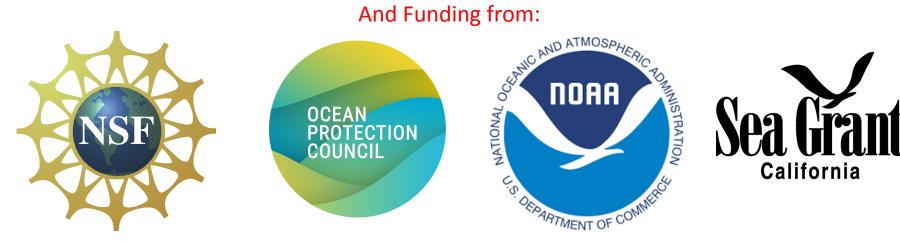
Climate change and aerobic habitat in the California Current System

Curtis Deutsch University of Washington

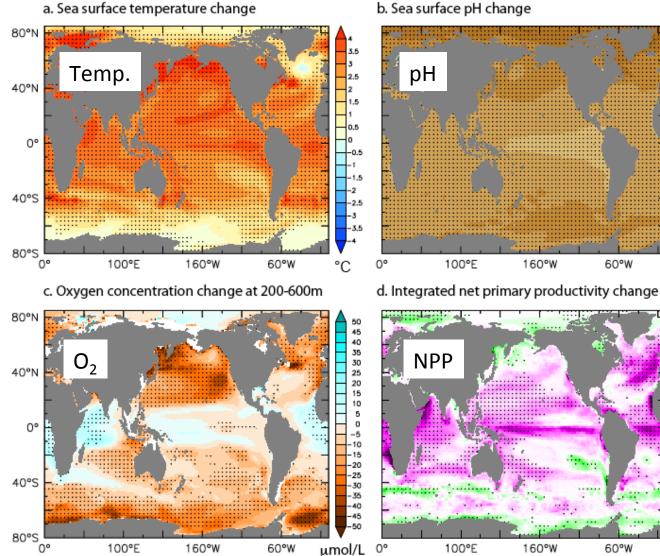
In Collaboration with:

Evan Howard, Jim McWilliams, Daniele Bianchi, Faycal Kessouri, Lionel Renault

And Funding from:



Global Ocean Climate Change



0.5 0.45 0.4

60°W

60°W

0.35

0.3

0.25

0.2 0.15 0.1 0.05

0 0.05

> 0.1 0.15 -0.2 0.25

0.3 -0.35

0.4 -0.45

200 180

160

140

120

100

80 60 40

20 0

-20 -40

-60 -80

100

120

140

160

180 200

gC/m2/v

North Pacific a region of strong ocean change.

However:

Coastal oceans poorly resolved in Global Earth System Models.

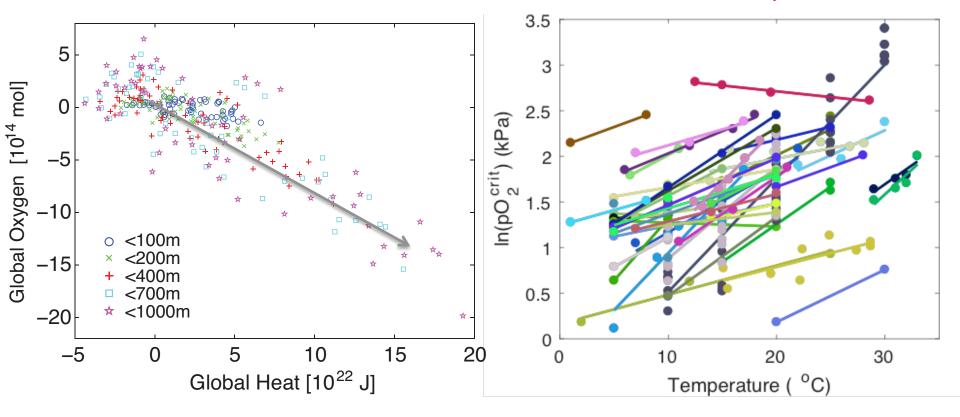
They disagree on magnitude and/or direction of change in key ecosystem properties, NPP and O_2 .

Bopp et al. [2013]

Temperature and O₂

Ocean O₂ content decreases with Temperature

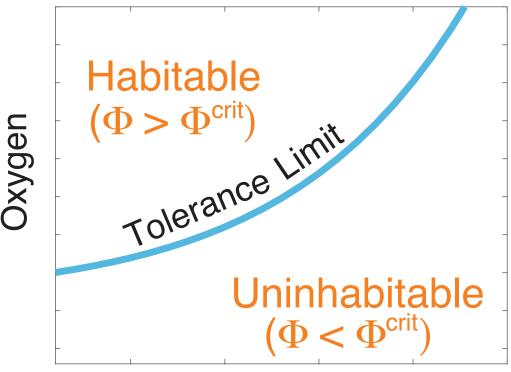
Organism O₂ requirement rises with Temperature



Ito et al. [2017]

Penn et al. [2018]

Habitat: A Metabolic Index



Temperature

The Metabolic Index, defined as ratio of potential O_2 supply to resting O_2 demand by organism:

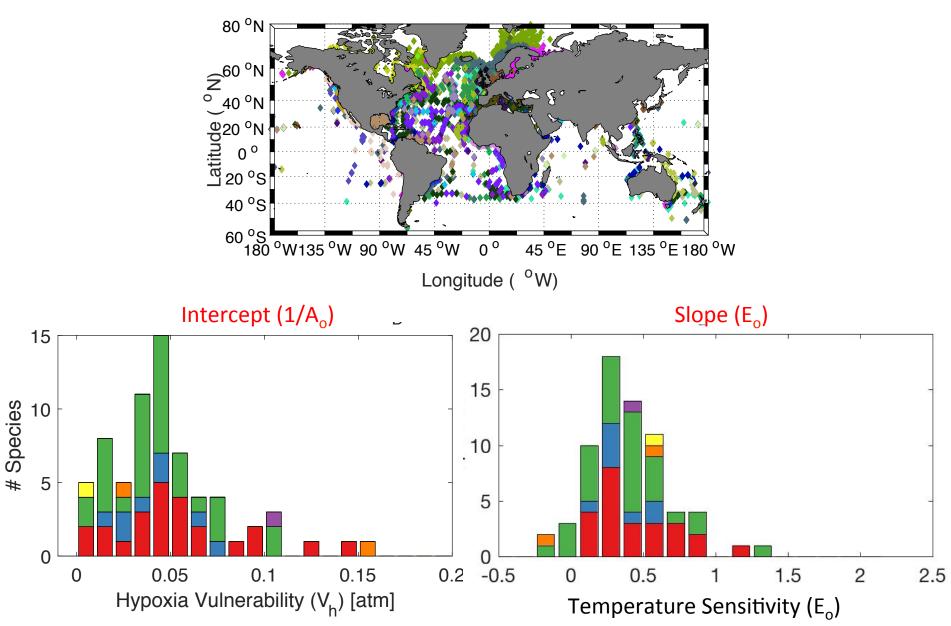
 $\Phi = A_o B^{\varepsilon} p O_2 \exp(E_o / k_B T)$

