmon species, such as Chum, Pink and Sockeye salmon (*O. keta*, *O. gorbuscha*, and *O. nerka*, respectively), have not experienced similar declines (Debertin et al. 2017), suggesting that life-history characteristics may also contribute to increased mortality for some species in this region. Teasing apart which factors have negatively impacted survival of juvenile salmon in marine waters is of concern to local, regional, and federal governments and other stakeholders (e.g., Salish Sea Marine Survival Project, <http://marinesurvivalproject.com/>) and management actions are being sought to lessen impacts and increase survival during this period.

Correlative studies of salmon abundance and environmental factors have been on-going in the greater region for many years (Pearcy 1988, Beamish et al. 2000, Teo et al. 2009, Burke et al. 2013). Additionally, food web models have been developed for the Strait of Georgia (Priekshot 2008) and Puget Sound (Harvey et al. 2012). These models show primary production as an important driver in the biological system, but also the effect of top predators in creating trophic cascades and influencing food web dynamics in the mid-trophic levels where time-series data are sparse (Harvey et al. 2012). Even with an understanding of the main variables in a given system, measuring abundances of each variable and the flux of energy among them often poses a logistical challenge (Christensen and Walters 2004). These models do not easily incorporate non-fisheries anthropogenic impacts, such as habitat loss or contaminant exposure,

yet we understand that in many systems diverse but cumulative impacts can play a role in species population change. For this reason, conceptually based models, incorporating a broader array of variables, are an important tool in providing an integrated picture of ecological and human drivers of ecosystem change.

One tool for evaluating the relative influence of ecosystem components is Qualitative Network Modeling (QNM, also called Qualitative Network Analysis, QNA, or Loop Analysis, Levins 1974, Puccia and Levins 1985, Raymond et al. 2011, Melbourne-Thomas et al. 2012, Harvey et al. 2016). This approach is advantageous for understanding a system of complex interactions which are not fully specified and when precise measurement is impossible, but when some mechanistic understanding of interactions exists. It allows for the testing of competing hypotheses, given different model structures or the invocation of perturbations to one or more of the model variables. QNM does not explicitly include non-linear direct effects, which occur in and influence social and ecological systems. However, QNM may help to determine the relative impacts of competing hypothesized factors or indicate where empirical work could be focused to improve system understanding (Levins 1974), guiding research efforts.

Researchers have used QNM for evaluating ecosystem response to ocean acidification in shellfish management (Reum et al. 2015) and the impacts of eutrophication and species management within a food web (Carey et al 2013), and for discerning the impact of management actions for species recovery (Harvey et al. 2016) in the Pacific Northwest. QNM is an important conceptual tool for discerning relative impacts of ecosystem components from which more complex, data-driven modeling efforts can stem. Here we apply this technique to

evaluate a suite of potential drivers thought to be contributing to increased early marine mortality in a group of Pacific salmon in the Salish Sea. Using a simulation framework, we invoke perturbations to each model variable and suites of variables based on salmon early marine survival hypotheses, and assess model response related to the salmon species of concern. This work is a foundational step in understanding the impacts of multiple drivers of marine survival declines in Salish Sea salmon.

Methods

We used QNM to address our primary question regarding the relative impacts of various factors on salmon early marine survival. Our analysis had three main steps: 1.) Construct an enhanced conceptual model showing positive, negative, and neutral relationships; 2.) Generate a pool of stable simulated models with random weights applied to each model linkage; and 3.) Invoke one or more perturbations based upon mechanistic understanding of the system and determine the model response. We describe these steps below.

Conceptual Model

To construct our conceptual model of the Salish Sea system, we gathered existing literature and experts on ecosystem components from the project technical team and iteratively developed a working conceptual model. We began by developing a list of over 40 possible variables drawn from hypotheses about the decline of Pacific salmon survival within the system (Salish Sea Marine Survival Project hypotheses, <http://marinesurvivalproject.com/the-project/key-hypotheses/>). The variables included: physical forcings, biological components from primary production to top predators and

competitors, and anthropogenic variables. We grouped model variables into several driver groups: environmental factors, primary production, food web interactions, and anthropogenic impacts (Table 1). While the conceptual model is not exhaustive, it does include many of the drivers identified in our working hypotheses and reflects known interactions within the ecosystem. The inclusion of model variables that are not biomass pools (e.g., temperature, habitat loss) highlights the flexibility of the qualitative modeling approach. Focusing on physical, bottom-up, top-down, and anthropogenic factors fit with the working hypotheses of the Marine Survival Project and enabled exploration of combinations of diverse variables.

The emphasis of the modeling effort was on understanding sources of decreased survival of the focal salmon species (Coho, Chinook, and Steelhead), which have shown an overall declining population trend (Figure 1, left panel, Appendix 1 for details), in addition to a decline in marine survival (Zimmerman et al. 2015, Ruff et al. in press, Kendall et al. in press). Central to our approach was specifying multiple salmon characteristics as modeled network nodes, namely: size, fitness, residence time, abundance, and survival. We used these traits as primary response variables throughout our analysis. While marine survival (herein “survival”) was our principal variable of interest, we included additional traits to evaluate the relative impact to metrics of salmon performance. We included “Other Salmon” as a model variable, representing Pink, Chum, and Sockeye salmon, because the migration timing of all Pacific salmon species means competitive interactions occur. However, the species represented by the “Other Salmon” variable have not experienced the same negative population trends (Irvine and Ruggerone 2016, Figure 1, right panel, and Appendix 1 for details) and are seen as important to the analysis but different from the focal species. While the emphasis was on representing the

most direct impacts to the focal salmon traits, we recognize that many of the model variables (e.g., temperature) could potentially have direct connections to other model nodes; we have included these where interactions were important for understanding implications for the focal salmon variables or where existing literature has shown strong connections.

We defined relationships among variables as positive, negative, or null based upon mechanistic understanding of the Salish Sea system. To implement the simulation modeling, we developed a conceptual digraph using the directed graphing software, Dia (v.0.97.2), to represent the model system and the interactions among variables. This diagram served as the foundation for our qualitative modeling.

Simulated Networks

We used the *QPress* package for Qualitative Network Analysis (Raymond et al. 2012) with custom modifications in R (R Core Team 2016) to interpret the conceptual digraph, construct simulated networks, and perform our analyses. The digraph is interpreted as an interaction matrix, **A**, where each directed pairwise interaction is represented as coefficients a_{ij} . **A** is treated similarly to a community interaction matrix, wherein the rate of change of any given node is a continuous function of all other interacting nodes (Levins 1974; Puccia and Levins 1985). The interacting components (i.e., model nodes) are set up as a series of differential equations:

$$\frac{dx_i}{dt} = f_i(x_1, x_2, \dots, x_n; c_1, c_2, \dots, c_m)$$

where x_i is the density of model component (population) i , the c s are growth parameters, and f_i is a function describing per capita growth-rate of that population (Raymond et al. 2011).

Therefore, the interaction coefficients a_{ij} describe the effect of a change in the level of component j on the level of component i , as defined by the partial derivative of f_i with respect to N_j : $a_{ij} = \partial f_i / \partial x_j$ evaluated at the equilibrium (Levins 1974, Raymond et al. 2011, Melbourne-Thomas et al. 2012).

Given a network model and corresponding interaction matrix, \mathbf{A} , the negative of the inverse community matrix ($-\mathbf{A}^{-1}$) yields estimated changes in the equilibrium abundances of each component x as a function of a sustained (press) perturbation of one or more system components (Puccia and Levins 1985). The *QPress* analysis package provides routines for evaluating the impact of a press perturbation to the system through simulation. For each simulation, a weight (drawn from a random uniform distribution, 0-1) was assigned to each linkage (edge). These weights were positive or negative depending upon the relationship between the two endpoints. If the resulting model with all assigned weights was stable (i.e., converged), the model was accepted. We simulated the network ~120,000 times to result in 10,000 stable simulated networks. We then assessed the proportion of model runs with positive, negative, and neutral responses given changes to particular nodes (see below).

We assessed the sensitivity and robustness of the model. We experimented with changing both the distribution and the variance of the weighting scheme, but did not find large differences in results, so maintained the default weighting for our analyses. We explored the weights of linkages in the balanced models to look for anomalies (methods and results provided in Appendix 2). Additionally, we calculated distance to the Survival node via pathways from each model variable to check for effects of model structure, described network properties such

as connectance and linkage density, and evaluated model behavior with the sequential addition of perturbed nodes and a set of “cumulative effects” of both influential and neutral nodes.

Invoking Perturbations

To test hypotheses regarding marine survival, we developed *a priori* perturbations to invoke upon each model node (Table 2). The direction of the perturbation (increase or decrease) was based upon our understanding of the system, changes that have occurred concomitant with declines in salmon marine survival (since the 1970s), and expected impacts as a result of anthropogenic change (Appendix 3). We employed several cumulative effects scenarios and modified existing software functions to meet our analytical objectives.

First, we perturbed each node individually and observed outcomes to all other model variables. This allowed for a simple comparison of impacts on the focal salmon metrics from each variable and the ability to compare the extent of the impact to that from any other variable. Second, we evaluated the relative effects of different groups of drivers (Table 3). For example, we were interested in food web effects, so we simultaneously decreased forage fish, increased marine mammals, decreased piscivorous fish, and increased gelatinous zooplankton—trends that have been observed in Puget Sound—and observed the impacts to the other model components. For each driver group, we selected four nodes to perturb, thereby standardizing the level of change invoked. By comparing impacts to salmon traits from primary production, food web, environmental, and anthropogenic drivers, we were able to query the relative impacts of each of these groups.

Finally, we developed scenarios based upon observed changes within three regions of Puget Sound to see how well the model reproduced cumulative impacts in terms of response to the focal salmon metrics, especially survival. The three regions were: a.) South Sound, with a known decline in salmon abundance and cumulative impacts including increased gelatinous zooplankton, nutrients, contaminants and hatchery production, and decreased forage fish abundance, b.) Hood Canal, which has had relatively stable salmon abundances, but impacts in oceanography, including increased stratification and temperature and low dissolved oxygen, and c.) Central Basin which has shown a decline in salmon abundance, relatively less than South Sound, but with a different suite of cumulative impacts including habitat loss, contaminant input, and decreased primary production. In reality, causes of declining survival are likely multifaceted, complex, and non-linear, and this modeling exercise allowed us to examine the relative influence of many factors within one modeling framework.

Results

Model

Our final conceptual model had 33 nodes including salmon traits and atmospheric, oceanographic, primary production, food web, and anthropogenic drivers (Table 1, Fig. 2, Appendix 3). There were a total of 148 linkages out of 1089 potential linkages within the model. This gives a network density or connectance (realized linkages/potential linkages) of 0.136 and a linkage density (average # of linkages/node) of 4.48. Connectance has been linked with network stability in ecological networks (Dunne et al. 2002b). The most highly connected nodes were Temperature, Diatoms, and (salmon) Abundance with a total of 11 connections each; the

Other Salmon node was both highly connected (10 linkages) and highly influenced by other variables (10 influencing nodes, Table 1). All nodes were a minimum distance of four nodes or fewer from survival, but the range of feedback linkages varied greatly, from 1 to 10. Each model node included a self-limiting loop to aid in model convergence; the exception was Survival, which was considered the primary variable of interest and was not constrained.

To assess whether the proximity of each model node to Survival influenced the outcomes of our analysis (i.e., are nodes more directly connected to Survival more likely to result in stronger outcomes?), we evaluated the proportion of negative results for Survival with the minimum nodal distance to Survival and found no relationship. Both closely connected nodes (minimum nodal distance of 1) and those more distant (≥ 2 nodes away), resulted in a range of negative responses (<20% to >95%) with respect to Survival. Thus, we do not believe the model structure strongly confounded our results.

Perturbations

The results of the press perturbation to each node showed that anthropogenic impacts resulted in the strongest negative responses in salmon traits, specifically Survival and Abundance (Table 2, Appendix 4). Here we use “strength of response” to refer to the relative proportion of outcomes that were positive or negative given a perturbation—a strong response was when proportionally more simulated models had positive or negative responses for the node of interest, while a weak response resulted when the simulated outcomes were equally positive/negative, or more neutral in outcome.

