

Regional state estimates

ECCO Summer School 2025

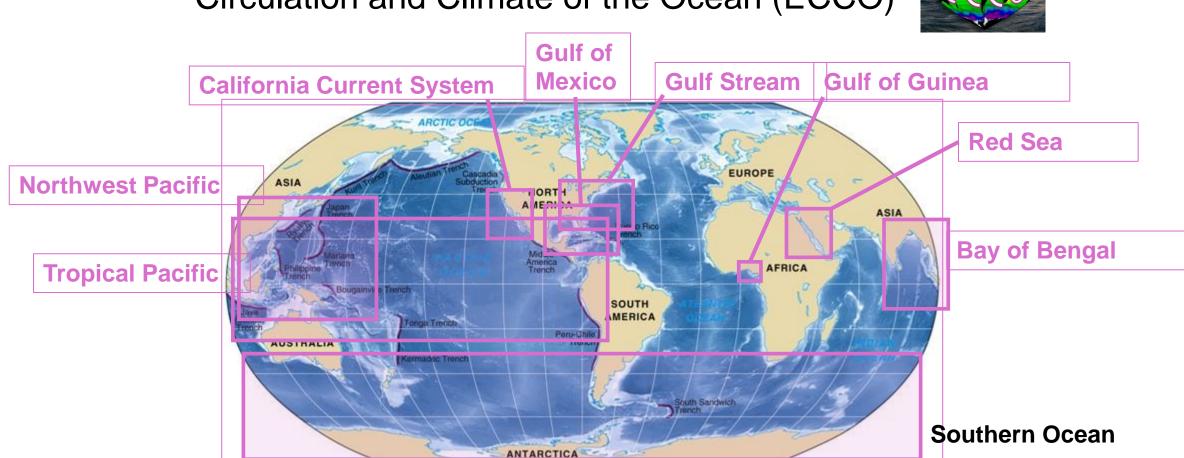
Matt Mazloff, Bruce Cornuelle, Ganesh Gopalakrishnan, Ariane Verdy, ECCO, et al.



Regional state estimation activities at Scripps Institution of Oceanography

Matt Mazloff, Ariane Verdy, Angela Kuhn, Ganesh Gopalakrishnan, Bruce Cornuelle

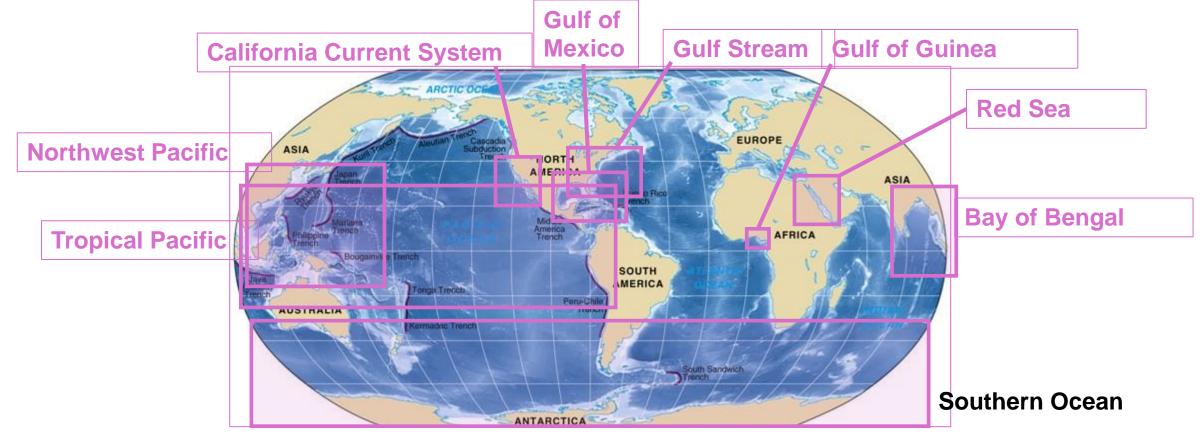
Software from the consortium for Estimating the Circulation and Climate of the Ocean (ECCO)



What is the utility of a regional state estimate? Why produce regional ocean data assimilations?

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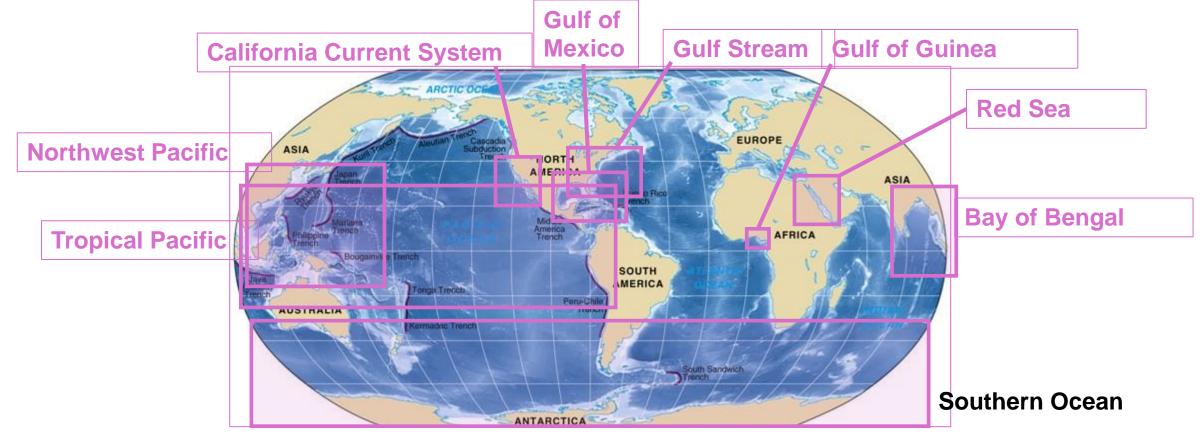
Who needs to know what the ocean is doing?

Pollutants, search and rescue, ship routing & navigation, acoustics, weather prediction, storm surge, wave state, ecosystem health, fish stock assessment, hydrological cycle, Etc.

