

The Near Real-time BGC/Physical 4D- Var DA system in the California Current

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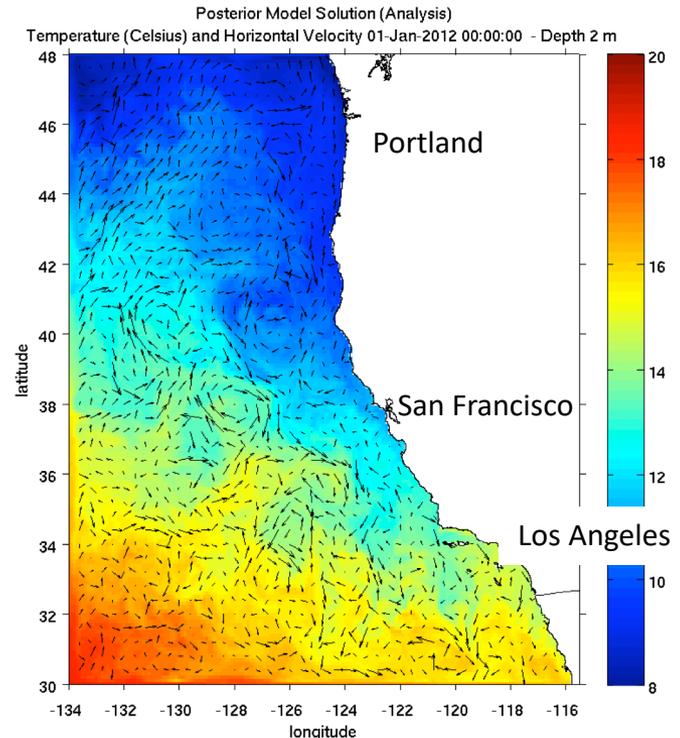
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University of California Santa Cruz

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Outline

- Review of L4D-Var for BGC models
- NRT system
- Ecosystem responses to physical drivers in coastal regions
 - Recent warm period in the CCS



BGC L4D-Var Review

- Gaussian data vs skewed data
- Positive and negative variables vs positive definite concentrations
- We assume lognormal variables
- For 4D-Var, requires additional linearizations

Logarithm transformation
Surface chl-a

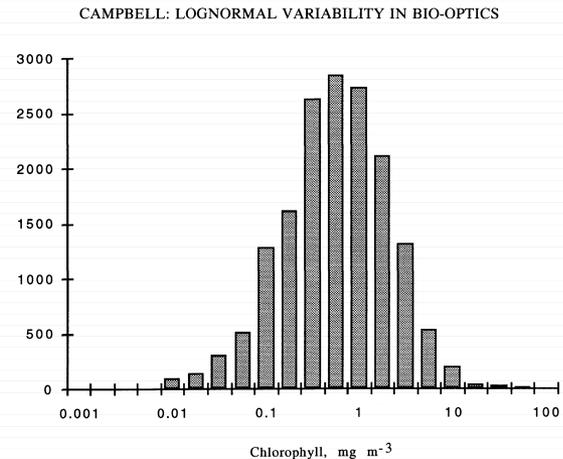


Figure 1. Histogram of 16,364 in situ measurements of ocean chlorophyll concentration from a compilation by *Balch et al.* [1992]. The data are global in scope, but sampling was concentrated at midlatitudes in the northern hemisphere, and central ocean gyre regions were undersampled.

Campbell (1995)



BGC Review – Fully Coupled G4DVar and L4DVar using augmented state vector

Gaussian Cost function

$$J_G(\delta \mathbf{x}_0) = \frac{1}{2} \delta \mathbf{x}_0^T \mathbf{B}^{-1} \delta \mathbf{x}_0 + \frac{1}{2} \sum_{i=1}^{N_0} (\mathbf{d}_i - \mathbf{H}_i \mathbf{M}_{i,0} \delta \mathbf{x}_0)^T \mathbf{R}_i^{-1} (\mathbf{d}_i - \mathbf{H}_i \mathbf{M}_{i,0} \delta \mathbf{x}_0),$$

Lognormal Cost function

$$J_L(\delta \mathbf{g}_0) = \frac{1}{2} \delta \mathbf{g}_0^T \mathbf{B}_L^{-1} \delta \mathbf{g}_0 + \frac{1}{2} \sum_{i=1}^{N_0} (\mathbf{p}_i - \mathbf{L}_i \mathbf{H}_i \mathbf{M}_{i,0} \mathbf{X}_{b,0} \delta \mathbf{g}_0)^T \mathbf{R}_{L,i}^{-1} (\mathbf{p}_i - \mathbf{L}_i \mathbf{H}_i \mathbf{M}_{i,0} \mathbf{X}_{b,0} \delta \mathbf{g}_0),$$

- Cost functions can be combined in terms of augmented state vector and error covariances

$$\delta \mathbf{z} = \begin{bmatrix} \delta \mathbf{x}_G \\ \delta \mathbf{g}_L \end{bmatrix}$$

$$\mathbf{B} = \begin{bmatrix} \mathbf{B}_G & \mathbf{0} \\ \mathbf{0} & \mathbf{B}_L \end{bmatrix}$$

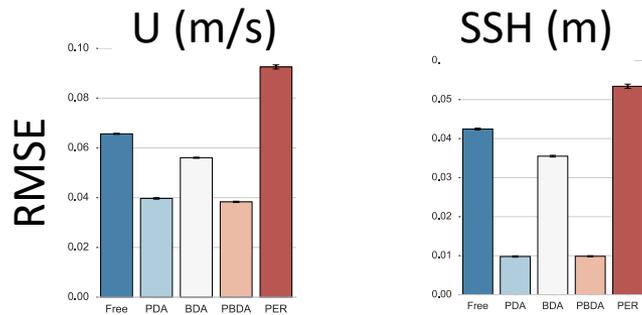
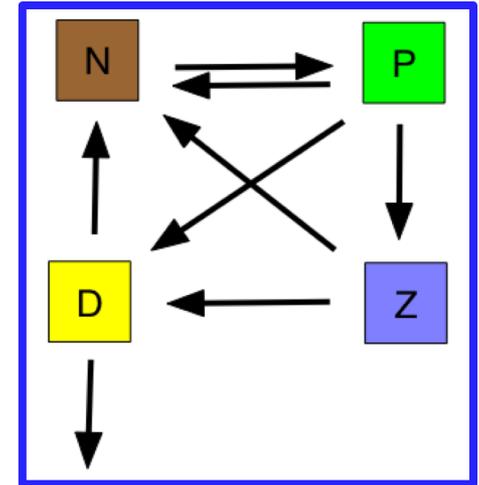
$$\mathbf{R} = \begin{bmatrix} \mathbf{R}_G & \mathbf{0} \\ \mathbf{0} & \mathbf{R}_L \end{bmatrix}$$



Fully coupled or not Fully coupled?

In principal, bio obs should feed back on physics and vice versa
 Demonstrated here in model twin experiment

- Statistics from 30 1-month runs.
- Assimilating physical data and surface Phytoplankton
- **Lowest error from combined PBDA**
 (not yet demonstrated with real data)



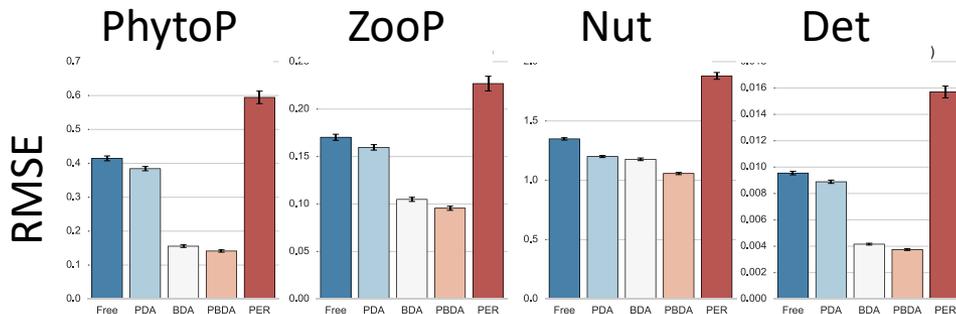
Free Run

Physical DA

Biological DA

Physical and biological DA

Persistence (1 month)