Individual perturbations showed an increase in CO₂ resulted in a strong positive response in Survival and Abundance; CO₂ positively influences Diatoms in the model, with positive effects cascading through the food web. Conversely, a decrease in Diatoms (primary production) resulted in a strong negative response in Survival and Abundance, as well as negatives outcomes to Zooplankton and Turbidity. A direct perturbation to Zooplankton (decrease) resulted in strong negative responses in Fitness and Size, but neutral results in Survival and Abundance. On the other hand, a direct decrease to Turbidity resulted in a strong positive response in the individual traits, Fitness and Size, but slightly negative responses in Survival and Abundance, population-level traits. This is despite the fact that Turbidity directly and positively affects Survival in the model; this relationship is a result of the association of Turbidity with primary production and the resulting predation dynamics in the model. A decrease in the predators (Piscivorous Fish and Birds) resulted in positive responses in Survival and Abundance. Marine Mammals, also known predators of salmon, but with increasing populations (and thus a positive perturbation), had a neutral response on all salmon response variables. Unexpectedly, increased Harvest had a positive effect on Survival; Harvest has a direct negative effect on Abundance within the model, but the feedback to Survival is mediated by the food web, specifically Forage Fish and Zooplankton, which may moderate the harvest impacts to Survival through reduced competition. This same result indicates that strong feedback mechanisms, like density dependence, were simulated by the model and highlights the importance of complex food web interactions for salmon survival.

The results of the driver group analysis, which evaluated cumulative impacts from one section of the network, showed anthropogenic impacts to have predominantly negative effects

on Survival, Abundance, and Fitness, with over 85% of the simulations having negative responses within these model groups (Fig. 3). For the environmental driver group, most simulation results were positive for Survival and Abundance, and for the food web and primary production groups the results were equivocal, with approximately half the responses negative and the other half positive. The Other Salmon category had a positive response in both the food web and anthropogenic drivers simulations, indicating that conditions that are less favorable within the food web for the focal salmon species may not be detrimental for other species (with diets that tend to be more planktivorous and rearing times within the Salish Sea that are generally shorter). Interestingly, the food web manipulation yielded strong negative responses on the salmon individual traits (Size and Fitness) but more moderate negative response on the population-level traits (Abundance and Survival).

For the regional differences, where we modeled three different regions of Puget Sound with respect to salmon survival, our model replicated the observed trends within these regions, with strong negative responses in focal salmon Survival, Abundance, and Fitness in both South Sound and Central Basin. While the results for Hood Canal were more weakly negative for salmon Survival and Growth than in the other regions, they were positive for Fitness and Size, suggesting that some of the oceanographic changes evidenced in Hood Canal may be beneficial for salmon. Responses of Other Salmon were positive in all three regions, reflecting observed population trends (Fig. 1). Therefore, although our model is a generalization of the processes occurring in the southern portion of the Salish Sea, it does replicate observed trends in the region.

We recognize that many of the perturbations invoked within the model are happening concurrently. To evaluate whether the model would maintain robustness when multiple interacting factors (e.g. “cumulative effects”) were invoked, we sequentially added disturbances to the model and evaluated the outcomes. When influential individual drivers were included (from Table 2), the results were strongly negative for Survival. We compared 10 of the strongest responding nodes with 10 that showed neutral influence on Survival and compared the results (Fig. 4). We observed that the influential individual nodes resulted in stronger negative impacts to Survival, while the response was neutral for the weaker suite of disturbances.

Discussion

Our model showed that a wide variety of the included drivers showed negative responses on early marine survival of Coho salmon, Chinook salmon and steelhead. Similarly, the model showed that the impacts to the Other Salmon variable (representing Pink, Chum, and Sockeye) were neutral or positive. With only five drivers (Precipitation, River Flow, Microbial Detritivores, Zooplankton, and Ichthyoplankton) negatively influencing the Other Salmon node in our simulations, the model structure seemed to capture the reduced impacts to these species within the Salish Sea (Fig. 1, Debertin et al. 2017). The combination of drivers having negative effects on attributes of salmon fitness and survival suggest that a single sector of the network is insufficient for explaining increased marine mortality and that feedbacks and complex interactions may both exacerbate and mediate the effects of individual drivers.

Anthropogenic factors induced negative responses in salmon traits, especially Survival and Abundance. The factors are both direct (e.g., Contaminants and Disease) and mediated by the food web (e.g., Hatcheries, with increased production leading to competitive interactions, and Habitat Loss, which has a negative effect on salmon Residency and Fitness, but also on Forage Fish, because nearshore habitat is critical to forage fish spawning). There are likely indirect connections that were unaccounted for in our model that may make these impacts even stronger in the real world. The food web components individually did not yield strong responses in focal salmon traits, but considering that many of these linkages are indirectly tied to these traits, the negative outcomes, even marginal, should be noted. Additionally, many of the feedbacks present in the model were through the food web. In the driver group analysis, the negative responses from food web changes to the individual traits of Size and Fitness is notable. Our model does not have a temporal component, but negative impacts to individual traits would likely manifest in survival and population declines over time.

The conceptual model exhibited particularly strong sensitivity to changes in key linkages: the effects of 1) fitness upon size and vice-versa and 2) survival upon abundance. These results suggest that processes influencing these factors will have strong influence upon marine survival and point to the importance of monitoring these pathways. Size and fitness are relatively easy-to-monitor characteristics measured in standard salmon sampling programs. The measurement of size, combined with techniques that can measure growth and fitness, including the use of chemical-based indicators of fitness, such as fatty acid biomarker characterization (Hook et al. 2014) or hormone markers (Beckman 2011) would provide some indication of how these attributes are changing over time. However, understanding mechanisms for changes in size and

318 fitness is more complex. As our model and existing literature suggest, several factors contribute
319 to changes in size over time, with food web alteration being one of the likely mechanisms.
320 However, these changes are not yet fully understood in the Salish Sea. Our model allowed for
321 comparison of multiple factors and showed that a decrease in primary production had the
322 strongest negative impact to salmon survival. Additional research on the impacts of changing
323 primary production in this system and the links to salmon condition would aid in further teasing
324 apart this relationship.

325 Our response metrics focused on model runs that converged on an equilibrium (i.e., only
326 balanced models were used in the perturbation scenarios). The number of model runs needed
327 (~122,000) to get a subset of converged models (10,000) suggests that most models did not
328 converge to a stable solution. The “real” Salish Sea is likely represented by one of the many
329 possible combinations and may in fact be unstable, not in a “converged” state, as our model
330 assumed. Such unstable states are predicted outcomes when ecosystems surpass tipping
331 points (Carpenter and Brock 2006, Samhouri et al. in press). Alternately, the Salish Sea
332 ecosystem may have reached a new equilibrium: analysis of marine survival trends indicate a
333 steep decline in the 1970s and 80s, leveling off at a low level that has persisted to the present
334 (Zimmerman et al. 2015). Hence, the simulation framework with multiple sets of initial
335 conditions may have allowed us to detect endpoints that include a new equilibrium for marine
336 survival of salmon.

337 The conceptual model underlying the analysis represents a complex set of feedbacks.
338 Additional interconnections that we did not represent would tend to stabilize the system even
339 more (Dunne et al. 2002a, Ives and Carpenter 2007). And while nonlinear properties are

fundamental to ecological systems, they are nearly always influenced by feedbacks, an essential component built into our model structure (DeAngelis and Waterhouse 1987, Scheffer et al. 2001). The lack of explicit spatial or temporal components within the model limits our ability to make predictions beyond a static snapshot. However, through our comparison of three sub-regions within our system, we were able to compare different starting conditions and gauge model response. While additional complexity in the temporal component would allow for detecting evidence of change over time, the lack of comprehensive empirical data across all ecosystem components currently limits the tractability of such a modeling approach. This model is forming the conceptual basis for complex end-to-end modeling efforts and multivariable time-series analyses related to this project, which will be undertaken with available data.

Our results suggest that teasing out the causes of declines in marine survival will be challenging, multi-faceted, and will involve both understood and unknown feedbacks. Multiple singular factors led to declines in most of the simulations, and suites of ecosystem components had strong effects on marine survival and other salmon population attributes. Nevertheless, our ability to distinguish causal factors will likely be improved by tracking multiple ecosystem indicators, especially those influencing salmon size and fitness. As with any model, ours is a simplified version of the ecosystem; however, our approach resulted in a complex representation of declining populations in an ecosystem context that served as a useful tool for identifying relative influences of numerous hypothesized drivers of marine mortality. Through this examination, we have identified some sectors of the ecosystem, such as the food web and anthropogenic impacts, which warrant further examination.

362 [Supplementary material](#)

363 For supplementary material accompanying this paper, visit:

364 <http://www.journals.cambridge.org/ENC>

365

366

Proof for Review

Acknowledgements

We thank members of the Salish Sea Marine Survival Project technical team for input to the conceptual model. B. Raymond provided assistance with modifying R code. I. Kaplan, C. Harvey, I. Perry, I. Pearsall, T. Curran, and I. Kemp provided insight on early versions of the modeling effort and results. A. Dufault and N. Kendall (Washington Department of Fish and Wildlife) assisted with data compilation for Fig. 1. The manuscript benefited from comments provided by I. Kaplan, C. Harvey, and R. Zabel (NOAA-Northwest Fisheries Science Center). Reviewers...

Financial Support

This work was paid for by the Pacific Salmon Commission's Southern Endowment Fund via Long Live the Kings. This is publication number X from the Salish Sea Marine Survival Project, an international research collaboration designed to determine the primary factors affecting the survival of juvenile salmon and steelhead in the Salish Sea.

Conflict of Interest

None

Human and/or Animal Experimentation

None

Literature Cited

Beamish, R.J., D.J. Noakes, G.A. McFarlane, W. Pinnix, R. Sweeting, and J. King. 2000. Trends in coho marine survival in relation to the regime concept. *Fisheries Oceanography* 9:114–119.

Beamish, R.J., R.M. Sweeting, K.L. Lange, D.J. Noakes, D. Preikshot, and C.M. Neville. 2010. Early marine survival of coho salmon in the Strait of Georgia declines to very low levels. *Marine and Coastal Fisheries: Dynamics, Management, and Ecosystem Science* 2: 424-439.

Beckman, B.R. 2011. Perspectives on concordant and discordant relations between insulin-like growth factor 1 (IGF1) and growth in fishes. *General and Comparative Endocrinology*, 170(2):233-252.

Burke, B.J., W.T. Peterson, B.R. Beckman, C. Morgan, E.A. Daly, et al. 2013. Multivariate models of adult Pacific salmon returns. *PLoS ONE* 8(1): e54134.doi:10.1371/journal.pone.0054134

Carey, M.P., P.S. Levin, H. Townsend, T.J. Minello, G.R. Sutton, T.B. Francis, C.J. Harvey, J.E. Toft, K.K. Arkema, J.L. Burke, C. Kim, A.D. Guerry, M. Plummer, G. Spiridonov, and M. Ruckelshaus. 2014. Characterizing coastal foodwebs with qualitative links to bridge the gap between the theory and the practice of ecosystem-based management. *ICES J Mar Sci* 71: 713–724.

Carpenter, S.R., and W.A. Brock. 2006. Rising variance: a leading indicator of ecological transition. *Ecology Letters*, 9(3), 311-318.

Christensen, V. and C.J. Walters. 2004. Ecopath with Ecosim: methods, capabilities and limitations. *Ecological Modelling* 172:109-139.

DeAngelis, D.L. and J.C. Waterhouse. 1987. Equilibrium and nonequilibrium concepts in ecological models. *Ecological Monographs* 57: 1:21.

Debertin, A.J., J.R. Irvine, C.A. Holt, G. Oka, and M. Trudel. 2017. Marine growth patterns of southern British Columbia chum salmon explained by interactions between density-dependent competition and changing climate. *Can. J. of Fish. and Aqua. Sci.* 10.1139/cjfas-2016-0265

Dunne, J. A., R.J. Williams, and N.D. Martinez. 2002a. Network structure and biodiversity loss in food webs: robustness increases with connectance. *Ecol. Lett.* 5: 558–567.