Environmental parameters: Oxygen pressure (pO₂) Temperature (T)

Physiological parameters: Oxygen sensitivity -Height of curve (1/A_o) Temperature sensitivity -Slope of curve (E_o)

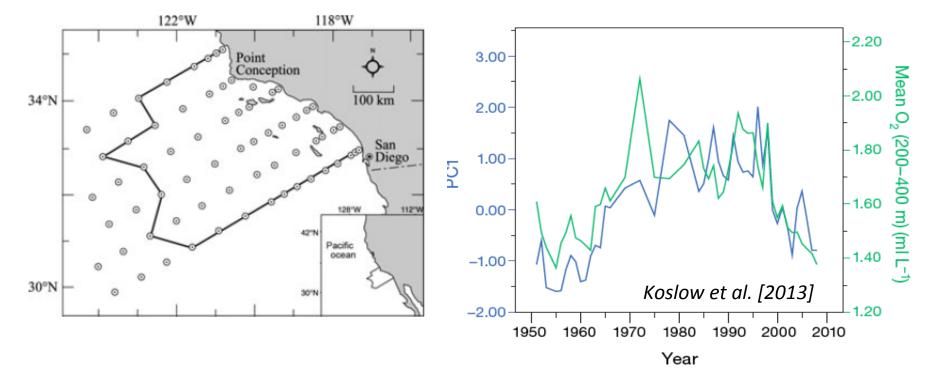
Deutsch et al. [2015]

Global Trait Diversity



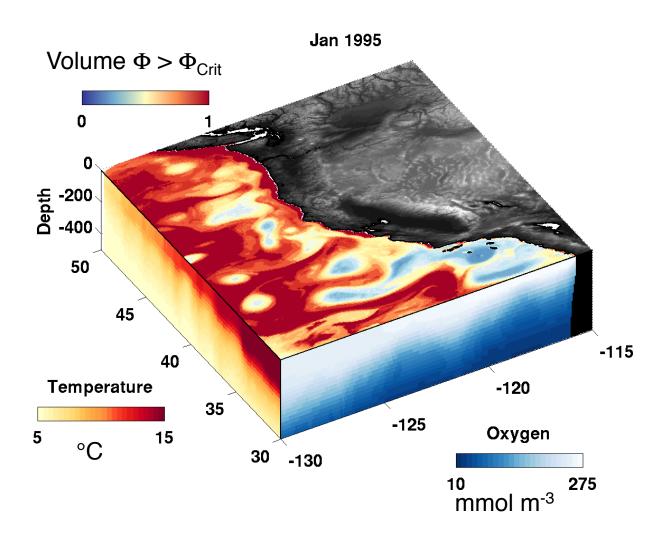
Historical Variability

In the Southern California Current, decadal changes in small pelagic fish abundance are correlated with regional O₂ changes.



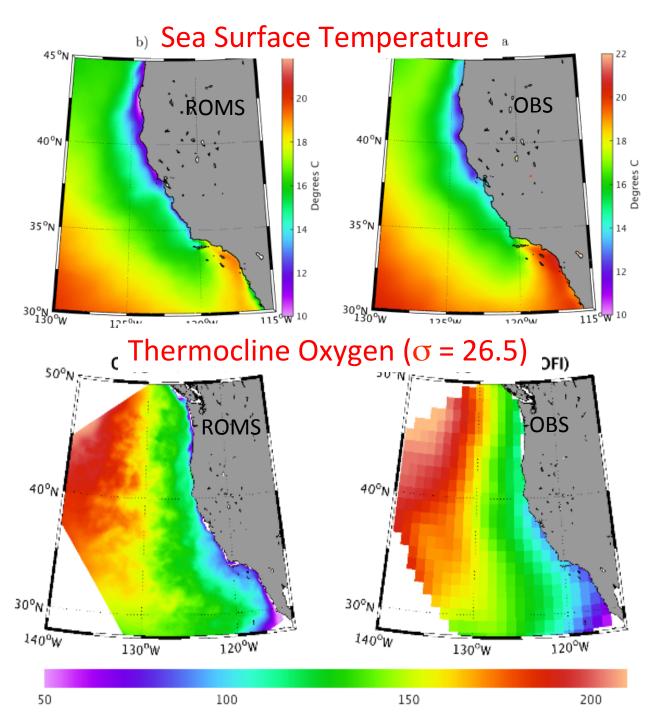
None of these species has measured Metabolic Index traits. Strategy: Determine species for which habitat boundaries and time variability can be explained by same set of traits (A_o, E_o).

Dynamic Habitat



How has (and will) climate change cause aerobic habitat change and associated shifts in species geographic range or abundance?

These questions require models that resolve coastal processes with high fidelity to (often sparse) observations.