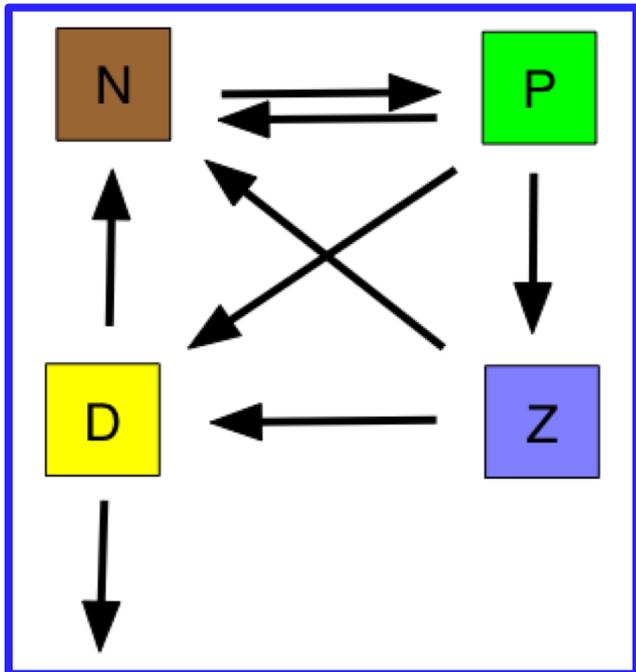
Song et al. (2016b)



We have developed L4D-Var for ROMS with two ecosystem models

NPZD

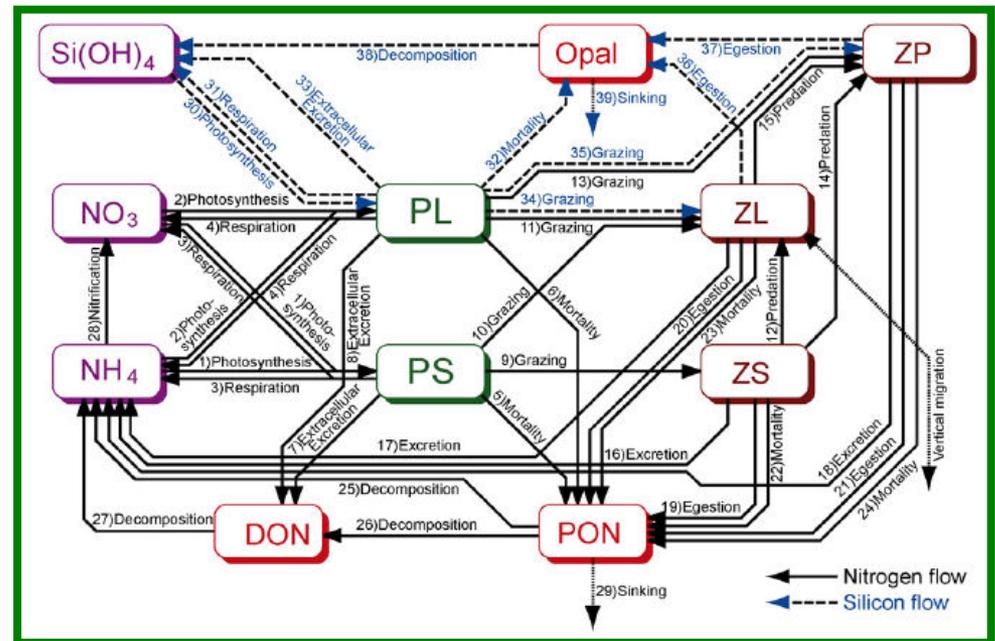
(Powell et al. 2006)



4 boxes

NEMURO

(Kishi et al. 2011)