Dunne, J.A., R. J. Williams, & N.D. Martinez. 2002b. Food-web structure and network theory: the role of connectance and size. *Proceedings of the National Academy of Sciences of the United States of America*. 99(20): 12917

- Greene, C.M., L. Kuehne, C.A. Rice, K.L. Fresh, D. Penttila. 2015. Forty years of change in forage fish and jellyfish abundance across greater Puget Sound, Washington (USA): anthropogenic and climate associations. *Marine Ecology Progress Series* 525: 153-170.
- Harvey, C.J., G.D. Williams, and P.S. Levin. 2012. Food web structure and trophic control in central Puget Sound. *Estuaries and Coasts* 35: 821. doi:10.1007/s12237-012-9483-1
- Harvey, C.J., J.C.P. Reum, M.R. Poe, G.D. Williams, and S.J. Kim. 2016. Using conceptual models and qualitative network models to advance integrative assessments of marine ecosystems. *Coastal Management*. 44: 486-503, DOI: 10.1080/08920753.2016.1208881
- Hoekstra, J. M., K. K. Bartz, M. H. Ruckelshaus, J. M. Moslemi, and T. K. Harms. 2007. Quantitative threat analysis for management of an imperiled species: Chinook salmon (*Oncorhynchus tshawytscha*). *Ecological Applications* 17:2061–2073.
- Hook, S., E. Gallagher, and G. Batley. 2014. The Role of Biomarkers in the Assessment of Aquatic Ecosystem Health. *Integrated Environmental Assessment and Monitoring*. 10(3):327-41.
- Holmes, E.E. 2001. Estimating risks in declining populations with poor data. *Proceedings of the National Academy of Sciences of the United States of America*. 98: 5072-5077.
- Irvine, J.R. and G.T. Ruggerone. 2016. Provisional estimates of numbers and biomass for natural-origin and hatchery-origin pink, chum, and sockeye salmon in the North Pacific, 1952-2015. NPAFC Doc. 1660. 45 pp. Fisheries and Oceans Canada, Pacific Biological Station and Natural Resources Consultants, Inc. (Available at <http://www.npafc.org>).
- Ives, A.R. and S.R. Carpenter. 2007. Stability and Diversity of Ecosystems. *Nature*. 317: 58-62.
- Johannessen, S.C. and B. McCarter. 2010. Ecosystem Status and Trends Report for the Strait of Georgia Ecozone. *DFO Can. Sci. Advis. Sec. Res. Doc.* 2010/010. vi + 45 p.
- Kendall, N.W., G.W. Marston, and M.M. Klungle. In Press. Declining patterns of Pacific Northwest steelhead trout (*Oncorhynchus mykiss*) adult abundance and smolt survival in the ocean. *Can. J. of Fish. and Aqua. Sci.*
- Levins, R. 1974. The qualitative analysis of partially specified systems. *Annals of the New York Academy of Sciences* 231: 123–138.
- Liu, J., T. Dietz, S.R. Carpenter, C. Folke, M. Alberti, C.L. Redman, S.H. Schneider, E. Ostrom, A.N. Pell, J. Lubchenco, W.W. Taylor, Z. Ouyang, P. Deadman, T. Kratz, and W. Provencher. 2007. Coupled human and natural systems. *AMBIO*. 236: 639–649.
- Mauger, G.S., J.H. Casola, H.A. Morgan, R.L. Strauch, B. Jones, B. Curry, T.M. Busch Isaksen, L. Whitely Binder, M.B. Krosby, and A.K. Snover, 2015. *State of Knowledge: Climate Change in*

- 467 *Puget Sound*. Report prepared for the Puget Sound Partnership and the National Oceanic and
 468 Atmospheric Administration. Climate Impacts Group, University of Washington, Seattle.
 469 doi:10.7915/CIG93777D
 470
- 471 May, R. M. 1974. Stability and Complexity in Model Ecosystems. *Monographs in Population*
 472 *Biology* 6, Princeton University Press, Princeton, New Jersey
 473
- 474 Melbourne-Thomas, J., S. Wotherspoon, B. Raymond, and A. Constable. 2012. Comprehensive
 475 evaluation of model uncertainty in qualitative network analyses. *Ecological Monographs*, 82(4):
 476 505–519.
 477
- 478 Ogden, A.D., J.R. Irvine, K.K. English, S. Grant, K.D. Hyatt, L. Godbout, and C.A. Holt. 2015.
 479 Productivity (recruits-per-spawner) data for sockeye, pink, and chum salmon from British
 480 Columbia. *Can. Tech. Rep. Fish. Aquat. Sci.* 3130: vi + 57 p.
 481
- 482 O'Neill, S.M., and West, J.E. 2009. Marine distribution, life history traits, and the accumulation
 483 of polychlorinated biphenyls in Chinook salmon from Puget Sound, Washington. *Transactions of*
 484 *the American Fisheries Society* 138(3): 616-632.
 485
- 486 Paine, R.T. 1966. Food web complexity and species diversity. *American Naturalist* 100: 65-76.
 487
- 488 Pearcy, W. G. 1988. Factors affecting survival of Coho Salmon off Oregon and Washington.
 489 Pages 67–73 in W. J. McNeil, editor. *Salmon production, management, and allocation*. Oregon
 490 State University Press, Corvallis.
 491
- 492 Pimm, S. L., J.H. Lawton and J.E. Cohen. 1991. Food web patterns and their consequences.
 493 *Nature* 350, 669–674.
 494
- 495 Preikshot, D.B. 2008. *Public Summary—Computer Modelling of Marine Ecosystems: Applications*
 496 *to Pacific Salmon Management and Research*. Vancouver, BC: Pacific Fisheries Resource
 497 Conservation Council.
 498
- 499 Puccia, C. J. and R. Levins. 1985. *Qualitative modeling of complex systems: An introduction to*
 500 *loop analysis and time averaging*. Cambridge, MA: Harvard University Press.
 501
- 502 Puget Sound Ecosystem Monitoring Program (PSEMP) Marine Waters Workgroup. 2016. *Puget*
 503 *Sound Marine Waters: 2015 Overview*. S. K. Moore, R. Wold, K. Stark, J. Bos, P. Williams, K.
 504 Dzinbal, C. Krembs and J. Newton (Eds). URL:
 505 www.psp.wa.gov/PSEMP/PSmarinewatersoverview.php.
 506
- 507 R Core Team. 2016. R: A language and environment for statistical computing. R Foundation for
 508 Statistical Computing, Vienna. www.r-project.org
 509

- Raymond, B., J. McInnes, J.M. Dambacher, S. Way, and D.M. Bergstrom. 2011 Qualitative modelling of invasive species eradication on subantarctic Macquarie Island. *Journal of Applied Ecology* 48: 181–191.
- Raymond, B, J. Melbourne-Thomas, S. Wotherspoon. 2012. QPress: Qualitative Network Analysis.
- Reum, J.C.P., B.E. Ferriss, P.S McDonald, D.M. Farrell, C.J. Harvey, T. Klinger, and P.S. Levin. 2015. Evaluating community impacts of ocean acidification using qualitative network models. *Marine Ecology Progress Series* 536: 11-24.
- Roberts, M., T. Mohamedali, and B. Sackmann, T. Khangaonkar, and W. Long. 2014. *Puget Sound and the Straits Dissolved Oxygen Assessment: Impacts of Current and Future Nitrogen Sources and Climate Change Through 2070*. Washington Department of Ecology, Publication No. 14-03-007. Olympia, WA 151 pp.
- Ruff, C.P., J.H. Anderson, I.M. Kemp, N.W. Kendall, P. Mchugh, A. Velez-Espino, C.M. Greene, M. Trudel, C.A. Holt, K.E. Ryding, and K. Rawson. In press. Salish Sea Chinook salmon exhibit weaker coherence in early marine survival trends than coastal populations. *Fisheries Oceanography*.
- Samhuri, J. F., K. S. Andrews, G. Fay, C. J. Harvey, E. L. Hazen, S. Hennessey, K. K. Holsman, M. E. Hunsicker, S. I. Large, K. Marshall, A. C. Stier, J. Tam, and S. Zador. In Press. Defining ecosystem thresholds for human activities and environmental pressures in the California Current. *Ecosphere*.
- Scheffer, M., S. Carpenter, J.A. Foley, C. Folke and B. Walker. 2001. Catastrophic shifts in ecosystems. *Nature* 413: 591-596.
- Teo, S. L. H., L. W. Botsford, and A. Hastings. 2009. Spatio-temporal covariability in Coho Salmon (*Oncorhynchus kisutch*) survival, from California to southwest Alaska. *Deep Sea Research Part II: Topical Studies in Oceanography* 56:2570–2578.
- Zimmerman, M., J. R. Irvine, M. O'Neill, J. H. Anderson, C. M. Greene, J. Weinheimer, M. Trudel, K. Rawson. 2015. Spatial and temporal patterns in smolt survival of wild and hatchery coho salmon in the Salish Sea. *Marine and Coastal Fisheries* 7(1):116-134.

Table and Figure Captions

Tables

Table 1. Descriptions of model nodes (variables) in the Salish Sea qualitative network model by driver group. Also shown are: connectivity among nodes, the number of influencing nodes, and the distance to the Survival node.

Table 2. Model variables in the Salish Sea qualitative network model with the direction of the press perturbation invoked in the simulations. The response of perturbations to each individual model node on the focal salmon traits (Survival, Abundance, Fitness, Size, Residence) and the Other Salmon model group are indicated by the colored boxes. The key to direction and strength of responses of model simulations is in the lower right portion of the table.

Table 3. Perturbations and responses by driver group. Total nodal distance is the sum of the nodal distances of each node to the Survival node.

Table 4. Salish Sea subbasin analysis with perturbations invoked and outcomes.

Figures

Figure 1. Salmon population trends within Puget Sound (left most subbasins), the Strait of Georgia (center subbasins) and the Pacific Coast (right) for species with generally decreasing trends (Chinook, Coho, and Steelhead, left) and increasing or stable populations trends (Chum, Pink, and Sockeye, right). Two abnormally high Pink salmon runs were omitted from the right panel for ease of presentation—these runs had trends of 1.05 and 0.43 and were both from Puget Sound subbasins. Data are from Washington Department of Fish and Wildlife, Pacific States Marine Fisheries Commission, Ogden et al. 2015, and Zimmerman et al. 2015.

Figure 2. Conceptual model of the Salish Sea related to salmon survival. Model variables (shapes) represent food web components (ovals), physical drivers (rectangles), anthropogenic impacts (diamonds) and salmon traits of interest (triangles). Survival is shown as a hexagon and was the primary variable of interest.

Figure 3. Results for the primary response variables for each of the four groups of drivers. Perturbations were according to Table 3 and in all four figures, bars represent the number of runs resulting in negative (dark bars) or positive (light bars) response in the response variable.

Figure 4. Comparison of multiple ($n=10$) strongly influential (top) and neutral (bottom) variables on salmon survival. The influential variables (resulting from the analysis presented in Table 2) were: CO₂, Contaminants, Diatoms, Disease, Forage Fish, Gelatinous Zooplankton, Habitat Loss, Hatcheries, Other Salmon, Temperature. The neutral variables were: Global Warming, Ichthyoplankton, Marine Mammals, Microbial Detritivores, Nutrients, Piscivorous Birds, Precipitation, Stratification, Sunlight, Zooplankton. The variables in each set were perturbed simultaneously to simulate cumulative impacts and assess model response. Dark bars show negative impacts to the response variable and light bars show positive responses.

Supplementary Material—Salish Sea Salmon Qualitative Network

Model

Appendix 1. Methods and Data Supporting Figure 1.

Methods

To arrive at population trends, we compiled available adult run sizes for salmon in rivers throughout the Salish Sea. Data were from Pacific Fishery Management Council Stock Assessment and Fishery Evaluation (SAFE) Documents (<http://www.pcouncil.org/salmon/stock-assessment-and-fishery-evaluation-safe-documents/review-of-2015-ocean-salmon-fisheries/>), Pacific Salmon Commission Technical Committee Reports (<http://www.psc.org/publications/technical-reports/technical-committee-reports/chinook/>), Washington Department of Fish and Wildlife, Ogden et al. 2015, and Zimmerman et al. 2015 (Table A1.1). We used the longest time series available (typically from 1970s to >2010) and used the methods of Holmes (2001) to estimate a population trend and confidence interval for each run. If total run size (TRS, a total of catch and escapement or number of spawners) was not available, we used escapement/spawner (E) data.

Table A1.1 Salmon runs used in the calculation of population trends. TRS=Total Run Size, E=Escapement.