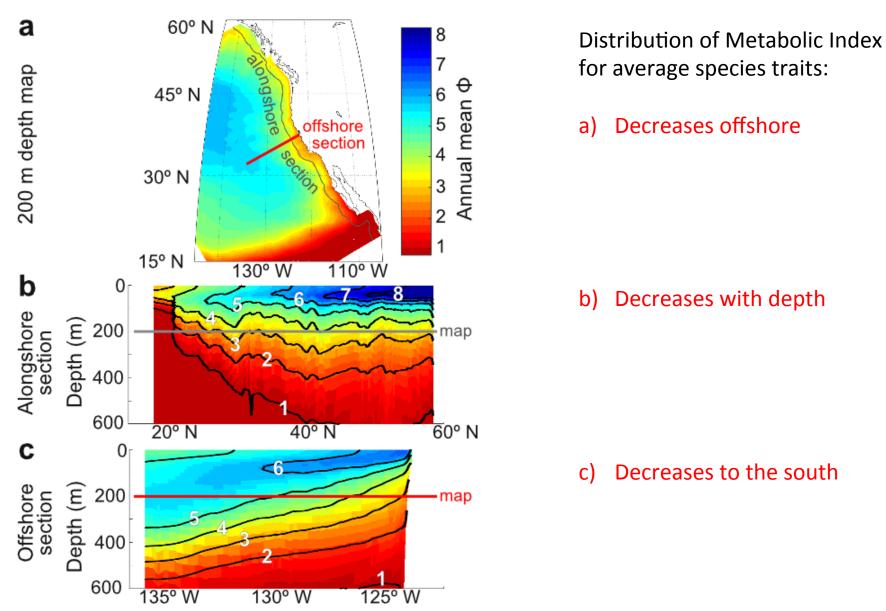
Model Validation

Model physical solution validated against historical hydrographic and satellite observations: Temp, Salinity, SSH, Currents, Eddy energy, nutrients, NPP, oxygen, CO2 parameters

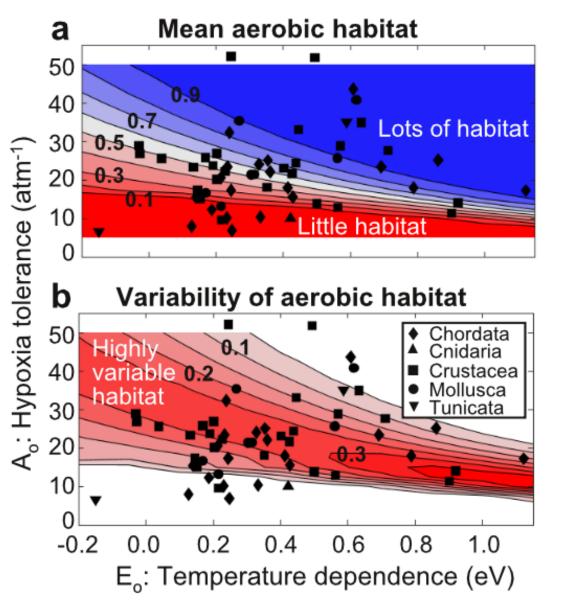
Biases generally low, particularly for well sampled fields needed for Metabolic Index.

Renault et al. [in review] Deutsch et al. [in review]

CCS Metabolic Index

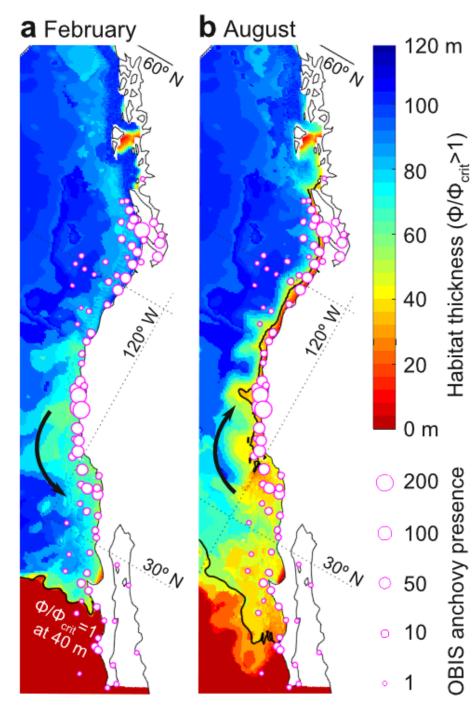


Aerobic Habitat



Most species in CCS likely experience strong variability in aerobic habitat.

Can we see this reflected in species abundance?

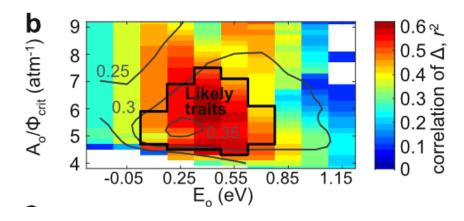


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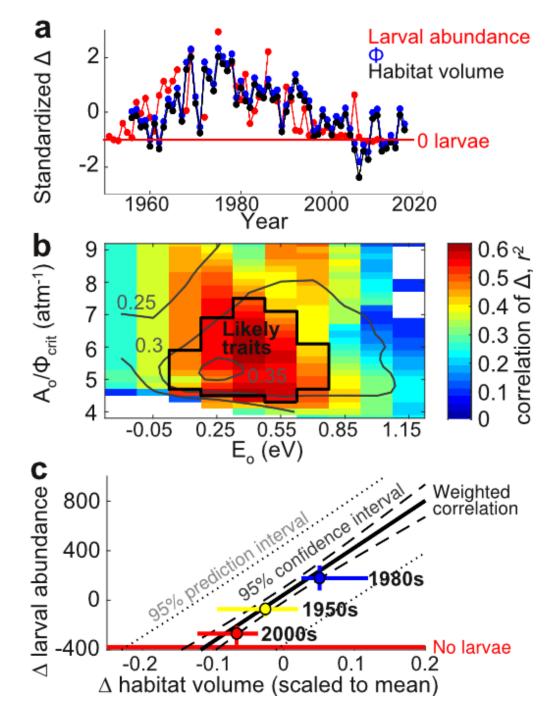
anchov

OBIS

Historical Anchovy distribution



Historical variation of anchovy abundance strongly correlated to aerobic habitat fluctuations from decadal climate variability.

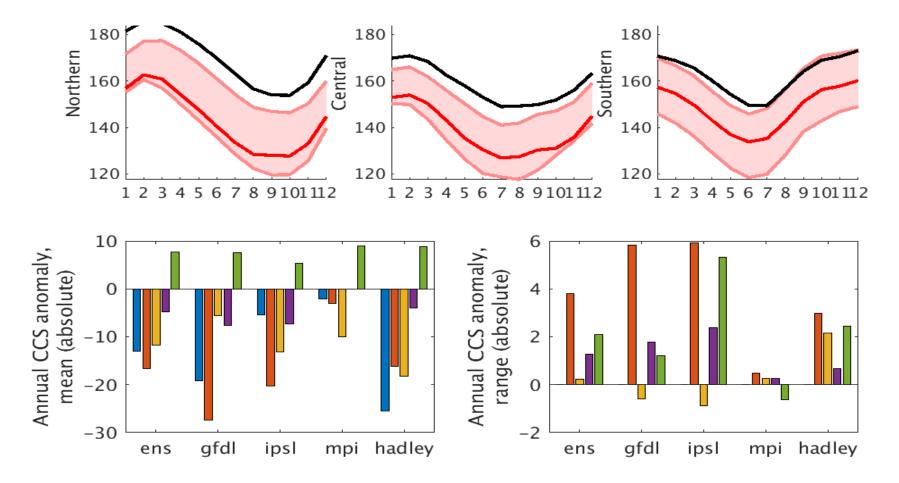


Decadal Anchovy variability

Historical variation of anchovy larval abundance in southern CCS strongly correlated to aerobic habitat fluctuations from decadal climate variability.