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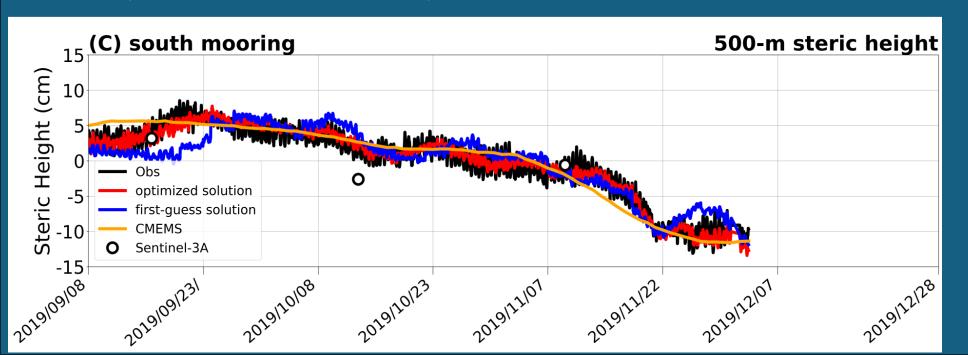
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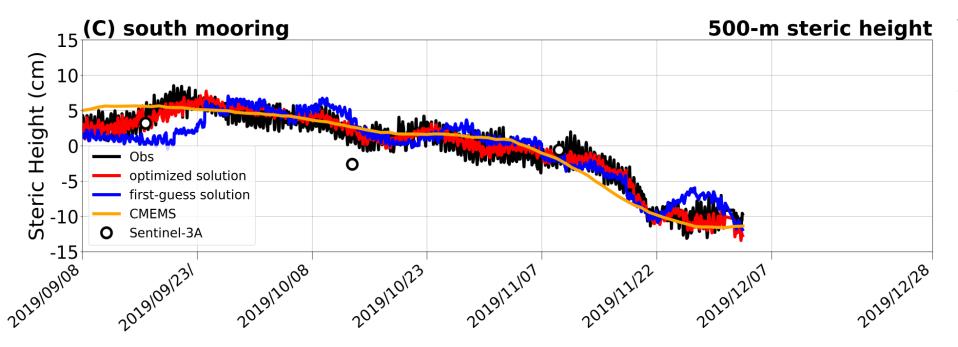


Why regional assimilations?

- Model setup and data assimilation machinery are tools.
- One often has a focused set of questions, and regional setups allows customized 'tools'.
- Targeted regional efforts can enable consistency with specific dynamical or system components (e.g. eddies, biomass, coastal, new obs platforms).
- Targeted regional efforts can enable consistency in specific regions or times, or with specific observational platforms.



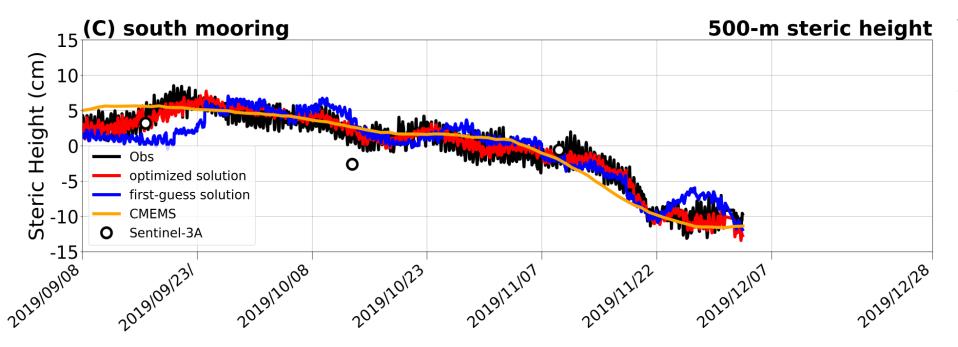
Tchonang et al 2024: Evaluation of a 4DVAR Assimilation System in the California Current at the SWOT Calibration/Validation Site. *JTECH*



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Targeted efforts can enable reproducing eddies

- Mapped products have formal mapping errors and representation error.
- Formal mapping error primarily from lack of info: sparse or noisy obs.
- Representation error comes from low resolution, missing physics, or errors in the model-data synthesis methodology.
- Mapping methods vary substantially, primarily with regards to how covariance estimates are determined and the level of physics included.
- Representation error: must choose what scales are signal or noise.



Tchonang et al 2024: Evaluation of a 4DVAR Assimilation System in the California Current at the SWOT Calibration/Validation Site. *JTECH*

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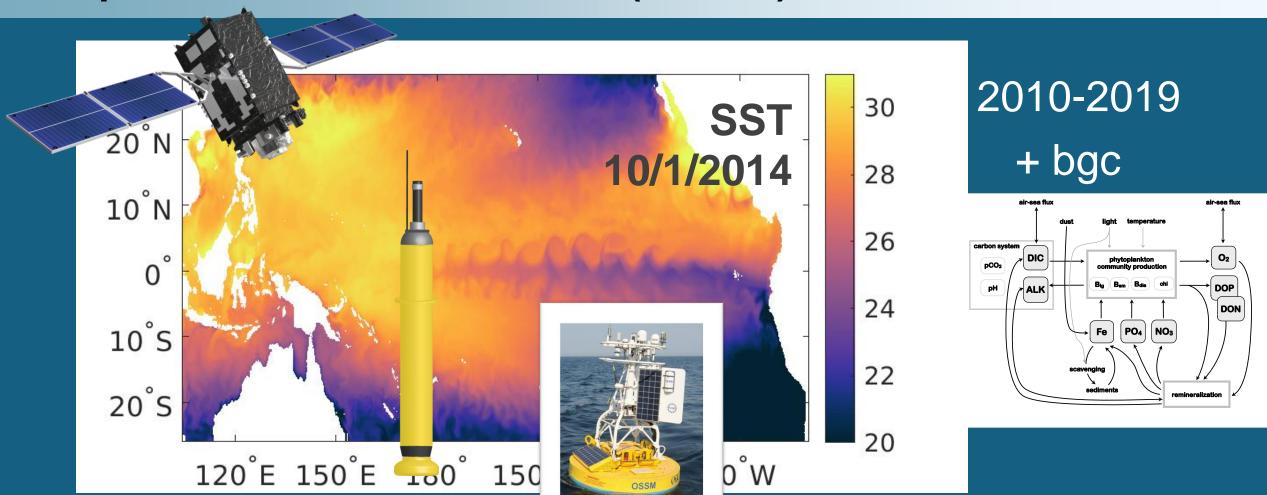
The science:

- The model is our hypothesis of how the world works.
- We make assumptions for computational efficiency.
- If we understand the ocean, those assumptions will give us the expected answers.
- If we don't understand, we must go back & make a new hypothesis a new model

How do we model ocean vertical transport? How do we forecast ocean sea level?