Run	Species	Data	Years	Data Source	Subbasin	Population Trend	95% Confidence Interval
Lake Washington	Chinook	TRS	1975-2014	PSC Joint Tech. Comm. 2016	Central Puget Sound	0.004	0.009
Green	Chinook	TRS	1975-2014	PSC Joint Tech. Comm. 2016	Central Puget Sound	-0.027	0.006
Snohomish	Chinook	TRS	1975-2014	PSC Joint Tech. Comm. 2016	Whidbey	-0.035	0.004
Hood Canal	Chinook	TRS	1981-2014	PFMC 2016	Hood Canal	0.035	0.010
Juan de Fuca	Chinook	TRS	1981-2014	PFMC 2016	Juan de Fuca	-0.006	0.006
South Puget Sound	Chinook	TRS	1981-2014	PFMC 2016	South Sound	0.013	0.005
Nooksack	Chinook	TRS	1981-2014	PFMC 2016	San Juan and Gulf Islands	-0.046	0.007
Skagit Spring	Chinook	TRS	1975-2014	PSC Joint Tech. Comm. 2016	Whidbey	0.014	0.006
Skagit Summer/fall	Chinook	TRS	1975-2014	PSC Joint Tech. Comm. 2016	Whidbey	-0.026	0.005
Stillaguamish	Chinook	TRS	1975-2014	PSC Joint Tech. Comm. 2016	Whidbey	-0.030	0.011

Willapa	Chinook	TRS	1976-2014	PFMC 2016	Pacific Coast	0.021	0.007
Grays Harbor Spring/Summer	Chinook	TRS	1976-2014	PFMC 2016	Pacific Coast	0.018	0.005
Grays Harbor Fall	Chinook	TRS	1975-2014	PSC Joint Tech. Comm. 2016	Pacific Coast	0.013	0.005
Queets Spr/Summer	Chinook	TRS	1976-2015	PSC Joint Tech. Comm. 2016	Pacific Coast	-0.025	0.007
Queets fall	Chinook	TRS	1976-2014	PSC Joint Tech. Comm. 2016	Pacific Coast	0.002	0.004
Hoh Spring/Summer	Chinook	TRS	1976-2015	PSC Joint Tech. Comm. 2016	Pacific Coast	-0.019	0.006
Hoh fall	Chinook	TRS	1976-2015	PSC Joint Tech. Comm. 2016	Pacific Coast	-0.002	0.004
Quillayute spring/summer	Chinook	TRS	1976-2015	PSC Joint Tech. Comm. 2016	Pacific Coast	-0.029	0.005
Quillayute fall	Chinook	TRS	1976-2015	PFMC 2016	Pacific Coast	-0.005	0.005
Hoko fall term	Chinook	TRS	1986-2015	PSC Joint Tech. Comm. 2016	Pacific Coast	0.021	0.009
Fraser Spring/summer	Chinook	TRS	1975-2015	PSC Joint Tech. Comm. 2016	Central Strait of Georgia	0.017	0.002
Harrison	Chinook	TRS	1984-2015	PSC Joint Tech. Comm. 2016	Central Strait of Georgia	0.050	0.004
Lake Shuswap	Chinook	E	1975-2015	PSC Joint Tech. Comm. 2016	Central Strait of Georgia	0.026	0.006
Nanaimo	Chinook	E	1975-2015	PSC Joint Tech. Comm. 2016	South Strait of Georgia, San Juan and Gulf Islands	-0.018	0.007
Cowichan	Chinook	E	1975-2015	PSC Joint Tech. Comm. 2016	South Strait of Georgia, San Juan and Gulf Islands	-0.006	0.007
Nimpkish	Chinook	E	1975-2015	PSC Joint Tech. Comm. 2016	Johnstone Strait	-0.052	0.015
Klinaklini	Chinook	E	1975-2015	PSC Joint Tech. Comm. 2016	Johnstone Strait	0.030	0.009
Kakweiken	Chinook	E	1975-2015	PSC Joint Tech. Comm. 2016	Johnstone Strait	-0.043	0.011
Kingcome	Chinook	E	1975-2015	PSC Joint Tech. Comm. 2016	Johnstone Strait	-0.064	0.018
Wakeman	Chinook	E	1975-2015	PSC Joint Tech. Comm. 2016	Johnstone Strait	-0.151	0.028
Marble	Chinook	E	1975-2015	PSC Joint Tech. Comm. 2016	Pacific Coast	0.013	0.005
Burman	Chinook	E	1975-2015	PSC Joint Tech. Comm. 2016	Pacific Coast	0.061	0.012
Tahsis	Chinook	E	1975-2015	PSC Joint Tech. Comm. 2016	Pacific Coast	0.024	0.014
Artlish	Chinook	E	1975-2015	PSC Joint Tech. Comm. 2016	Pacific Coast	0.029	0.016
Kaouk	Chinook	E	1975-2015	PSC Joint Tech. Comm. 2016	Pacific Coast	0.041	0.026
Tahsish	Chinook	E	1975-2015	PSC Joint Tech. Comm. 2016	Pacific Coast	0.004	0.011
Cedar River Winter	Steelhead	TRS	1987-2013	N. Kendall, WDFW pers. comm	Central Puget Sound	-0.220	0.023
Green River Winter	Steelhead	TRS	1978-2013	N. Kendall, WDFW pers. comm	Central Puget Sound	-0.053	0.004
Puyallup/Carbon Winter	Steelhead	TRS	1983-2013	N. Kendall, WDFW pers. comm comm	Central Puget Sound	-0.077	0.006
White River (Puyallup) Winter	Steelhead	TRS	1986-2013	N. Kendall, WDFW pers. comm	Central Puget Sound	-0.042	0.008
Englishman River winter	Steelhead	E	1982-2015	K. Pellett, pers. comm	Central Strait of Georgia	-0.063	0.011

East Hood Canal Winter	Steelhead	TRS	1989-2013	N. Kendall, WDFW pers. comm	Hood Canal	0.088	0.016
Skokomish Winter	Steelhead	TRS	1985-2006	N. Kendall, WDFW pers. comm	Hood Canal	-0.020	0.013
South Hood Canal Winter	Steelhead	TRS	1988-2013	N. Kendall, WDFW pers. comm	Hood Canal	-0.004	0.009
West Hood Canal Winter	Steelhead	TRS	2003-2013	N. Kendall, WDFW pers. comm	Hood Canal	0.015	0.021
Tsitika River summer	Steelhead	E	1976-2015	K. Pellett, pers. comm	Johnstone Strait	0.008	0.007
Salmon River winter	Steelhead	TRS	1981-2015	N. Kendall, WDFW pers. comm	Johnstone Strait	0.021	0.036
Sequim and Discovery Bays Tributaries Winter	Steelhead	TRS	1987-2013	N. Kendall, WDFW pers. comm	Juan de Fuca	-0.101	0.100
Strait of Juan de Fuca Independent Tributaries Winter	Steelhead	TRS	1991-2009	N. Kendall, WDFW pers. comm	Juan de Fuca	-0.028	0.022
Glendale winter	Steelhead	E	1992-2014	K. Pellett, pers. comm	Pacific Coast	-0.178	0.085
Ahnuhati winter	Steelhead	E	1992-2014	K. Pellett, pers. comm	Pacific Coast	-0.186	0.065
Kakweiken winter	Steelhead	E	1992-2014	K. Pellett, pers. comm	Pacific Coast	-0.206	0.063
Atwaykellesse winter	Steelhead	E	1992-2014	K. Pellett, pers. comm	Pacific Coast	-0.116	0.062
Wahpeeto winter	Steelhead	E	1992-2014	K. Pellett, pers. comm	Pacific Coast	-0.201	0.066
Heber River summer	Steelhead	E	1975-2015	K. Pellett, pers. comm	Pacific Coast	-0.012	0.004
Gordon River summer	Steelhead	E	1998-2015	K. Pellett, pers. comm	Pacific Coast	0.030	0.042
Nisqually Winter	Steelhead	TRS	1980-2013	N. Kendall, WDFW pers. comm	South Sound	-0.107	0.016
Samish Winter	Steelhead	TRS	1979-2013	N. Kendall, WDFW pers. comm	South Strait of Georgia	0.039	0.011
Pilchuck Winter	Steelhead	E	1981-2014	N. Kendall, WDFW pers. comm	Whidbey	-0.020	0.009
Skagit River Summer and Winter	Steelhead	TRS	1978-2013	N. Kendall, WDFW pers. comm	Whidbey	-0.034	0.007
Snohomish System Winter	Steelhead	TRS	1987-2013	N. Kendall, WDFW pers. comm	Whidbey	-0.048	0.012
Snohomish/Skykomish Winter	Steelhead	TRS	1987-2013	N. Kendall, WDFW pers. comm	Whidbey	-0.061	0.010
Snoqualmie Winter	Steelhead	TRS	1987-2013	N. Kendall, WDFW pers. comm	Whidbey	-0.056	0.008
Stillaguamish Winter	Steelhead	TRS	1987-2013	N. Kendall, WDFW pers. comm	Whidbey	-0.075	0.022
Green	Coho	TRS	1973-2010	Zimmerman et al. 2015	Central Puget Sound	0.039	0.014
Puyallup	Coho	TRS	1974-2010	Zimmerman et al. 2015	Central Puget Sound	-0.038	0.008
Big Qualicum	Coho	TRS	1973-2010	Zimmerman et al. 2015	Central Strait of Georgia	-0.143	0.011
Inch	Coho	TRS	1984-2010	Zimmerman et al. 2015	Central Strait of Georgia	0.103	0.023
Big Beef Creek	Coho	TRS	1977-2010	Zimmerman et al. 2015	Hood Canal	-0.025	0.009
Quilcene	Coho	TRS	1979-2010	Zimmerman et al. 2015	Hood Canal	0.039	0.014
Skokomish	Coho	TRS	1973-2010	Zimmerman et al. 2015	Hood Canal	0.013	0.008
Straits	Coho	TRS	1985-2010	Zimmerman et al. 2015	Juan de Fuca	-0.034	0.012
Black	Coho	TRS	1978-2010	Zimmerman et al. 2015	North Strait of Georgia	-0.052	0.021
Puntledge	Coho	TRS	1974-2010	Zimmerman et al. 2015	North Strait of Georgia	-0.151	0.020

Quinsam	Coho	TRS	1978-2004	Zimmerman et al. 2015	North Strait of Georgia	-0.104	0.010
Bingham Creek	Coho	TRS	1976-2010	Zimmerman et al. 2015	Pacific Coast	0.017	0.014
Cowlitz	Coho	TRS	1982-2010	Zimmerman et al. 2015	Pacific Coast	-0.126	0.030
Grays	Coho	TRS	1982-2010	Zimmerman et al. 2015	Pacific Coast	-0.036	0.009
Quinault	Coho	TRS	1977-2010	Zimmerman et al. 2015	Pacific Coast	0.079	0.015
Robertson	Coho	TRS	1975-2010	Zimmerman et al. 2015	Pacific Coast	-0.014	0.013
Satsop	Coho	TRS	1973-2010	Zimmerman et al. 2015	Pacific Coast	0.036	0.010
Washougal	Coho	TRS	1976-2010	Zimmerman et al. 2015	Pacific Coast	-0.102	0.023
Deschutes	Coho	TRS	1977-2008	Zimmerman et al. 2015	South Sound	-0.048	0.041
Chilliwack	Coho	TRS	1982-2004	Zimmerman et al. 2015	South Strait of Georgia	-0.066	0.060
Nooksack	Coho	TRS	1976-2009	Zimmerman et al. 2015	South Strait of Georgia	-0.071	0.028
Skagit	Coho	TRS	1991-2010	Zimmerman et al. 2015	Whidbey	-0.049	0.014
Stilliguamish	Coho	TRS	1981-2014	Zimmerman et al. 2015	Whidbey	-0.013	0.013
Snohomish	Coho	TRS	1981-2014	Zimmerman et al. 2015	Whidbey	-0.026	0.010
Baker	Coho	TRS	1983-2010	Zimmerman et al. 2015	Whidbey	-0.048	0.018
Skykomish	Coho	TRS	1978-2010	Zimmerman et al. 2015	Whidbey	0.012	0.010
Tulalip Bay	Coho	TRS	1974-2010	Zimmerman et al. 2015	Whidbey	0.011	0.017
Green River (Area 10A)	Pink	TRS	2001-2013	A. Dufault, WDFW, pers. comm.	Central Puget Sound	1.053	0.364
Puyallup River (Area 11A)	Pink	TRS	1959-2013	A. Dufault, WDFW, pers. comm.	Central Puget Sound	0.133	0.038
Chambers	Pink	TRS	1959-2013	A. Dufault, WDFW, pers. comm.	Central Puget Sound	0.096	0.206
Nisqually R. Drainage	Pink	TRS	1959-2013	A. Dufault, WDFW, pers. comm.	Central Puget Sound	0.167	0.088
Area 13A streams	Pink	TRS	1959-2013	A. Dufault, WDFW, pers. comm.	Central Puget Sound	-0.029	0.054
Area 13B streams	Pink	TRS	1959-2013	A. Dufault, WDFW, pers. comm.	Central Puget Sound	0.036	0.035
Port Gamble	Pink	TRS	1959-2013	A. Dufault, WDFW, pers. comm.	Hood Canal	0.425	0.593
Dosewallips	Pink	TRS	1959-2013	A. Dufault, WDFW, pers. comm.	Hood Canal	-0.095	0.028
Duckabush	Pink	TRS	1959-2013	A. Dufault, WDFW, pers. comm.	Hood Canal	-0.025	0.054
Hamma	Pink	TRS	1959-2013	A. Dufault, WDFW, pers. comm.	Hood Canal	0.091	0.052
Hoodspout Hatchery	Pink	TRS	1959-2013	A. Dufault, WDFW, pers. comm.	Hood Canal	0.063	0.024
Misc 12C	Pink	TRS	1959-2013	A. Dufault, WDFW, pers. comm.	Hood Canal	-0.002	0.041
Skokomish R. Drainage	Pink	TRS	1959-2013	A. Dufault, WDFW, pers. comm.	Hood Canal	0.093	0.055
Dungeness River	Pink	TRS	1959-2013	A. Dufault, WDFW, pers. comm.	Juan de Fuca	-0.009	0.053
Elwha River	Pink	TRS	1959-2013	A. Dufault, WDFW, pers. comm.	Juan de Fuca	-0.182	0.093
Nooksack River	Pink	TRS	1959-2013	A. Dufault, WDFW, pers. comm.	South Strait of Georgia	-0.012	0.035
Samish River	Pink	TRS	1959-2013	A. Dufault, WDFW, pers. comm.	South Strait of Georgia	0.221	0.068

