

Evaluating uncertainty in estimates of how climate change may impact Northeast Pacific marine ecosystems



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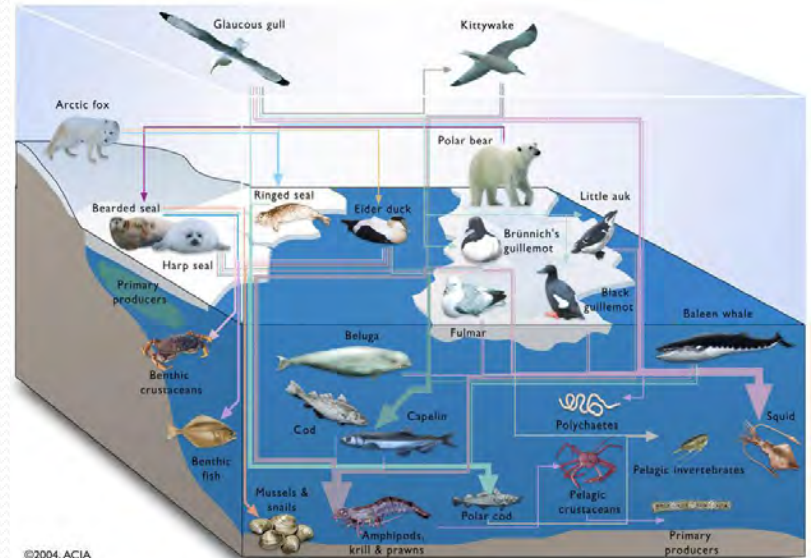
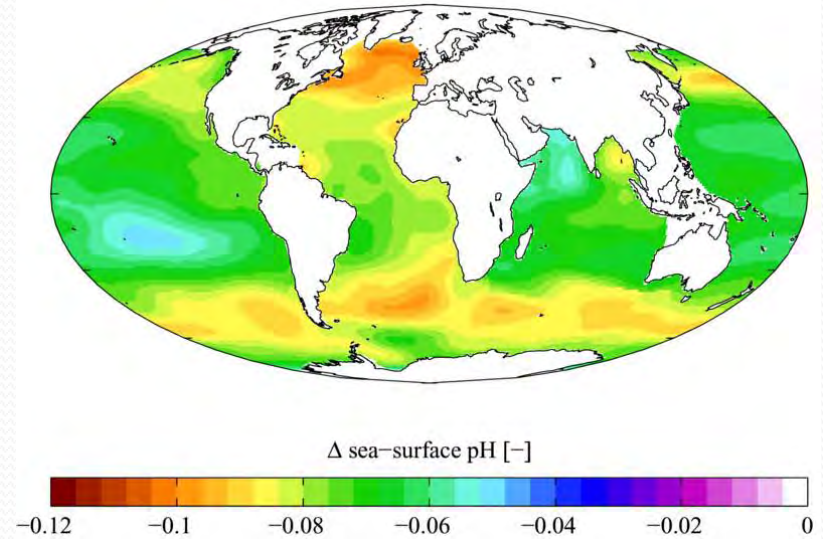
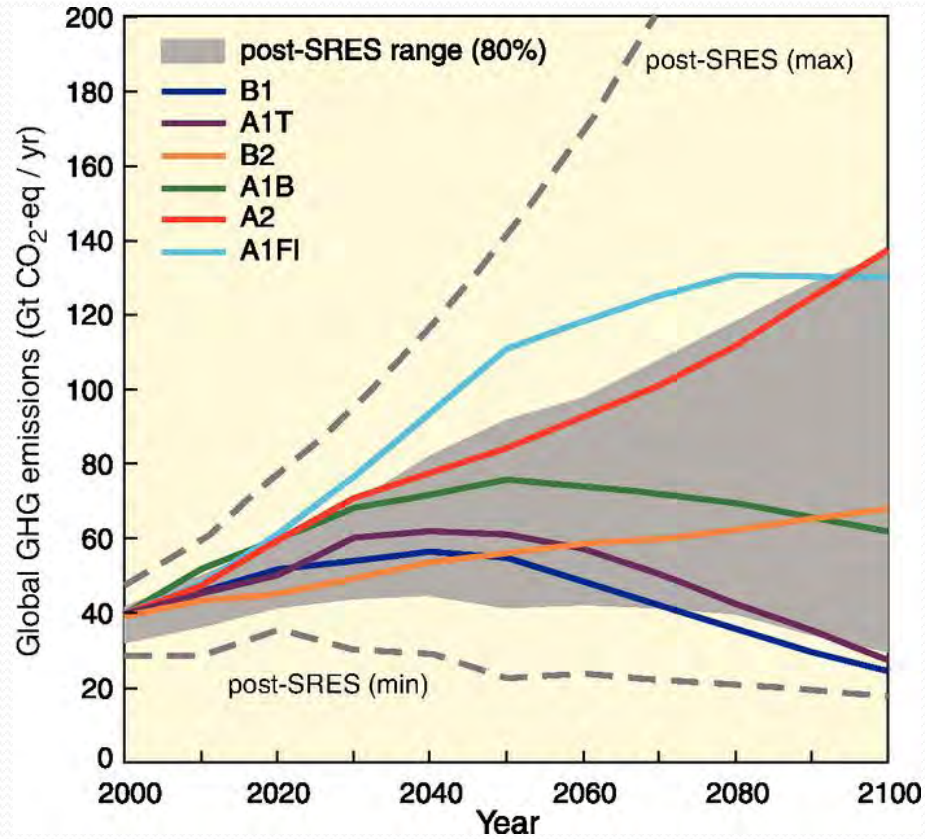
4. NOAA – Geophysical Fluid Dynamics Lab

5. WCVI Aquatic Management Board



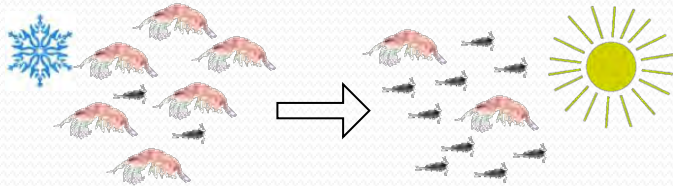
*West Coast
Vancouver
Island Aquatic
Management
Board*

Uncertainty in estimating response to climate change

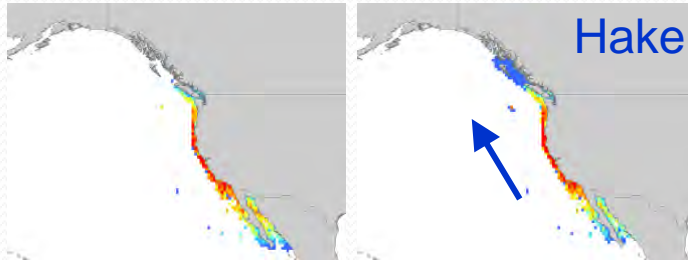


Biological impacts

Plankton community structure

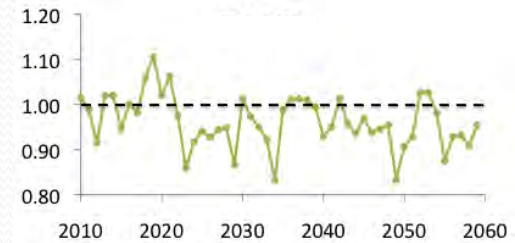


Species range shifts



Cheung et al. 2009 (Fish and Fisheries)