Once we have a good model (i.e. the forward problem), we can better use observations (i.e. the inverse problem)

Tropical Pacific State Estimate (TPOSE)



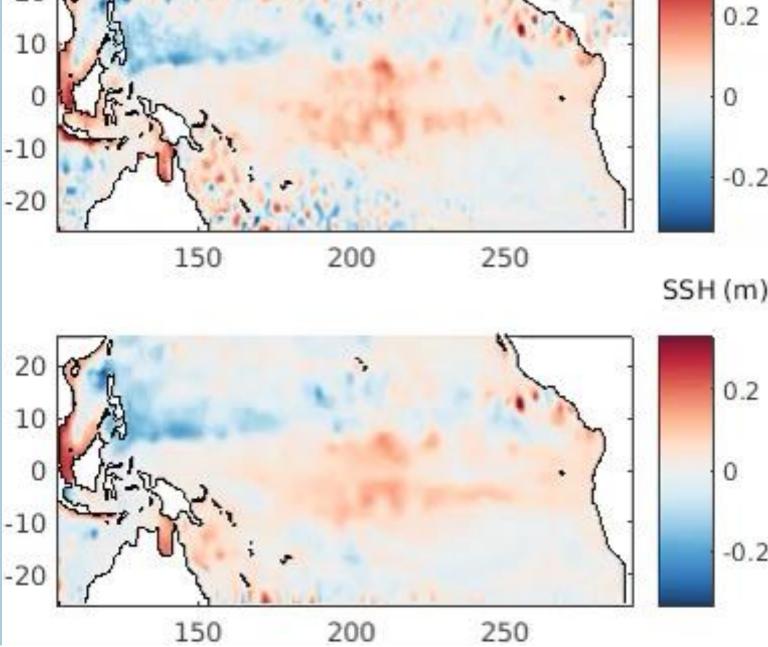
Exploring connections between surface forcing and interior dynamics through mixing

Sea Surface Height (SSH) anomaly

AVISO mapped

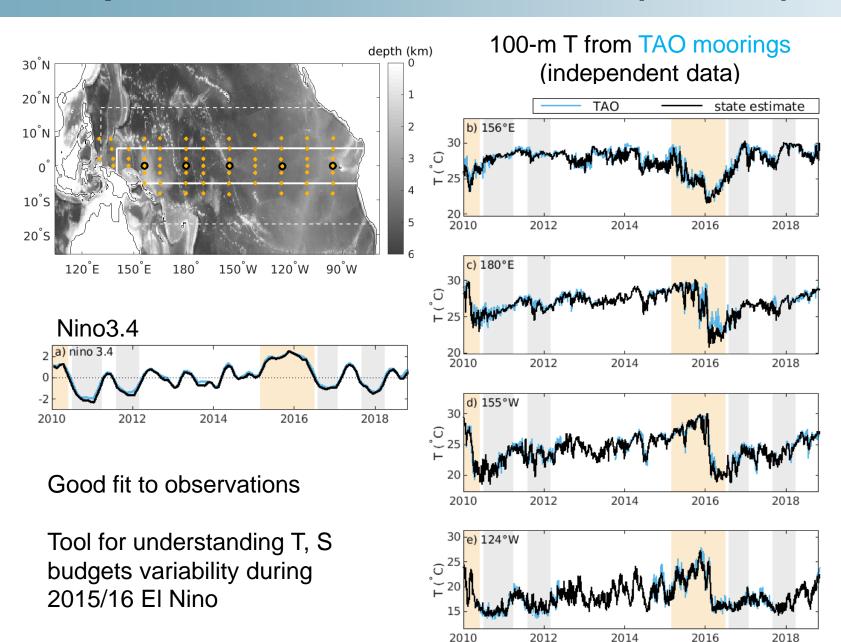
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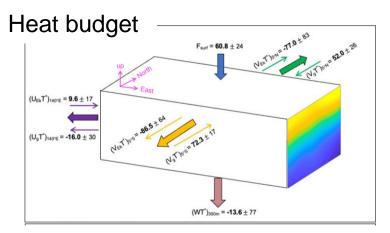


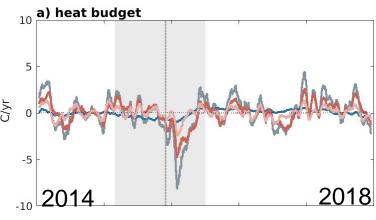


2015/1/1

Tropical Pacific State Estimate (TPOSE)





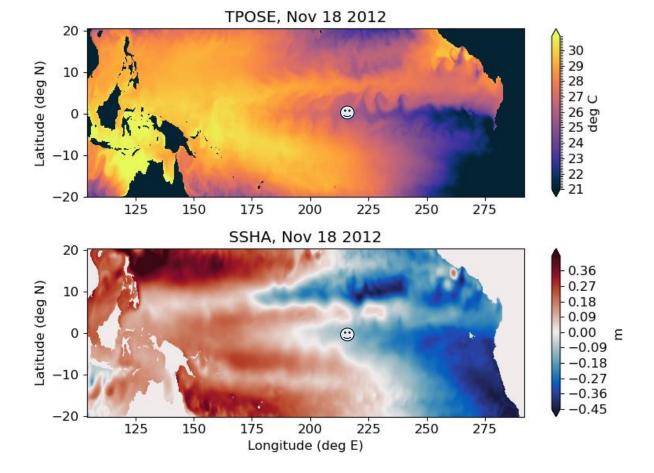


Verdy et al. (2023). Balancing
Volume, Temperature, and Salinity
Budgets During 2014-2018 in the
Tropical Pacific Ocean State
Estimate. JGR-Oceans

