The MARACOOS ROMS 4D-Var "Doppio" Operational Data Assimilative Ocean Forecast System for Northeast U.S. Coastal Waters John Wilkin

with: Hernan Arango, Eli Hunter, Julia Levin, Alex Lopez, Andrew Moore and Javier Zavala-Garay

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Rutgers

2019 ROMS 4D-Var Training Workshop NOAA Center for Weather and Climate Prediction (NCWCP), July 29 to August 2, 2019

Regional Ocean Modeling System

www.myroms.org

The coast is at the center of human-ocean interactions

Safety



(photo: USCG/ T Sperduto)

- Ocean rescue
- Storm surge
- Oil spill tracking



Oil-covered rocks at Refugio State Beach (photo: NOAA)

Economy

Maritime economy contributed \$282 billion & 2.8 million jobs to the U.S. economy in 2011.

- Sustainable fisheries
- Harmful Algal Bloom tracking and prediction
- Offshore energy design, development and operation
- Navigation systems prevent groundings in busy ports
- Forecast ocean conditions for the fishing fleet (safety, harvest) and fishing surveys
- Predicting ocean acidification events for shellfish hatcheries

Environment



"Chá bă," a new buoy off the coast of Washington, contributes better information about the ocean conditions that oysters can and cannot tolerate. (photo: J Payne, Pacific Ocean Shelf Tracking Project)

IOOS Integrated Ocean Observing System

The coast is at the center of human-ocean interactions

Safaty

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Economy

When operated as real-time now-cast or forecast systems ... high-resolution coastal models that <u>downscale global analyses</u> and incorporate <u>added dynamics</u>, forcing, <u>data assimilation</u>, and inter-disciplinary interactions (e.g. sediments, ecosystems)

- provide <u>actionable guidance</u> for decisionmaking related to water quality, public
 health, coastal flooding, shipping,
 maritime safety, and other applications.
 - Predicting ocean acidification events for shellfish hatcheries

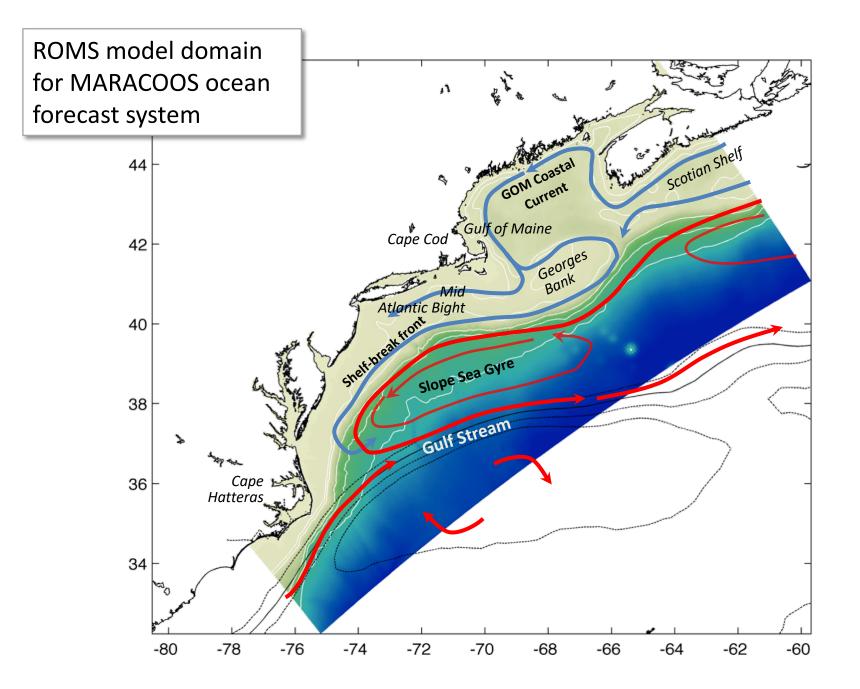
Fnvironment

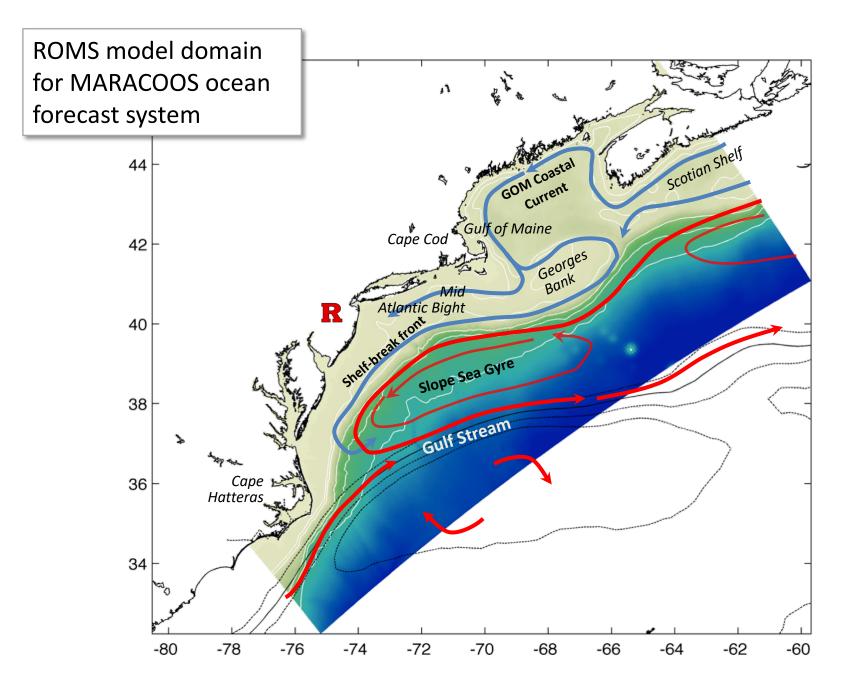


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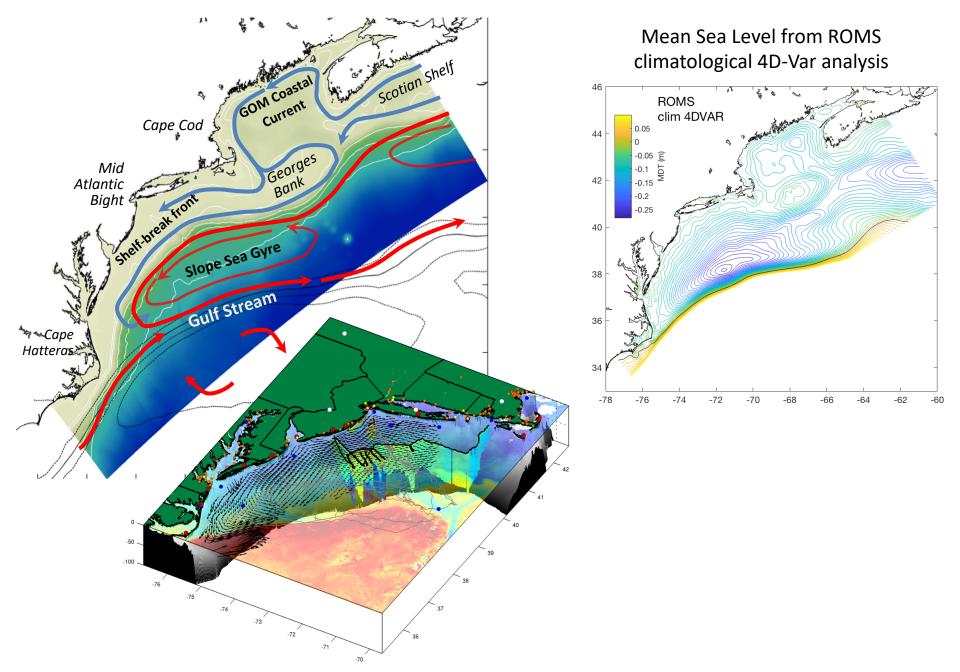
Oil-covered rocks at Refugio State Beach (photo: NOAA)



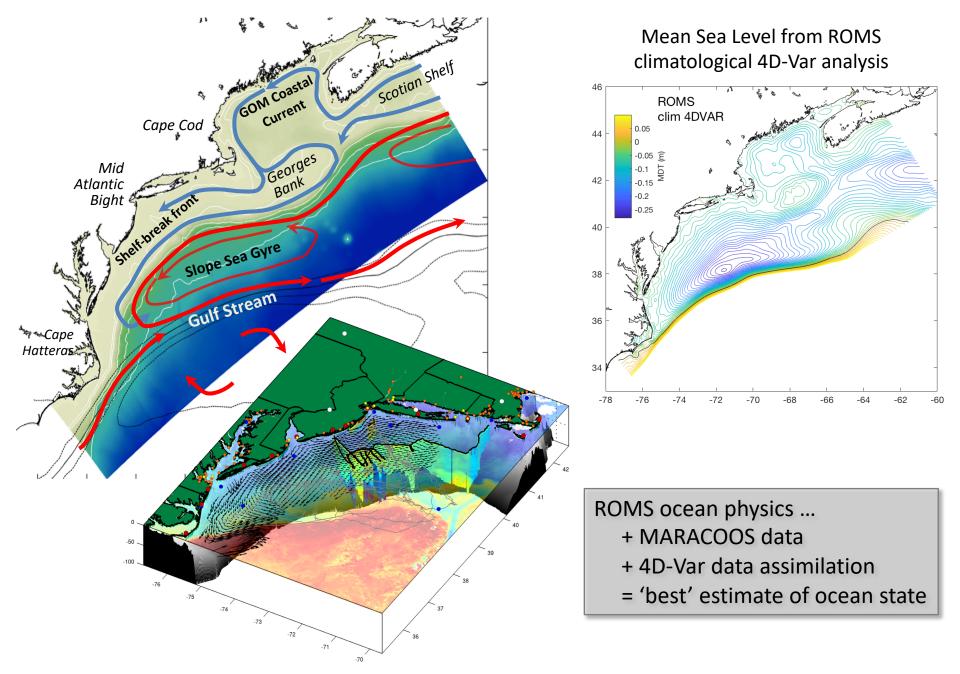




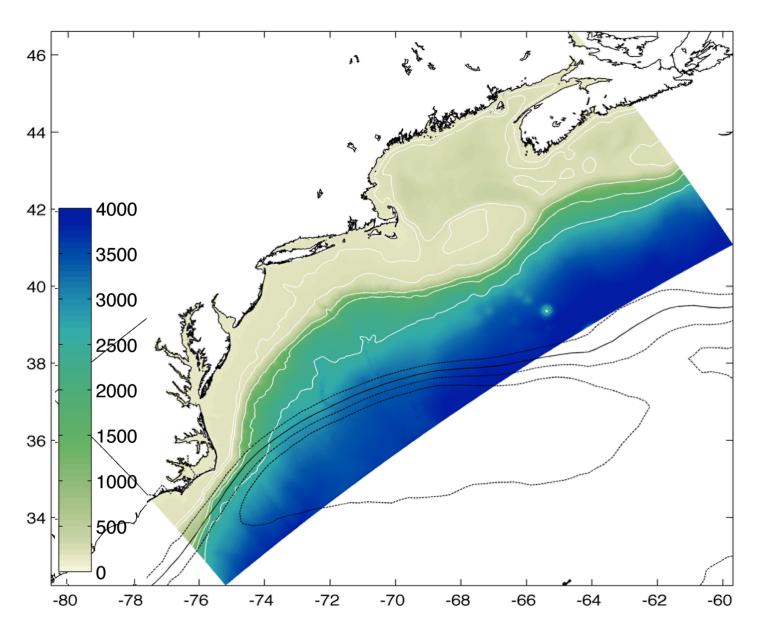
MARACOOS ROMS: Real-time forecast for Mid-Atlantic Bight and Gulf of Maine