Fraser	Pink	TRS	1959-2011	Ogden et al. 2015	South Strait of Georgia	0.070	0.030
Skagit River	Pink	TRS	1959-2013	A. Dufault, WDFW, pers. comm.	Whidbey	0.004	0.030
Snohomish River	Pink	TRS	1959-2013	A. Dufault, WDFW, pers. comm.	Whidbey	0.080	0.025
Stillaguamish River	Pink	TRS	1959-2013	A. Dufault, WDFW, pers. comm.	Whidbey	0.000	0.027
Lake Washington	Sockeye	E	1972-2015	A. Dufault, WDFW, pers. comm.	Central Puget Sound	-0.005	0.006
Fraser	Sockeye	TRS	1960-2012	Ogden et al. 2015	South Strait of Georgia	0.003	0.004
Baker Lake	Sockeye	E	1970-2015	A. Dufault, WDFW, pers. comm	Whidbey	0.044	0.012
Misc 10 -Seattle	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	Central Puget Sound	0.084	0.009
Green-Duwamish River	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	Central Puget Sound	0.135	0.014
Misc 10e -Port Orchard	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	Central Puget Sound	0.022	0.007
Misc 11 -Tacoma	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	Central Puget Sound	0.020	0.006
Puyallup River	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	Central Puget Sound	0.062	0.012
Port Gamble Pens (9a)	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	Hood Canal	-0.016	0.009
N. Hood Canal (12)	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	Hood Canal	-0.015	0.008
Misc 12b -Hood Canal	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	Hood Canal	0.013	0.007
Quilcene River (12a)	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	Hood Canal	0.034	0.011
Misc 12c -Hood Canal	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	Hood Canal	0.063	0.006
Hoodsport Hatchery	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	Hood Canal	0.049	0.010
Skokomish River	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	Hood Canal	0.073	0.007
Misc 12d -Hood Canal	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	Hood Canal	0.025	0.007
Dungeness River	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	Juan de Fuca	0.052	0.012
Elwha River	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	Juan de Fuca	-0.061	0.014
Misc Strait Streams	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	Juan de Fuca	0.023	0.009
Grays Harbor Wild Runsize	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	Pacific Coast	-0.006	0.006
Willapa Bay Wild And Hatchery Escapement And Total Runsize	Chum	TRS	1980-1999	A. Dufault, WDFW, pers. comm	Pacific Coast	-0.010	0.006
Inside South Coast (Non-Fraser)	Chum	TRS	1953-2012	Ogden et al. 2015	Pacific Coast	0.000	0.005
Hecate Lowlands	Chum	E	1954-2012	Ogden et al. 2015	Pacific Coast	-0.014	0.004
North Haida Gwaii	Chum	E	1954-2012	Ogden et al. 2015	Pacific Coast	-0.028	0.009
Portland Inlet	Chum	E	1954-2012	Ogden et al. 2015	Pacific Coast	-0.006	0.005
Rivers Inlet	Chum	E	1954-2012	Ogden et al. 2015	Pacific Coast	-0.009	0.014
Skidegate	Chum	E	1954-2012	Ogden et al. 2015	Pacific Coast	-0.013	0.006
Misc 13 -South Puget Sound	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	South Sound	-0.018	0.050

Chambers Creek	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	South Sound	0.004	0.032
Nisqually River	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	South Sound	0.026	0.054
Misc 13a -Minter Creek	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	South Sound	0.062	0.008
Misc 13b -Olympia	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	South Sound	0.029	0.003
Nooksack River	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	South Strait of Georgia	0.019	0.005
Samish River	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	South Strait of Georgia	0.007	0.010
Misc 7b Streams	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	South Strait of Georgia	0.082	0.011
Skagit River	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	Whidbey	-0.026	0.006
Tulalip Tribe	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	Whidbey	0.063	0.013
Snohomish River	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	Whidbey	-0.001	0.011
Stillaguamish River	Chum	TRS	1968-2009	A. Dufault, WDFW, pers. comm	Whidbey	-0.011	0.008

Appendix 2. Model Sensitivity Analysis

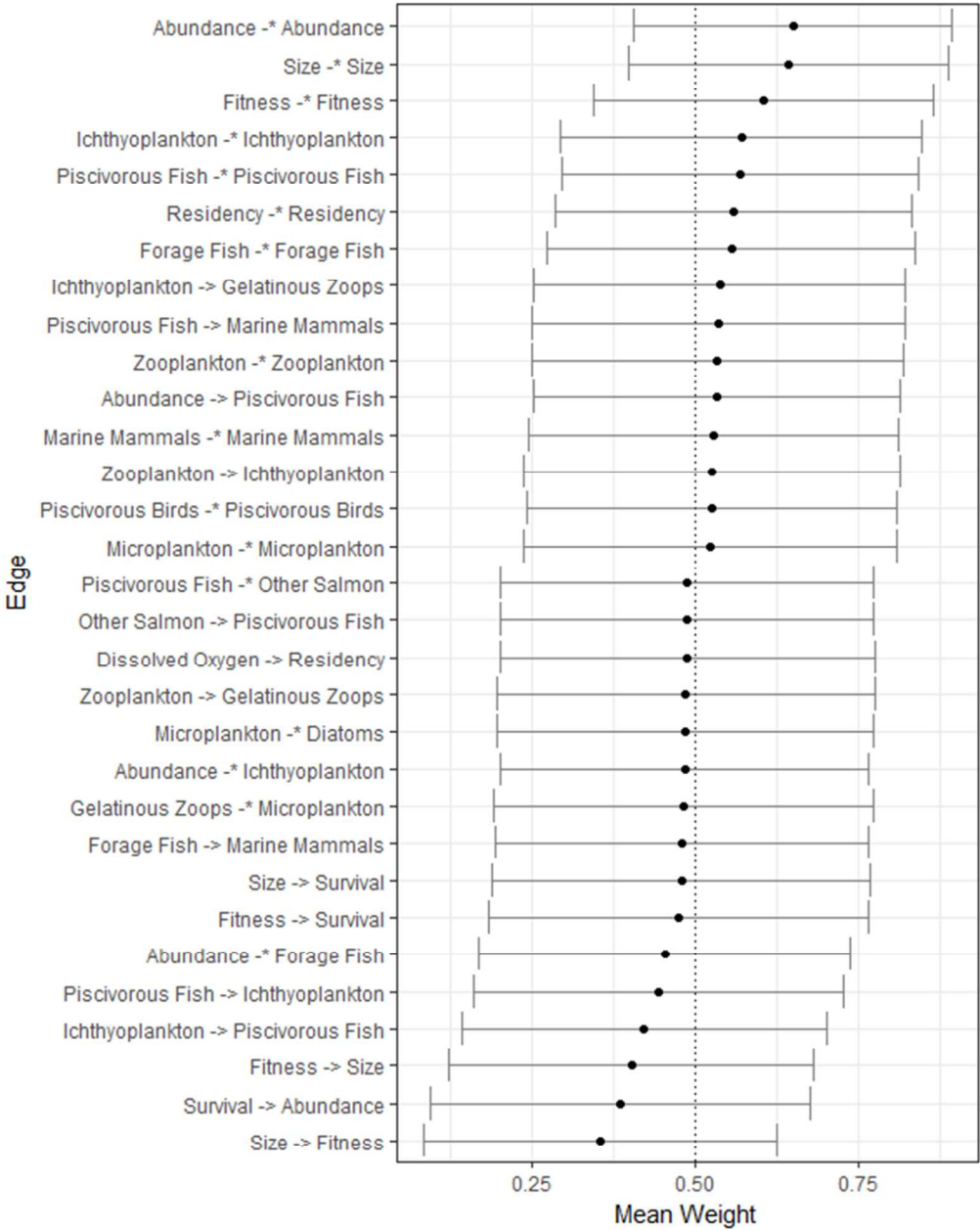
Methods

To assess the sensitivity of the model linkages in the simulations, we calculated means and standard deviations of the weights for all linkages from the pool of accepted models. Our hypothesis was that some linkages would be more influential in model stability and that those with mean weights different than the expected mean ($\mu=0.5$, given assignments that were random ($U(0,1)$)) would provide some indication of linkage sensitivity.

Results

Our sensitivity analysis showed that most model edges (linkages between groups) were stable with regard to the weights applied in the simulation routine, with means from the pool of balanced models very close to the expected mean of 0.5 (>75% of 148 edges). However, some model linkages in the pool of accepted models had mean weights above or below the expected mean and we considered these model groups more sensitive (Fig. A2.1). Linkages among salmon traits were most sensitive, with means in the balanced models considerably higher or lower than the expected mean (e.g., Fitness to Size and Survival to Abundance, self-limiting loops on salmon traits). However, some food web components such as Ichthyoplankton and Forage Fish were also in the group of lower-than-expected model means, suggesting that certain food web pathway weights were sensitive in the model simulations. The links with bidirectional connectivity (positive feedbacks, e.g. Fitness to Size, arrows shown in royal blue in Fig. 2) tended to be the most sensitive to the weighting scheme, with lower-than-average weights resulting in balanced models. Model linkages with mean weights above the expected mean included many high trophic level linkages (e.g., Piscivorous Fish to Marine Mammals), as well as self-limiting loops (salmon traits especially), highly connected nodes (e.g. Zooplankton), and top-level predators (Marine Mammals and Piscivorous Birds). The linkages with both higher- and lower-than-expected means (that were not self-limiting loops) tended to be highly connected food web components rather than anthropogenic impacts or physical drivers.

46



47

48 Figure A2.1. Calculated means and standard deviations from all model runs for edges (model
49 linkages) with mean values different from the expected mean (0.5). The expected mean is
50 denoted by the black dashed line. Positive linkages are indicated by an arrow and negative
51 linkages are indicated by a dash and asterisk.

52

53

54 Appendix 3. Model Perturbations and References

Drivers	Variables	Reference	Invoked Perturbation
<i>Environmental</i>	Sunlight	Littell et al. 2009	↑
	Winter Storms	Littell et al. 2009	↑
	Precipitation	Mauger et al. 2015, Littell et al. 2009	↑
	Upwelling ¹	Mauger et al. 2015	↓
	Stratification ²	Mauger et al. 2015	↑
	Temperature	Littell et al. 2009, Field et al. 2006, Hollowed et al. 2001	↑
	River Flow	Littell et al. 2009 ³	↑
	Turbidity	PSEMP 2016, WADOE 2017	↓
	Dissolved Oxygen	Roberts et al. 2014	↓
<i>Production</i>	Nutrients	Roberts et al. 2014	↑
	Microplankton	Moore et al. 2015	↑
	Microbial Detritivores	PSEMP 2016	↑
	Diatoms	Brandenberger 2008, WADOE 2017	↓
<i>Food web</i>	Zooplankton ⁴	Li et al. 2013	↓
	Gelatinous Zooplankton	Greene et al. 2015	↑
	Forage Fish	Greene et al. 2015	↓
	Ichthyoplankton	Palsson et al. 1998	↓
	Other Salmon	PFMC 2016, Figure 1	↑
	Piscivorous Fish	Palsson et al. 1998	↓
	Piscivorous Birds	Gaydos and Pearson 2011, Anderson et al. 2009	↓
	Marine Mammals	Chasco et al. 2017	↑
<i>Anthropogenic</i>	Hatcheries	Christie et al. 2012, Waples 1999	↑
	Harvest ⁵		↑
	Habitat Loss	Puget Sound Water Quality Action Team 2002	↑
	CO2	Feely et al. 2010	↑
	Global Warming	IPCC 2014	↑
	Contaminants	O'Neill et al. 2009, Meador et al. 2006, Crecelius et al. 1995	↑
	Disease	SSMSP Hypotheses	↑

55 Literature Cited

¹ While impacts to upwelling are unknown, we invoked a decrease in upwelling because it would reduce the delivery of nutrients to Puget Sound and potentially disrupt primary production.

² Decreased mixing as a result of changes in freshwater flow could lead to increased stratification; while the exact response is unknown (see Mauger et al. 2015), we have invoked an increase in stratification because decreased mixing would likely lead to declines in primary productivity.

³ River flow is projected to be higher during the winter/spring period of salmon outmigration, but lower in the summers due to decreased precipitation and higher temps. during this season (Littell et al. 2009).

⁴ While it is unknown if total abundance of zooplankton has decreased, there is some indication that the community has shifted. We invoked a decrease in zooplankton to reflect this shift.