Mechanisms of Zonal Momentum Transport at 0N, 140W in an Ocean State Estimate

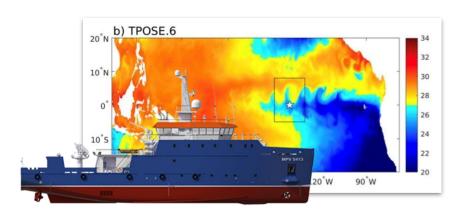
Ellen Davenport¹, Ariane Verdy¹, Matt Mazloff¹, Bruce Cornuelle¹, Amy Waterhouse¹, and Dan Whitt²

¹Scripps Institution of Oceanography ²NASA Ames

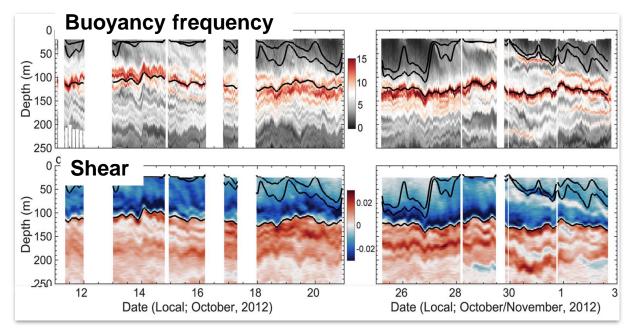






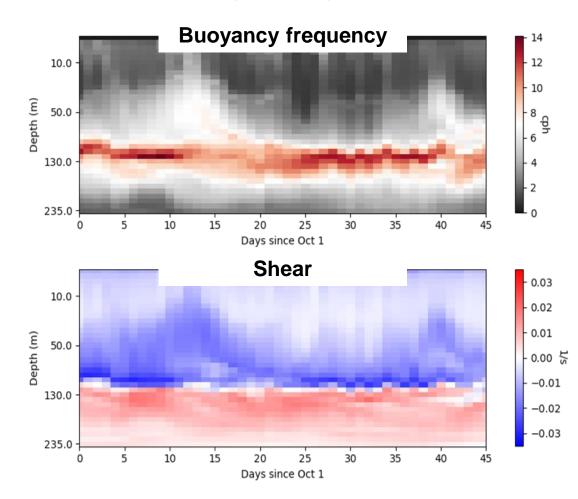


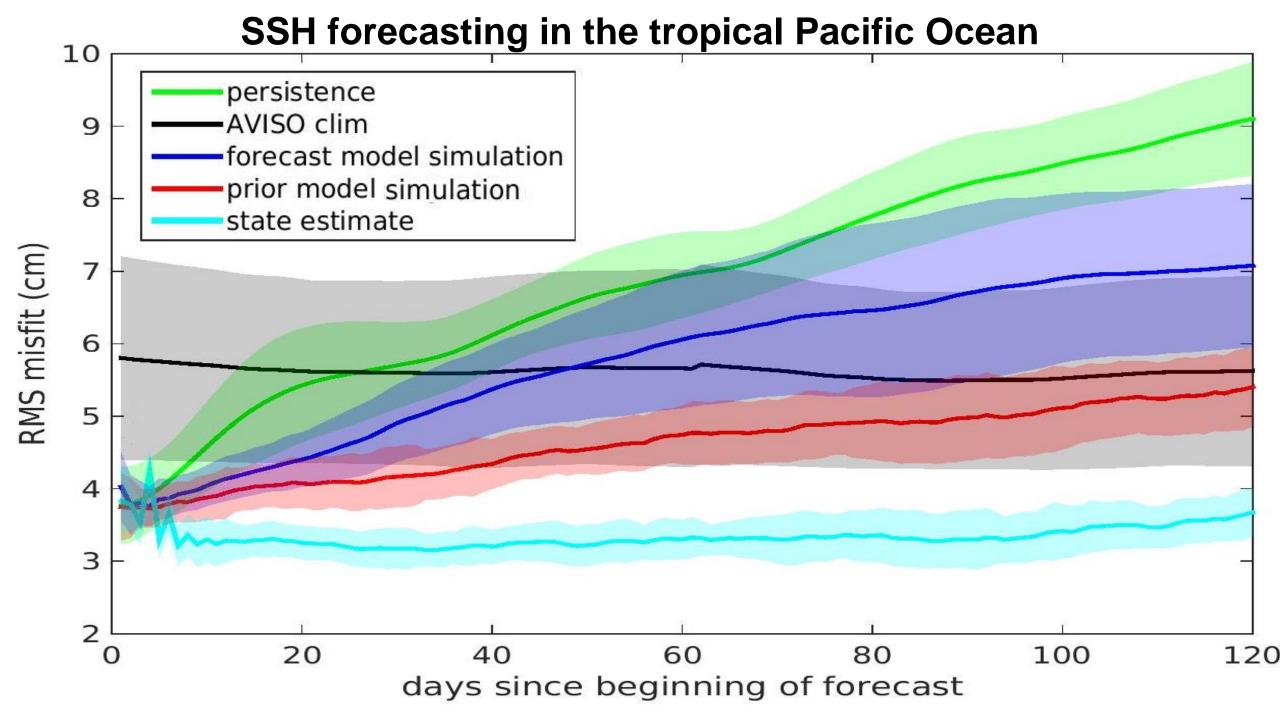
EquatorMix: 0°N,140°W, October 2012



Pinkel et al. (2023) GRL

TPOSE: 0°N,140°W, October 2012



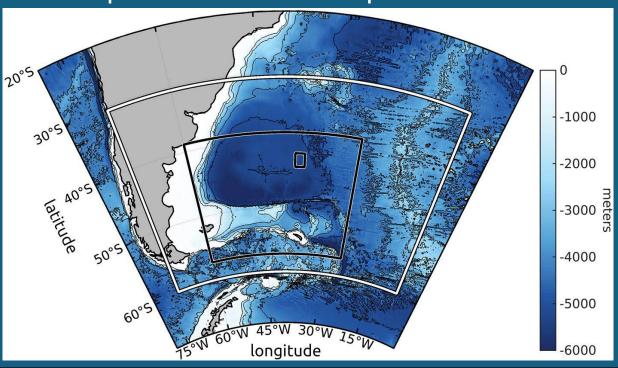


Sources of forecast error: model params, initial conditions, boundary conditions

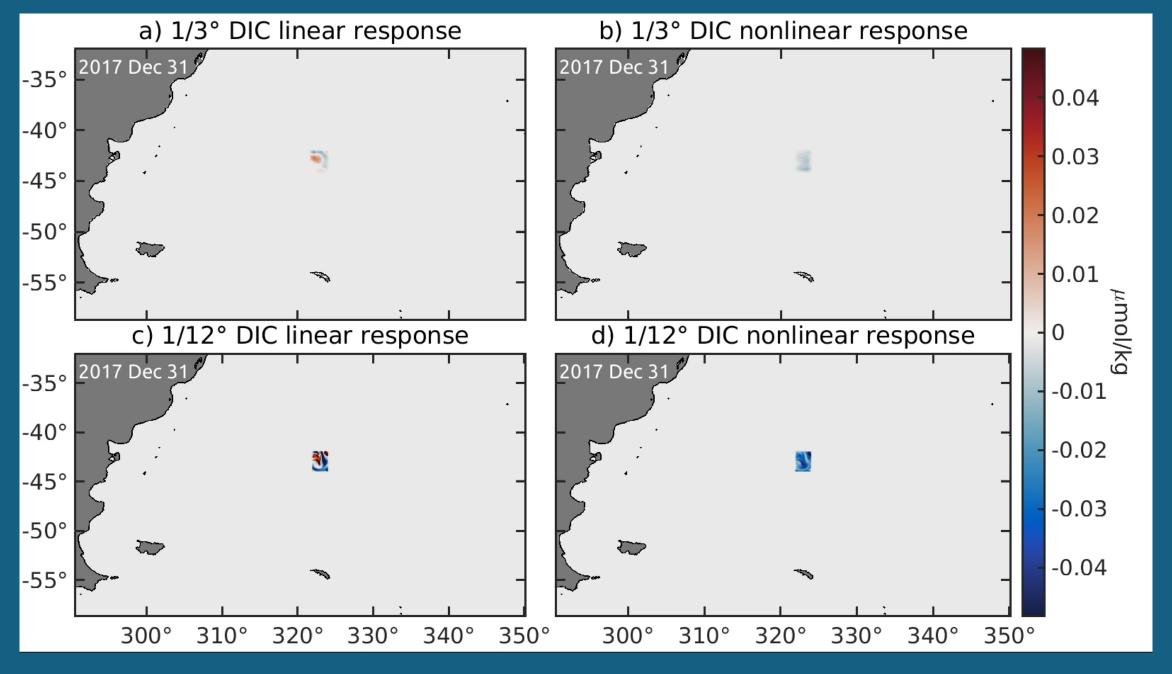
Forecast errors due to amplification of initial condition errors by flow instabilities.

- What skill can be achieved for what lead times?
- May be addressed by increasing or reallocating observations, which requires understanding their structure. What and where to observe to maximize predictability?