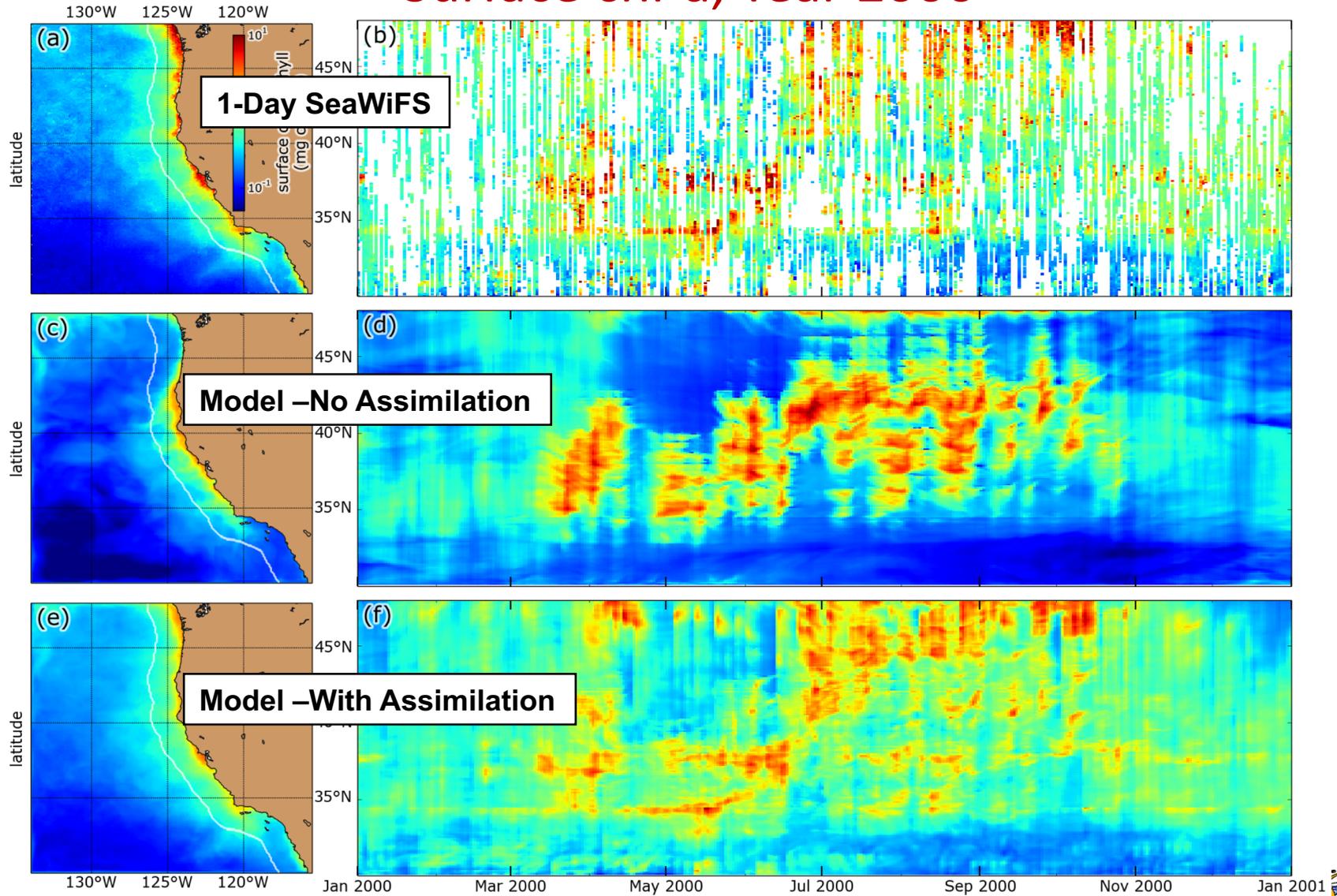


11 boxes



Fully coupled 4D-Var using NEMURO

Surface chl-a, Year 2000



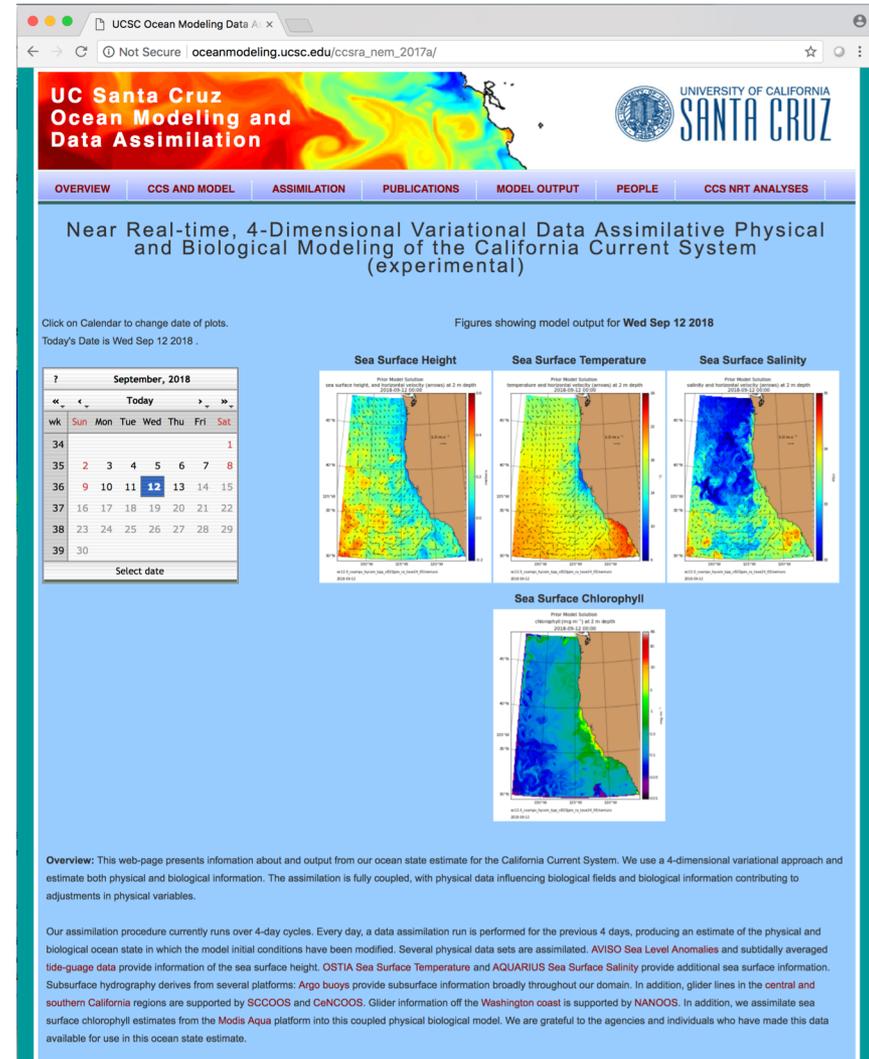
Mattern et al. (2017)



UCSC Coupled Physical/Biogeochemical System ROMS 4D-Var, 2011-present

(http://oceanmodeling.ucsc.edu/ccsra_nem_2017a/)

- 1/10° CCS ROMS configuration
- Online since July 2014
- 4-day assimilation cycles
- Assimilates SST, SSH, SCHL, glider T/S, Argo T/S, HF RADAR velocities
- Model output available on a TDS
- Figures of model fields posted
- Calendar searchable



An application: The recent warm blob in the Pacific

- $\sim 3^{\circ}\text{C}$ multiannual marine heatwave
- What are the ecosystem impacts of this anomalous event?

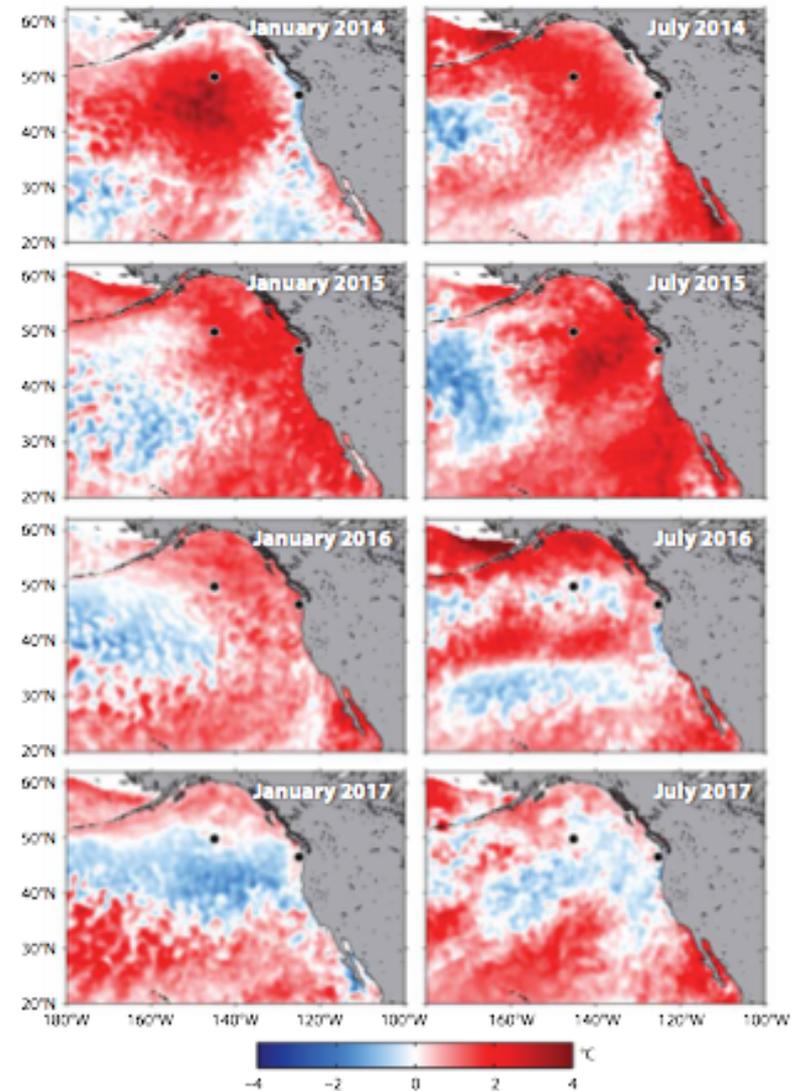
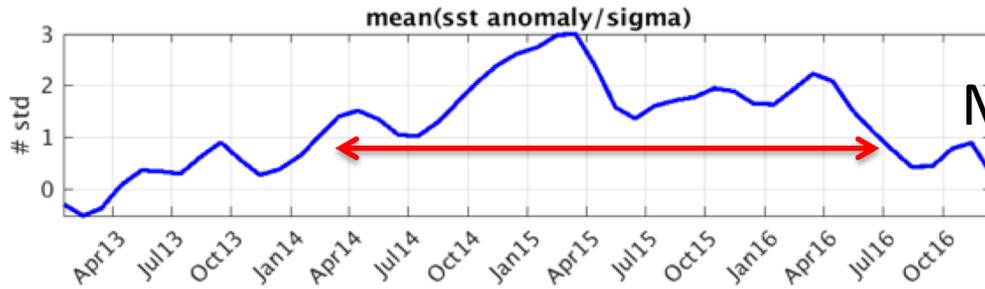
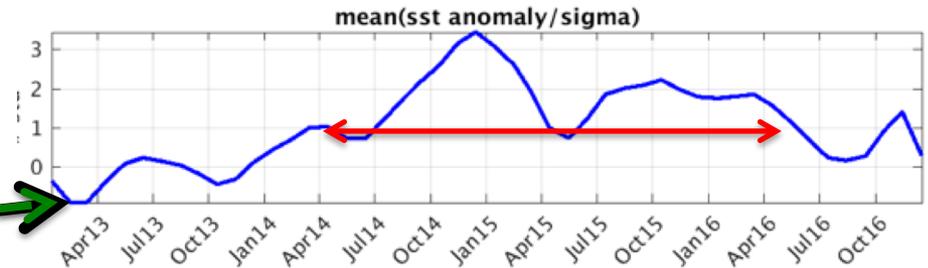
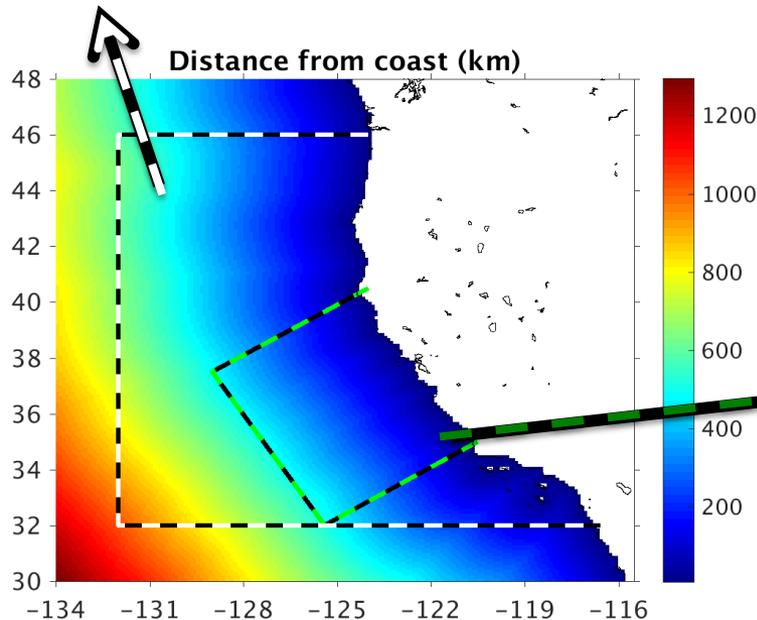