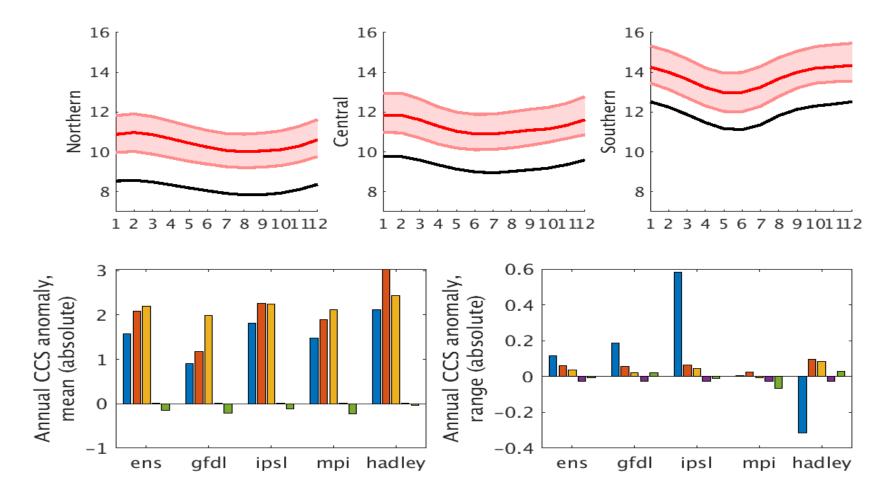
Oxygen loss in the CCS

Climate models show large and consistent reductions of O₂ globally and in the CCS.
 Dynamical downscaling with ROMS suggest they are even underestimated.
 The role of winds opposes warming trend, but is small compared to remote effects.

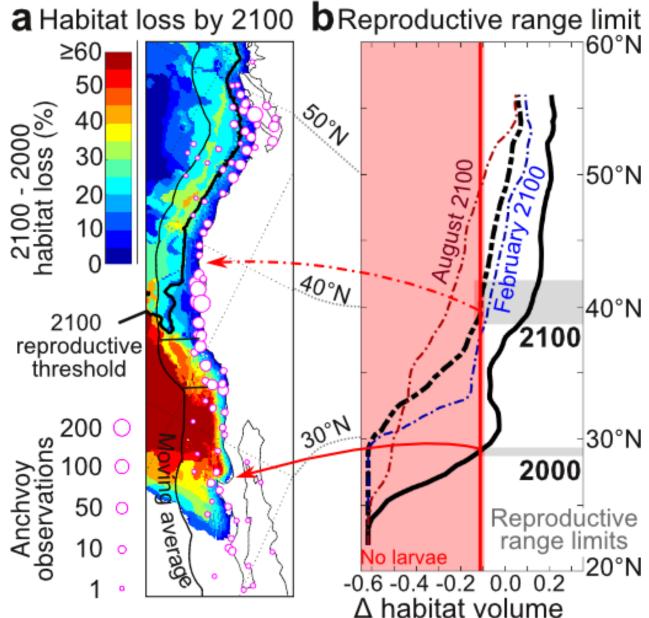


Warming of the CCS

Warming in upper 200m consistent across models. Magnitude from Stratification forcing is slightly reduced by local wind changes.



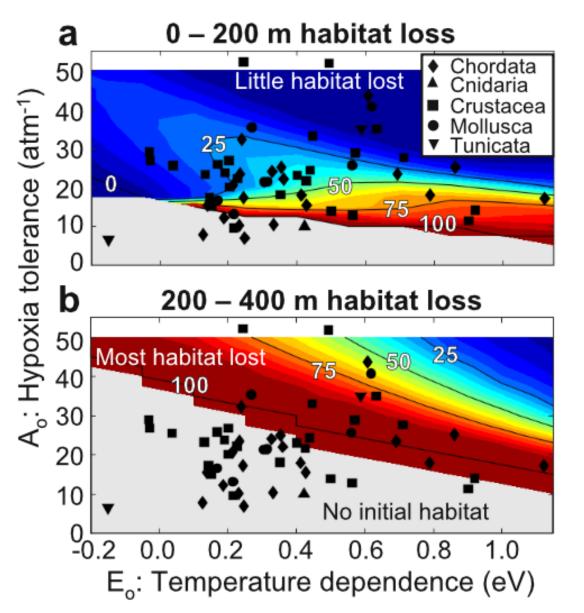
Geographic Range Contraction



Based on historical observations, projected warming and O_2 loss in the California Current System will lead to extirpation of anchovy throughout the southern 1/3 of their range.

Howard et al. [in review]

Ecosystem Implications



Future warming and O₂ loss in the California Current (projected by ROMS downscaling) implies large losses of habitat, but strongly dependent on species traits and habitat depth.

This is likely to reorganize species interactions and food webs.

Outlook

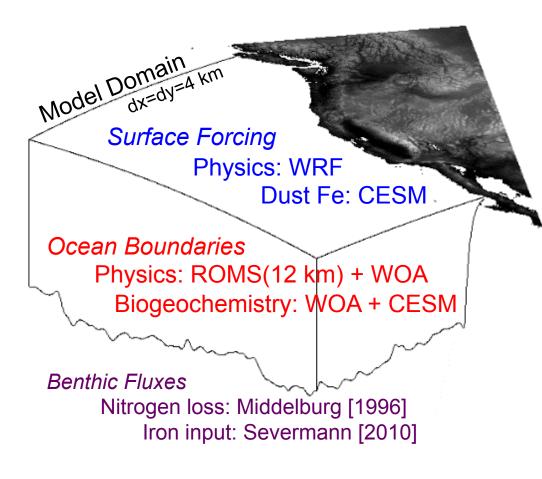
- New project to measure the Metabolic Index parameters for >10 CCS species
- Collaboration with Brad Seibel (USF) and Martha Sutula (SCCWRP)
- Funding from NOAA and CA SeaGrant
- Strong connection to management agencies via SCCWRP

Conclusions

- 1) Historical reconstructions of the California Current with eddyresolving hindcast model reproduce long-term mean and interannual variability in key ecosystem variables.
- 2) Downscaling projections of future climate change with global Earth System Models show large-scale wind changes to have a modest impact on bulk ecosystem metrics, including NPP. Caveat: potential changes in coastal wind drop-off?
- 3) Climate forcing of sub-surface ocean properties, e.g. Oxygen and pH (not shown) derive primarily from basin-scale impacts on stratification and properties imported into CCS.
- 4) Intensification of temperature-dependent hypoxia explains large past changes in Northern Anchovy, and appears to be largest perturbation to CCS ecosystem, with large but differential impacts on species habitability.