- Can we control/damp the chaotic behavior and then parameterize the impact?
- Can we derive a nudging in the form of a parameterization of the chaotic processes?

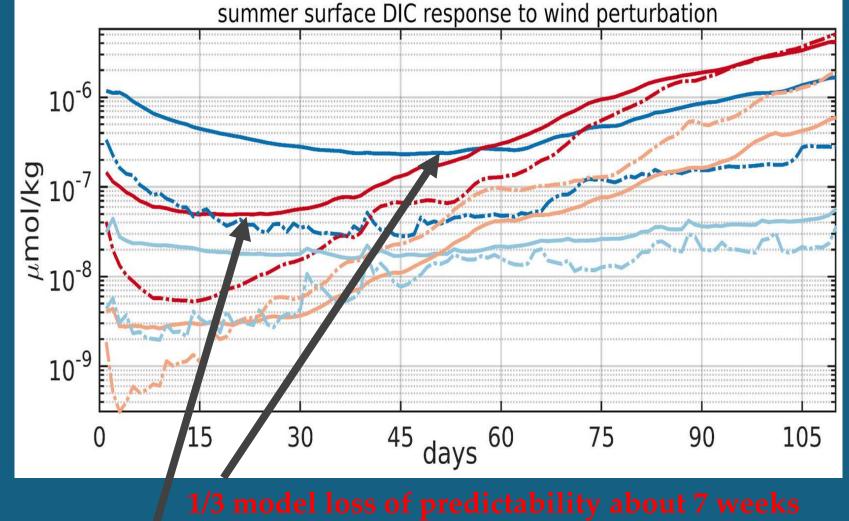
Investigating Predictability of DIC and SST in the Argentine Basin Through Wind Stress Perturbation Experiments Swierczek et al, 2021



Sources of error: Forecast errors due to amplification of IC errors by flow instabilities.



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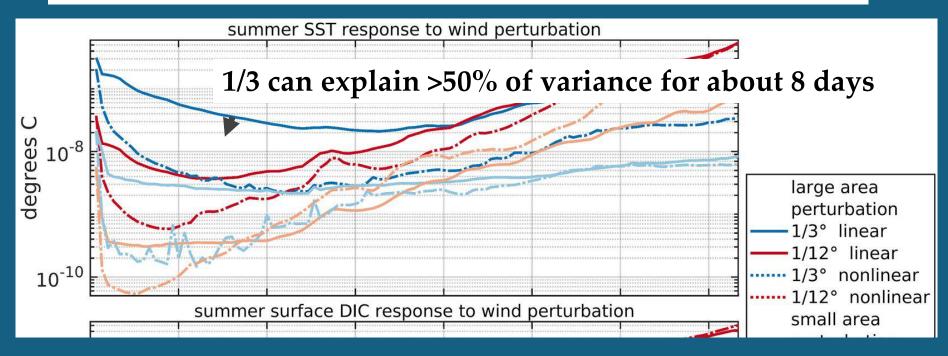
1/3° nonlinear
1/12° nonlinear
small area
perturbation
1/3° linear
1/12° linear
1/3° nonlinear
1/12° nonlinear

-1/12° linear

1/3 model loss of predictability about 7 weeks
1/12 model loss of predictability about 3 weeks

Sources of error: Forecast errors due to amplification of IC errors by flow instabilities.