FIGURE 2. North Pacific sea surface temperature anomaly showing the evolution of the "warm blob" from its origination in winter 2013-2014 through the following four years. Satellite temperature data are from AVHRR only Optimum Interpolation Sea Surface Temperature (OISST, <https://www.ncdc.noaa.gov/oisst/data-access>), and anomalies are computed relative to a 30 year climatology constructed from 1982 to 2011. The locations of the OOI Washington Offshore and the OOI Station Papa (50°N, 145°W) mooring are shown as filled black circles.

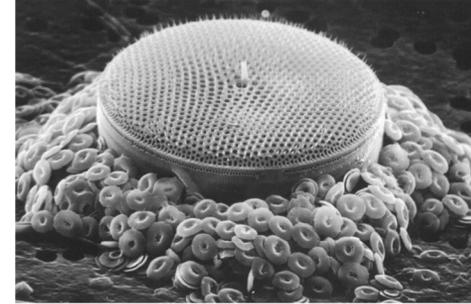
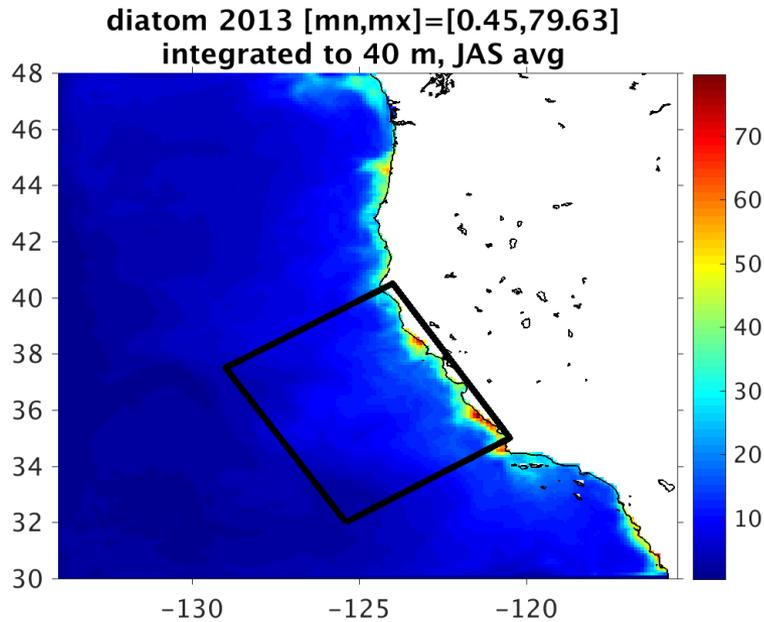
The Blob and El Niño as seen by a bgc/physical reanalysis



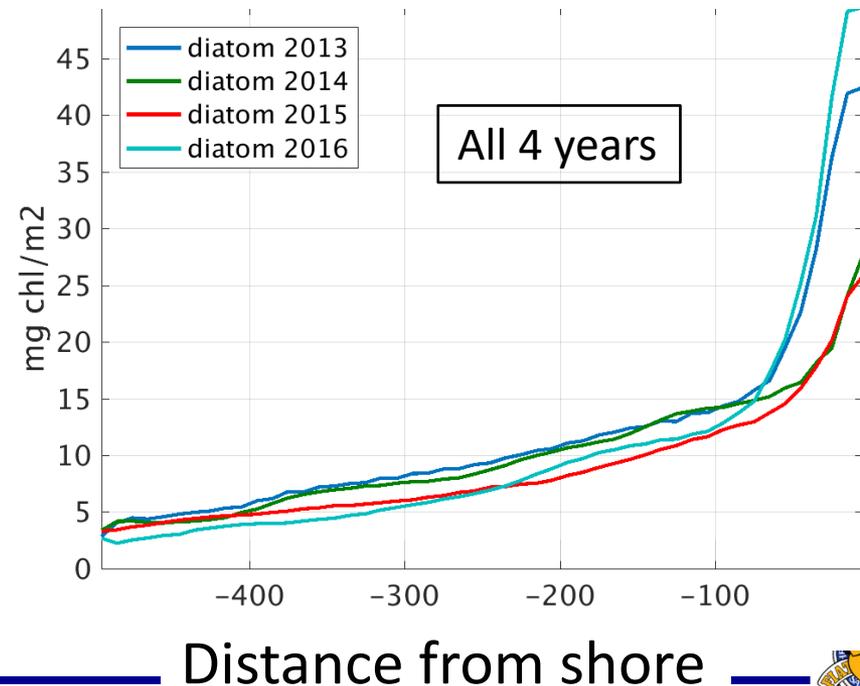
SST > 1 σ
March 2014 until June 2016



Central CCS Diatom fields, July-Sept Averages



Average in time (JAS) and
in space (cross-shore distance)



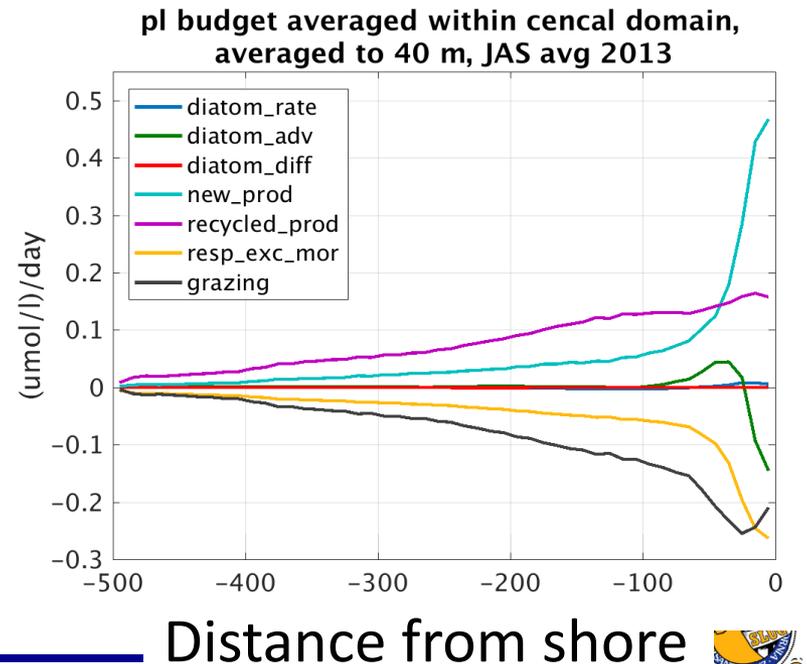
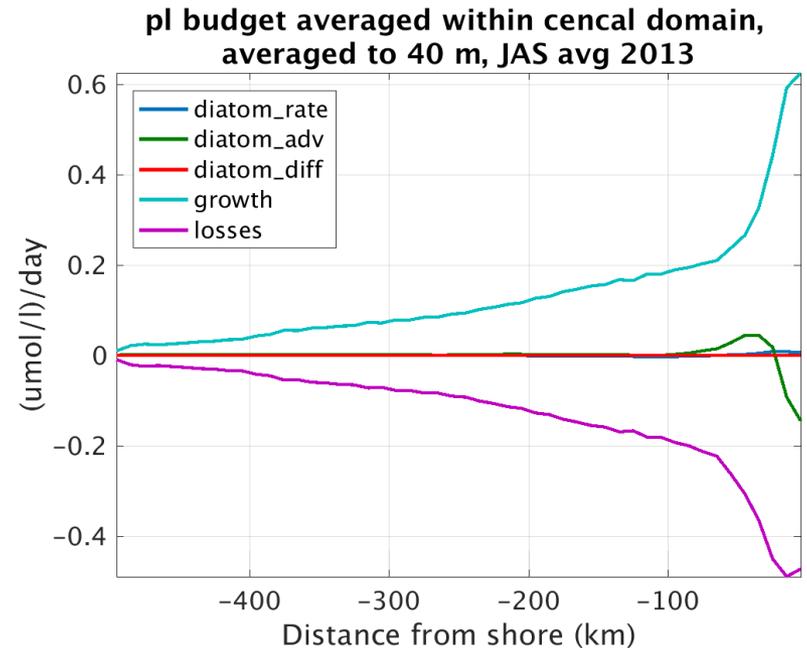
Annual average
diatom stock low
during 2014-2015