EXTRA SLIDES

Historical Hindcast & Validation



Regional Ocean Modeling System (ROMS)

- Eddy-resolving ocean model
- 2-decade Hindcast 1994-2013

Forcing

- Atmospheric model (WRF) 6km atmospheric winds, radiation, E-P
- Physical boundary conditions from global reanalysis (Mercator)
- Biogeochemical boundary conditions from climatology

Validation

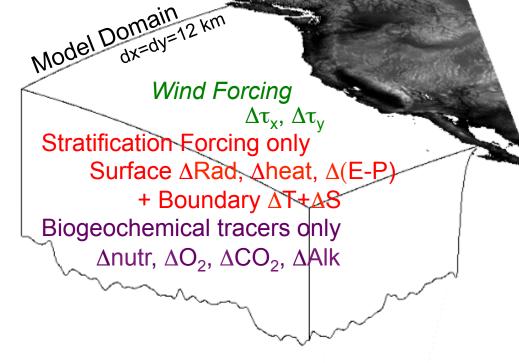
- Historical hydrography
- Satellite remote sensing

Ocean ecosystem model (BEC)

- 3 phytoplankton, 1 zooplankton
- Biogeochemical cycles of C, O₂, P, N, Fe, Si

Future Projection & Attribution

By isolating responses to distinct climate forcings (wind, warming, chemistry), we can attribute the leading causes of CCS ecosystem change



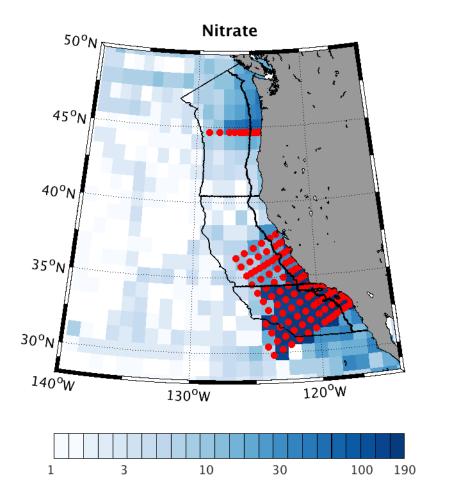
Howard, Frenzel, Deutsch, et al. [in prep]

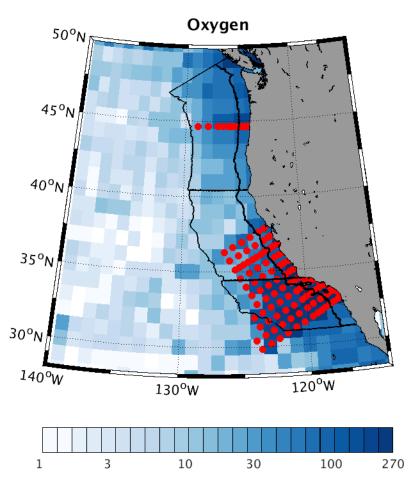
Downscale Global Model projections by applying anomalies at surface and open boundaries, based on Earth System Models

5 Global Models (CMIP5): GFDL, MPI, IPSL, Hadley, NCAR, + Model mean Attribution experiments: Stratification only (S) Winds only (W) BGC tracers only (B) All forcings (A)

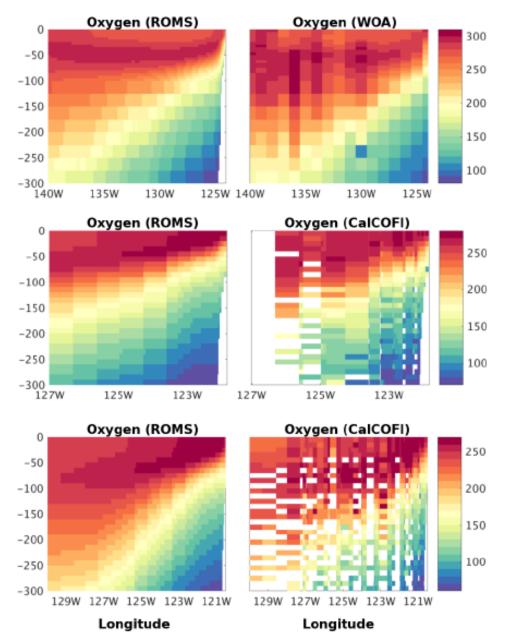
Historical Validation

We compare model solutions to mean and variability of historical hydrographic profiles. The California Current is far less under-sampled than most of the ocean.



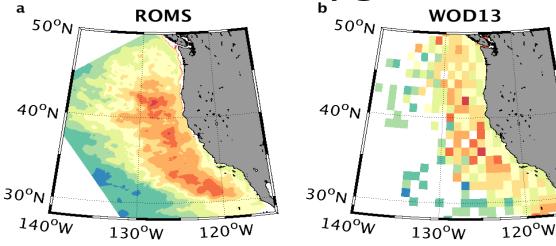


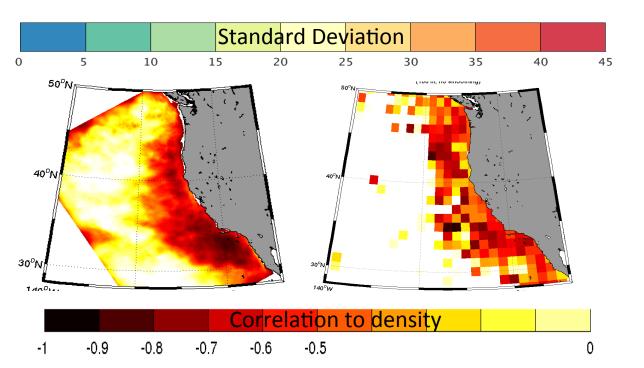
Oxygen and Upwelling



Vertical sections of annual mean O_2 , from ROMS (right columns) and WOD (right columns) along three repeat hydrography lines that span the latitude range of the CCS reveal the role of upwelling in bringing low O_2 into shallow environments on the continental shelf.