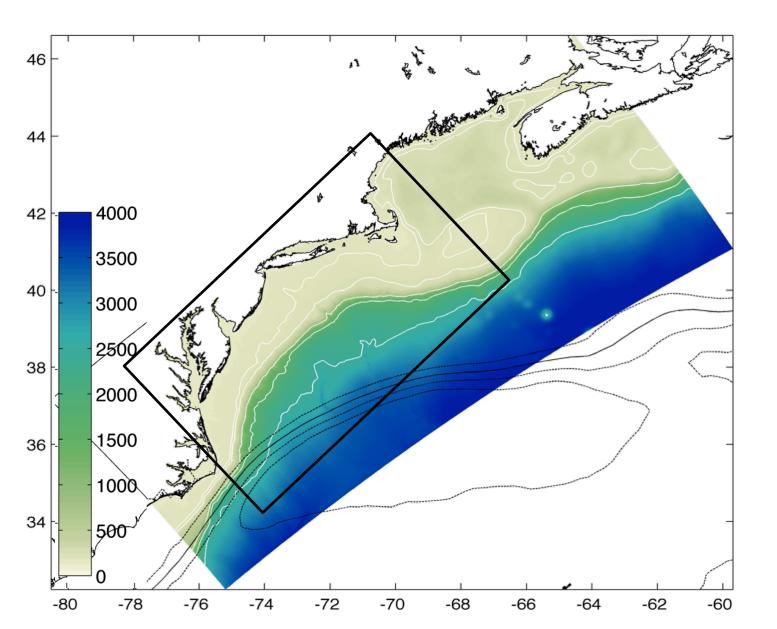
MARACOOS ROMS: Real-time forecast for Mid-Atlantic Bight and Gulf of Maine



Doppio* real-time and reanalysis ROMS system for the Mid-Atlantic Bight and Gulf of Maine



Doppio* real-time and reanalysis ROMS system for the Mid-Atlantic Bight and Gulf of Maine



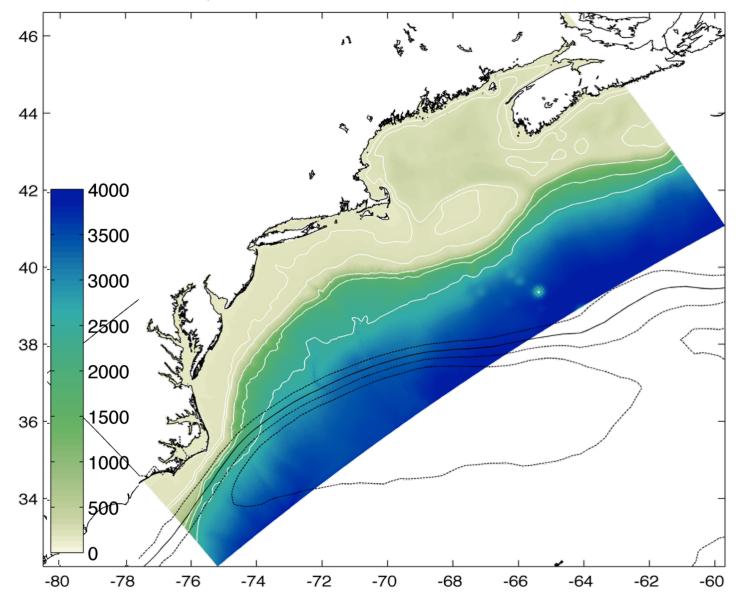


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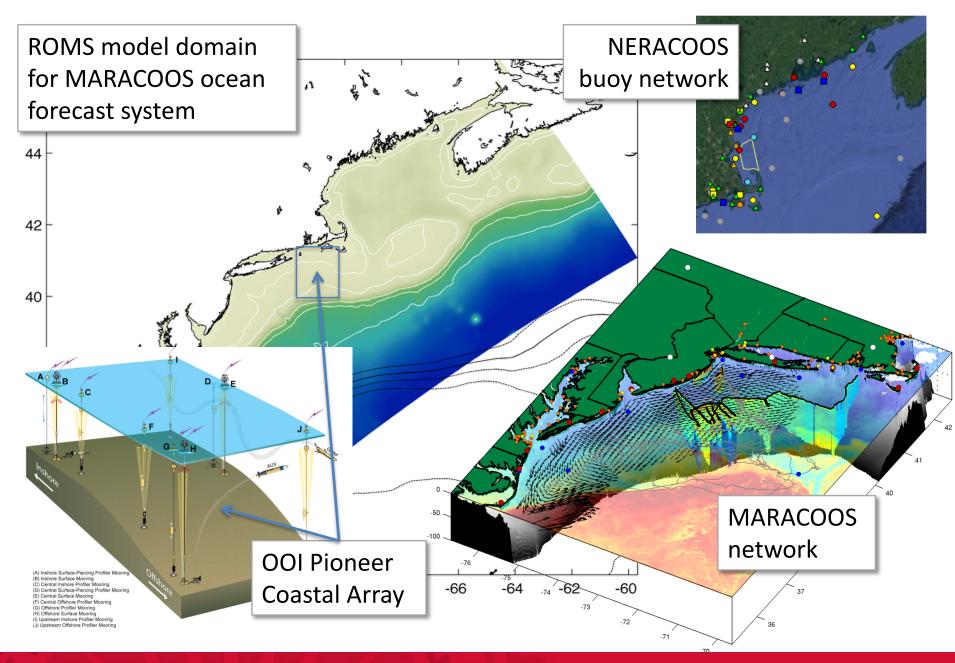
**Experimental System for Predicting Shelf and Slope Optics

Doppio* real-time and reanalysis ROMS system for the Mid-Atlantic Bight and Gulf of Maine

* It's a double espresso

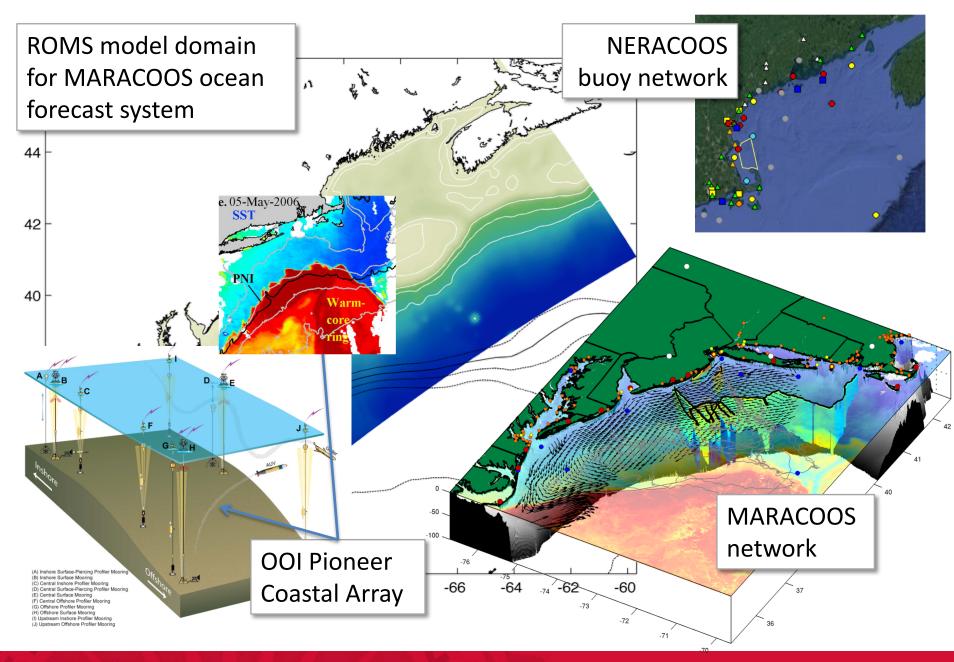






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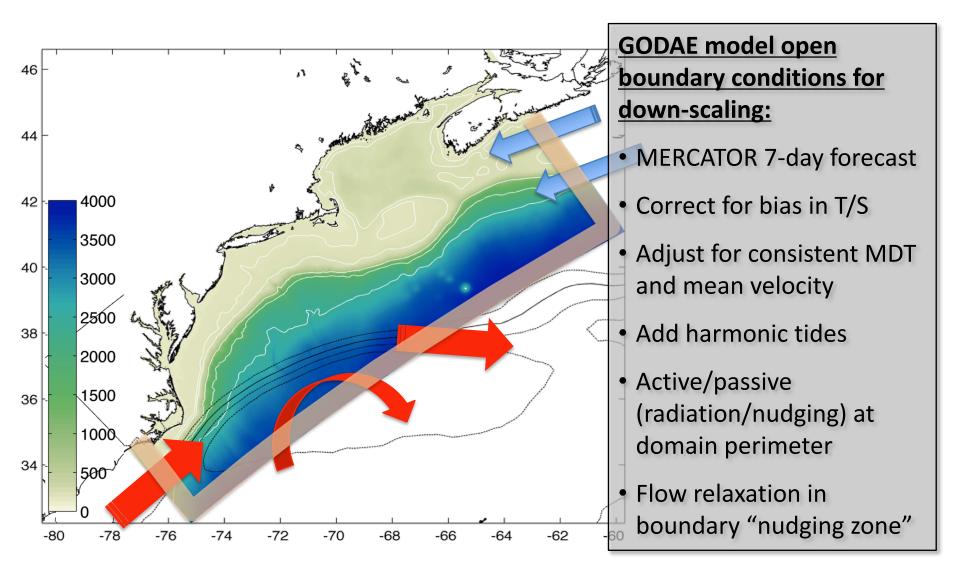
maracoos.org | oceansmap.maracoos.org Mid-Atlantic Regional Association of Coastal Ocean Observing Systems