Diatom budget (summertime average)

$$dP/dt = \text{advection} + \text{diffusion} + \text{growth} + \text{losses}$$

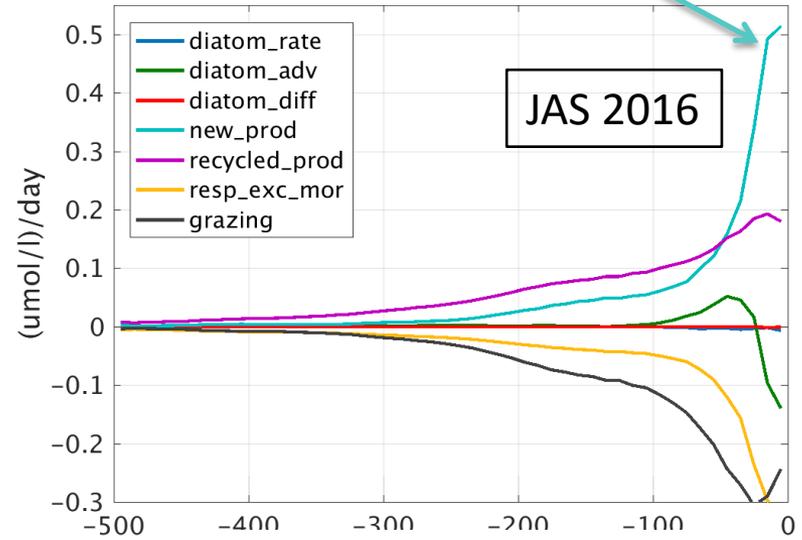
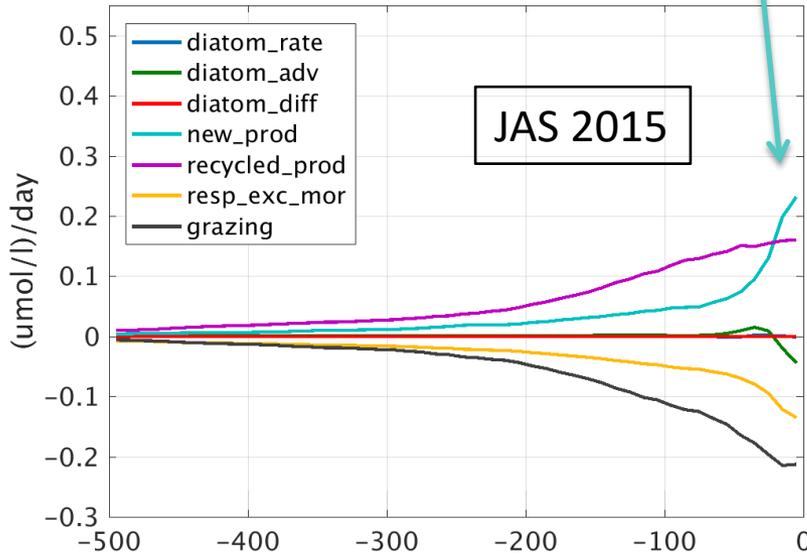
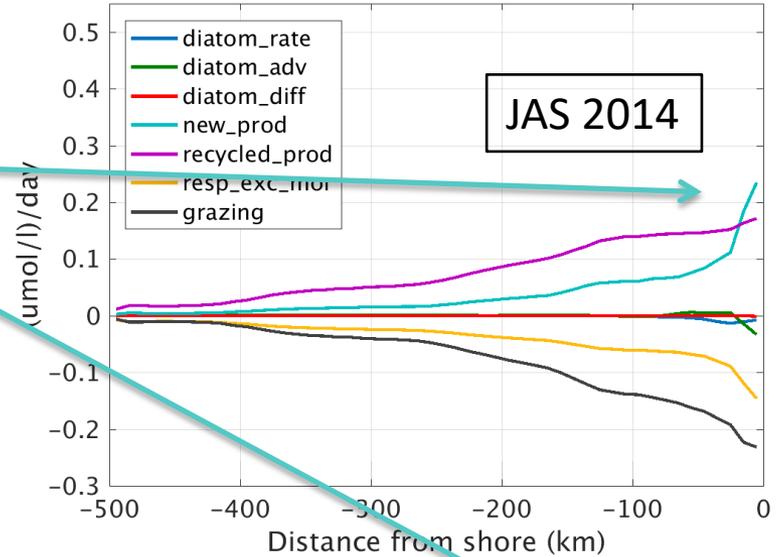
Mostly a balance between growth and losses with a small contribution by advection nearshore.

$$dP/dt = \text{advection} + \text{diffusion} + \text{new production} + \text{recycled production} + \text{grazing} + \text{excretion} + \text{respiration} + \text{mortality}$$



Summertime dynamical balances reveal starkly different new primary production

$dP/dt = \text{advection} + \text{diffusion} + \text{new production}$
 $+ \text{recycled production} + \text{grazing} + \text{excretion}$
 $+ \text{respiration} + \text{mortality}$



Distance from shore

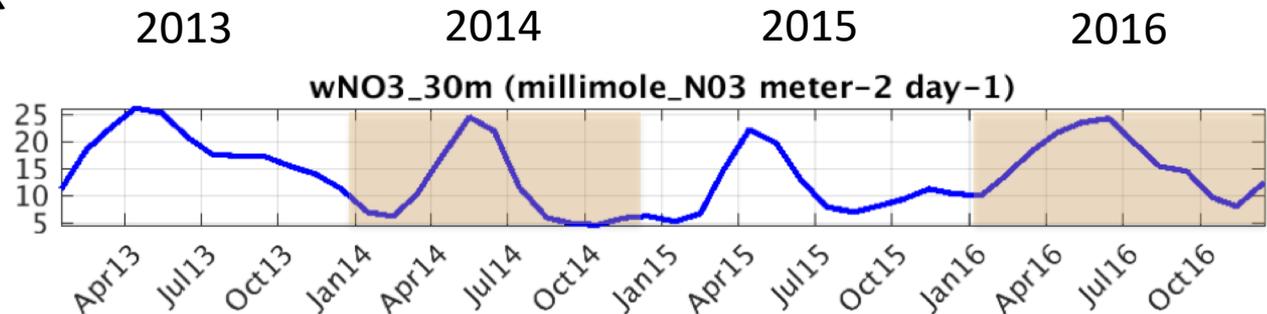
Distance from shore



Explanation

- Changes in new production can be due to changes in
 - light
 - temperature
 - higher temp -> higher growth rates
 - nutrient flux