Primary productivity

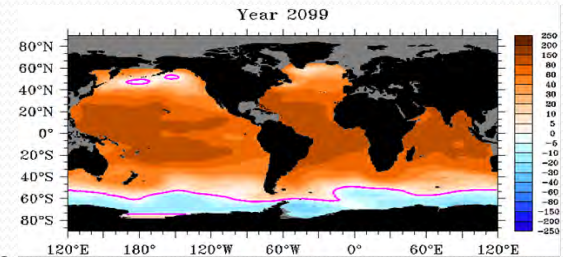


Geophysical Fluid Dynamics Laboratory CM2.1

Deoxygenation

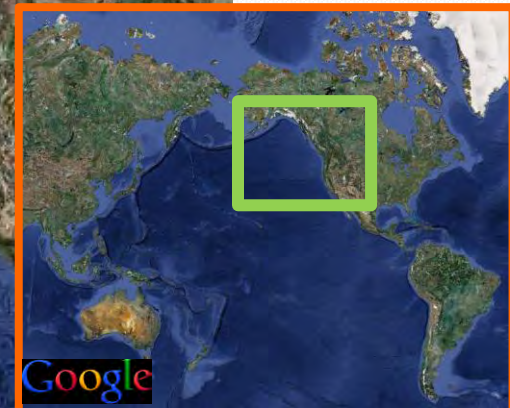
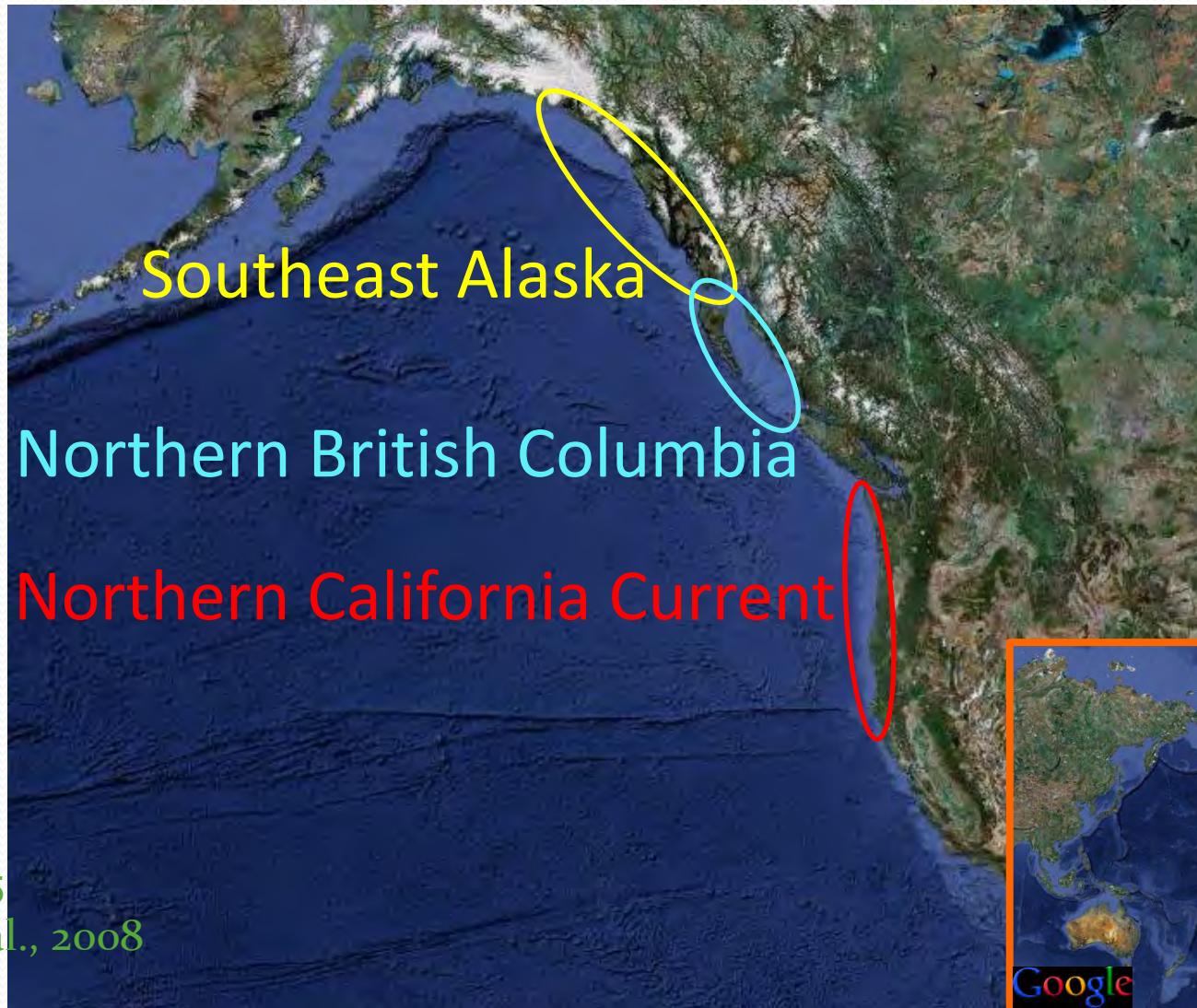


Ocean acidification



Orr et al 2005

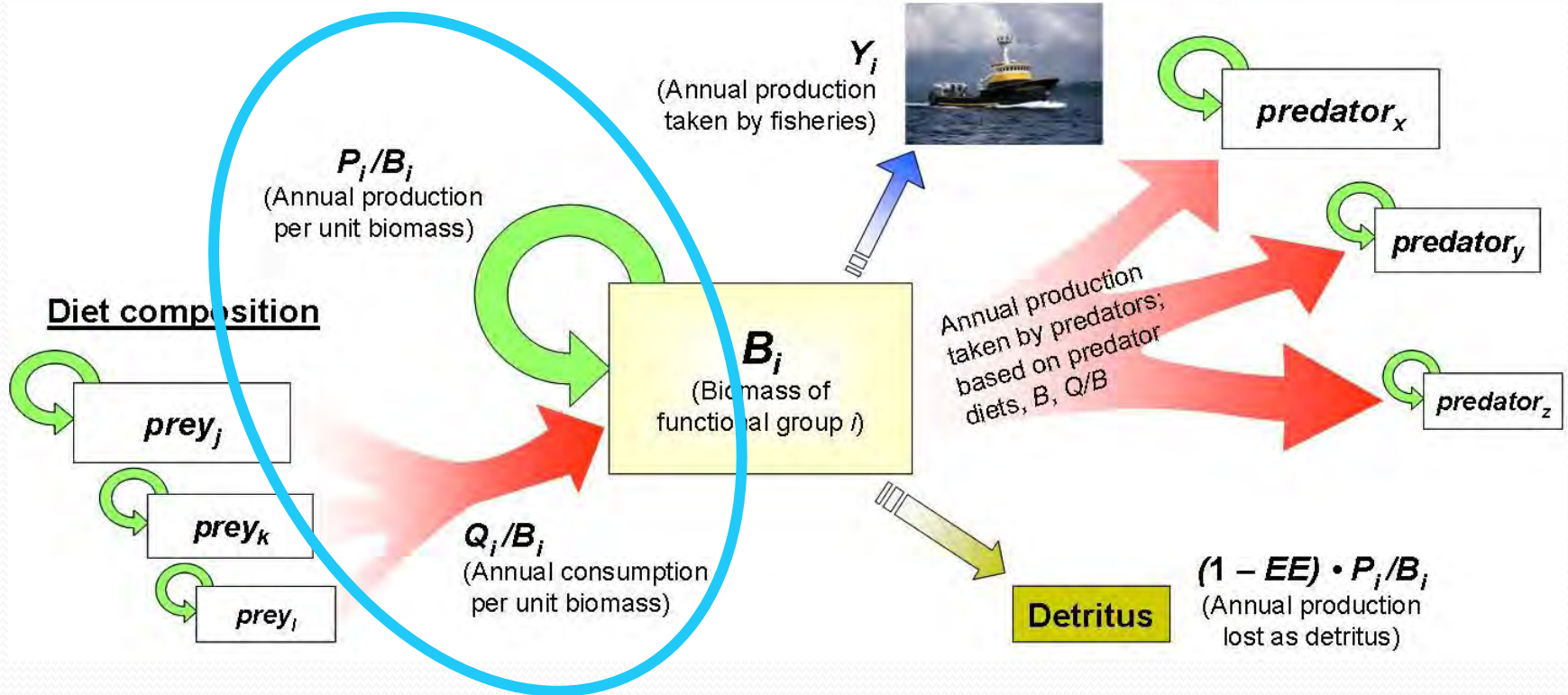
Three NE Pacific food web models



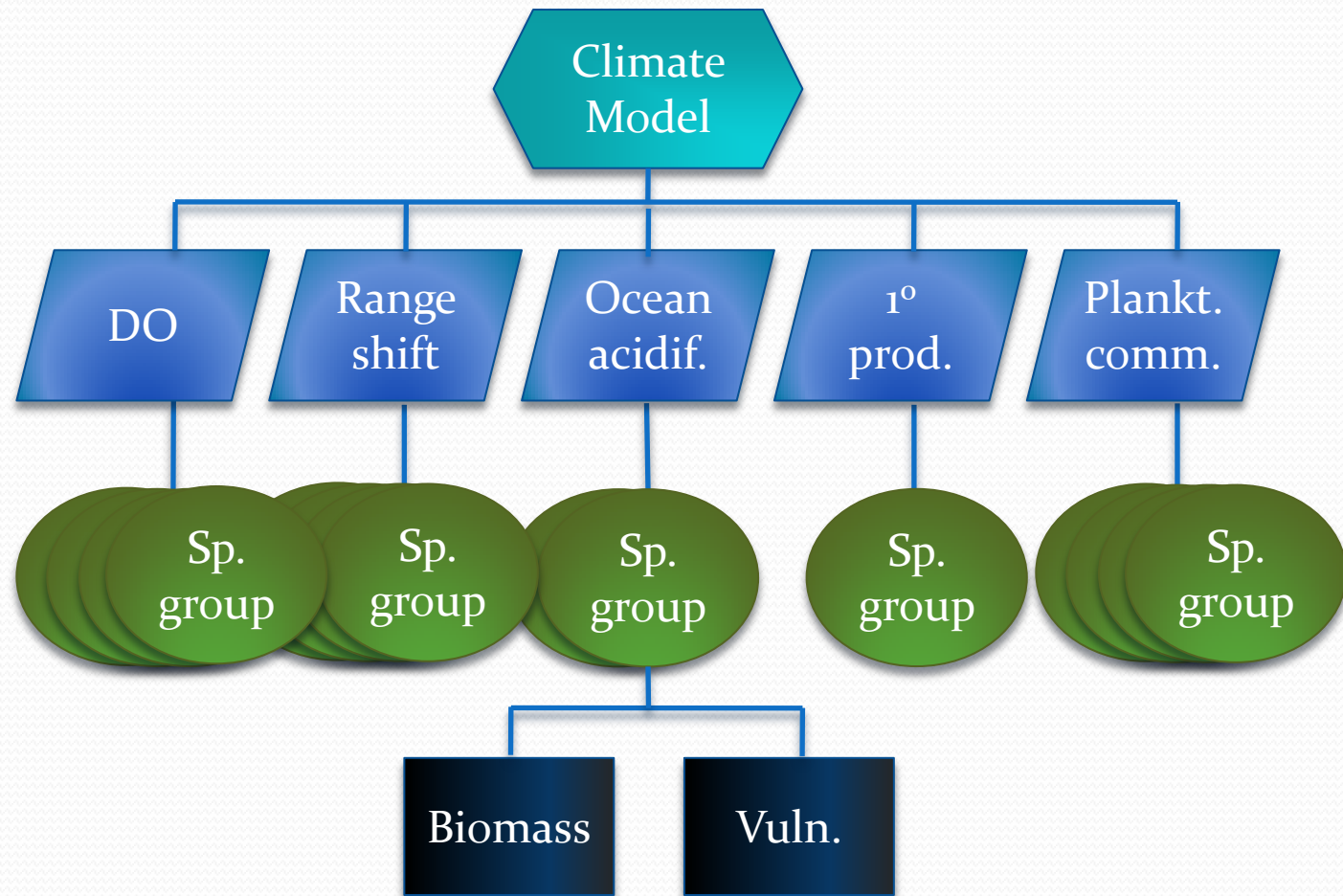
Guénette 2005
Ainsworth et al., 2008
Field 2004