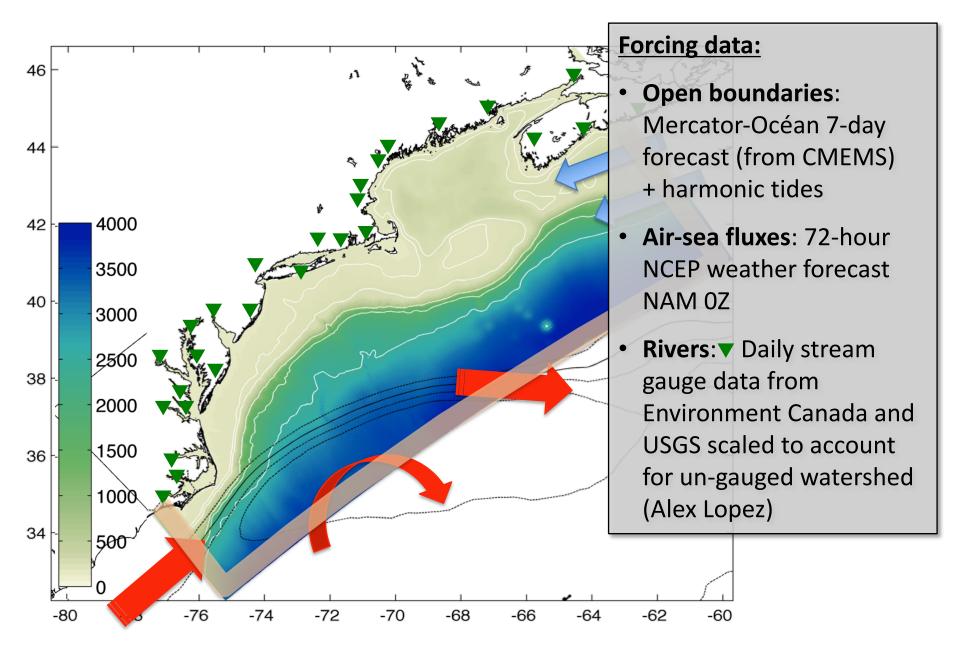
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maracoos.org | oceansmap.maracoos.org Mid-Atlantic Regional Association of Coastal Ocean Observing Systems

MARACOOS ROMS: Real-time forecast for Mid-Atlantic Bight and Gulf of Maine



MARACOOS ROMS: Real-time forecast for Mid-Atlantic Bight and Gulf of Maine



<u>MARACOOS ROMS</u>: Data streams routinely assimilated into the near-real-time 4D-Var 'best' analysis of Mid-Atlantic Bight and Gulf of Maine ocean conditions

Observation type and platform	Source	Sampling frequency and resolution	Latency
AVHRR infrared SST	MARACOOS.org and NOAA CoastWatch	4 passes per day, 1 km	2 hr
GOES infrared SST	NOAA Coast- Watch	hourly, 6 km	12 hr
AMSR2 and WindSat microwave SST	NASA JPL PO- DAAC	daily, 15 km	24 hr
SSH: 4 satellite altimeters: Jason, AltiKa, CryoSat and Sentinel-3 with coastal corrections	Radar Altimeter Database System at TU Delft	~1 pass daily in do- main, ~4 km	4 hr
in situ T, S on GTS from National Data Buoy Center buoys, Argo floats, shipboard XBT, surface drift- ers	NOAA Observing System Monitor- ing Center (OSMC)	varies with platform	~12 hr
Surface currents from CODAR HF- radar	MARACOOS.org	hourly, 1km	4 hr
Glider T, S \sim 1-2 deployments per month in domain by MARACOOS	IOOS Glider Data Assembly Center	dense along trajectory	2 hr

Table 21.1. A summary of the observational data streams accessed for the ROMS Doppio near real time data assimilation system.

<u>MARACOOS ROMS</u>: Data streams routinely assimilated into the near-real-time 4D-Var 'best' analysis of Mid-Atlantic Bight and Gulf of Maine ocean conditions

Typically 200,000 platform observations in a 3-day analysis cycle sst	Source	Sampling frequency and resolution	Latency
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Some details ...

Bias removal

- Removing bias from open boundary conditions is crucial
- 4D-Var will not converge if it cannot reconcile model and data error
- Co-variances embodied in the Adjoint and Tangent Linear physics are incorrect if the background state is biased

Open Boundary Conditions

 (T and S) from MERCATOR are adjusted by matching mean to regional climatology (MOCHA)

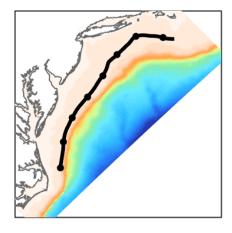
4DVAR analysis of mean climatological

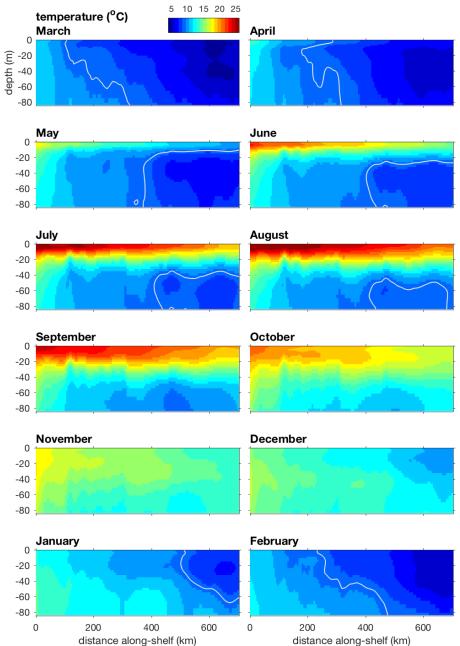
ocean state

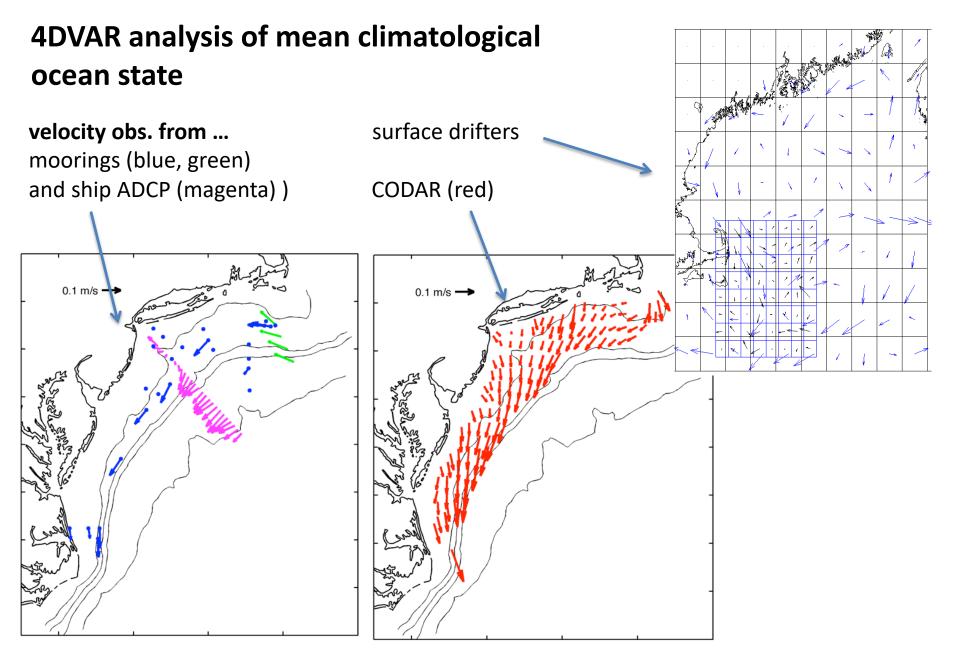
Combine ...

High-res regional T/S climatology (MOCHA)

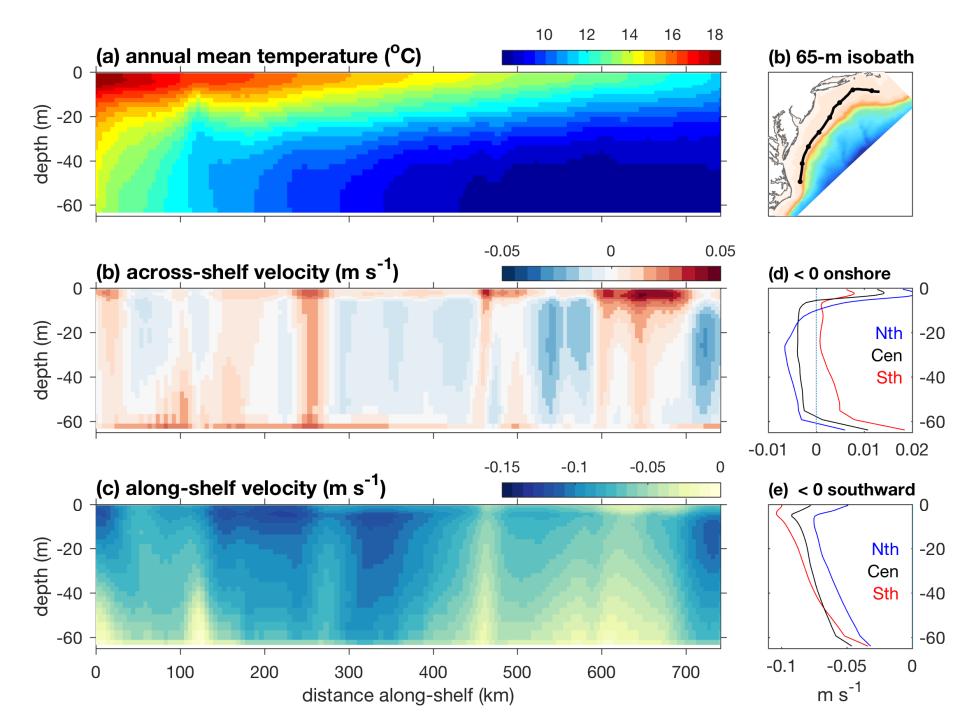
with ...