Oxygen Variability



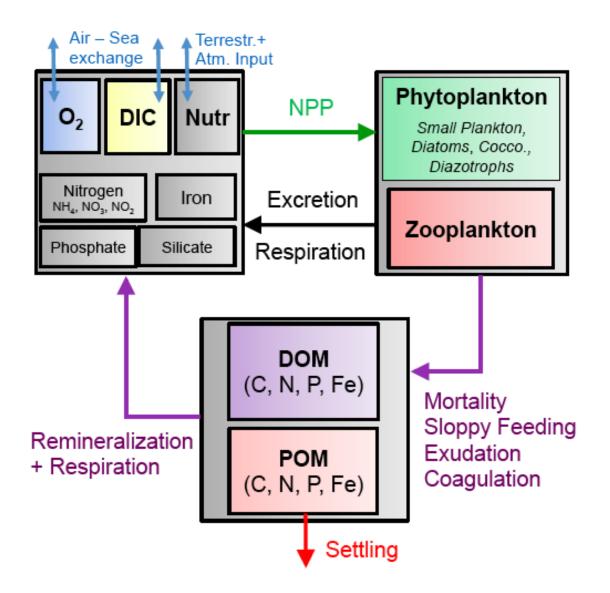


Variability of subsurface (100m) O_2 is high throughout the CCS, and is well reproduced by hindcast simulations.

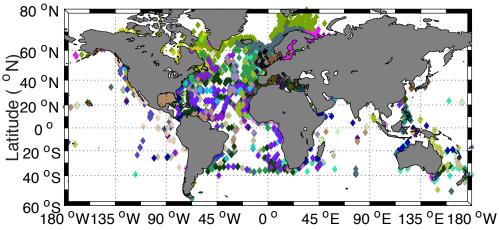
The primary factor driving these changes is fluctuations in the depth of the pycnocline.

This is revealed by the strong correlation between O_2 and density in both observations and simulations.

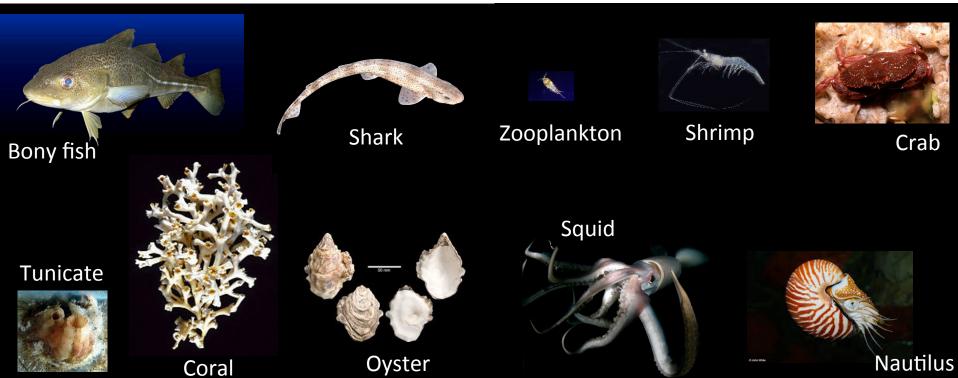
Ecosystem Model



Species Data

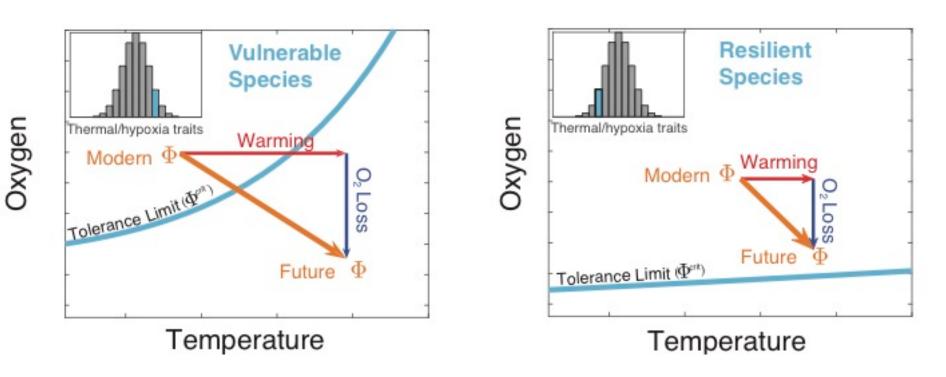


Labotatory measurements of Metabolic rate and critical O_2 (P_{crit}) at multiple temperatures. Database: >70 diverse species, globally distributed, from 5 phyla, wide size range.

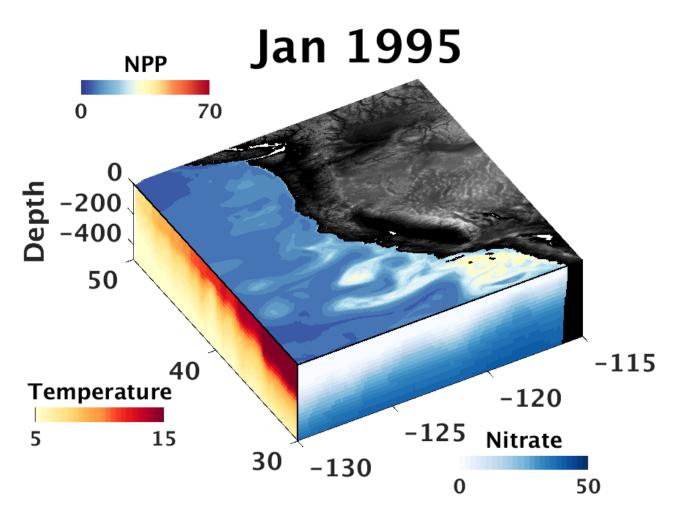


Vulnerable vs Resilient

Changing Habitability depends on:
1) Species sensitivity (Slope, Intercept)
2) Environmental change (Temp, O₂)