Vertical nitrate flux
across 30 m depth

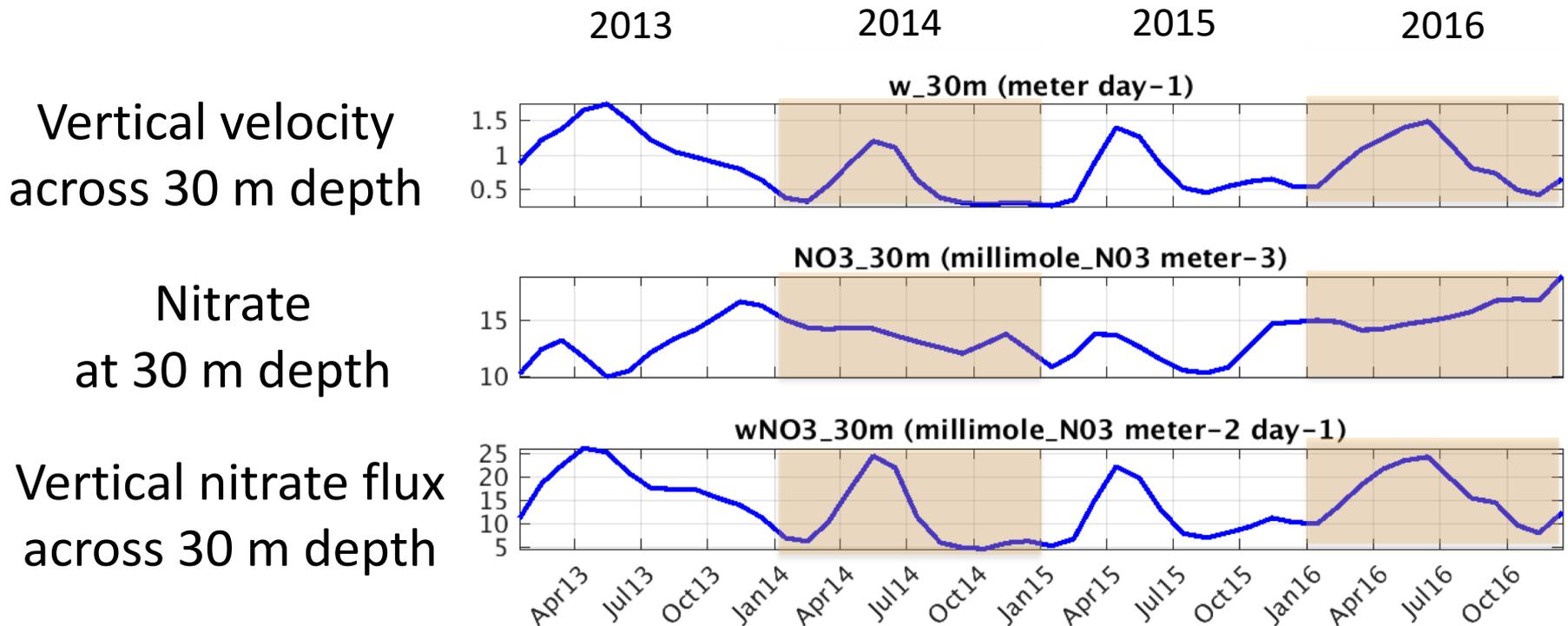


- Note anomalous summer lows in 2014/2015



Physical nitrate transport

(monthly averages, averaged to 50km from shore)



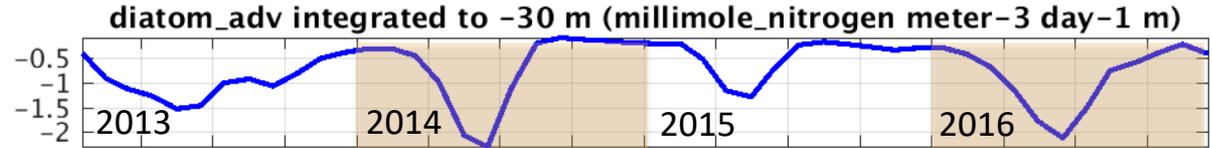
Vertical nitrate flux predominantly results from vertical velocity (94% of variance)



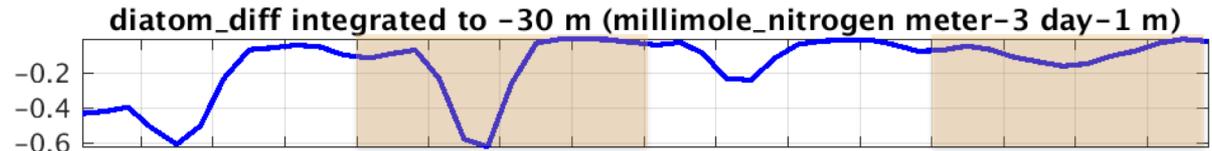
Diatom budget terms

(monthly averages, averaged to 50km from shore)

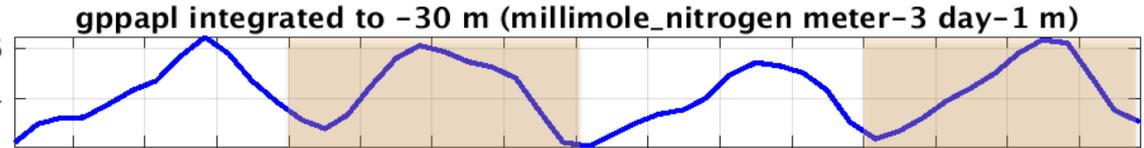
advection



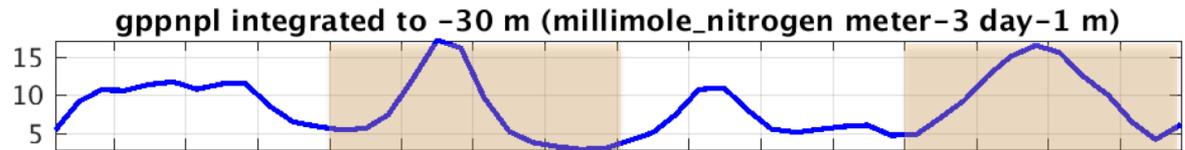
diffusion



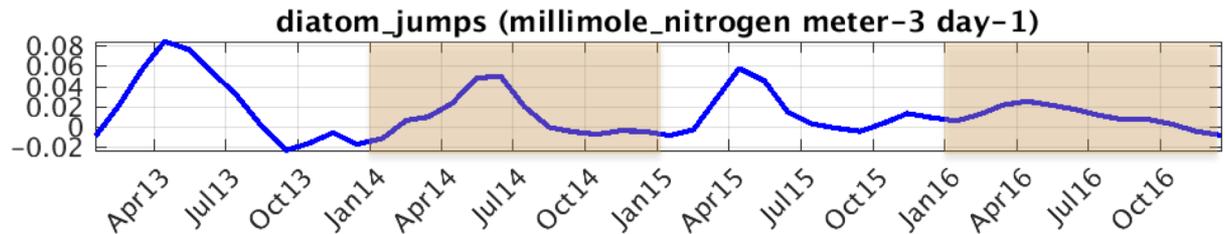
Recycled production



New production



Non-conservative changes due to assimilation cycles



79% of variance in new production accounted for by nitrate flux, 67% by wind stress



Summary

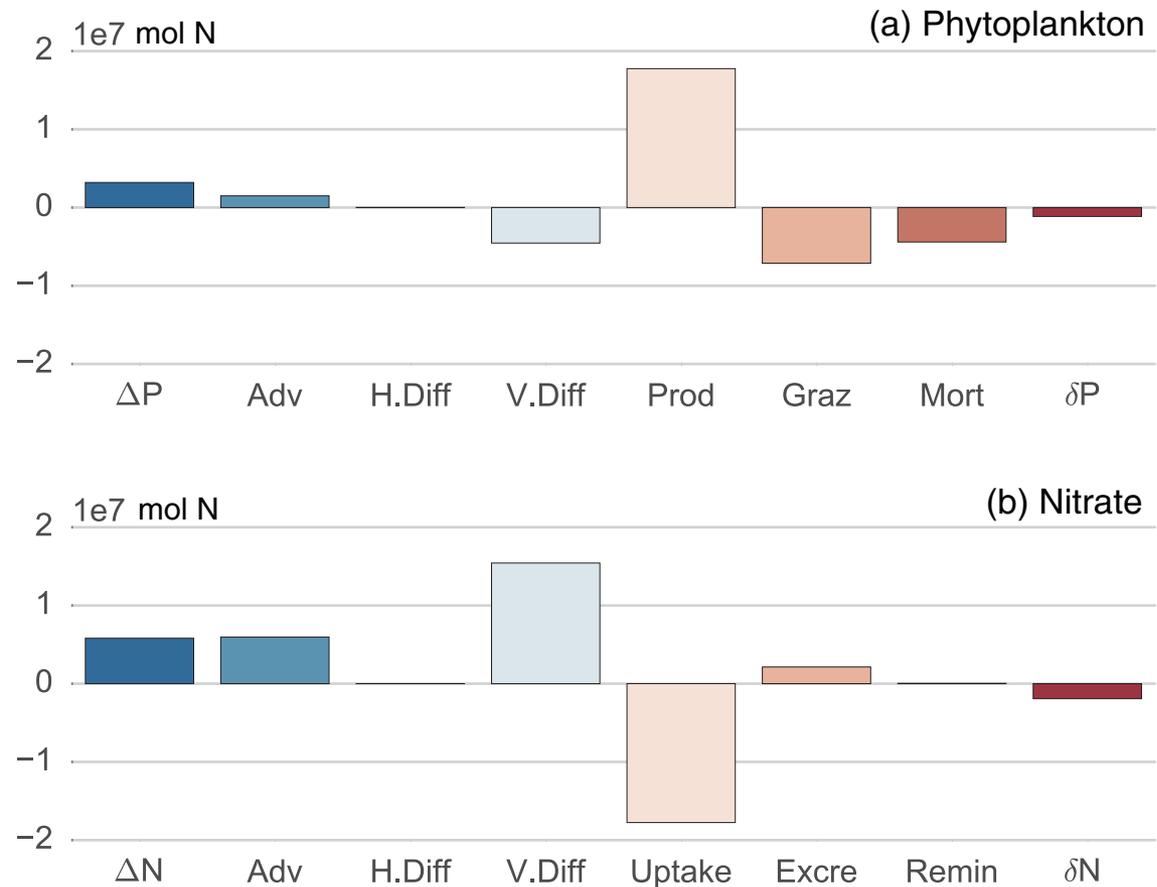
- NRT L4D-Var, physical and BGC data assimilation underway in the CCS for years.
- Reanalysis provides a sensible dynamical interpolation from sparse data and offers a platform for dynamical analysis to understand ecosystem impacts of physical drivers.
- During 2014-2015 (Blob)
 - **Low Diatom** annual average concentration
 - Recycled production not particularly anomalous.
 - **New production** was anomalous.
 - **Springtime** new production not significantly impacted
 - **Summertime** new production significantly lowered
 - Vertical nitrate flux dominated by **vertical velocity** (and wind stress), not nutricline depth