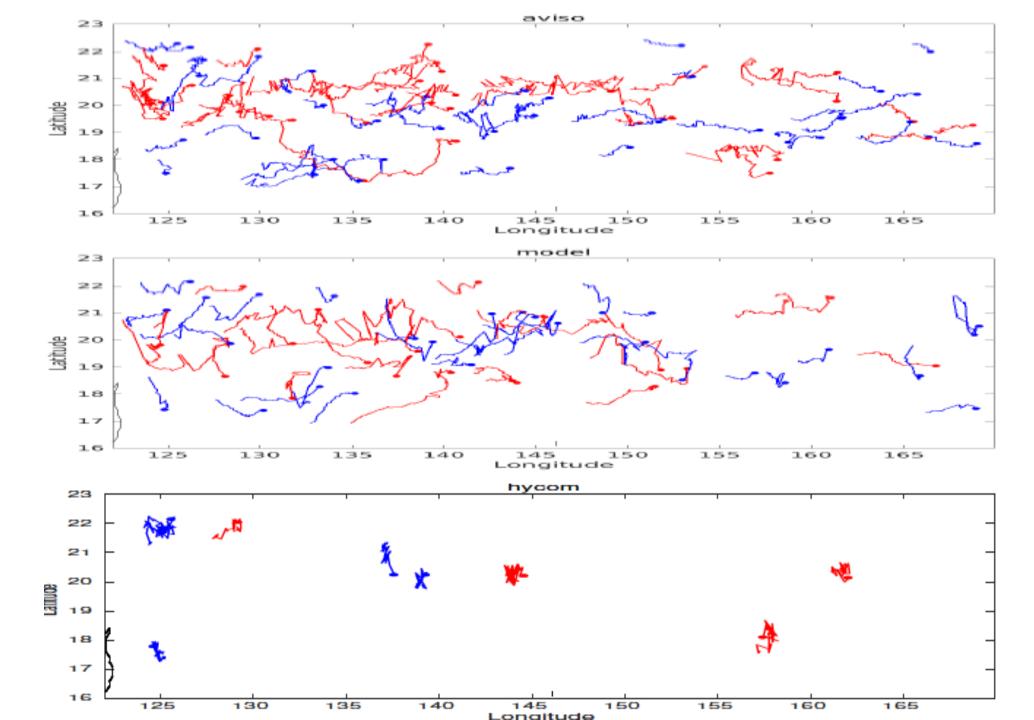
Fraction of variance explained of 1/12 response by 1/3 model

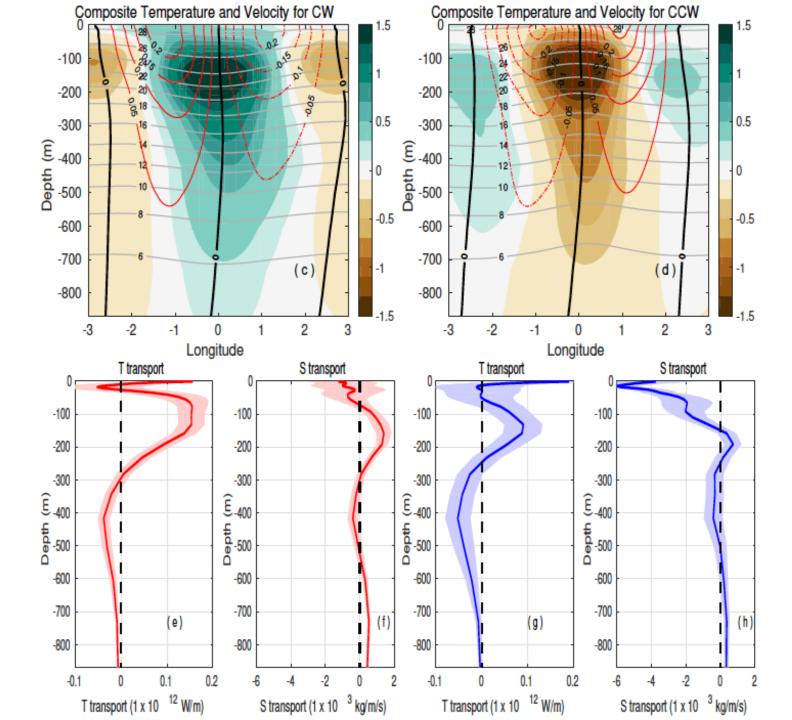


- The 1/3 is more predictable.
- But the 1/3° model response is only consistent with the 1/12° model for about 8 days calling into question the potential predictive skill of the coarser model at longer lead times.

Can we derive a nudging in the form of a parameterization of the chaotic processes?

HYCOM/NCODA MIT state estimates **AVISO** April 2011 Sea surface height (m) along 20°N. Eddy detection applied with criteria that lifetime greater than 4 weeks. The eddy locations -0.2 shown in blue (CCW) and black -0.4 (CW) April 2010 Longitude (deg) Longitude (deg) Longitude (deg)

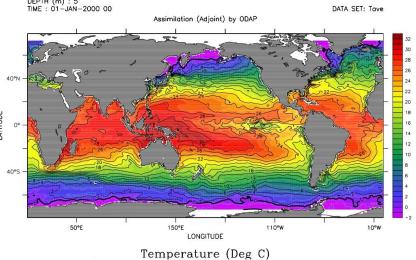


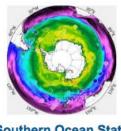


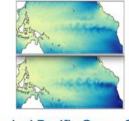


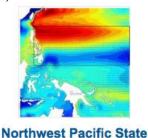
ECCO software enables numerous applications

Flagship: 1992 to present production run









Estimate (NWPac)

Southern Ocean State Estimate (SOSE)