Ecopath with Ecosim

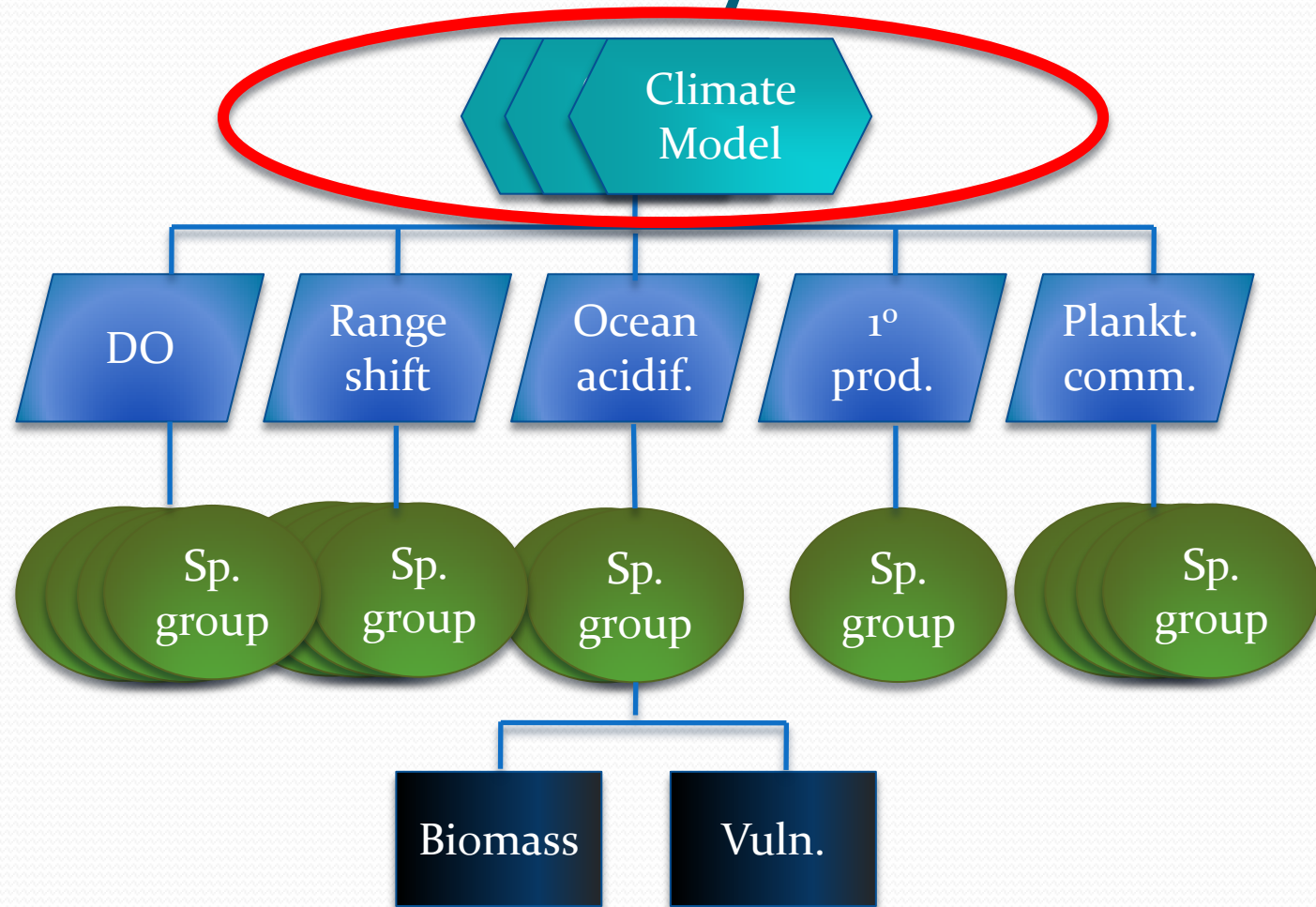
Polovina, Christensen, Pauly, Walters



Addressing uncertainty



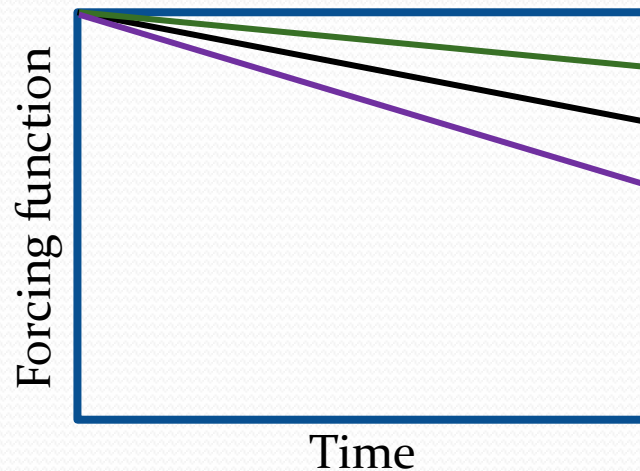
Model uncertainty



Model uncertainty

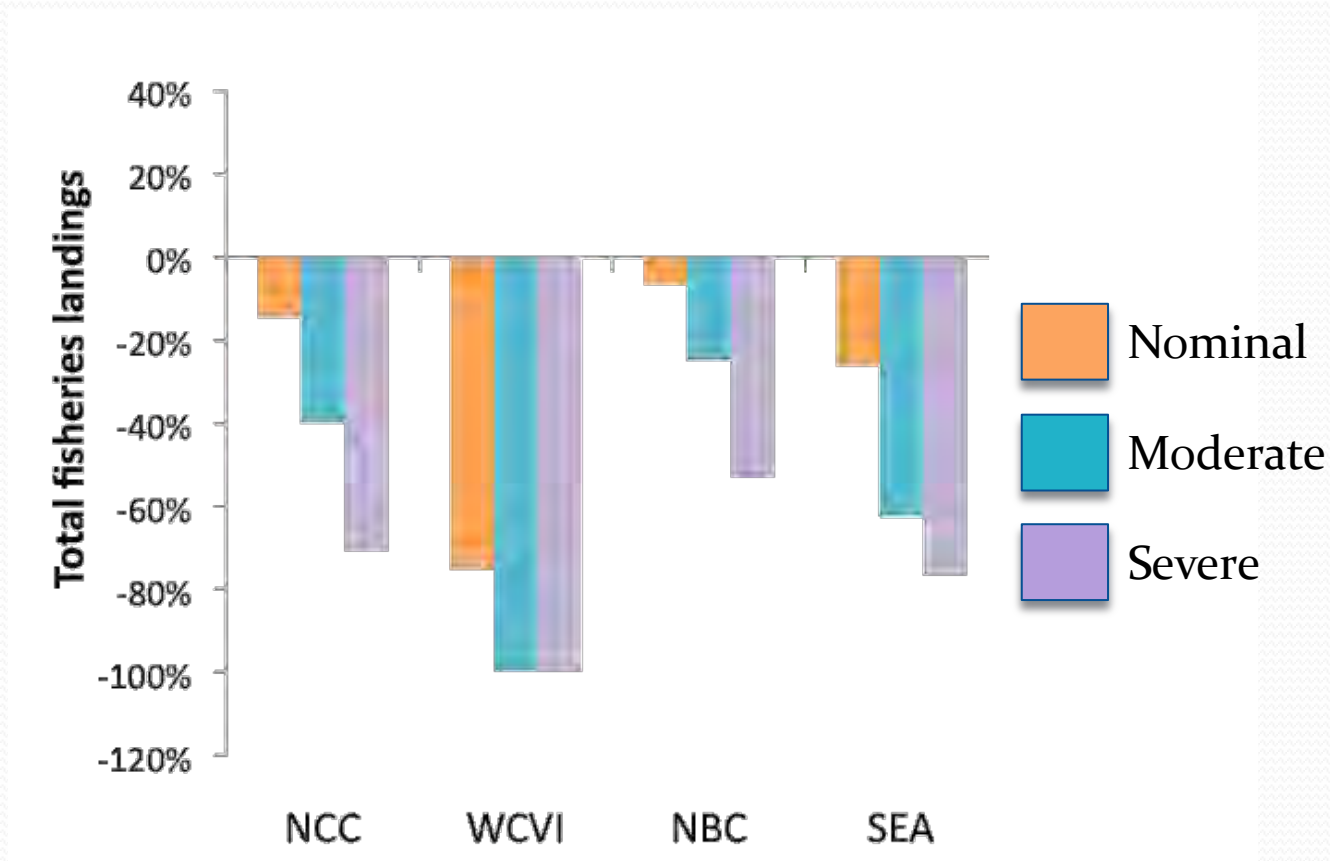
For all impacts, we ran scenarios of 3 different strengths:

<i>Scenario strength</i>		
Nominal	Moderate	Severe
Effect - 50%	Effect	Effect + 50%



Model uncertainty

Models vary in their sensitivity to climate scenarios



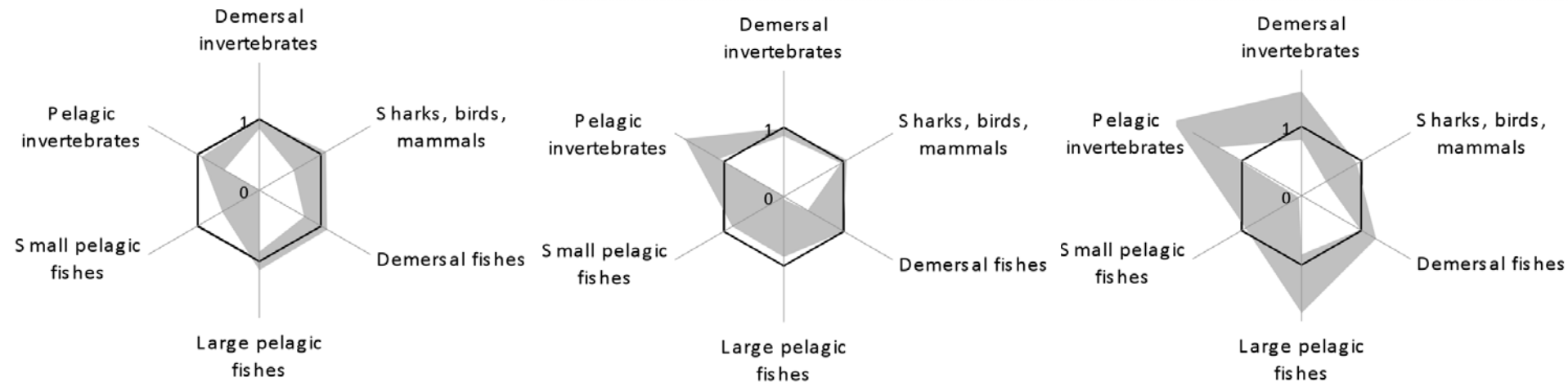
Model uncertainty

Functional groups vary in their sensitivity to climate scenarios

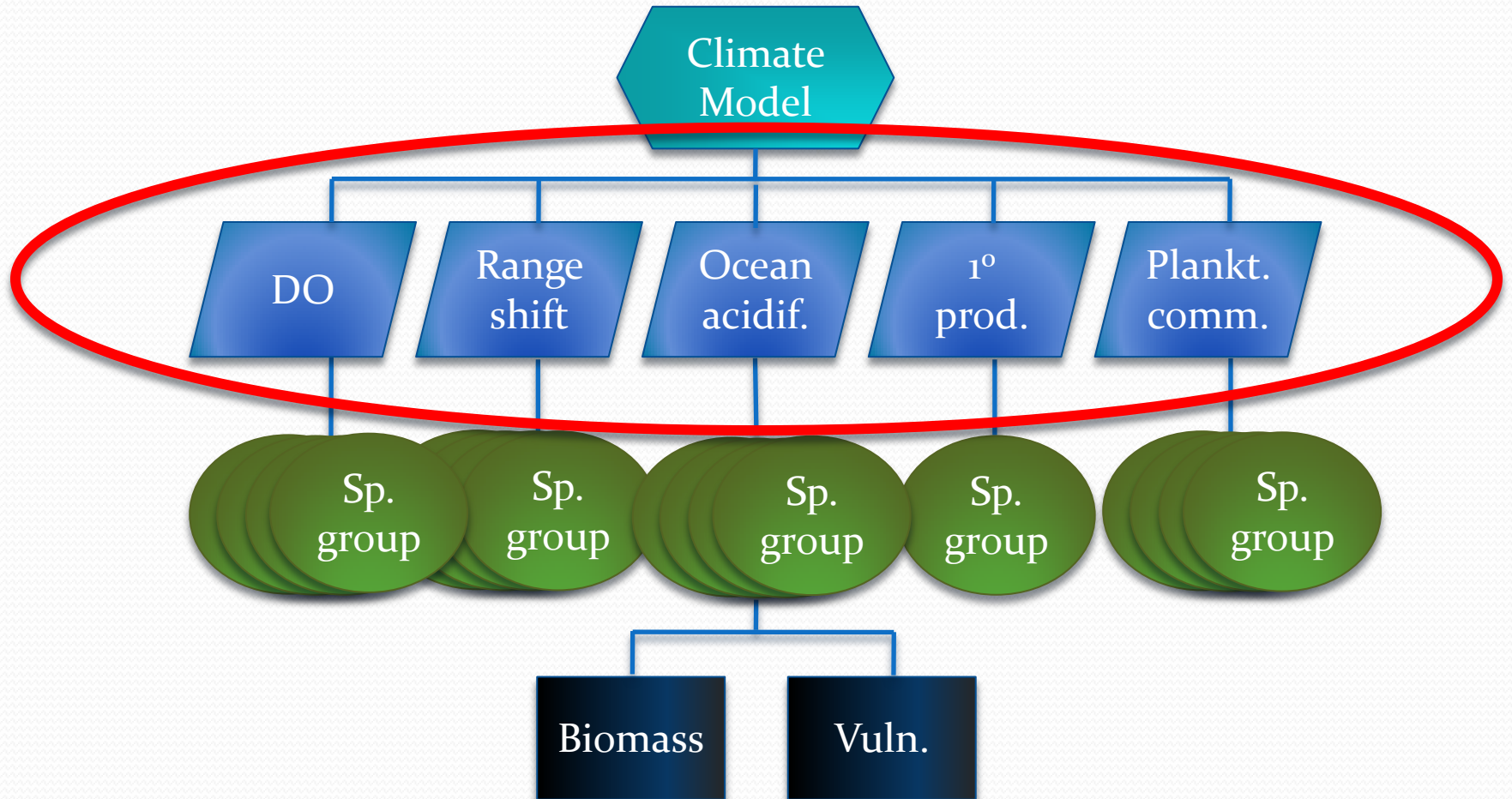
Southeast Alaska

N. British Columbia

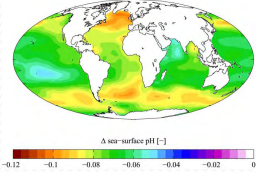
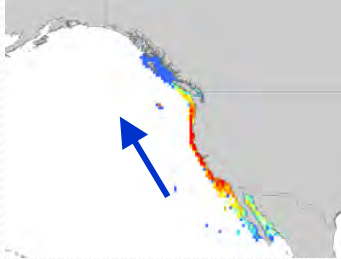