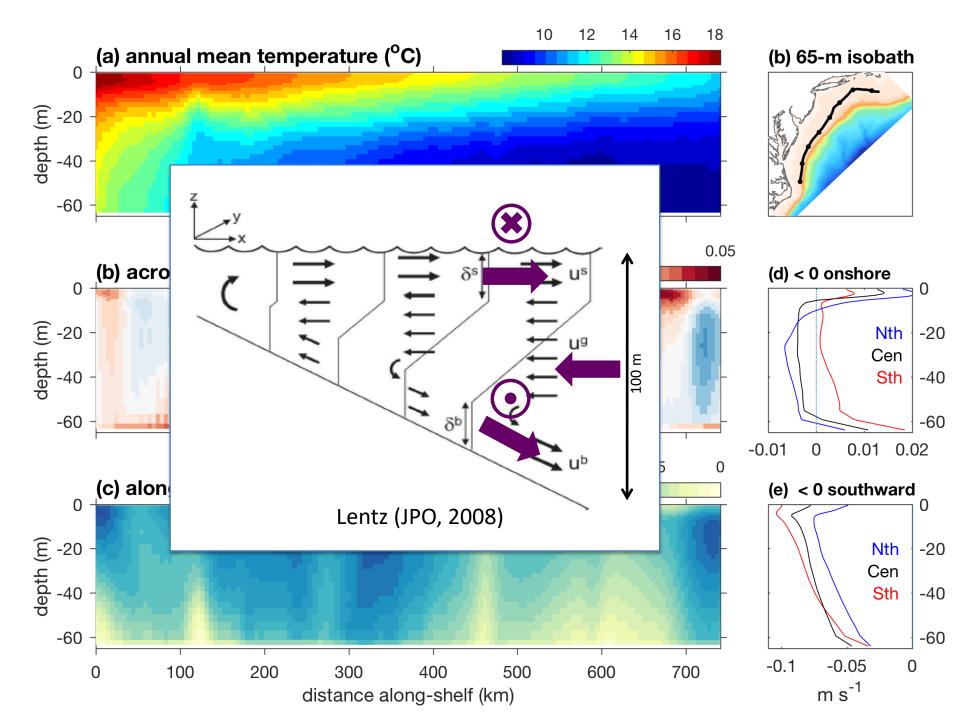




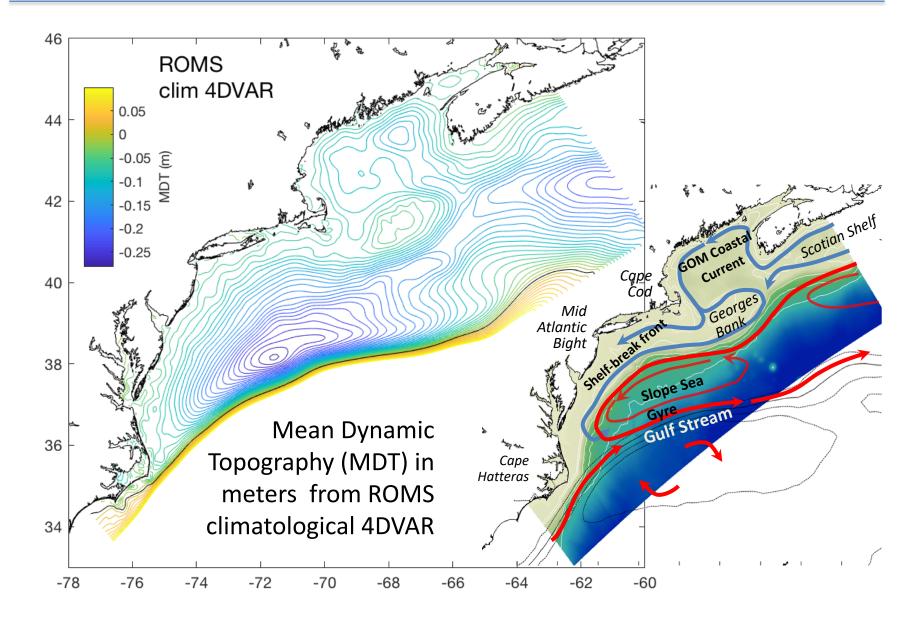


... to get ... >>

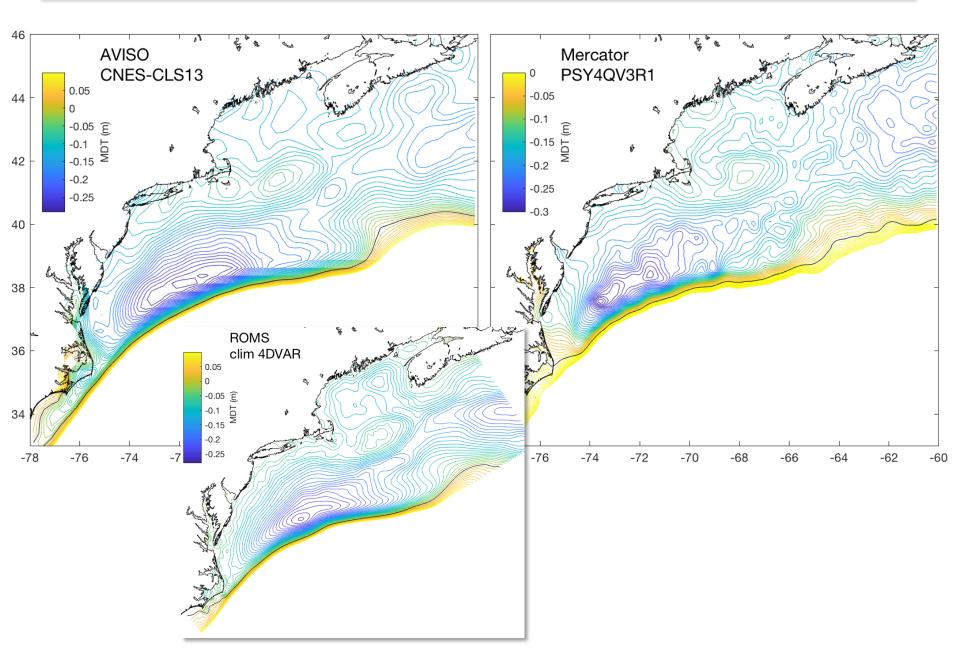




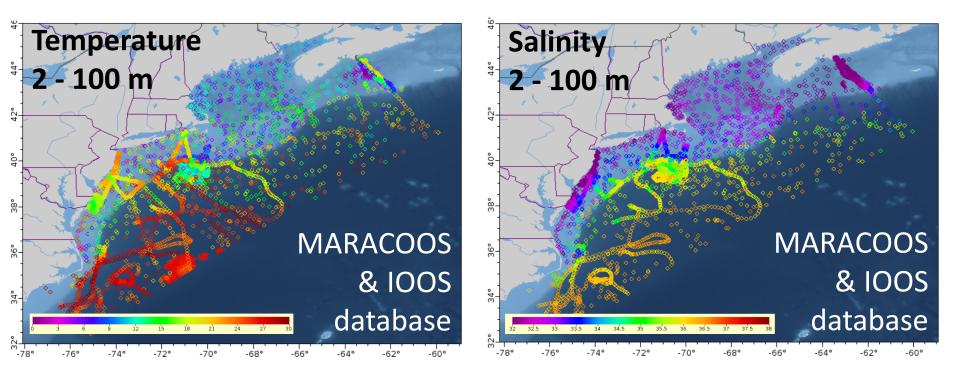
Mean Dynamic Topography of the MAB and GoM

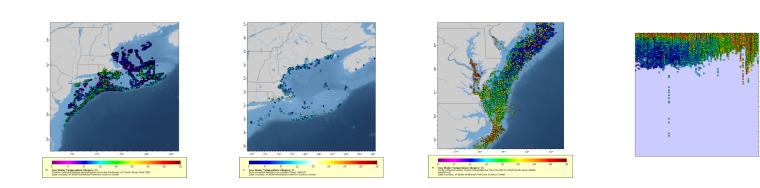


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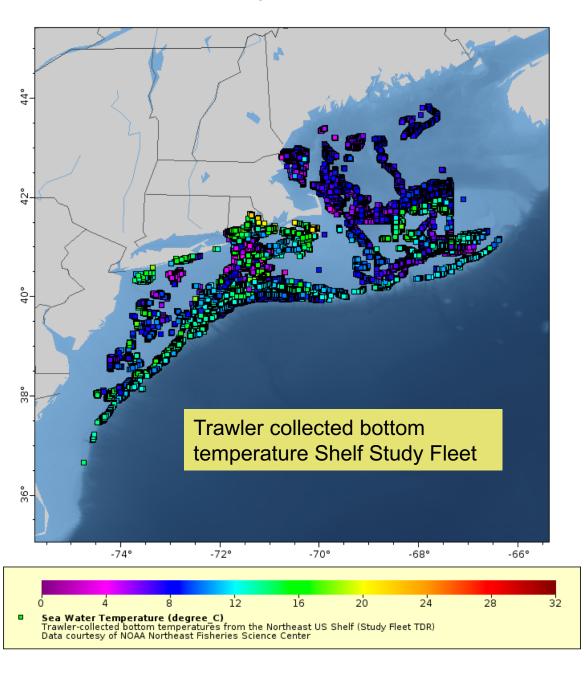


Subsurface in situ ocean observations for the northeast U.S. in 2015

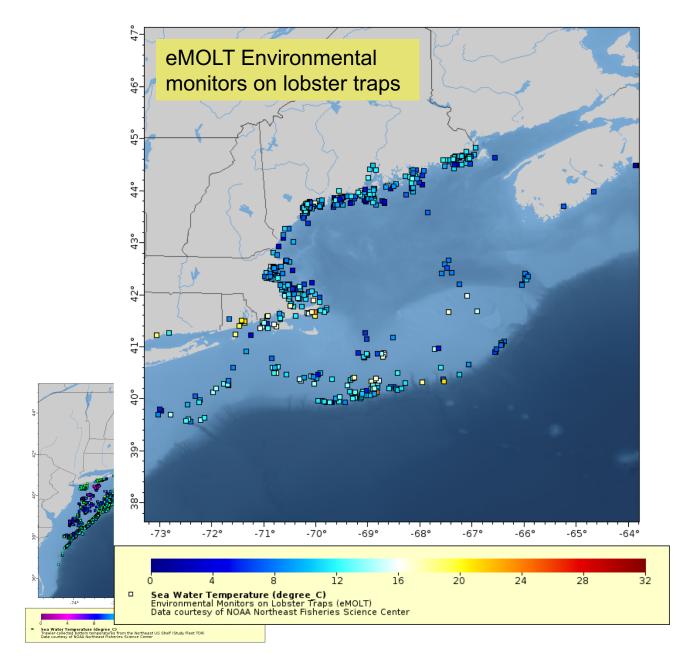




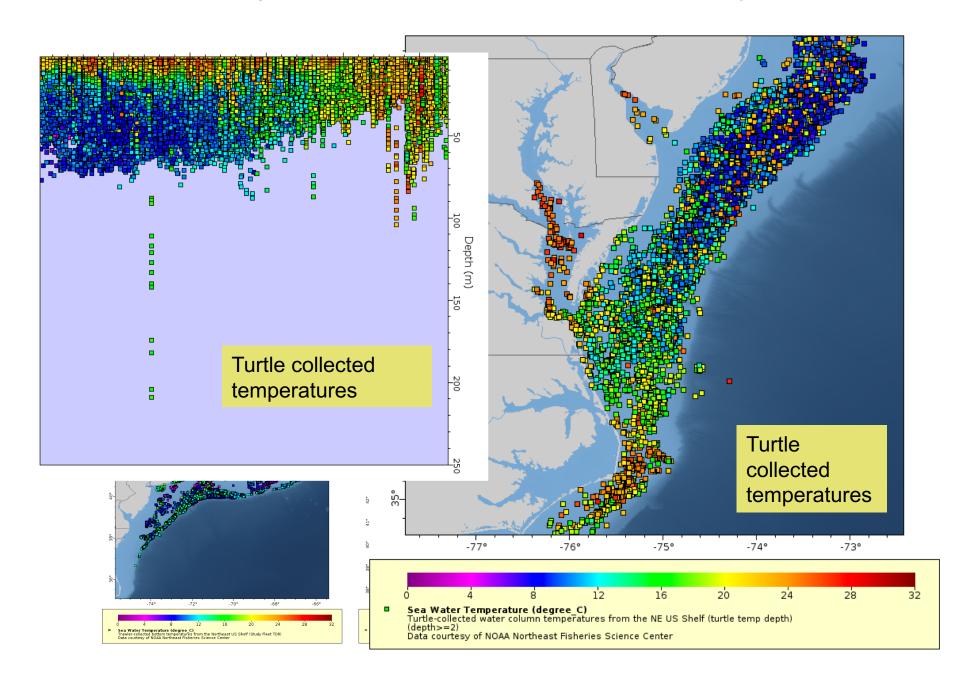
Subsurface in situ ocean observations - delayed