⁵ Harvest has decreased over the time period we used to frame this analysis; however, we were interested in the impacts of increased harvest on salmon survival so invoked an increase in harvest as the perturbation

- Anderson, E.M., Bower, J.L., Nysewander, D.R., Evenson, J.R. and Lovvorn, J.R. 2009. Changes in avifaunal abundance in a heavily-used wintering and migration site in Puget Sound, Washington during 1966–2007. *Marine Ornithology* 37: 19-27.
- Burke, B. J., B. R. Beckman, W. T. Peterson, C. Morgan, E. A. Daly, M. Litz. 2013. Multivariate methods to forecast Pacific salmon returns. *PLoS ONE* 8(1): e54134.
- Chasco, B. I. Kaplan, A. Thomas, A. Alejandro-Gutiérrez, D.P. Noren, M.J. Ford, M.B. Hanson, J. Scordino, S.J. Jeffries, S.F. Pearson, K. Marshall, E.J. Ward. 2017. Estimates of Chinook salmon consumption in Washington State inland water by four marine mammal predators from 1970-2015. *Can. J. of Fish. and Aqua. Sci.* 10.1139/cjfas-2016-0203
- Christie, M.R., M.L. Marine, R.A. French, R.S. Waples and M.S. Blouin. 2012. Effective size of a wild salmonid population is greatly reduced by hatchery supplementation. *Heredity* 109: 254-260.
- Crecilius, E.A., V.I. Cullinan, L.F. Lefkovitz and C. Peden. 1995. Historical trends in the accumulation of chemicals in Puget Sound. 825-831, in Volume 2, Puget Sound Research '95 Proceedings, Puget Sound Water Quality Authority, Olympia.
- Feely, R. A., S. R. Alin, J. Newton, C. L. Sabine, M. Warner, A. Devol, C. Krembs, and C. Maloy. 2010. The combined effects of ocean acidification, mixing, and respiration on pH and carbonate saturation in an urbanized estuary. *Estuarine, Coastal and Shelf Science*, 88, 442-449, doi:10.1016/j.ecss.2010.05.004
- Field, D., D. Cayan, and F. Chavez. 2006. Secular warming in the California current and North Pacific. *California Cooperative Oceanic Fisheries Investigations Reports*, 47, 92-108.
- Hoekstra, J. M., K.K. Bartz, M.H. Ruckelshaus, J.M. Moslemi, and T.K. Harms. 2007. Quantitative threat analysis for management of an imperiled species: Chinook salmon (*Oncorhynchus tshawytscha*). *Ecological Applications*, 17: 2061-2073.
- Hollowed, A. B., S. R. Hare, and W. S. Wooster. 2001. Pacific Basin climate variability and patterns of Northeast Pacific marine fish production. *Progress in Oceanography*, 49, 257-282, doi:10.1016/S0079-6611(01)00026-X.
- Gaydos, J.K. and S.F. Pearson. 2011. Birds and mammals that depend on the Salish Sea: A compilation. *Northwest Naturalist*: 92:79–94.
- Intergovernmental Panel on Climate Change. 2014. Climate Change 2014: Synthesis Report. Contribution of Working Groups I, II and III to the Fifth Assessment Report of the Intergovernmental Panel on Climate Change [Core Writing Team, R.K. Pachauri and L.A. Meyer (eds.)]. IPCC, Geneva, Switzerland, 151 pp.

Li, L., D. Mackas, B. Hunt, J. Schweigert, E. Pakhomov, I. Perry, M. Galbraith, T.J. Pitcher. 2013. Zooplankton communities in the Strait of Georgia, British Columbia, track large-scale climate forcing over the Pacific Ocean. *Progress in Oceanography*. doi:<http://dx.doi.org/10.1016/j.pocean.2013.05.025>

Littell, J.S., M. McGuire Elsner, L.C. Whitely Binder, and A.K. Snover (eds). 2009. *The Washington Climate Change Impacts Assessment: Evaluating Washington's Future in a Changing Climate*. Climate Impacts Group, University of Washington, Seattle, Washington. Available at: www.cses.washington.edu/db/pdf/wacciaexecsummary638.pdf

Mauger, G.S., J.H. Casola, H.A. Morgan, R.L. Strauch, B. Jones, B. Curry, T.M. Busch Isaksen, L. Whitely Binder, M.B. Krosby, and A.K. Snover. 2015. State of Knowledge: Climate Change in Puget Sound. Report prepared for the Puget Sound Partnership and the National Oceanic and Atmospheric Administration. Climate Impacts Group, University of Washington, Seattle. doi:10.7915/CIG93777D

Meador, J.P., F.C. Sommers, G.M. Ylitalo, C.A. Sloan. 2006. Altered growth and related physiological response in juvenile Chinook salmon (*Oncorhynchus tshawytscha*) from dietary exposure to polycyclic aromatic hydrocarbons (PAHs). *Can. J. of Fish. and Aqua. Sci.* 63: 2364-2376.

Moore, S.K. et al., 2011. Past trends and future scenarios for environmental conditions favoring the accumulation of paralytic shellfish toxins in Puget Sound shellfish. *Harmful Algae*, 10, 521-529.

Moore, S.K. et al., 2015. Present-day and future climate pathways affecting the harmful algal blooms species *Alexandrium catenella* in Puget Sound, WA, USA. *Harmful Algae*, 48, 1-11. 25

O'Neill, S.M., and West, J.E. 2009. Marine Distribution, Life History Traits, and the Accumulation of Polychlorinated Biphenyls in Chinook Salmon from Puget Sound, Washington. *Transactions of the American Fisheries Society* 138(3): 616-632.

Pacific Fishery Management Council. 2016. Review of 2015 Ocean Salmon Fisheries: Stock Assessment and Fishery Evaluation Document for the Pacific Coast Salmon Fishery Management Plan. Pacific Fishery Management Council, Portland, Oregon. http://www.pcouncil.org/wp-content/uploads/2016/02/Review_of_2015_Salmon_Fisheries_FullDocument.pdf

Palsson, W.A., T.J. Northrup, and M.W. Baker. 1998. Puget Sound Groundfish Management Plan. Washington Department of Fish and Wildlife. Olympia, WA.

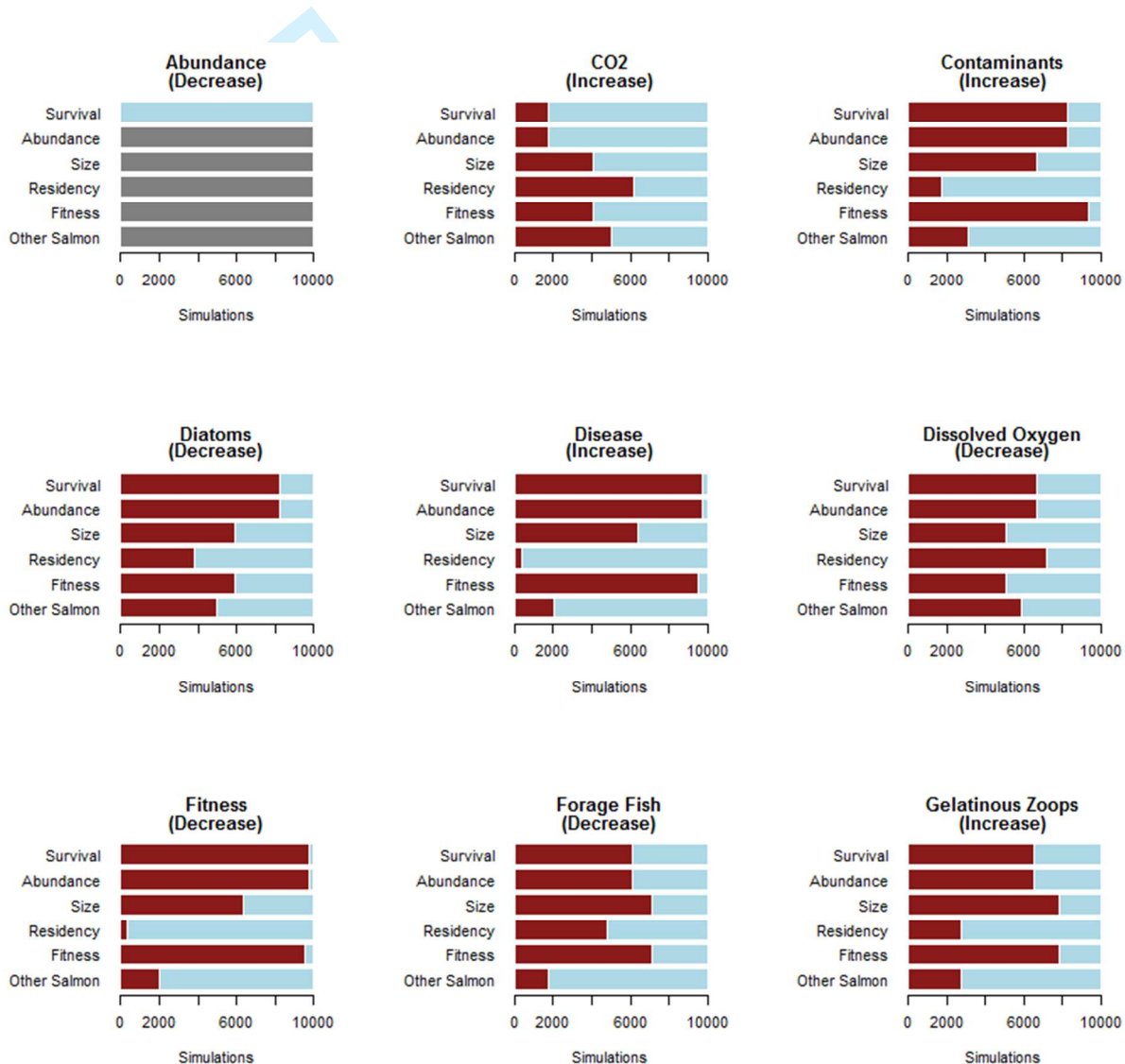
PSEMP Marine Waters Workgroup. 2016. Puget Sound marine waters: 2015 overview. S. K. Moore, R. Wold, K. Stark, J. Bos, P. Williams, K. Dzinbal, C. Krembs and J. Newton (Eds). URL: www.psp.wa.gov/PSEMP/PSmarinewatersoverview.php

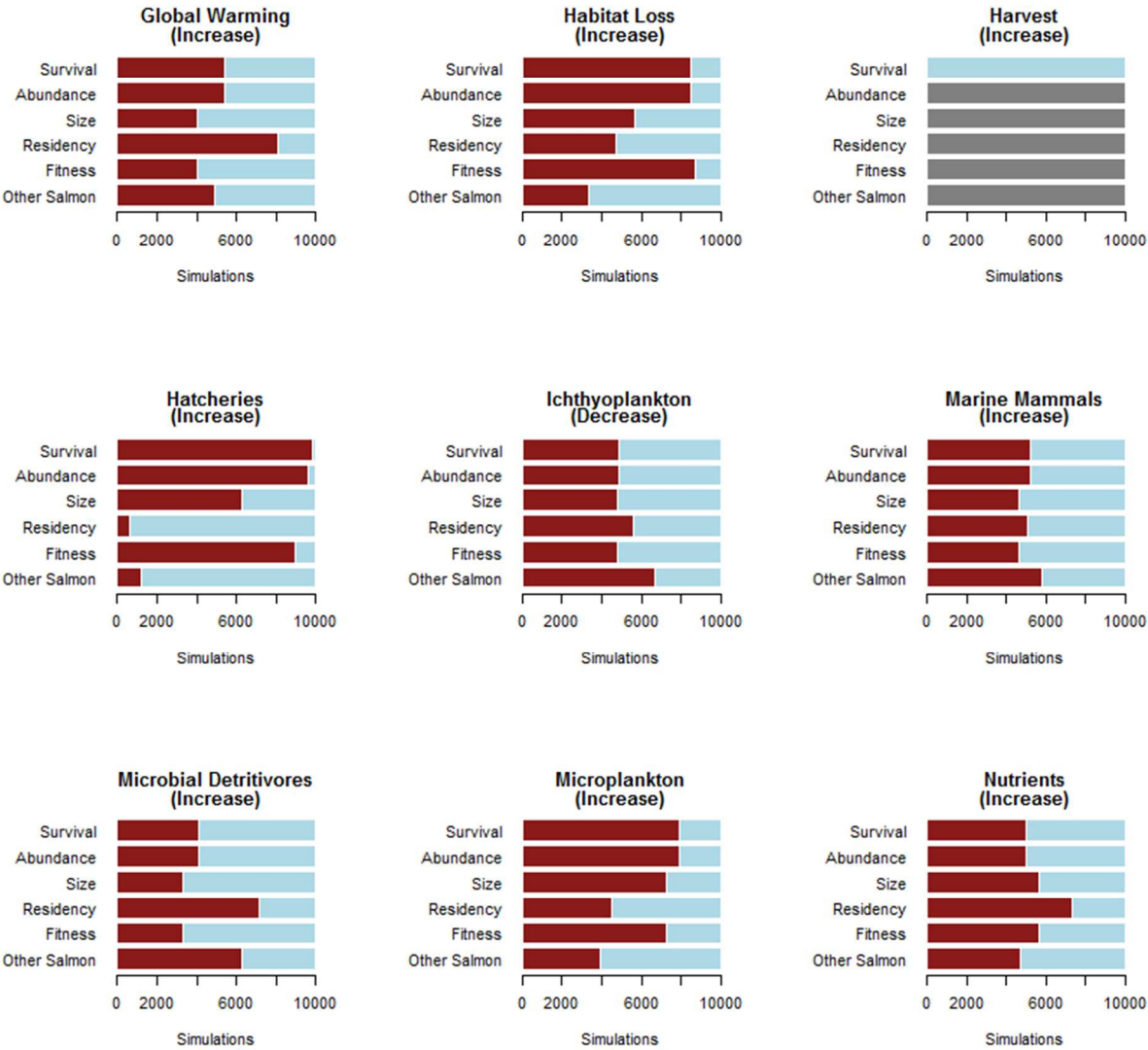
144
145 Puget Sound Water Quality Action Team. 2002. *2002 Puget Sound Update: Eighth Annual*
146 *Report of the Puget Sound Ambient Monitoring Program*. Puget Sound Water Quality Action
147 Team. Olympia, Washington.
148
149 Roberts, M., T. Mohamedali, B. Sackmann, T. Khangaonkar, W. Long. 2014. Puget Sound and
150 the Straits Dissolved Oxygen Assessment Impacts of Current and Future Human Nitrogen
151 Sources and Climate Change through 2070. Washington State Department of Ecology,
152 Publication No. 14-03-007. Olympia, WA.
153
154 Waples, R.S. 1999. Dispelling some myths about hatcheries. *Fisheries* 24: 12–21.
155
156 WADOE. 2017. Marine Waters Program, Washington Department of Ecology.
157 http://www.ecy.wa.gov/programs/eap/mar_wat/
158