Flux estimates (npzd model)

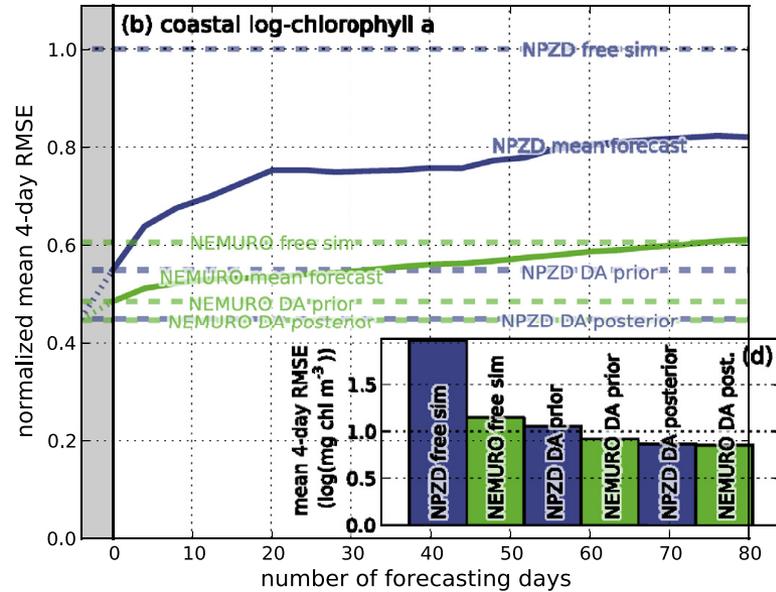
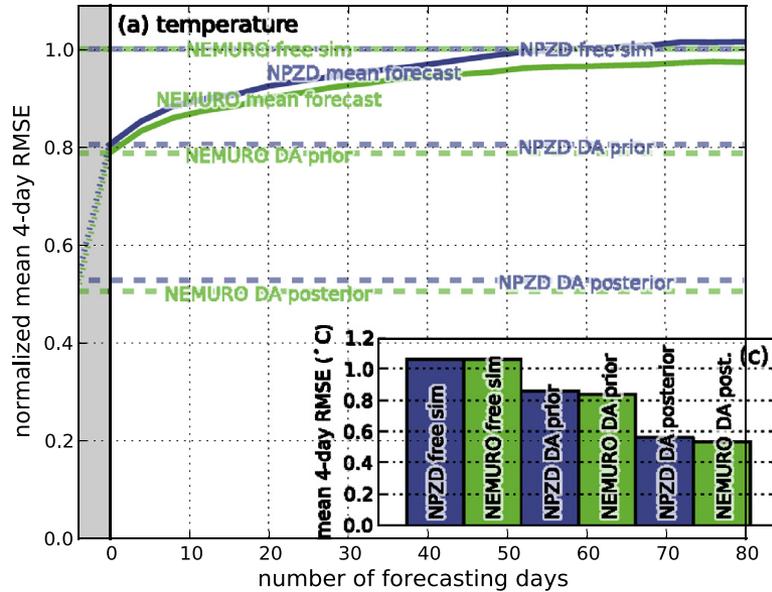
(coastal zone average, April-September)

- δP and δN are adjustments due to assimilative jumps
- Small compared to dominant biological and physical processes



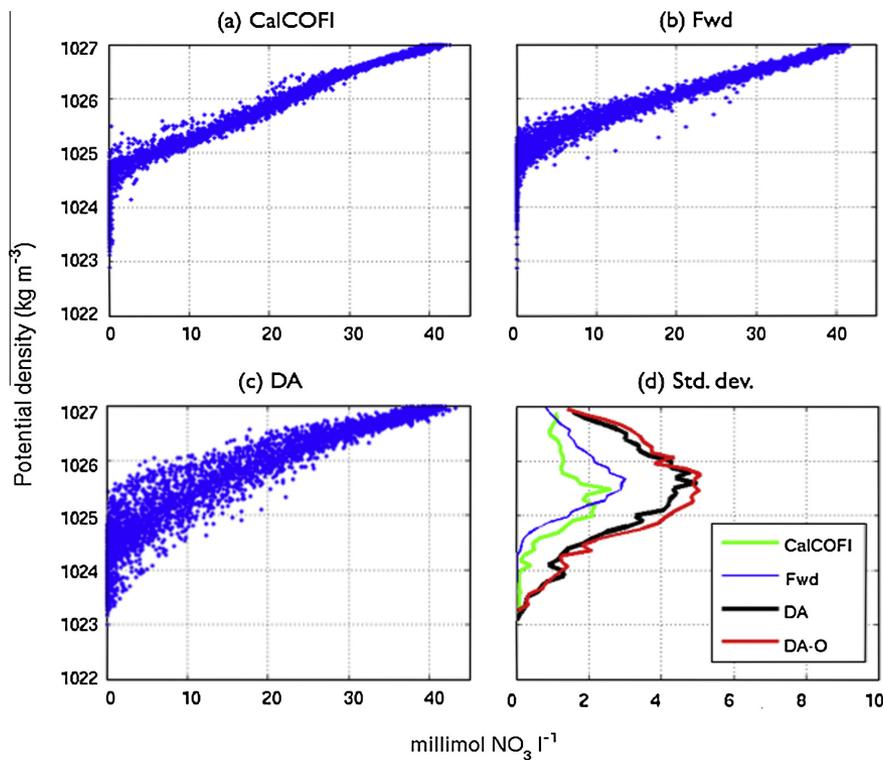
Forecasting skill

Improvements out to at least 10 arguably 30 days

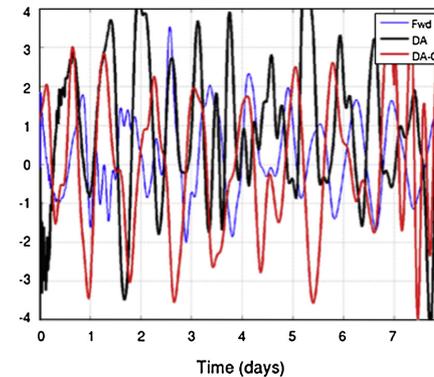


2 negative impacts of physical assimilation on ecosystem response

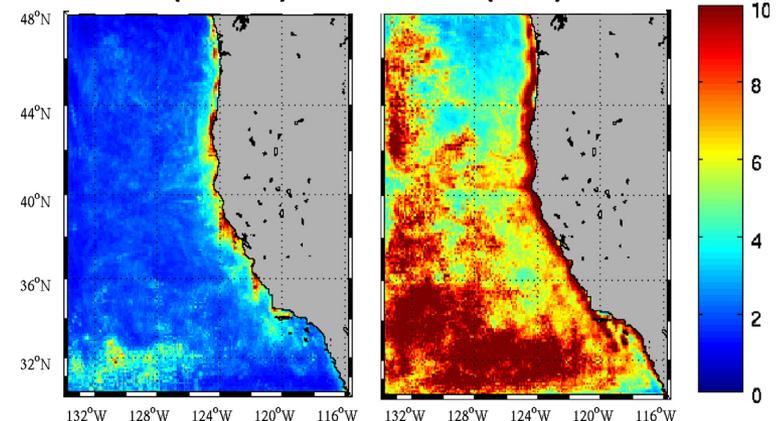
Nutrient/density relation



w @ 50 m (m/day)