California Current System State Estimate (CASE)

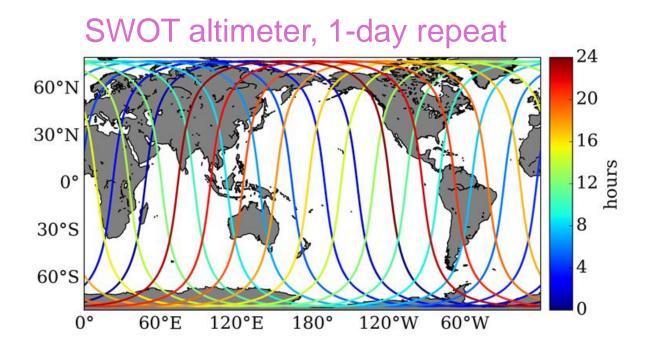
Estimate (TPOSE)

Tropical Pacific Ocean State Gulf of Mexico State Estimate (GoM)

Targeted efforts can enable

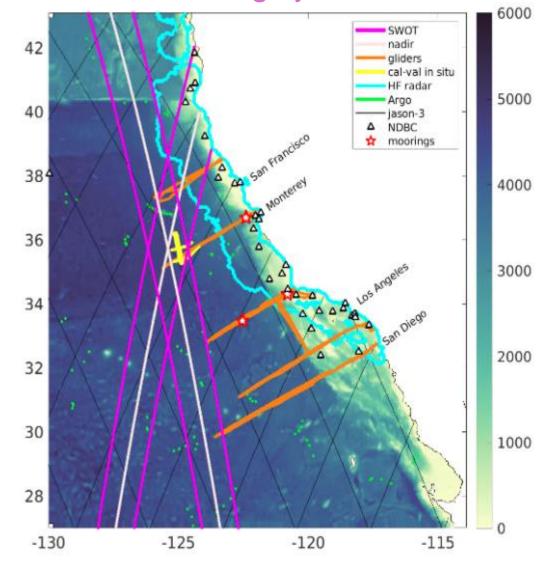
- Reproducing eddies: currently intrinsic eddies only controllable with initial conditions.
- Consistency with specific system components (e.g. BGC, coast, internal waves)
- Consistency in specific regions or times, or with specific obs platforms. Here the optimization problem can be framed to prioritize consistency
- Observing system design
- Forecasting: goal of state estimate is to provide forecast initialization
- Development

California Current System (CCS)



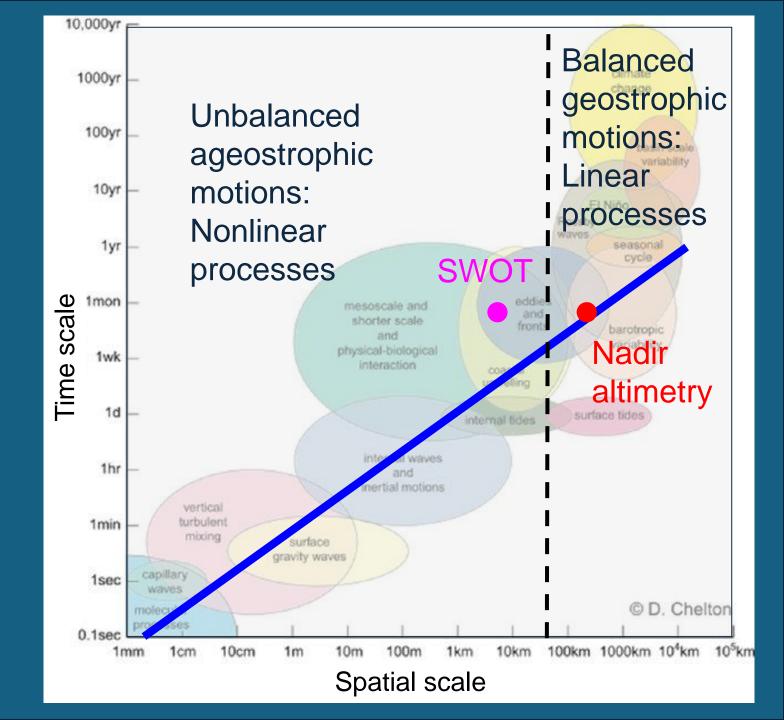
4D-Var assimilation for SWOT: from geostrophic balance to km-scale dynamics

Observing system

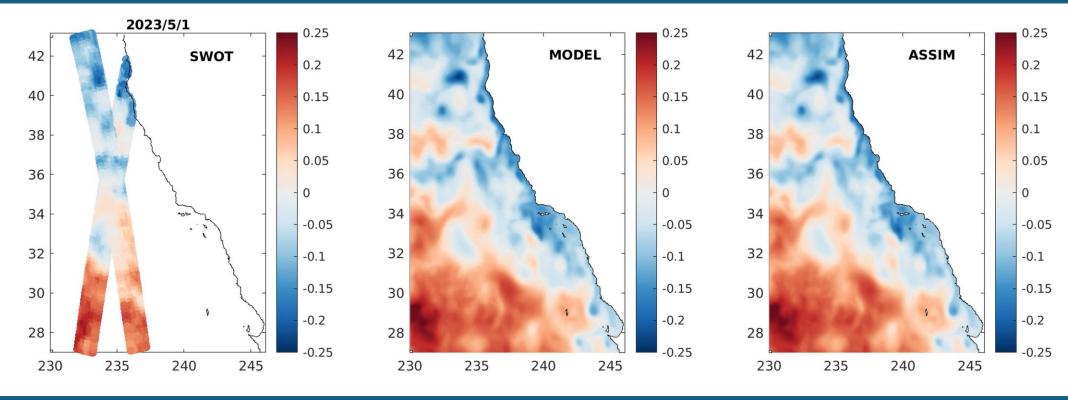


The SWOT Challenge

- How do we take advantage of SWOT's spatial resolution, without corresponding temporal resolution?
- Our goal: with 4D-Var
 assimilation, use data to
 constrain dynamics