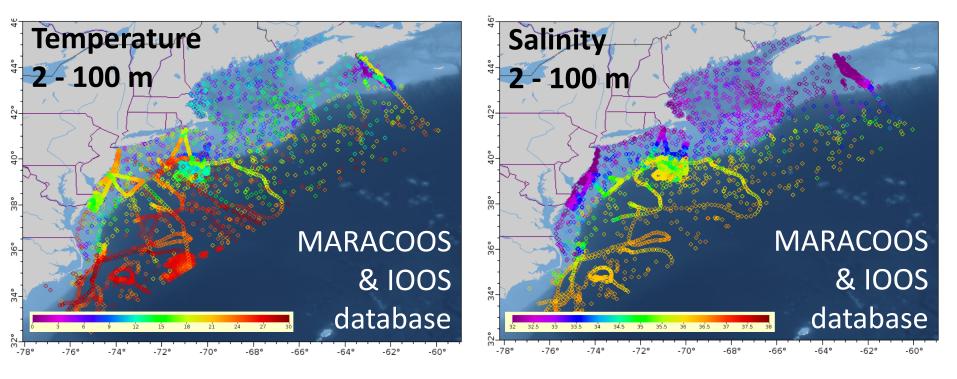
Subsurface in situ ocean observations - delayed

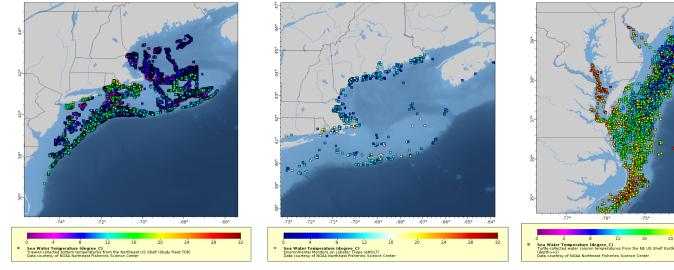


Subsurface in situ ocean observations - delayed

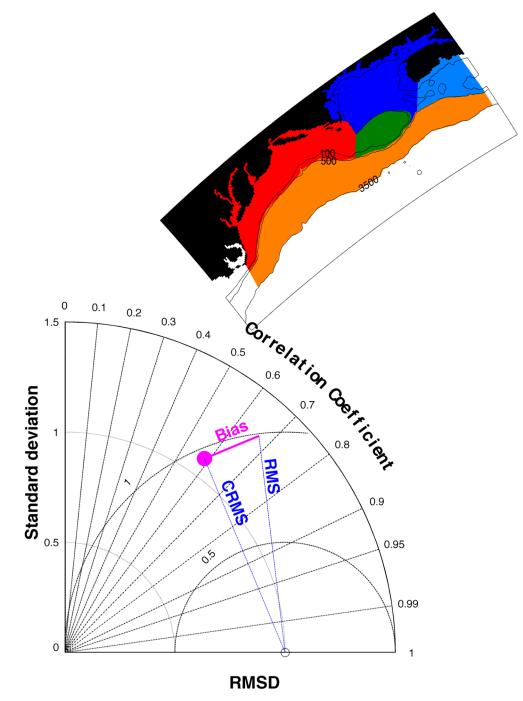


Subsurface in situ ocean observations – model skill assessment

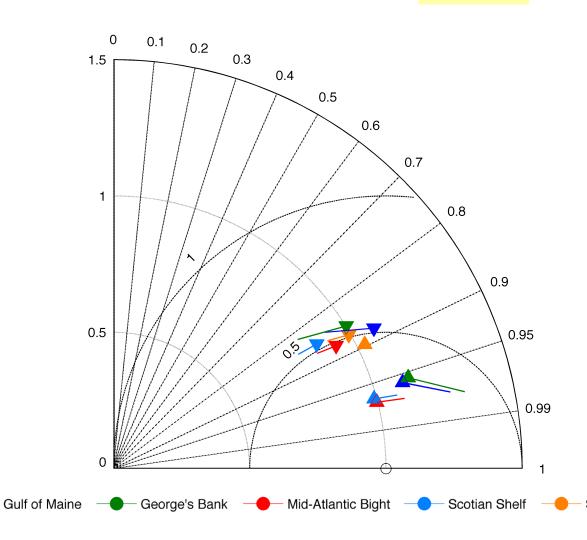




Run	Explanation
Control	The optimal configuration, used as benchmark for other cases
TRMM	Uses precipitation from NASA's Tropical Rainfall Measuring Mission instead of from NCEP's NARR
Unbias corrected Mercator	Uses open boundary files from Mercator-ocean without bias correction to annual mean
Uwind Minus Current	Surface stress calculated from wind relative to surface current
Unbias corrected HYCOM	Uses open boundary files from HYCOM without bias correction to the annual mean



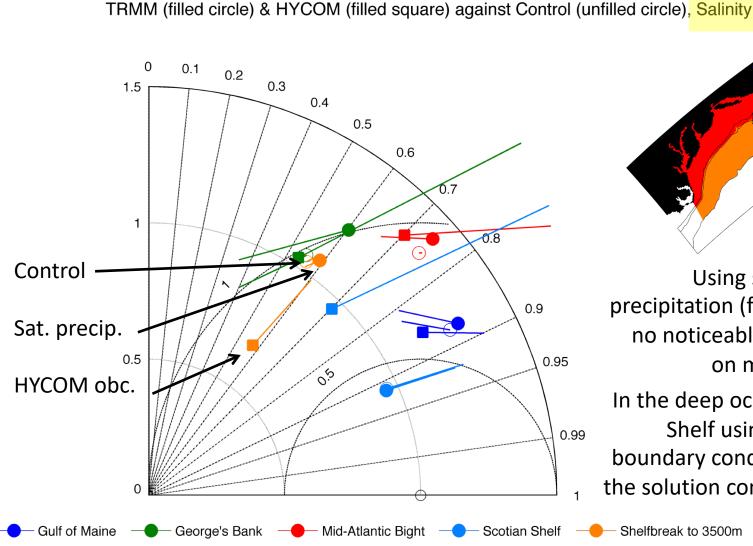
Control (filled), SST (Δ) & Subsurface Temperature (∇)



100 × 0 500 × 0

The control case best demonstrates DOPPIO skill when it comes to temperature, both at the sea surface and below. The number of available observations between the two varies by orders of magnitude Shelfbreak to 3500m

A. Lopez

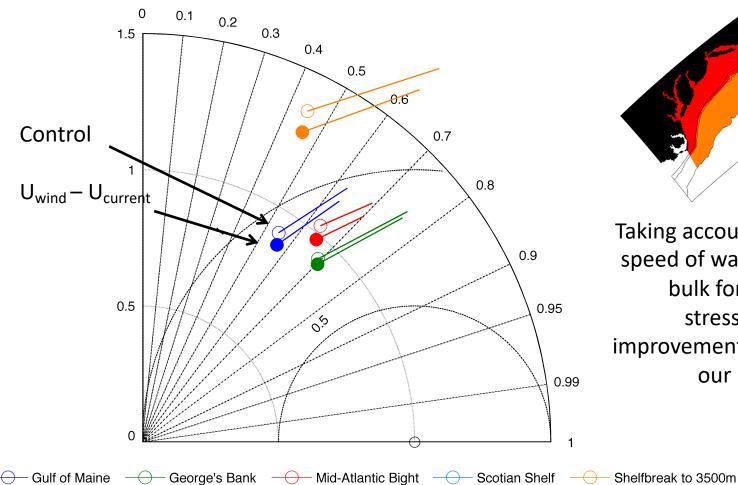


Using satellite derived precipitation (filled circles) has no noticeable positive affect on modeled salinity.

In the deep ocean and Scotian Shelf using HYCOM open boundary conditions degrades the solution compared to using MERCATOR