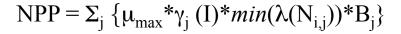
Part I Lower Trophic Level

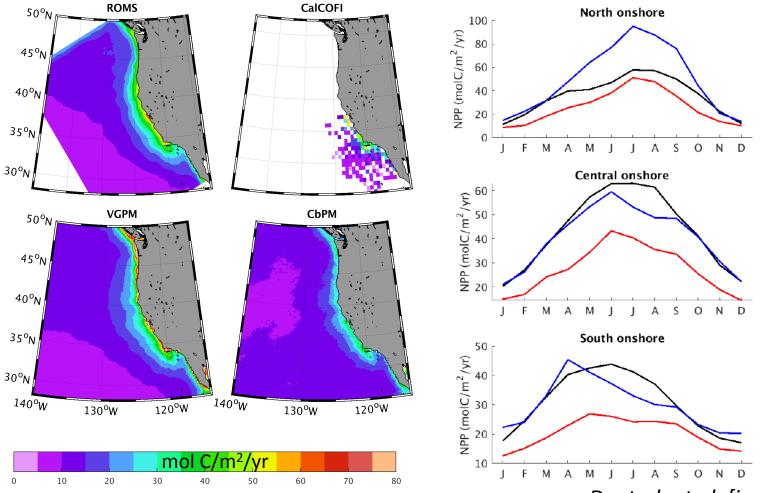


How does Primary Productivity of the California Current respond to climate variations and change?

Net Primary Productivity

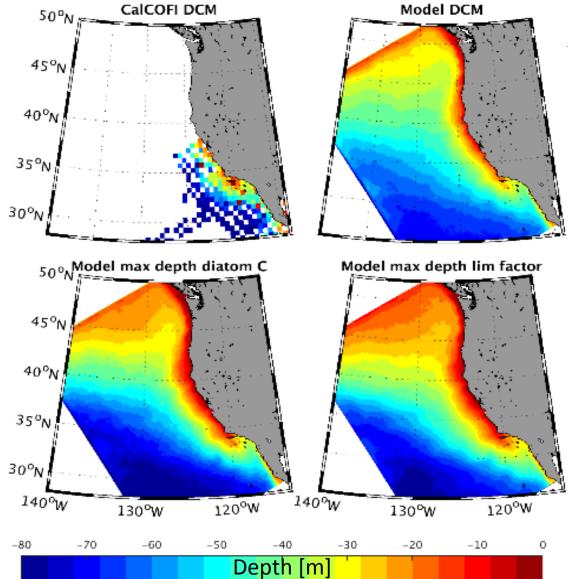
Spatial distribution and Seasonal Cycle of Net Primary Productivity.





Deutsch et al. [in review]

Nutrient vs Light Limitation



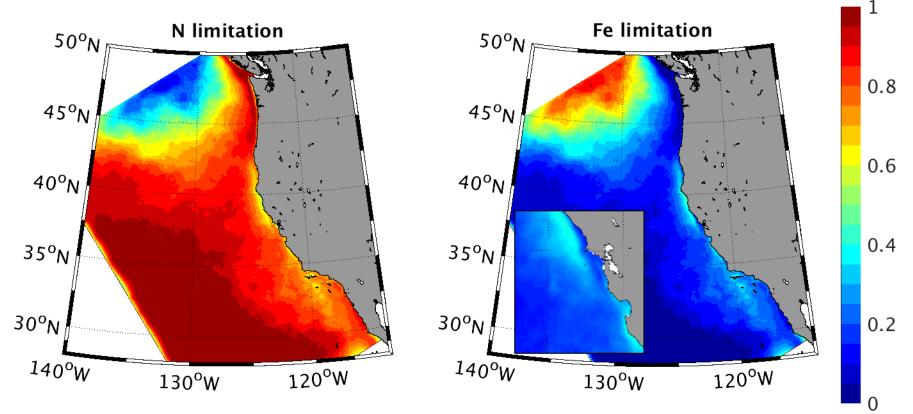
NPP =
$$\Sigma_j \{\mu_{\max} * \gamma_j (I) * min(\lambda(N_{i,j})) * B_j\}$$

Depth of the vertical maximum of Chlorophyll Reflects the trade-off between light (high at surface) and nutrients (high at depth).

Model fidelity to observations implies the model captures a realistic trade-off between nutrient limitation and light limitation.

Nitrogen vs Iron Limitation

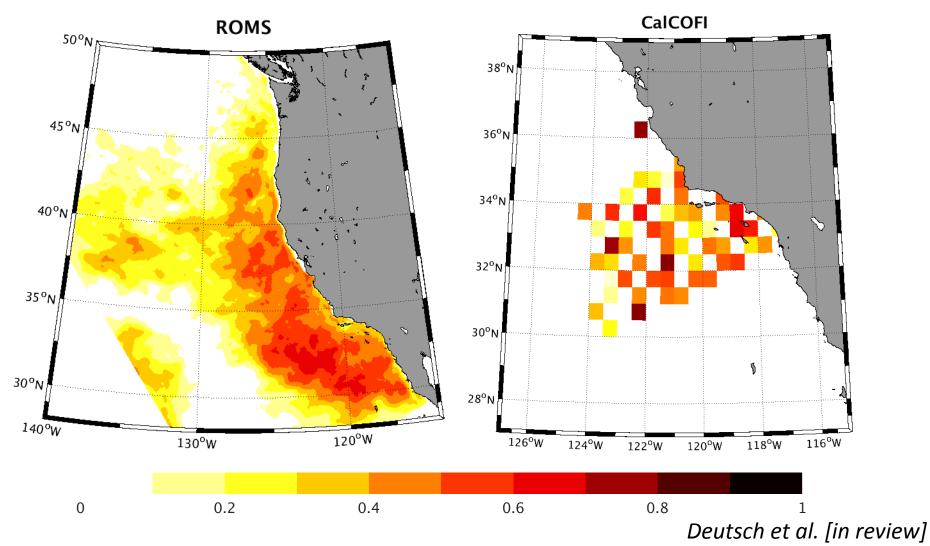
NPP =
$$\Sigma_j \{ \mu_{max} * \gamma_j (I) * min(\lambda(N_{i,j})) * B_j \}$$