N. Cali. Current



Process uncertainty



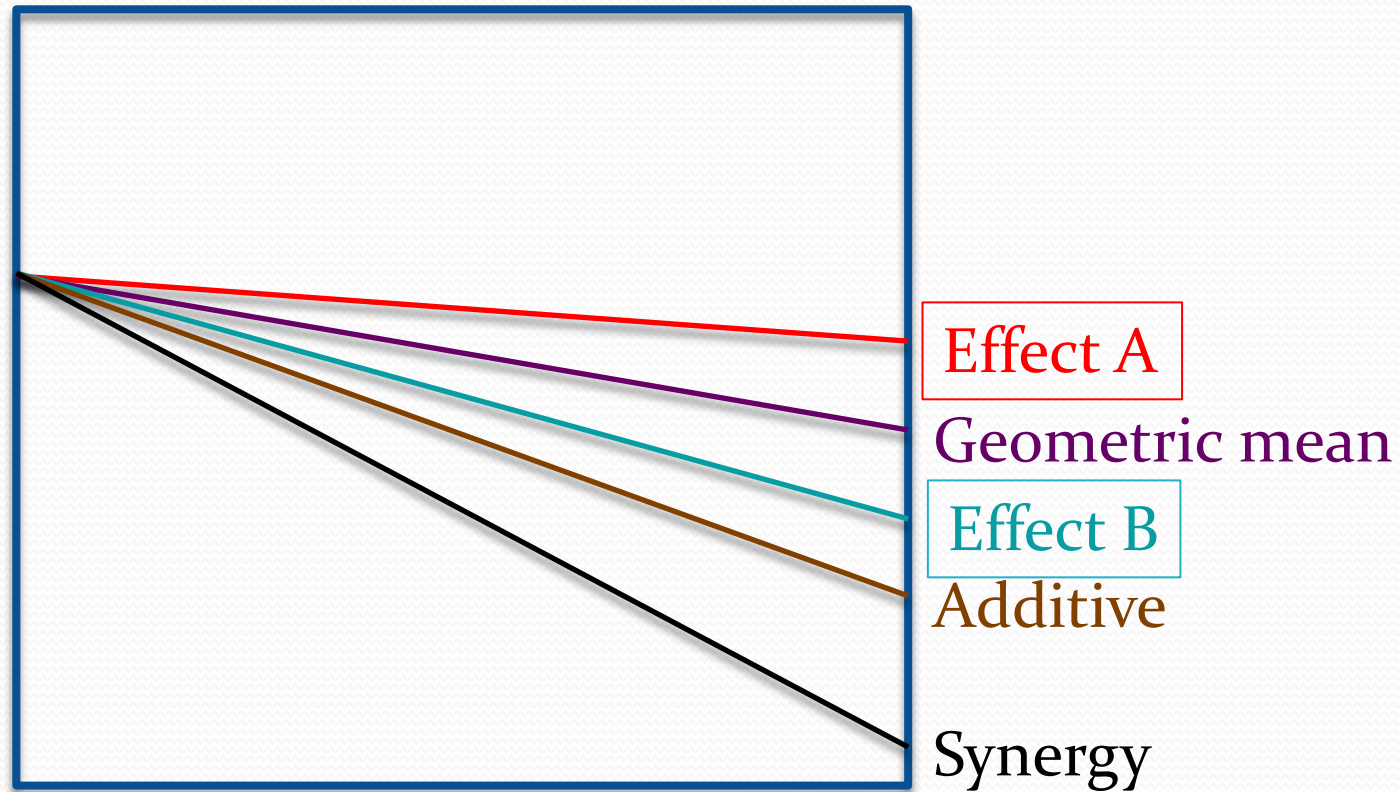
Cumulative effects

Additive =  +  + 

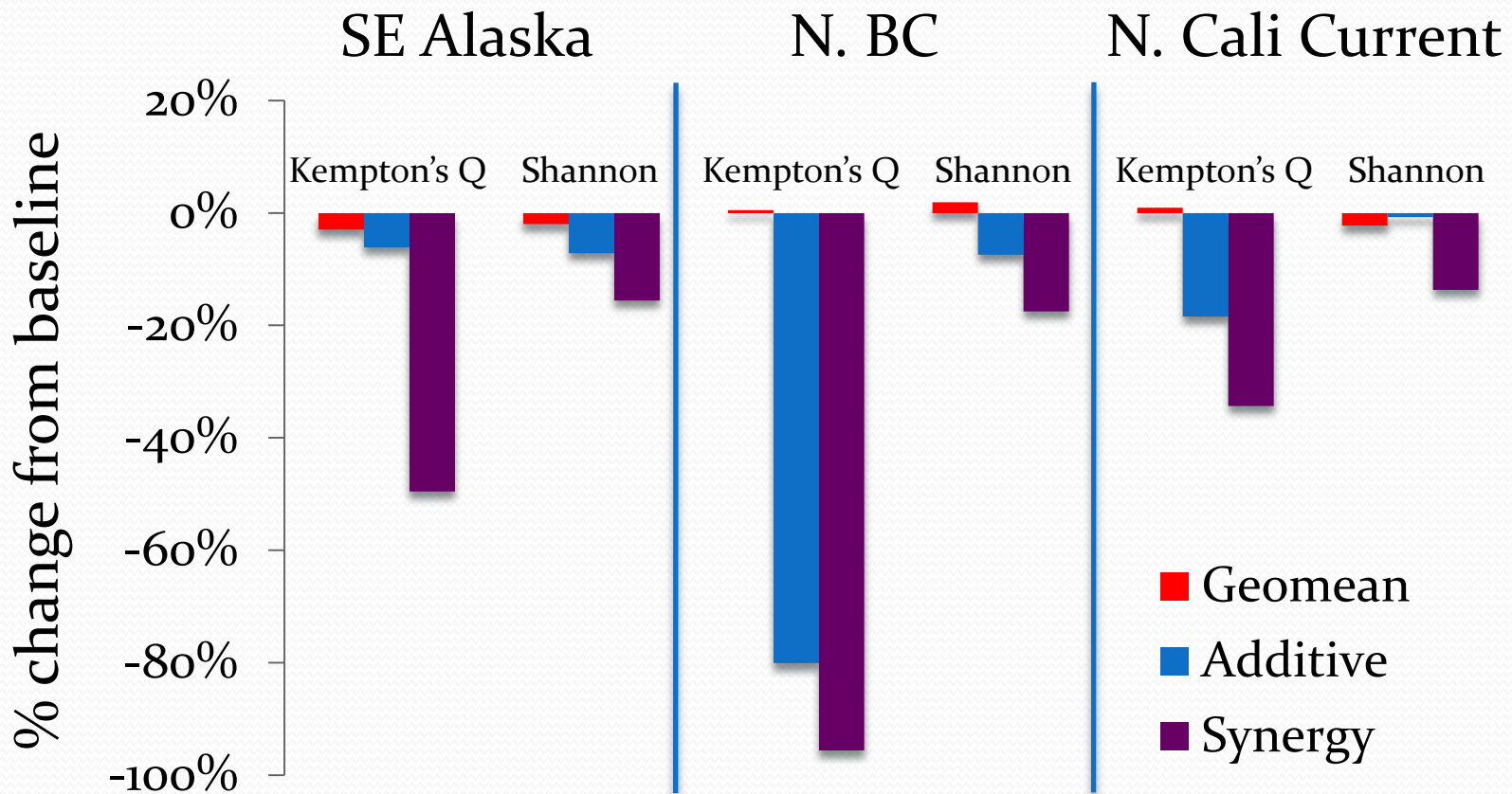
Geomean = $\left(\text{World map} \times \text{Regional map} \times \text{Photograph} \right) \cdot 3$

Synergy = Additive + Geomean

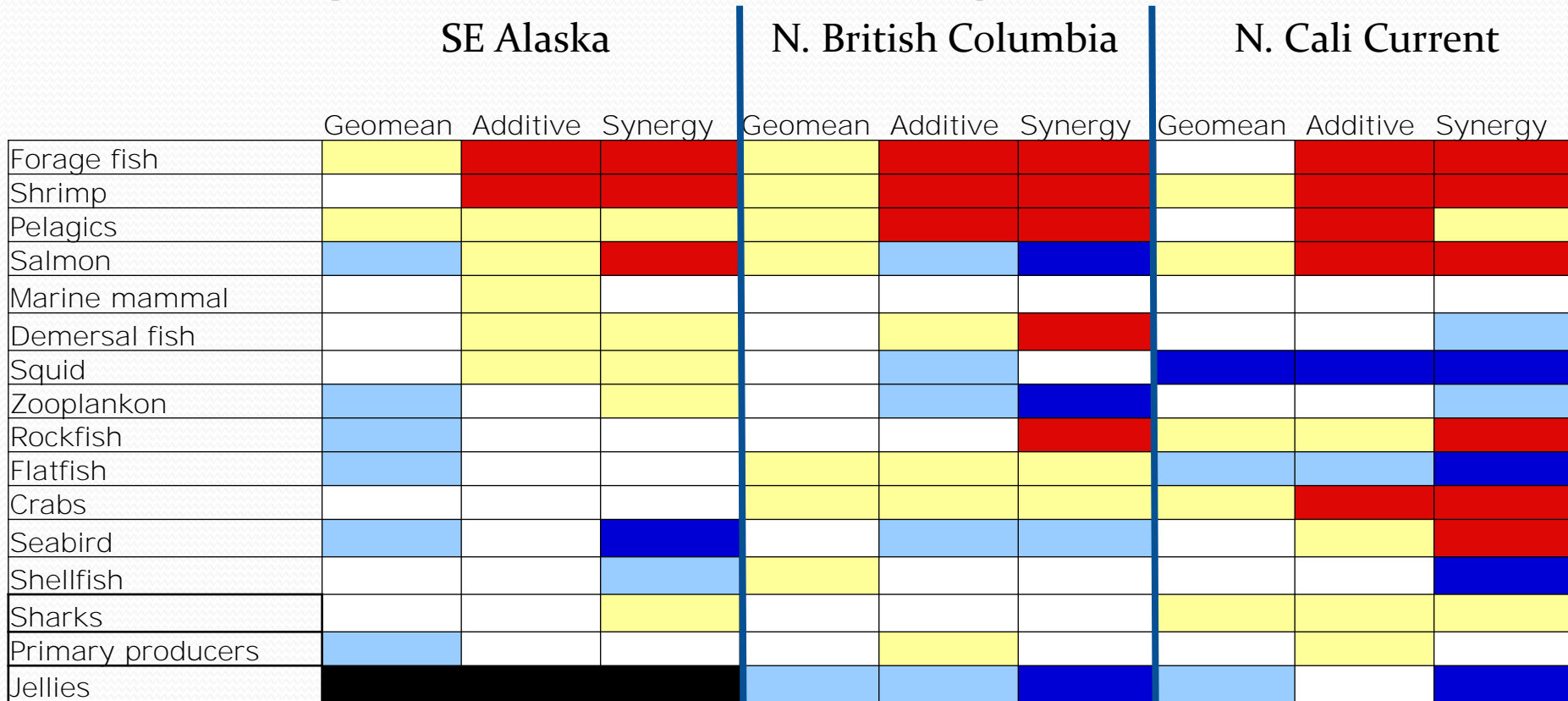
Cumulative effects



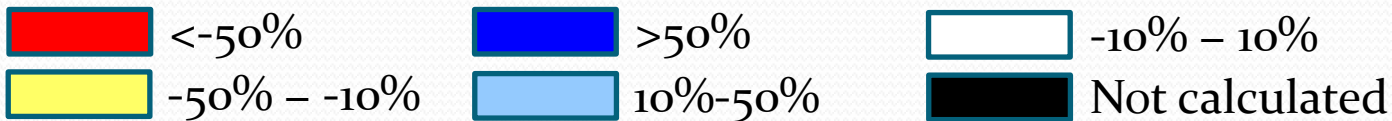
Diversity scores impacted by summation technique