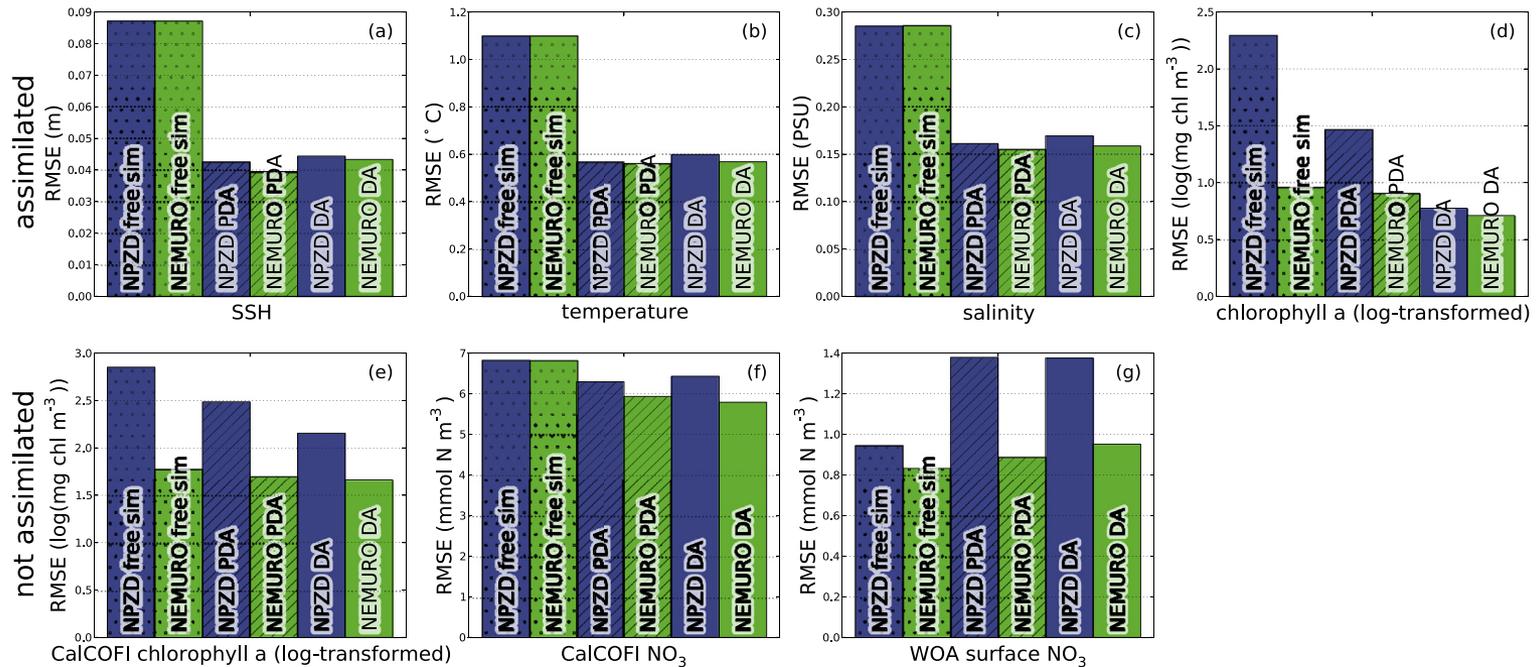
RMS w @ 50 m (m/day)
(FWD) (DA)



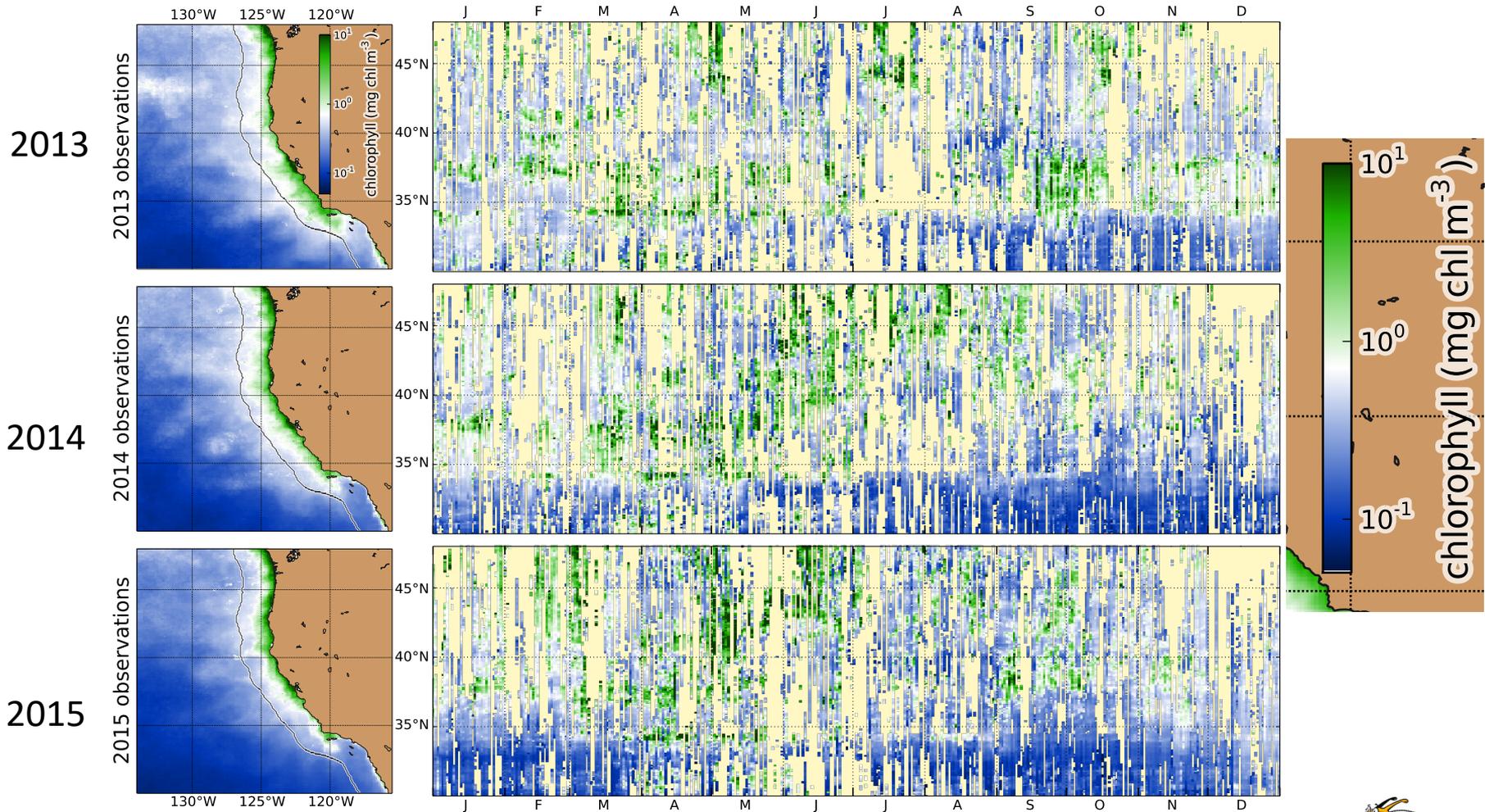
Raghukumar et al. (2015)



Assimilation improvement for assimilated and unassimilated data



2013-2015 MODIS AQUA CCS Surface chlorophyll-a (mg chl-a/m³)



2013-2015 Reanalysis CCS Surface chlorophyll-a (mg chl-a/m³)