A. Lopez

Uwind Minus Current (filled) against Control (unfilled), Complex Velocity

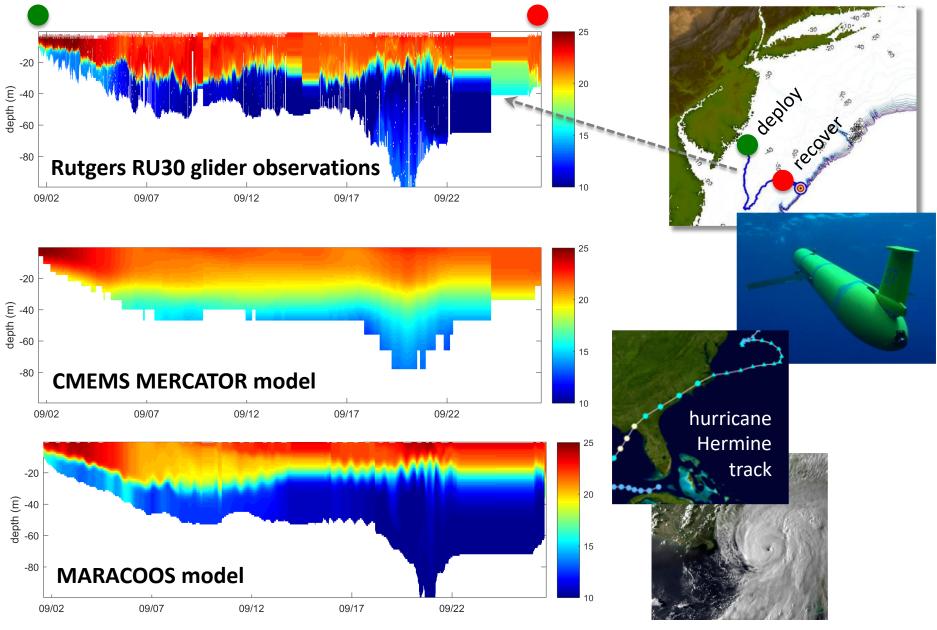


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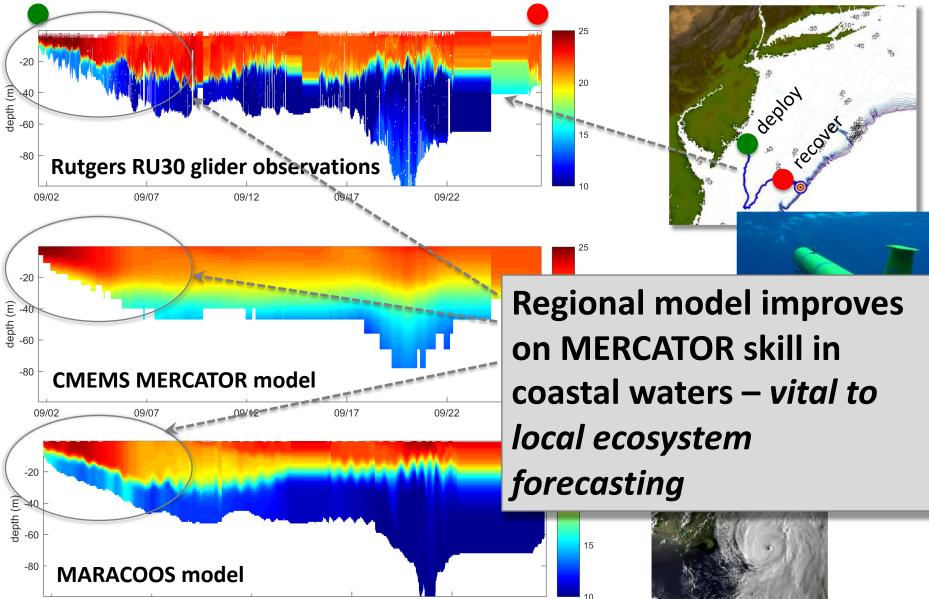
Taking account of the relative speed of water and air in the bulk formula for surface stress brings a modest improvement, so this became our new control case

A. Lopez

Autonomous underwater glider observations and model predictions during hurricane Hermine, August 2016



Autonomous underwater glider observations and model predictions during hurricane Hermine, August 2016



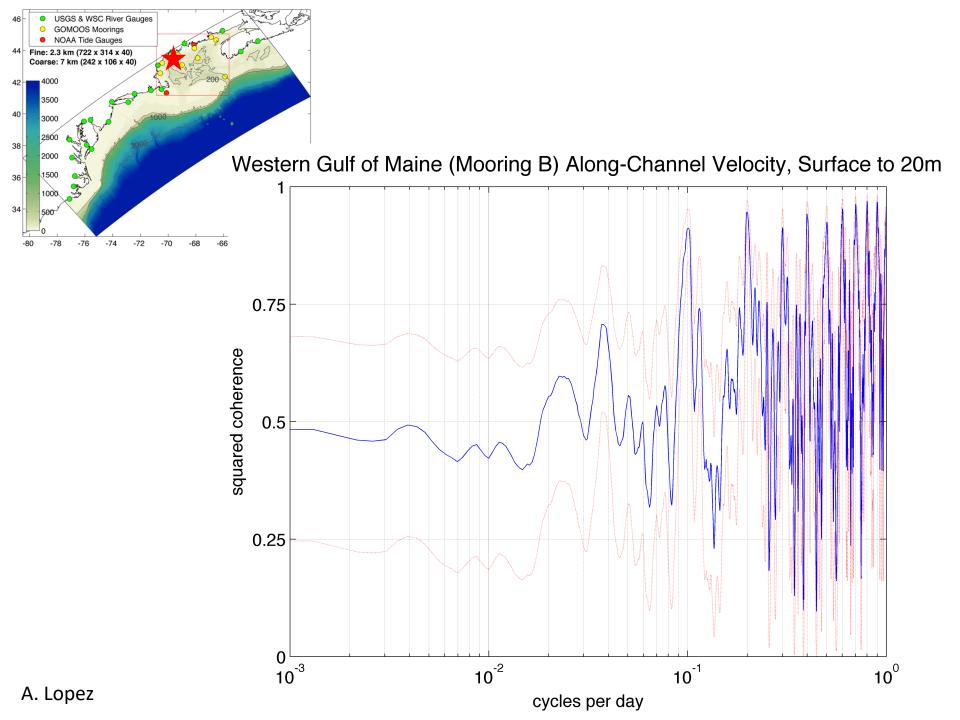
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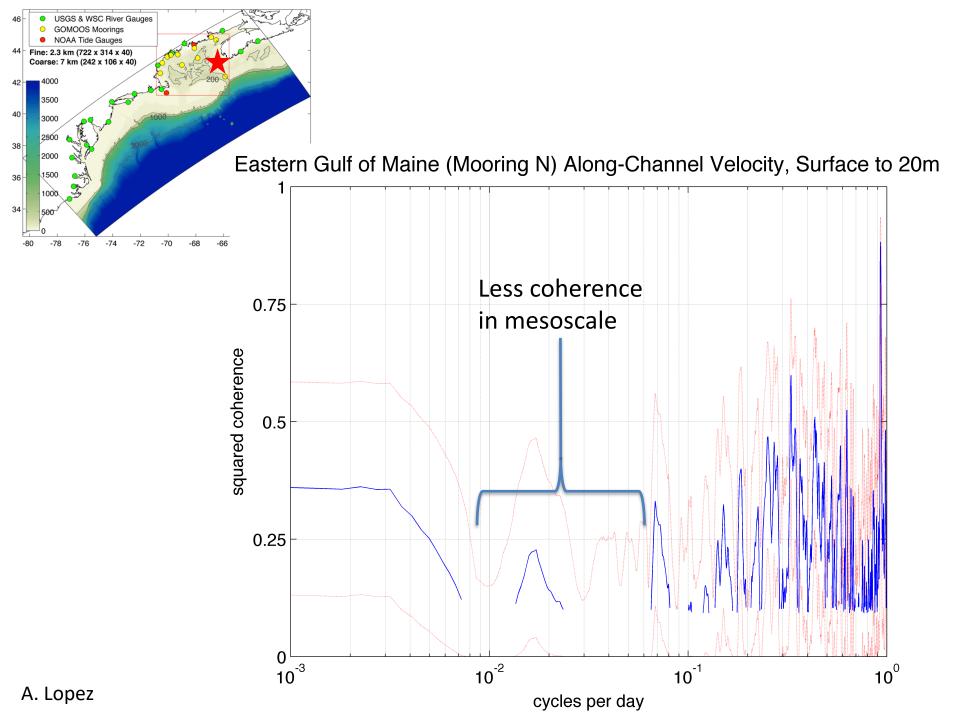
09/12

09/07

09/17

09/22





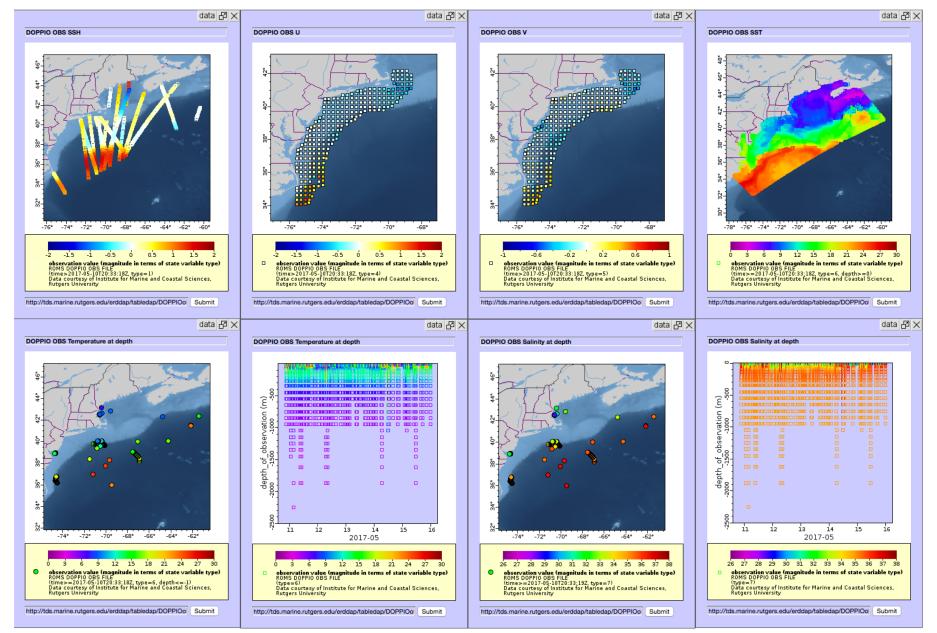
Easier access to scientific data

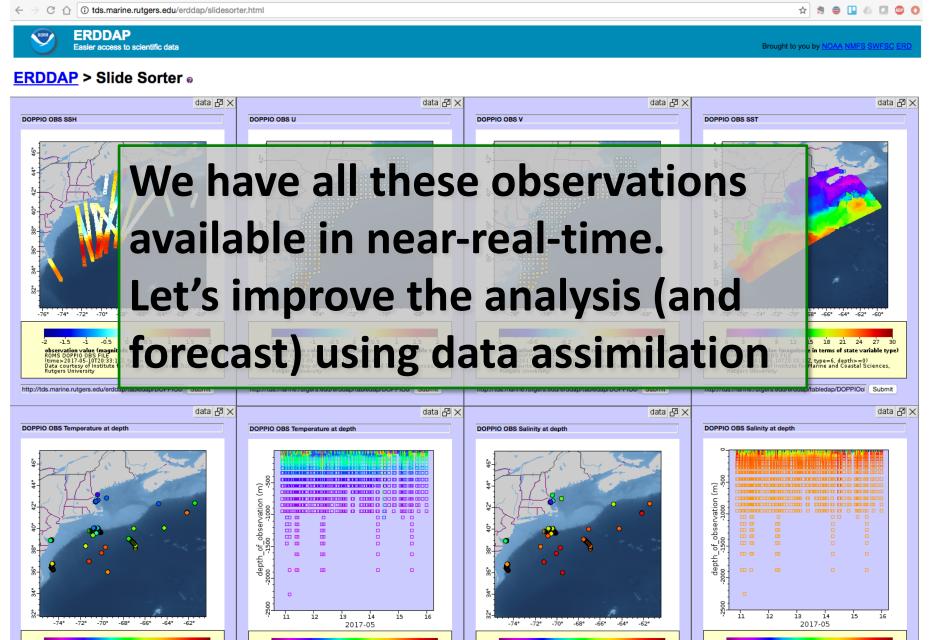
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28 29 30 31 32 33 34 35 36 37 38

observation value (magnitude in terms of state variable type)

ROMS DOPPIO OBS FILE (time>=2017-05-10T20:33:19Z, type=7) Data courtesy of Institute for Marine and Coastal Sciences.

http://tds.marine.rutgers.edu/erddap/tabledap/DOPPIOo Submit

27

Rutgers University

26 27 28 29 30 31 32 33 34 35 36 37 38

(type=7) Data courtesy of Institute for Marine and Coastal Sciences, Rutgers University

http://tds.marine.rutgers.edu/erddap/tabledap/DOPPIOol Submit

observation value (magnitude in terms of state variable type) ROMS DOPPIO OBS FILE

9 12 15 18 21 24 27 30

observation value (magnitude in terms of state variable type) ROMS DOPPIO OBS FILE

Data courtesy of Institute for Marine and Coastal Sciences.