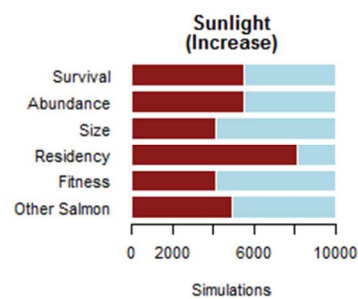
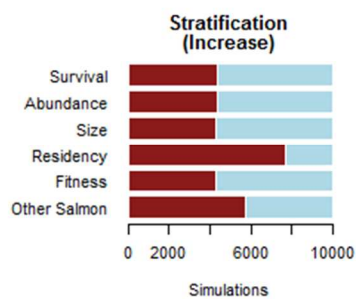
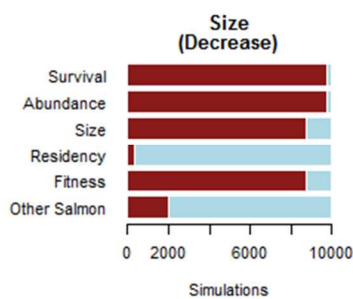
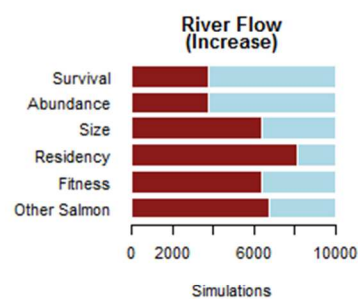
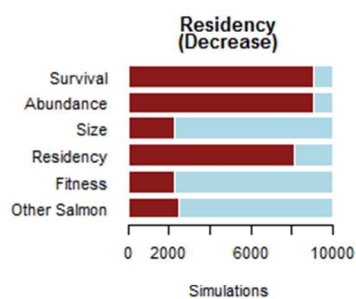
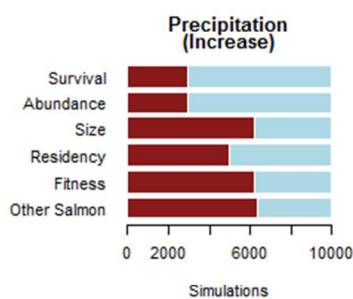
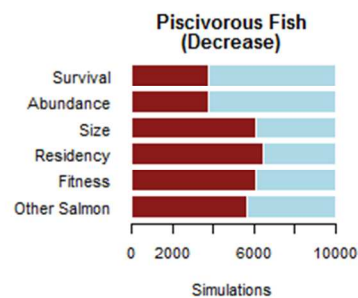
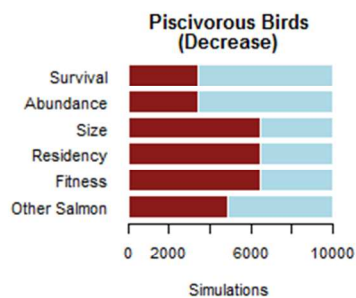
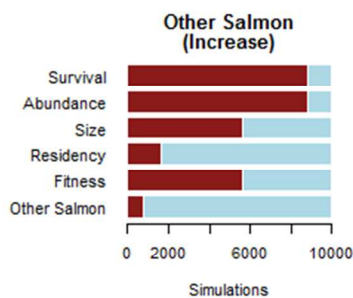
159

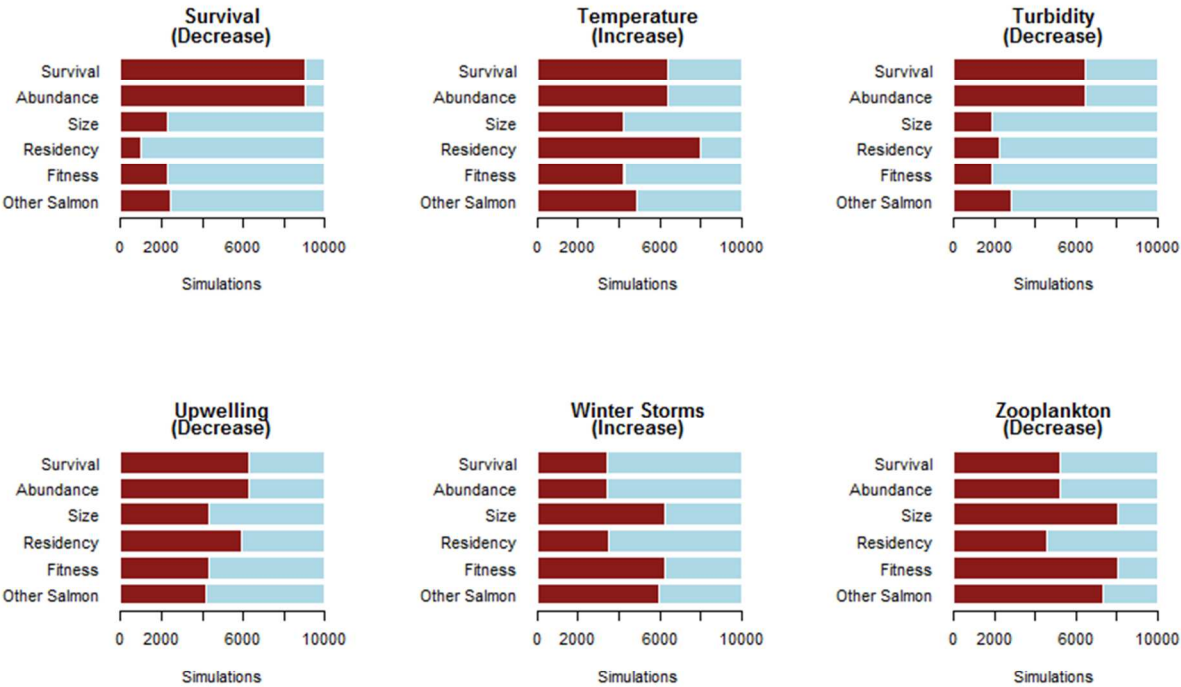
Appendix 4. Model Output for Individual Perturbations

Model output showing 6 model nodes of interest: Salmon Survival, Salmon Abundance, Salmon Size, Residency, Fitness and Other Salmon. Other Salmon refers to the populations (chum, pink, and sockeye) which have not seen a noticeable decline in survival in recent decades. In each plot box, the model node that was perturbed is shown in the title, with the direction of the press (Increase or Decrease) shown below. The bar graphs indicate the proportion of model simulations with negative (red bars) and positive (blue bars) outcomes for that model node, given the invoked press perturbation. Where the bars are dark gray, there was no impact to those nodes.

















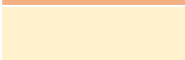



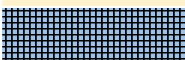








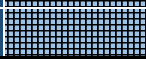
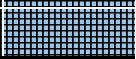
173
174
175

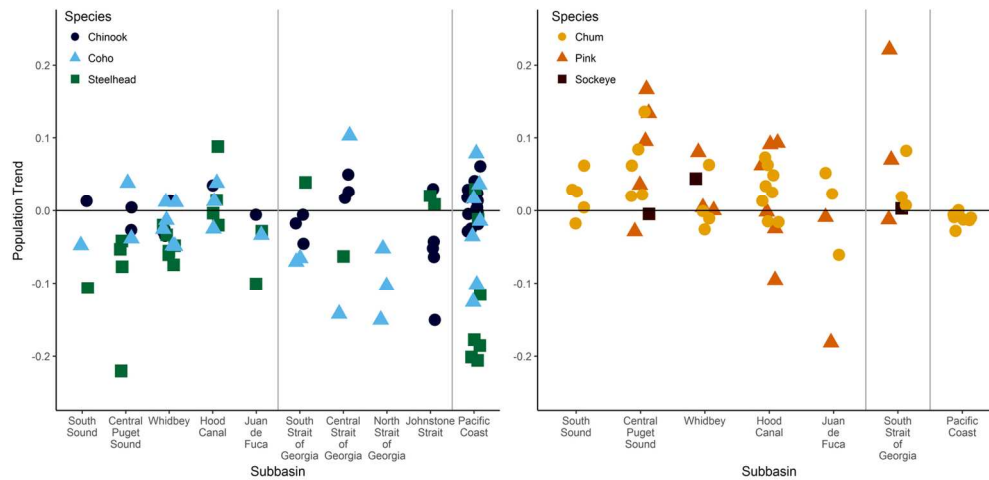


<i>Drivers</i>	<i>Variables</i>	<i>Total # of Connected Nodes</i>	<i>Number of Influencing Survival</i>	<i>Minimum Nodal Distance to Survival</i>	<i>Description</i>
<i>Environmental</i>	Sunlight	4	2	3	Solar radiation
	Winter Storms	5	1	4	Commonly occurring winter storm events
	Precipitation	5	2	3	Annual total precipitation
	Upwelling	5	2	3	Oceanographic upwelling driven by wind and currents in the coastal waters
	Stratification	8	6	3	Formation of layers in the water column resulting from ocean conditions
	Temperature	11	5	2	Water temperature within the Salish Sea (generalized, but upper portion of the water column where salmon occur)
	River Flow	4	2	2	Annual streamflow
	Turbidity	4	3	1	Relative clarity of the water within the Salish Sea
	Dissolved Oxygen	9	7	2	Amount of oxygen available in Salish Sea waters
<i>Production</i>	Nutrients	5	5	4	Total nutrients (generalized to be anthropogenic sources of N)
	Microplankton	8	5	4	Dinoflagellates, e.g. <i>Noctiluca</i> spp.
	Microbial Detritivores	7	6	3	Generalized microbes, including bacteria
	Diatoms	11	9	3	Autotrophic phytoplankton
<i>Food web</i>	Zooplankton	9	8	2	Energy-rich zooplankton (e.g., copepods, krill, amphipods)
	Gelatinous Zooplankton	6	5	3	Zooplankton including ctenophores, medusae, and salps
	Forage Fish	9	9	2	Herring, smelt, and other small-bodied fishes
	Ichthyoplankton	7	6	2	Immature stages of fish, residing in the water column
	Other Salmon	10	10	2	Chum, Pink, and Sockeye Salmon
	Piscivorous Fish	7	6	1	Any fish-eating fish; characterized by gadids and scorpaenids in the Salish Sea
	Piscivorous Birds	5	4	1	Any fish-eating bird, such as cormorants and auklets
	Marine Mammals	7	6	1	Generally harbor seals, sea lions, orcas, and dolphins
<i>Anthropogenic</i>	Hatcheries	4	1	2	Production, through human intervention, of large numbers of juvenile fish through breeding programs, specifically salmon
	Harvest	2	1	3	Catch of fish, specifically Steelhead, Coho, and Chinook salmon; generalized to include both recreational and commercial take
	Habitat Loss	5	1	2	Loss on intertidal and subtidal habitats for spawning or rearing
	CO2	5	5	4	Input of carbon dioxide via anthropogenic activities
	Global Warming	3	1	2	The general warming trend of the earth's atmosphere
	Contaminants	6	1	2	Exposure to common toxins like PCBs, PBDE, etc., as well as contaminants of emerging concern (e.g. pharmaceuticals)
	Disease	3	2	2	Exposure to diseases such as <i>Nanophyetus</i> and bacterial kidney disease
<i>Salmon Traits</i>	Residence Time	6	5	1	The amount of time an outmigrating salmon spends in the Salish Sea
	Size	6	5	1	Overall size of salmon
	Fitness	7	6	1	Overall health of salmon
	Abundance	11	4	2	Number or biomass of salmon
	Survival	8	7	-	Successful completion of the marine life stage by individuals of a population