(Dis)agreement among techniques on magnitude of change

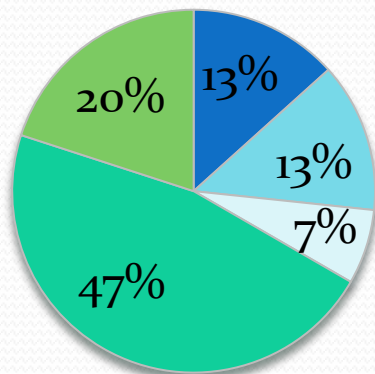


% change from baseline

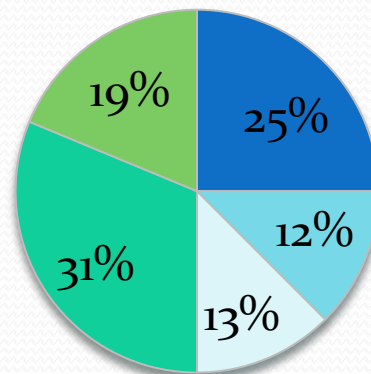


(Dis)agreement among techniques on magnitude of change

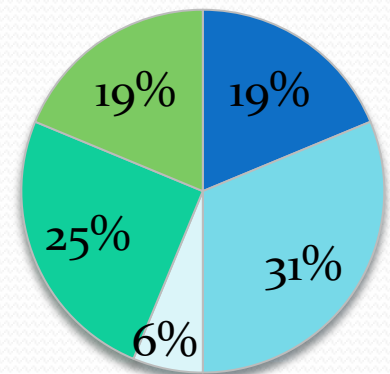
SE Alaska



N. British Columbia



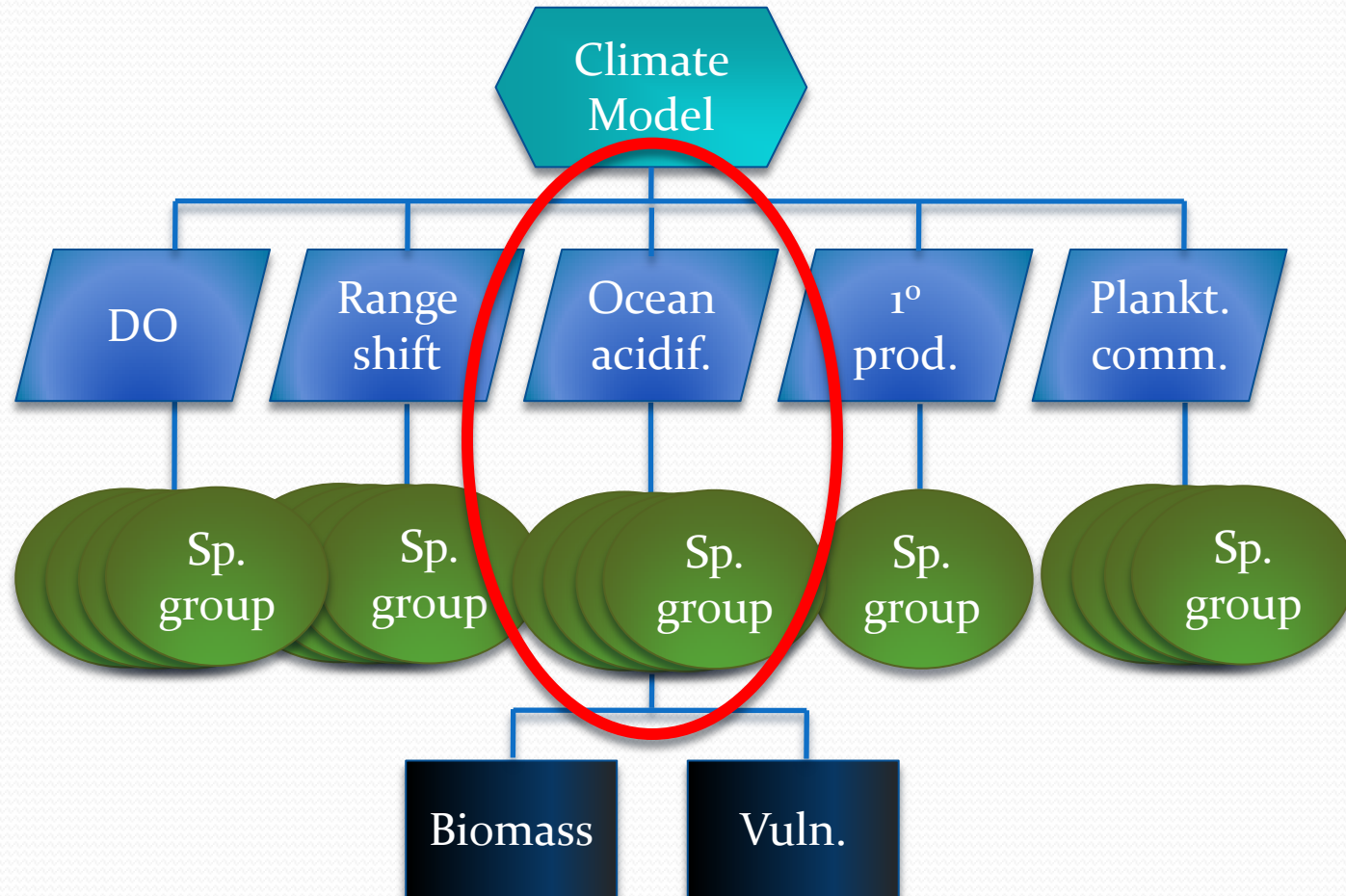
N. Cali. Current



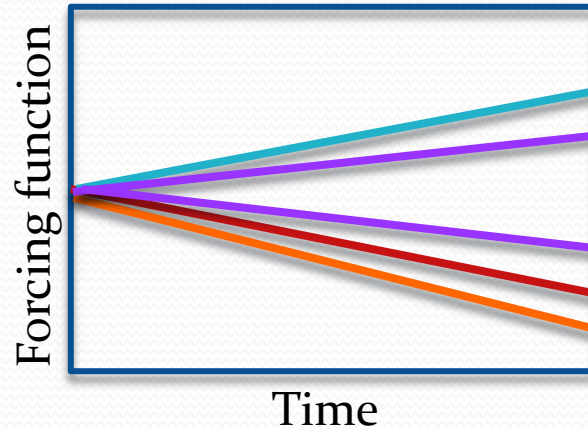
Process uncertainty

- Combinatorial techniques greatly affect magnitude of predicted change
- Agreement among techniques when results ranked is slightly better
- Must consider how well combination techniques capture physiological and population-level process of interest

Effect size uncertainty



Ocean acidification scenarios



Certain

Shrimp

Euphausiids

“Epifauna”

“Infauna”

“Carn. epibenthic inverts”

Uncertain – unidirection

Crabs

Copepods

“Infaunal detrital inverts”

“Epibenthic inverts”

“Carnivorous zooplankton”

“Small zooplankton”

Marine plants

Uncertain – multidirectional

Jellyfish

Phytoplankton

“Microzooplankton”

“Large zooplankton”

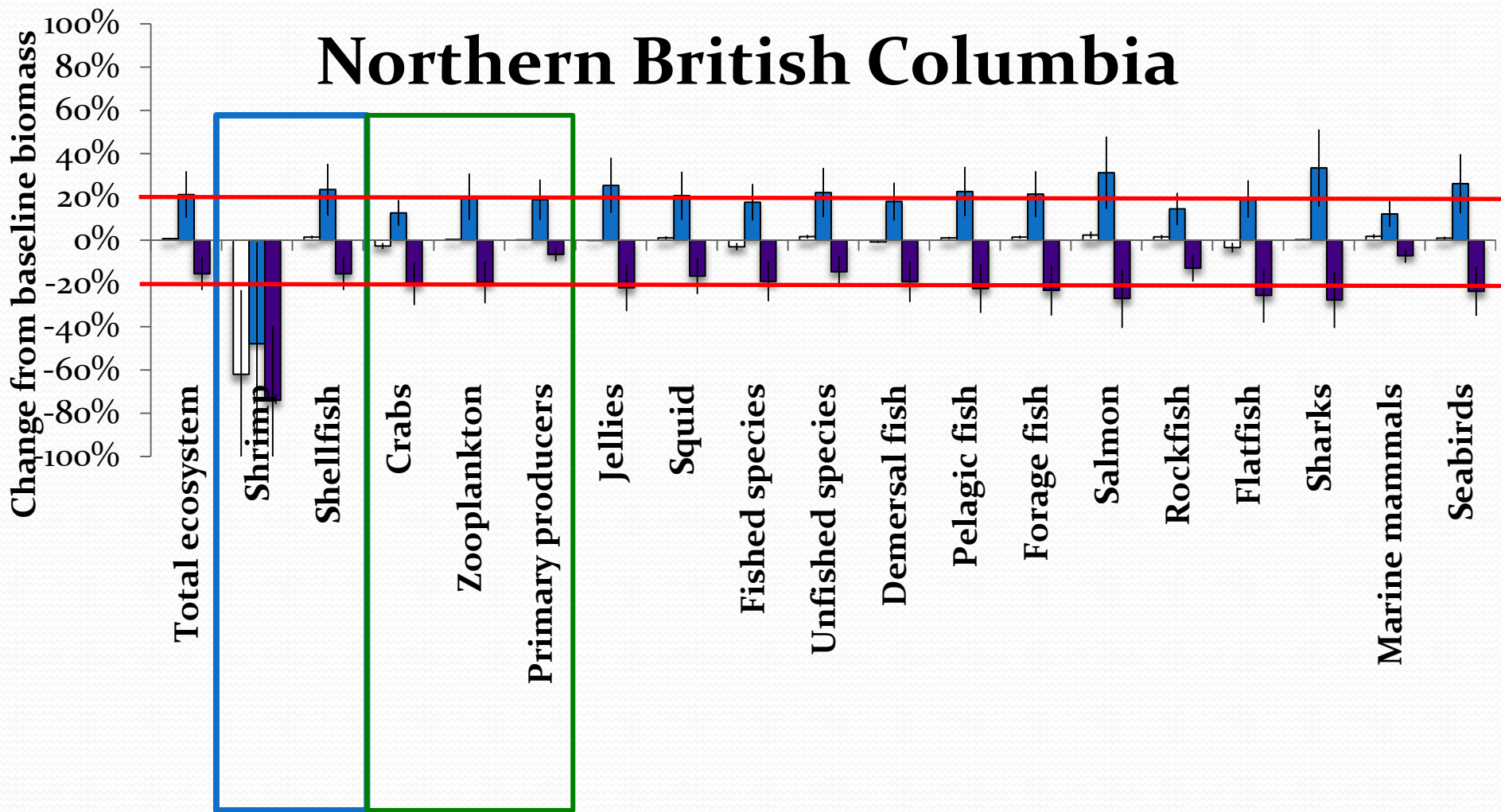
Ocean acidification scenarios

Linear change over 50 yrs.