Observed and modeled SSH



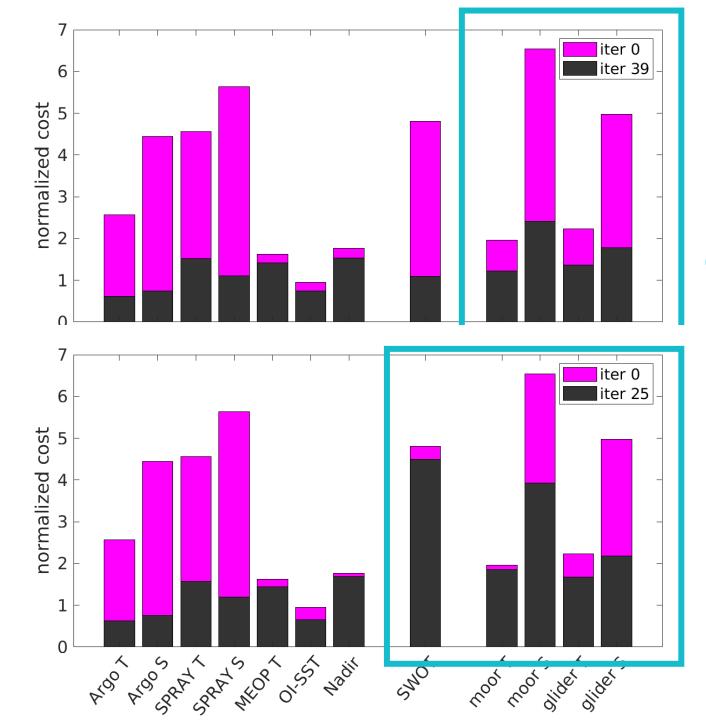
Observations

Forward model run (iteration 0)

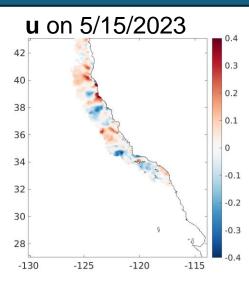
State estimate (iteration 39)

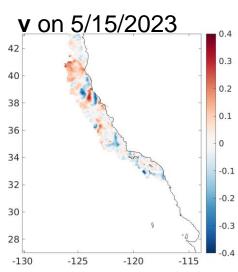
SWOT assimilated

SWOT withheld

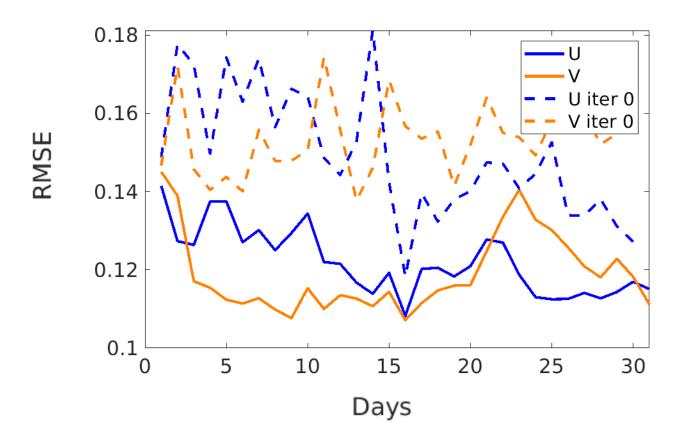


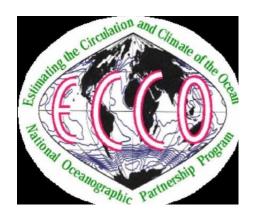
Independent data (not assimilated)





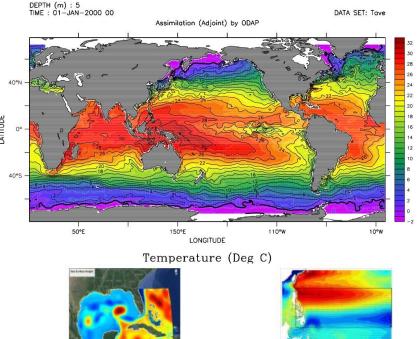
HF radar misfits vs time

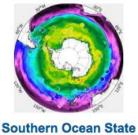


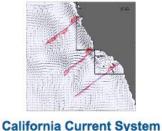


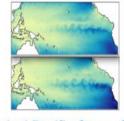
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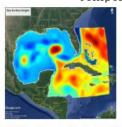
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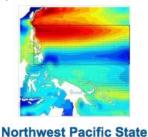












Estimate (NWPac)

State Estimate (CASE) Estimate (SOSE)

Estimate (TPOSE)

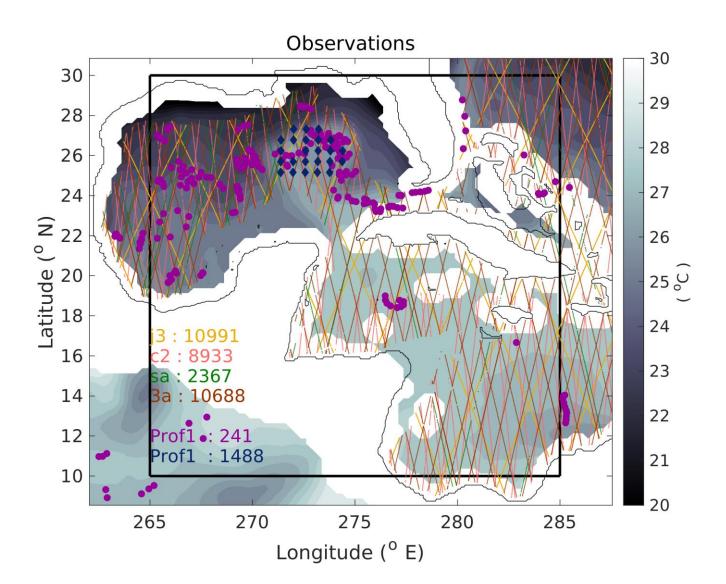
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Gulf of Mexico State Estimation and Prediction



Observing system design for loop current and eddy shedding

- SSH along-track anomalies
- Gridded OI-SST
- Argo program T/S profiles
- CPIES T/S profiles

Forecasting the loop current Gopalakrishnan et al 2013