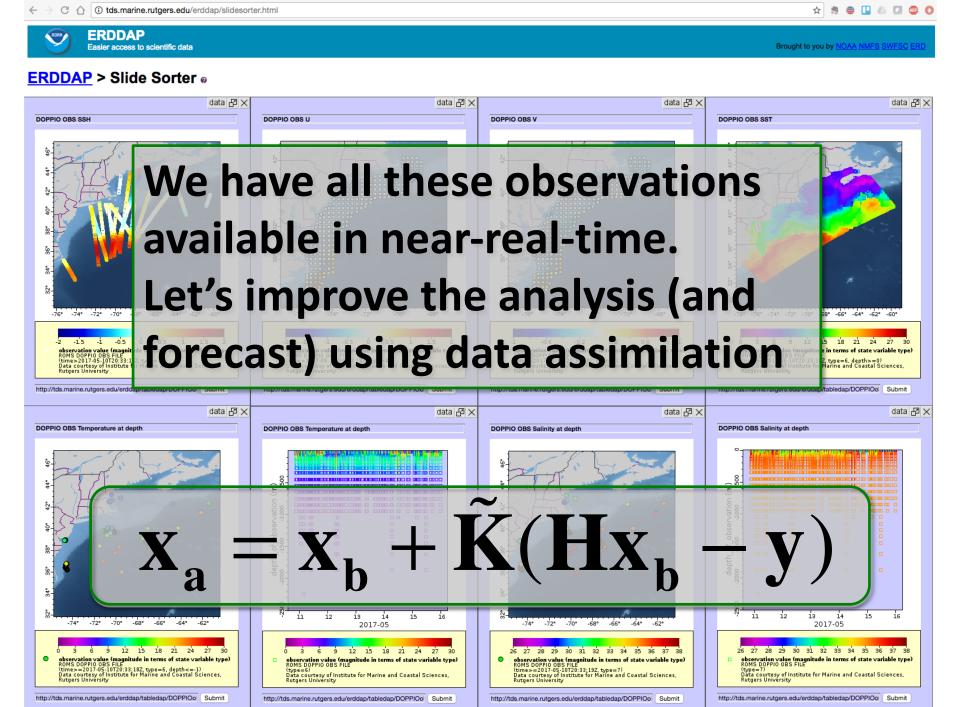
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(type=6)

Rutgers University

9 12 15 18 21 24 27 observation value (magnitude in terms of state variable type) ROMS DOPPIO OBS FILE (time>=2017.05.10720:33;18Z, type=6, depth<=-1) Data courtesy of Institute for Marine and Coastal Sciences, Rutgers University

http://tds.marine.rutgers.edu/erddap/tabledap/DOPPIOol Submit



ROMS includes three variants of 4D-Var data assimilation*

- A primal formulation of incremental strong constraint 4DVar (I4DVAR)
- A dual formulation based on a physical-space statistical analysis system (4D-PSAS)
- A dual formulation
 Representer-based variant of
 4DVar (R4DVAR)

- 4DVar can adjust initial, boundary, and surface conditions
- In reanalysis we adjust <u>initial</u> <u>conditions</u>, <u>open boundary</u> <u>conditions</u> and <u>surface fluxes</u> using I4DVAR in a 3-day cycle not overlapping
- In operational MARACOOS forecast system we adjust <u>initial</u> <u>conditions</u> <u>and surface fluxes</u> using 4DVAR PSAS during 3-day analysis, then run 3-day forecast, repeated daily (analysis windows overlap)
- * Moore, A. M., H. Arango, G. Broquet, B. Powell, A. T. Weaver, and J. Zavala-Garay (2011), The Regional Ocean Modeling System (ROMS) 4-dimensional variational data assimilations systems, Part I System overview and formulation, *Prog. Oceanog.*, *91*(34-39).

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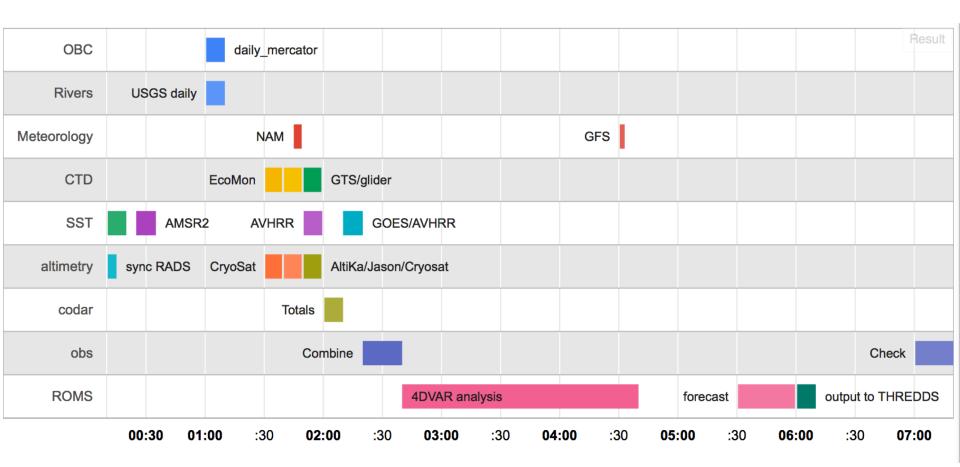
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ROMS includes three variants of 4D-Var data assimilation*

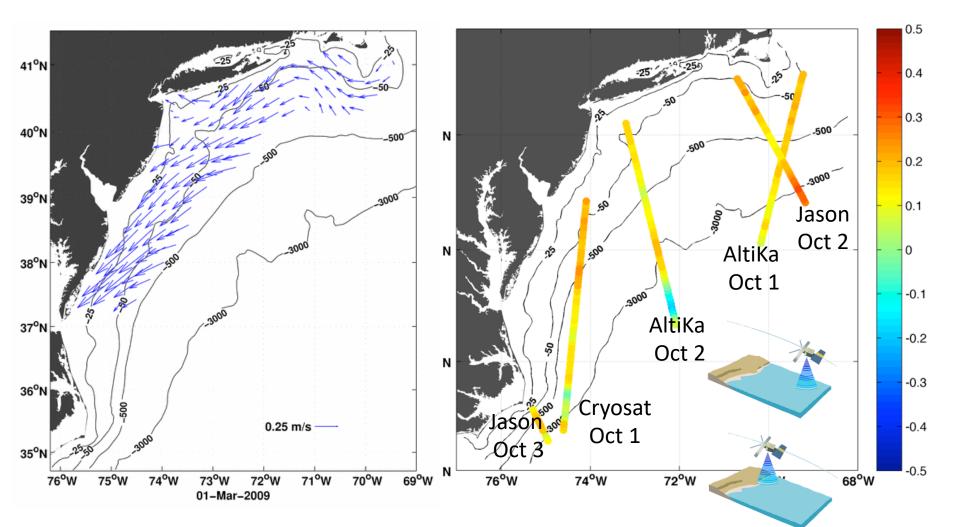
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- * Moore, A. M., H. Arango, G. Broquet, B. Powell, A. T. Weaver, and J. Zavala-Garay (2011), The Regional Ocean Modeling System (ROMS) 4-dimensional variational data assimilations systems, Part I - System overview and formulation, *Prog. Oceanog.*, 91(34-39).

Data pre-processing flow for MARACOOS ROMS 4DVar

Scripts run daily (cron jobs) and use Matlab, Python, perl and NCO tools



All times New York local (U.S. EST)

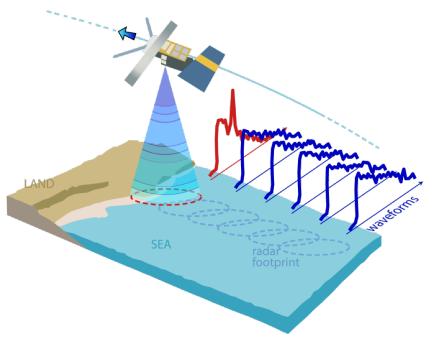


Example of CODAR data after quality control, binning and decimation to a set of independent observations. Example of along-track altimeter sea level anomaly data during a single 3-day analysis window.

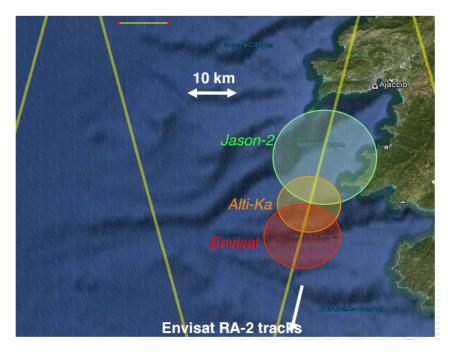
Coastal Satellite Altimetry

As satellite approaches the coast

- radar reflection from land contaminates the range signal
- microwave emissivity of land contaminates the radiometer water vapor range correction



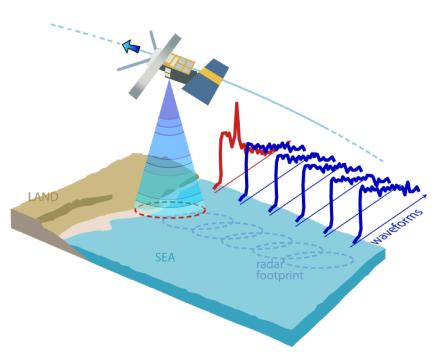




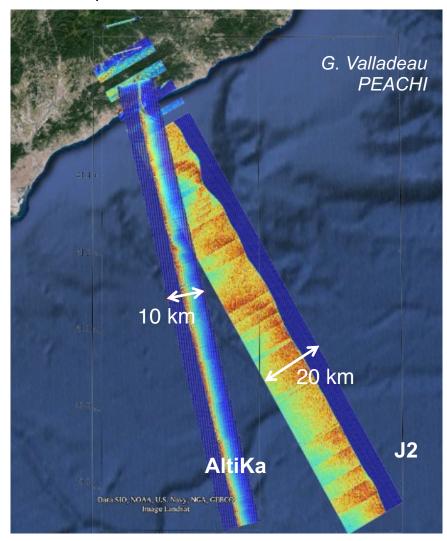
Credits: www.coastalt.eu

Coastal Satellite Altimetry

Technical advances in coastal altimetry now allow for correction of these problems, and altimetry can be used within a few km of the coast