<i>Effect size</i>	<i>Climate scenario</i>		
	Nominal	Moderate	Severe
Small	5%	10%	15%
Large	25%	50%	75%

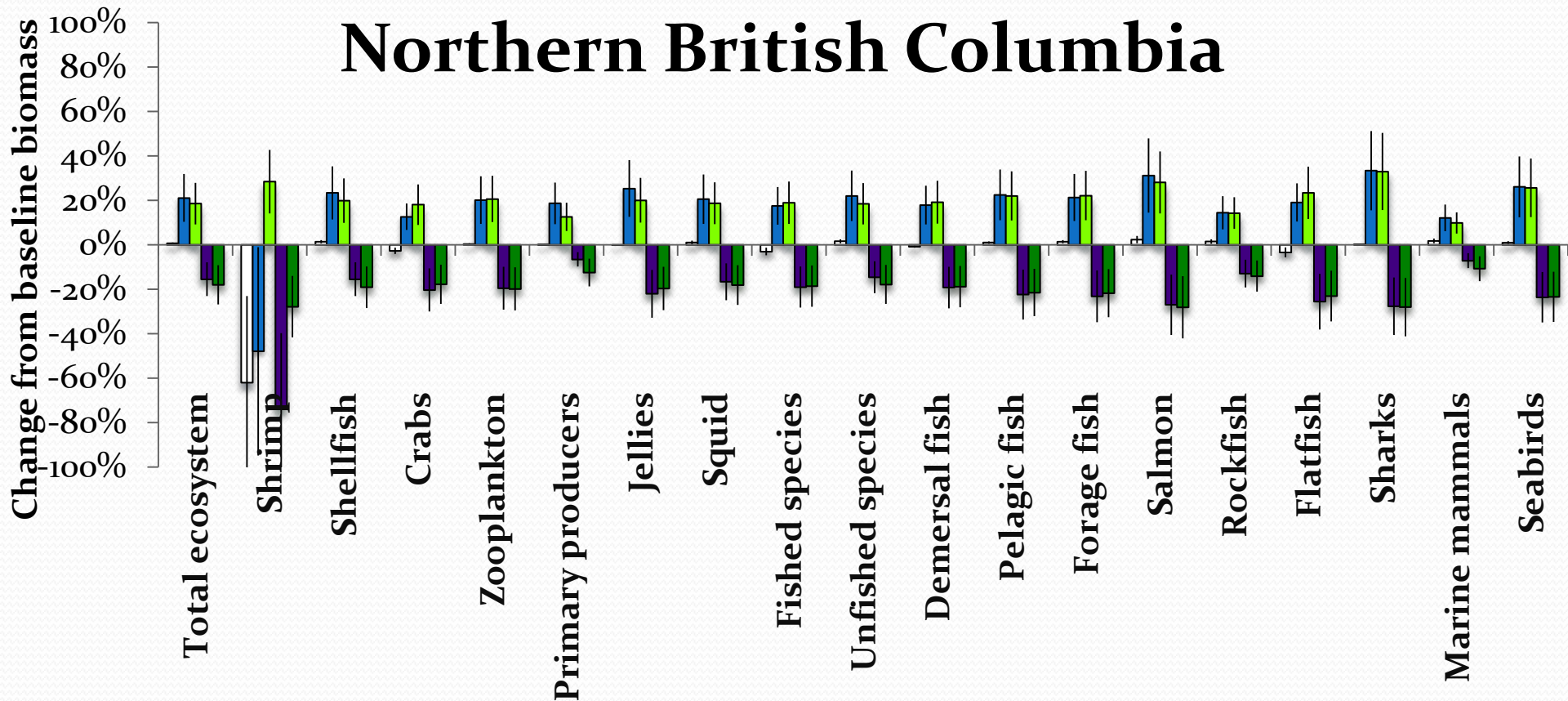
Certain groups
 Uncertain positive
 Uncertain negative

Northern British Columbia





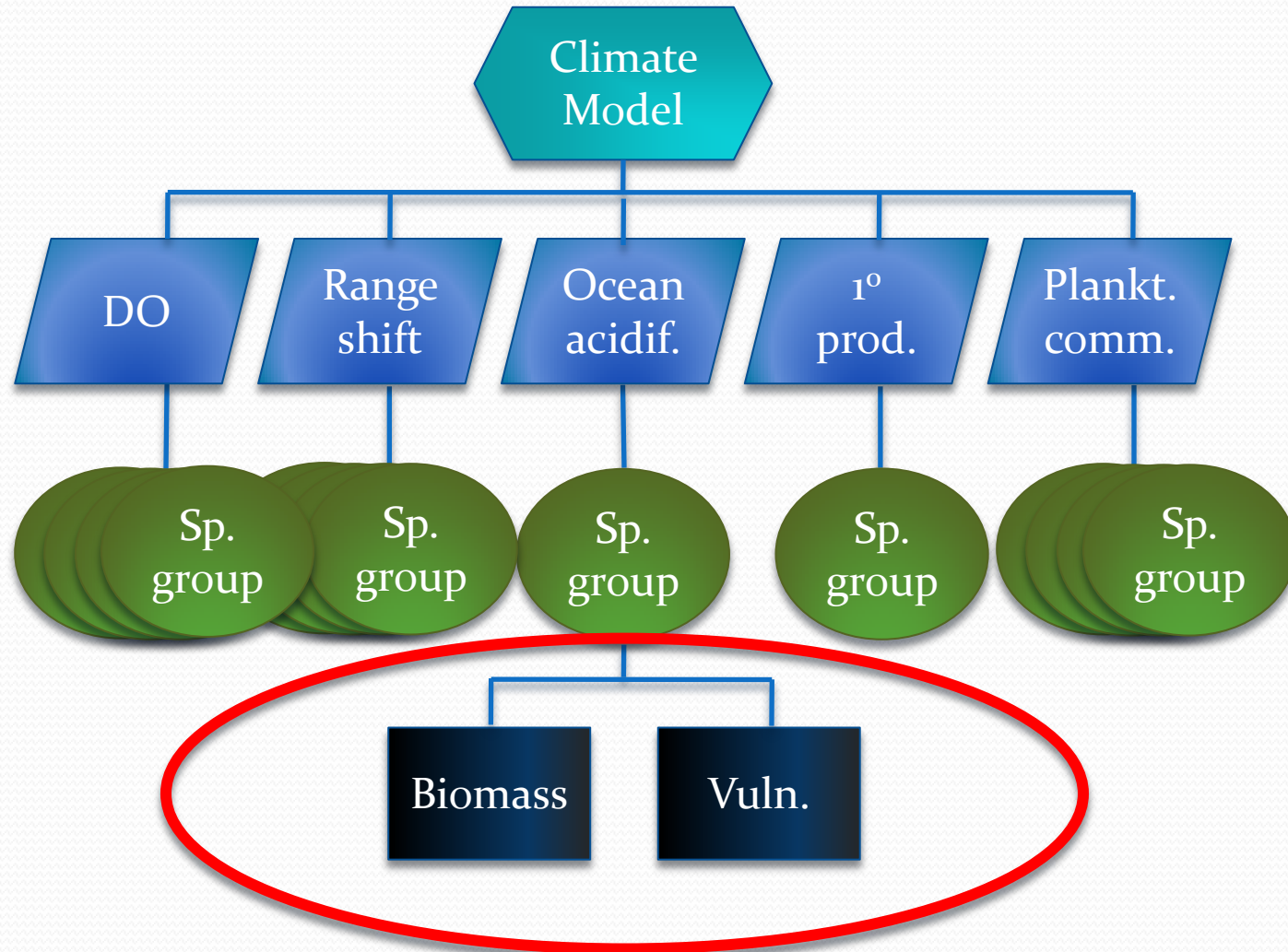
Northern British Columbia



Effect size uncertainty

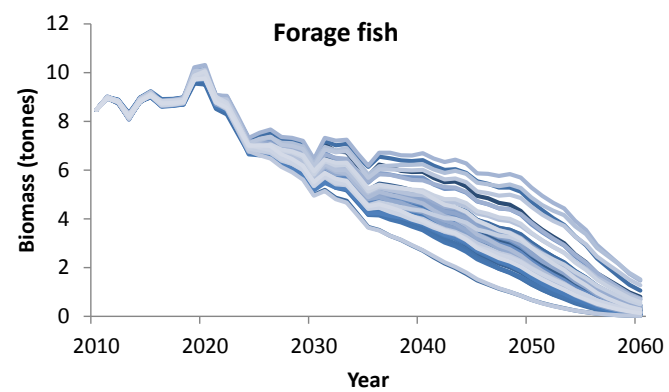
- Importance of understanding climate change impacts on primary producers
- Ecosystem impacts of change in non-primary producers can be dwarfed by change in primary producers