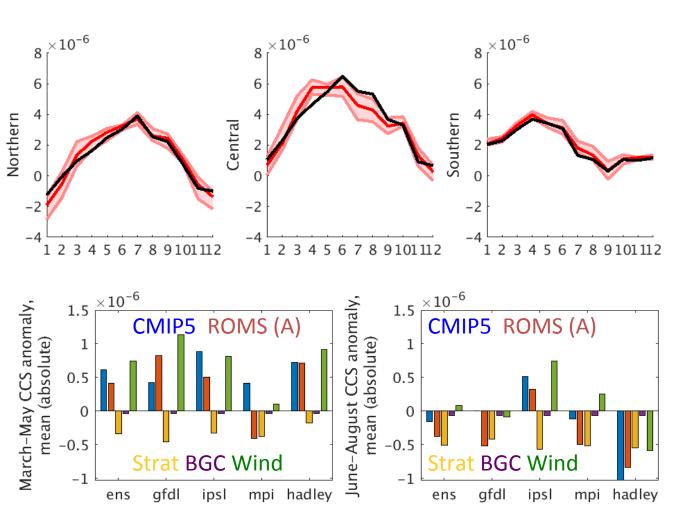
Frequency of limitation by Nitrogen (left) and Iron (right) for the model's dominant primary producer, diatoms. The frequency of limitation is based on 5-day average output, and weighted by biomass. Inset shows offshore band of relatively frequent Fe limitation along the central CCS, as observed by Firme et al. [2008].

NPP: Interannual Variability

Model exhibits a high correlation between subsurface density and NPP, indicating the strong influence of pycnocline depth on nutrient supply. Similar correlations seen in observations.



Changing Upwelling



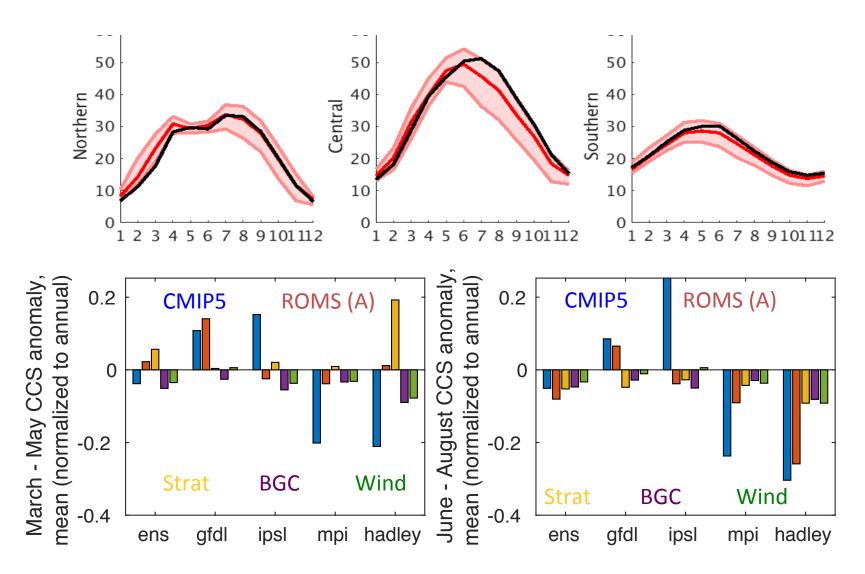
Upwelling changes are subtle, and mostly seasonal, not annual mean.

Robust tendency for more in spring and less in summer.

Stratification significantly offsets wind effect in spring, and is decisive in summer!

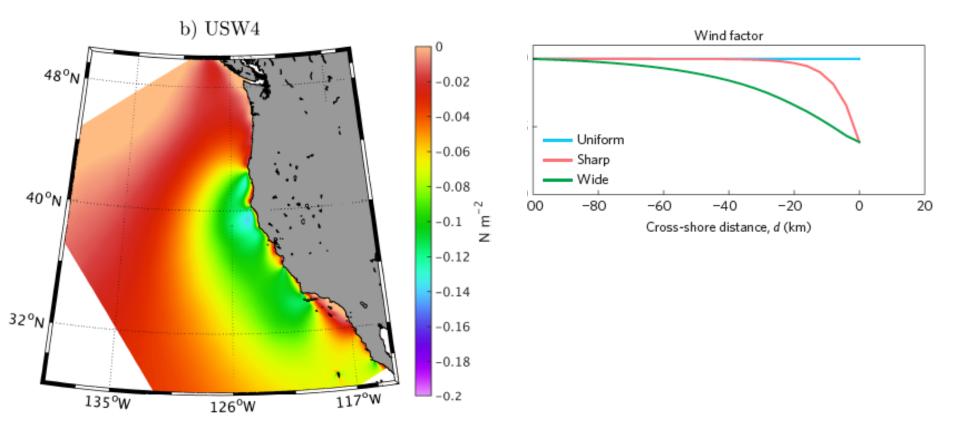
Changing NPP

Like wind, changes in NPP also modest (<5%) and mostly seasonal shifts. Remote nutrient redistribution and stratification (+Temp) more important than winds!



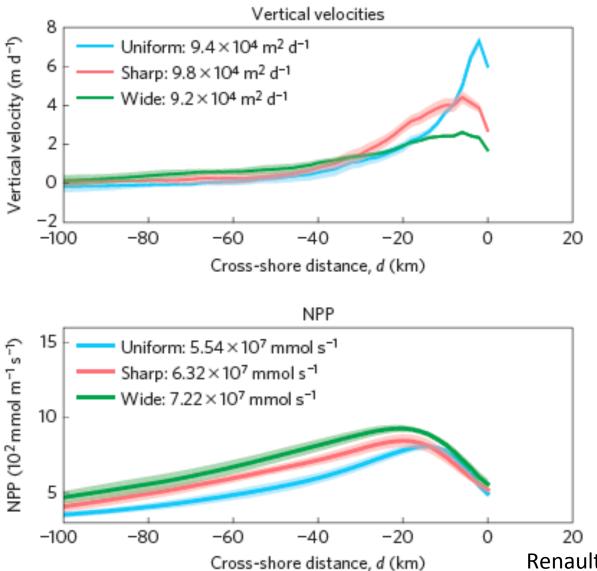
The role of wind structure

Global models capture large-scale wind changes, but not regional details. Could those matter? Experiments with the drop-off in wind strength along the coast...



Renault et al. [2016] Nature Geoscience

NPP Response to Wind Pattern



Coastal wind drop-off doesn't change the total upwelling, but shifts the balance between coastal and offshore, consistent with expectations.

Despite the same total upwelling, NPP is impacted by wind structure. The reduction is mediated by eddies.

Renault et al. [2016] Nature Geoscience