Drivers	Variables	Invoked Perturbation	Response Variables					
			Survival	Abundance	Fitness	Size	Residence	Other Salmon
Environmental	Sunlight	↑						
	Winter Storms	↑						
	Precipitation	↑						
	Upwelling	↓						
	Stratification	↑						
	Temperature	↑						
	River Flow	↑						
	Turbidity	↓						
	Dissolved Oxygen	↓						
Production	Nutrients	↑						
	Microplankton	↑						
	Microbial Detritivores	↑						
	Diatoms	↓						
Food web	Zooplankton	↓						
	Gelatinous Zooplankton	↑						
	Forage Fish	↓						
	Ichthyoplankton	↓						
	Other Salmon	↑						
	Piscivorous Fish	↓						
	Piscivorous Birds	↓						
	Marine Mammals	↑						
Anthropogenic	Hatcheries	↑						
	Harvest	↑						
	Habitat Loss	↑						
	CO2	↑						
	Global Warming	↑						
	Contaminants	↑						
	Disease	↑						
Salmon Traits	Residence	-	Strong Neg. Effect (>80% of runs negative)					
	Size	-	Weak Neg. Effect (60-80% of runs negative)					
	Fitness	-	Neutral (40-60% of runs positive/negative)					
	Abundance	-	Weak Pos. Effect (60-80% of runs positive)					
	Survival	-	Strong Pos. Effect (>80% of runs positive)					

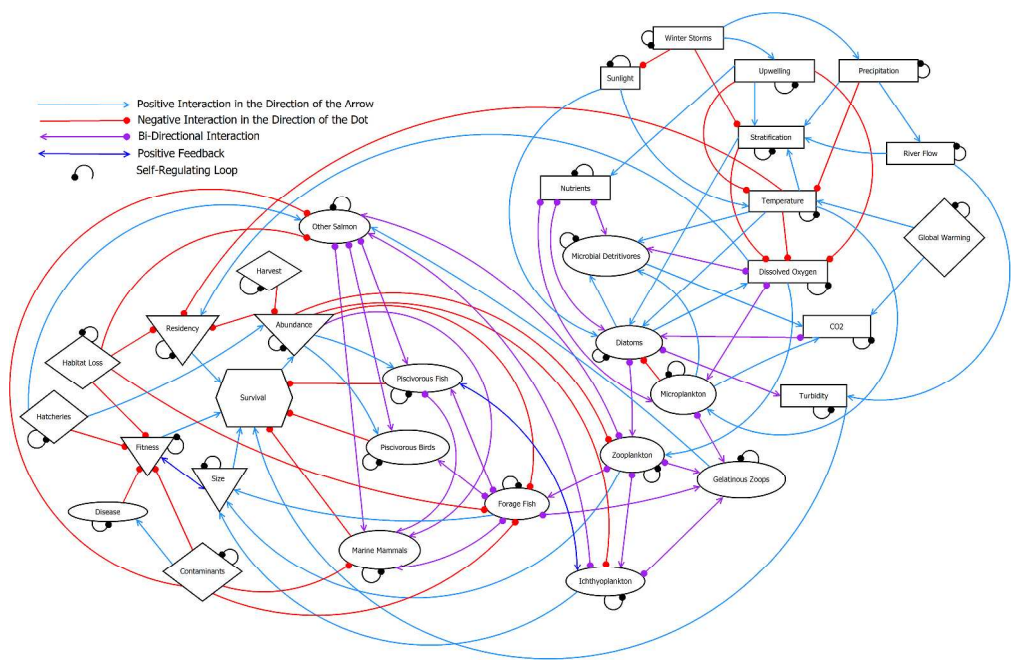
Variable	Driver Group			
	Environmental	Primary Production	Food web	Anthropogenic
Winter Storms	↑			
Precipitation	↑			
Temperature	↑			
Dissolved Oxygen	↓			
Nutrients		↑		↑
Microplankton		↑		
Microbial Detritivores		↑		
Diatoms		↓		
Gelatinous Zooplankton			↑	
Forage Fish			↓	
Piscivorous Fish			↓	
Marine Mammals			↑	
Hatcheries				↑
Habitat Loss				↑
Contaminants				↑
Total Nodal Distance	11	14	8	11

<i>Drivers</i>	<i>Perturbations</i>	<i>References</i>	<i>South Sound</i>	<i>Hood Canal</i>	<i>Central Basin</i>
<i>Oceanographic</i>	Nutrients	Roberts et al. 2014	↑		
	Stratification	Mauger et al. 2015		↑	
	Dissolved Oxygen	Roberts et al. 2014		↓	
	Turbidity	PSEMP 2016		↓	
	Temperature	PSEMP 2016		↑	
<i>Food web</i>	Diatoms	PSEMP 2016			↓
	Gelatinous Zooplankton	Greene et al. 2015	↑		↑
	Forage Fish	Greene et al. 2015	↓		↓
	Other Salmon	Fig. 1, this paper		↑	
<i>Anthropogenic Impacts</i>	Contaminants	O'Neill et al. 2009	↑		↑
	Habitat Loss	Hoekstra et al. 2007			↑
	Hatcheries	Hoekstra et al. 2007	↑		
<i>Response Key</i>		<i>Responses</i>	<i>South Sound</i>	<i>Hood Canal</i>	<i>Central Basin</i>
Strong Neg. Effect (>80% of runs were negative)		Survival			
Weak Neg. Effect (60-80% of runs were negative)		Abundance			
Neutral (40-60% of runs positive/negative)		Fitness			
Weak Pos. Effect (60-80% of runs were positive)		Size			
Strong Pos. Effect (>80% of runs were positive)		Residency			
		Other Salmon			



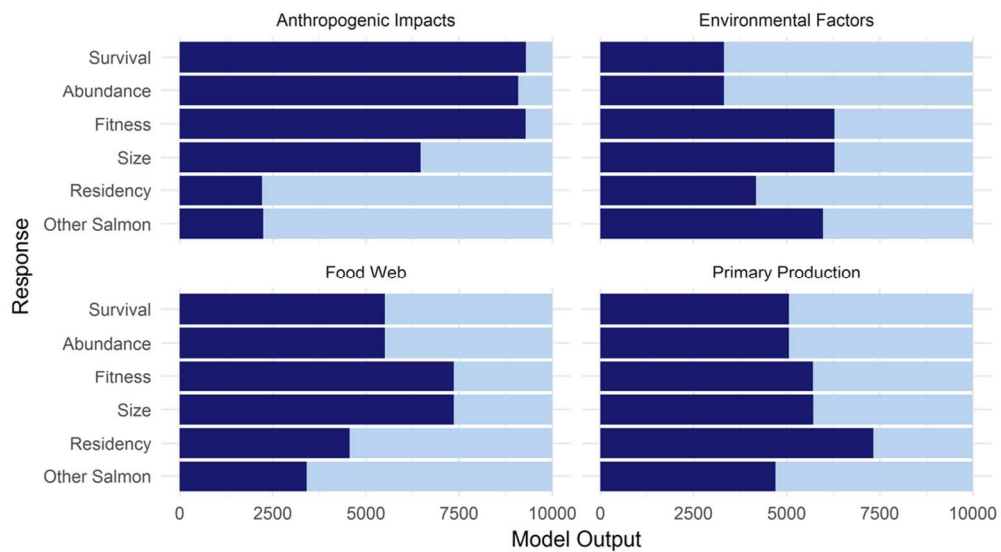
Salmon population trends within Puget Sound (left most subbasins), the Strait of Georgia (center subbasins) and the Pacific Coast (right) for species with generally decreasing trends (Chinook, Coho, and Steelhead, left) and increasing or stable populations trends (Chum, Pink, and Sockeye, right). Two abnormally high Pink salmon runs were omitted from the right panel for ease of presentation—these runs had trends of 1.05 and 0.43 and were both from Puget Sound subbasins. Data are from Washington Department of Fish and Wildlife, Pacific States Marine Fisheries Commission, Ogden et al. 2015, and Zimmerman et al. 2015.

139x66mm (300 x 300 DPI)



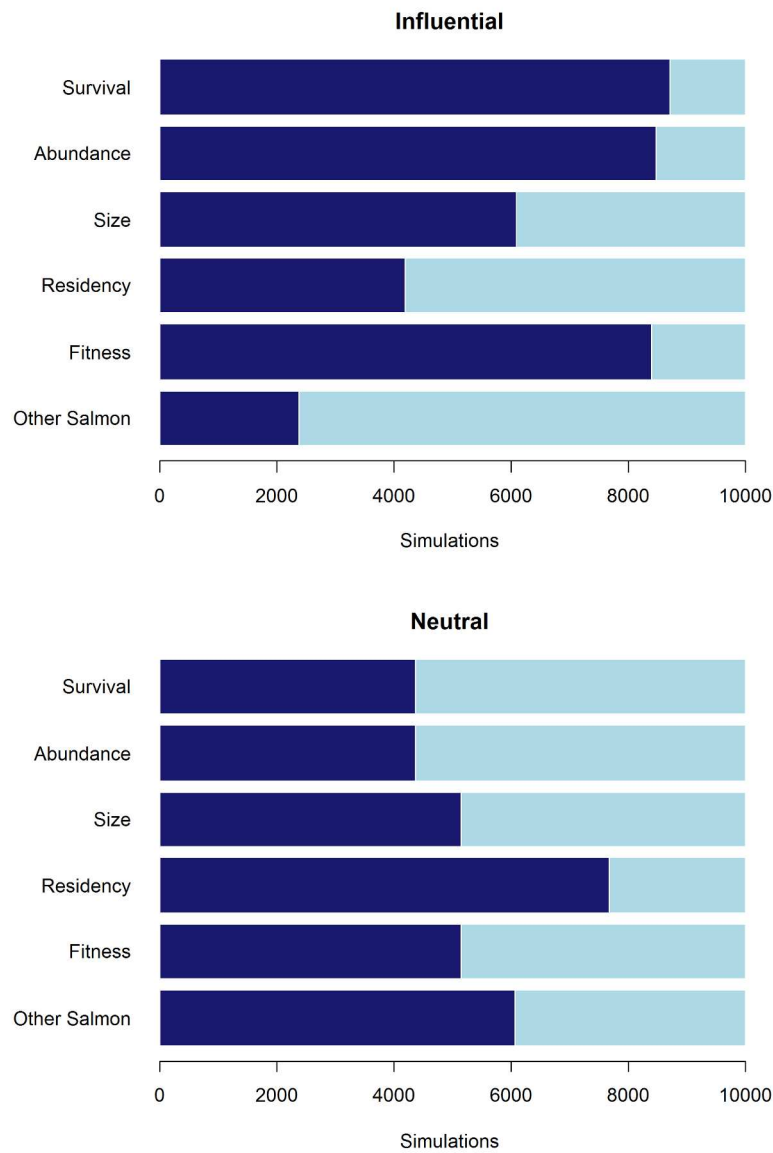
Conceptual model of the Salish Sea related to salmon survival. Model variables (shapes) represent food web components (ovals), physical drivers (rectangles), anthropogenic impacts (diamonds) and salmon traits of interest (triangles). Survival is shown as a hexagon and was the primary variable of interest.

1377x891mm (96 x 96 DPI)



Results for the primary response variables for each of the four groups of drivers. Perturbations were according to Table 3 and in all four figures, bars represent the number of runs resulting in negative (dark bars) or positive (light bars) response in the response variable.

101x58mm (300 x 300 DPI)



Comparison of multiple (n=10) strongly influential (top) and neutral (bottom) variables on salmon survival. The influential variables (resulting from the analysis presented in Table 2) were: CO2, Contaminants, Diatoms, Disease, Forage Fish, Gelatinous Zooplankton, Habitat Loss, Hatcheries, Other Salmon, Temperature. The neutral variables were: Global Warming, Ichthyoplankton, Marine Mammals, Microbial Detritivores, Nutrients, Piscivorous Birds, Precipitation, Stratification, Sunlight, Zooplankton. The variables in each set were perturbed simultaneously to simulate cumulative impacts and assess model response. Dark bars show negative impacts to the response variable and light bars show positive responses.

177x266mm (300 x 300 DPI)