Parameter uncertainty



Monte Carlo methods

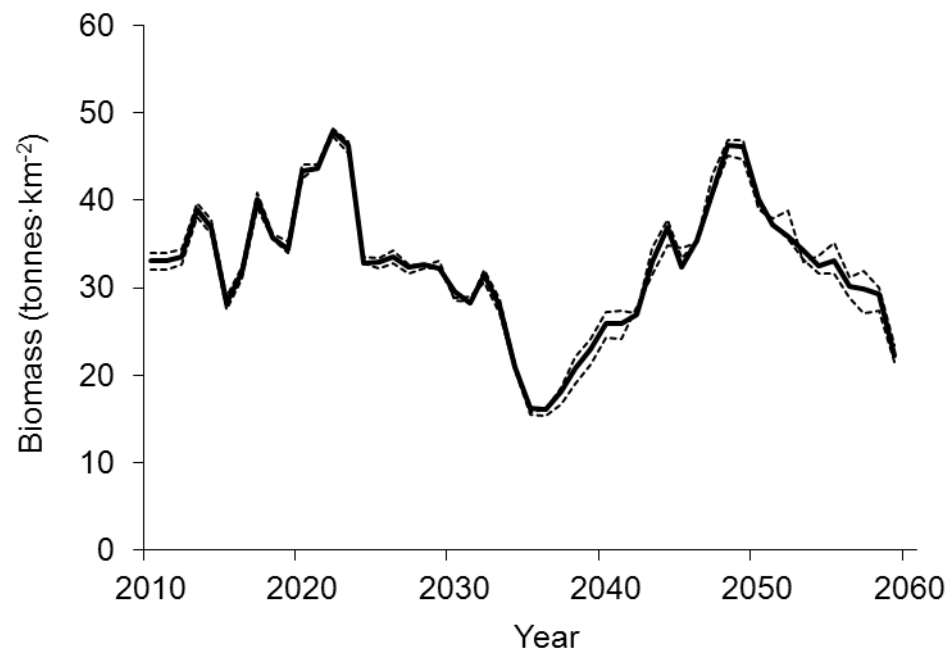
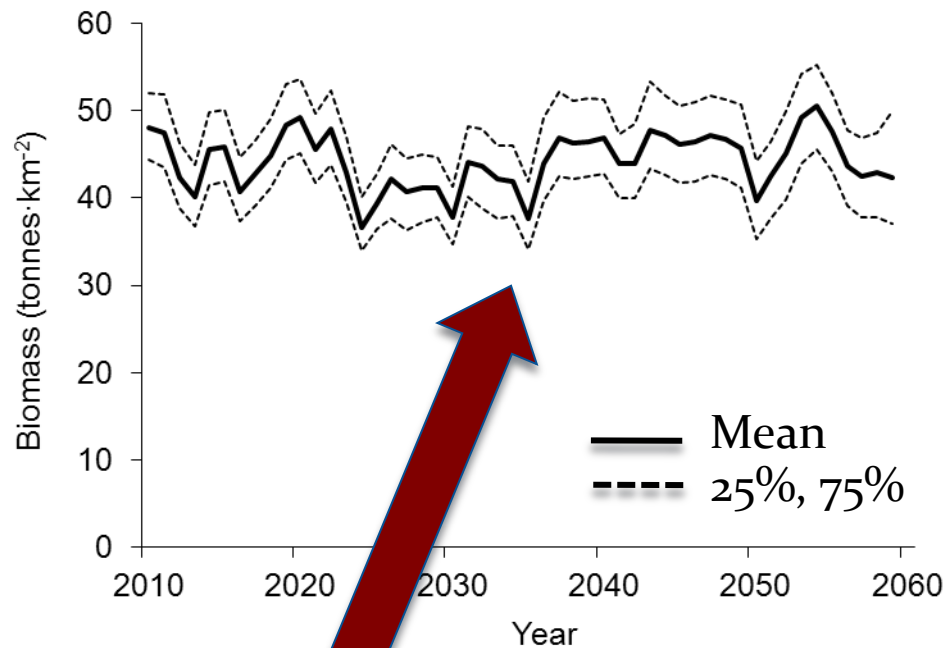
- Biomass
 - Ecopath with Ecosim Monte Carlo routine
 - Coefficient of variation dependent on functional group
 - 15% phytoplankton, zooplankton & benthic meiofauna
 - 5% all others
 - Uniform distribution, 200 trials
- Vulnerability
 - Manually implemented in Ecosim
 - Every predator-prey interaction varied independently
 - Coefficient of variation: 50% all groups
 - Uniform distribution, 200 trials
- Results: 1) functional group, 2) aggregated functional group



Monte Carlo results

Biomass Monte Carlo:
Southeast Alaska, benthic invertebrates

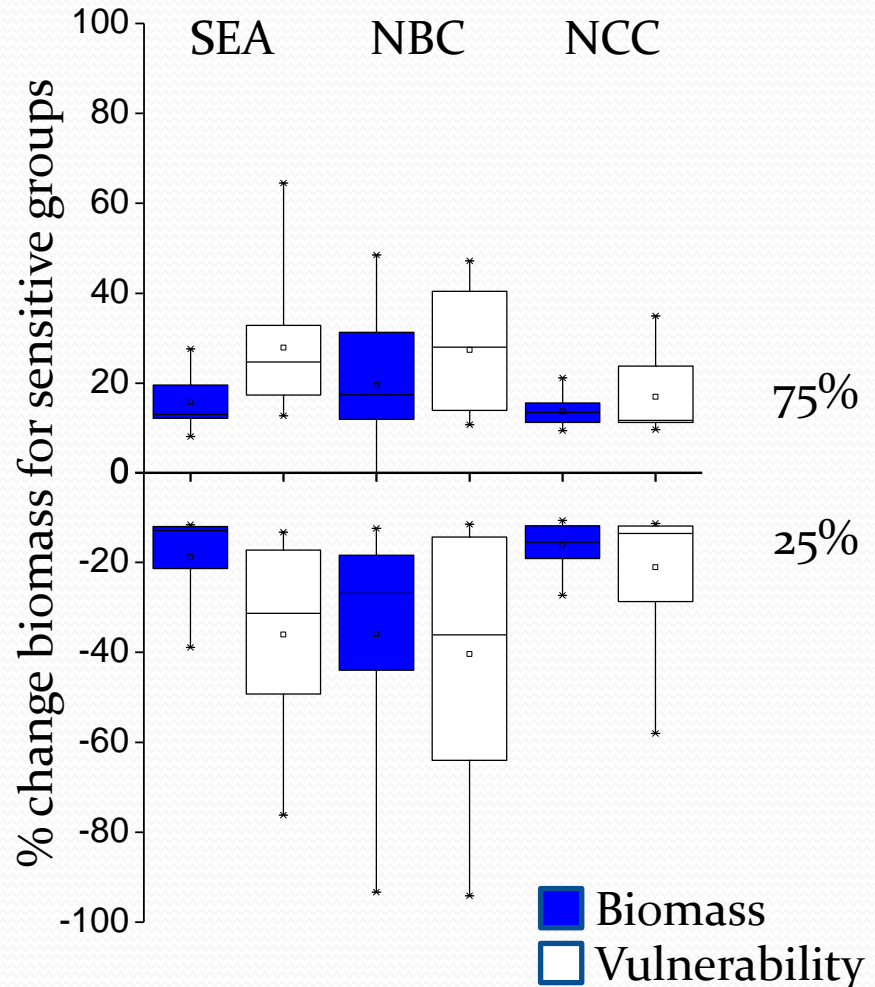
Biomass Monte Carlo:
N. Cali Current, small pelagic fish



Compounding impacts of trophic dynamics

Monte Carlo results

- About 30% of the functional groups and aggregated functional groups have >10% change in biomass
- Biomass change for vulnerability Monte Carlos less than forced variation
 - Magnitude of change
 - Number of groups



Sensitive functional groups: biomass change >10%

Functional groups

All models

Salmon

Pacific Ocean perch

Flatfish

Halibut

Crabs

Shrimp

Epifaunal inverts

Infaunal detritivores

2 models

Sharks

Forage fish

Herring

Rockfish

Pacific cod

Arrowtooth

Phytoplankton

Aggregated groups

All models

Benthic inverts

2 models

Large and small pelagics

Birds

Flatfish

1 model

Elasmobranches



Parameter uncertainty

- Knowing the biomass and vulnerability of some species groups matters more than others
- Sensitivity to biomass estimates is greater than to vulnerability estimates
- Sensitivity of specific functional groups to parameter uncertainty is fairly well captured in aggregated groups

Conclusions

- Modeling ecosystem impacts of multiple climate change impacts is messy
- Understanding how uncertainty impacts our results can help us target assumptions and data inputs that may matter most
- Recognize that modeling exercises tell us the **type of changes** we may expect, not the magnitude or the specific effects