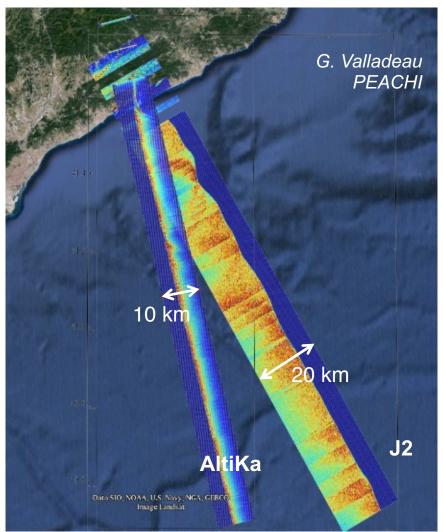
AltiKa / Jason-2 in west Mediterranean Sea



Credits: www.coastalt.eu

Coastal Satellite Altimetry: Technical developments

- Re-tracking of waveforms recovers valid data close to coast
- AltiKa small footprint is advantageous near to coast
- CryoSat-2 *Delayed Doppler* SAR gives refined along-track resolution ~700 m
- Range corrections are customized to coastal setting (e.g. water vapor wet troposphere; local tides)
- Appropriate physics (e.g. DAC)



AltiKa / Jason-2 in west Mediterranean Sea

Credits: www.coastalt.eu

Altimetry in 'conventional' mesoscale models

Mesoscale operational forecast models (e.g. Mercator-Océan, HYCOM) typically apply

- Dynamic Atmosphere Correction (DAC)
 - remove Inverted Barometer effect (no dynamic P_{air} in model)
 - remove high-frequency variability that would be aliased by ~10 day satellite sampling interval by using the solution from a highresolution wind forced barotropic model (T-UGO)
- Remove tides (not modeled)
- Compute anomalies with respect to a Mean Sea Surface (MSS) to remove uncertain geoid
 - 20-year mean on reference mission tracks
 - gridded analysis for others (CryoSat, geodetic, S-3)
- Add Mean Dynamic Topography (MDT) from global analysis
 - 20-year mean from in situ and other data
- Radiometer wet troposphere range correction
- Other standard corrections (sea state bias, earth tides, ionosphere ...)

Using coastal altimetry with regional models

Model dynamics and altimeter "corrections" should be consistent <u>Our approach</u>:

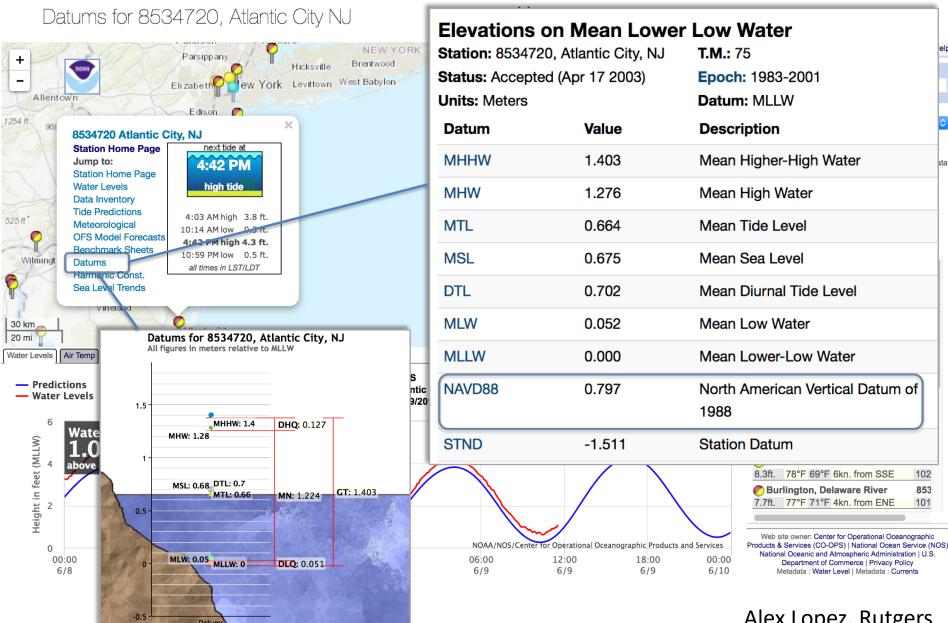
- Air pressure from met. forecast is imposed at model sea surface
 - so we do *not* apply Inverted Barometer correction
- The model simulates CTW generation and propagation within the domain (including stratification influence on dispersion relation)
 - so we do <u>not</u> apply DAC
- Mean Sea Surface: Global products are probably inadequate for coastal applications for non-reference missions track altimeters
 - active area of research and experimentation
- Add MDT from regional data assimilation analysis
 - best estimate from long term observations, constrained by coastal model kinematics (effects of bathymetry and coastline)
- Wet troposphere: extrapolate from open ocean or use NWP model
- Other standard corrections (sea state bias, earth tides, ionosphere ...)

Other practical considerations...

<u>Our approach:</u>

- Tides: phase errors between regional model and data can dominate model-data misfit
 - remove tides from altimetry but restore from harmonics of the regional model
- 4D-Var (in our configuration) does not impose time correlation of low frequency sea level anomaly
 - we duplicate altimeter data +/- 3 hours of pass (with appropriate tide signal) which suppresses projection onto fast barotropic waves
- Sea level datum
 - reconcile with local tide gauges referenced to NAVD88 to give
 Total Water Level above datum useful for storm surge inundation

Mean Sea Level datum



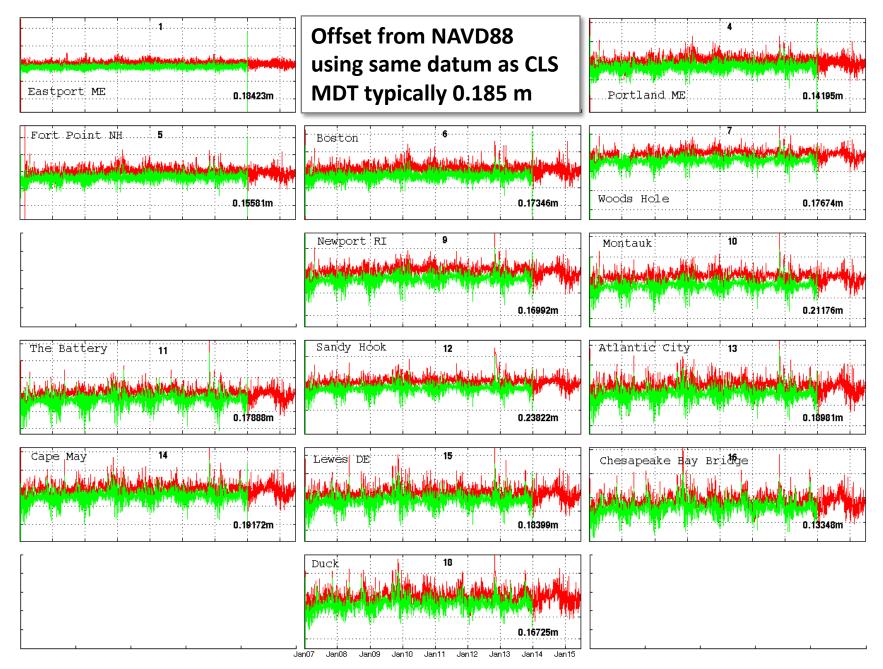
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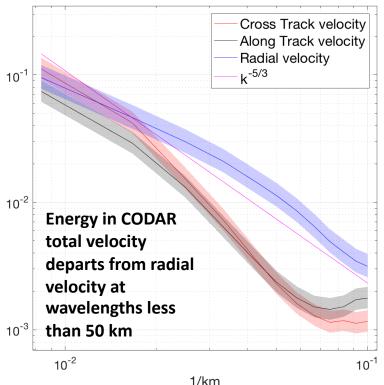
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Alex Lopez, Rutgers

Other practical considerations... Our approach:

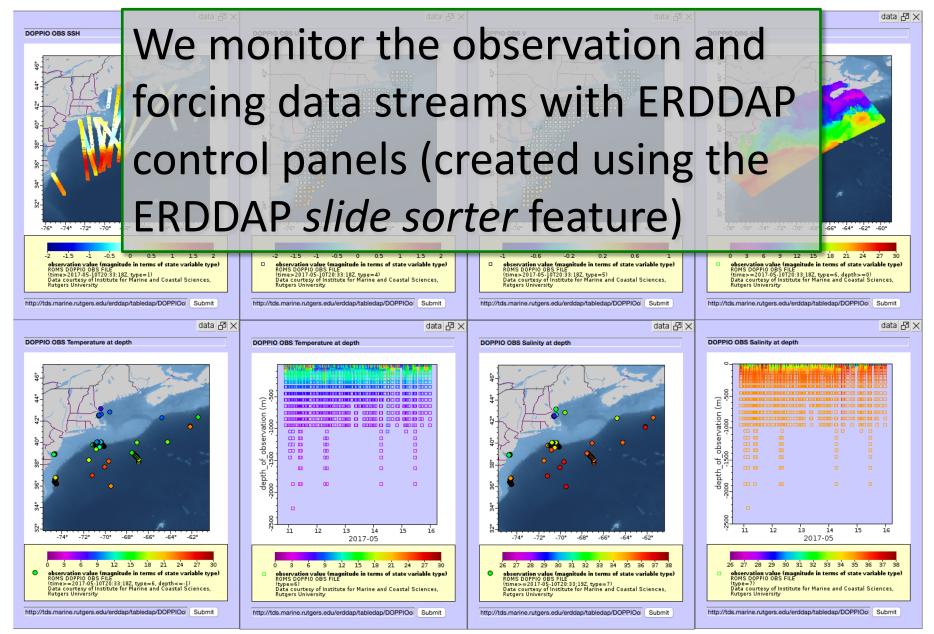
- Hourly HR-radar total vectors are assimilated without any adjustment for tides, i.e. the observed tide signal is retained in the data
- MARACOOS total vectors are computed by an optimal interpolation combiner step that filters wavelengths < 50 km
- Direct assimilation of *radial component* velocity data, via an appropriate observation operator, would allow 4D-Var to effectively perform the combination constrained by ocean dynamics
- This allows attribution of data impact of individual radar sites in the network



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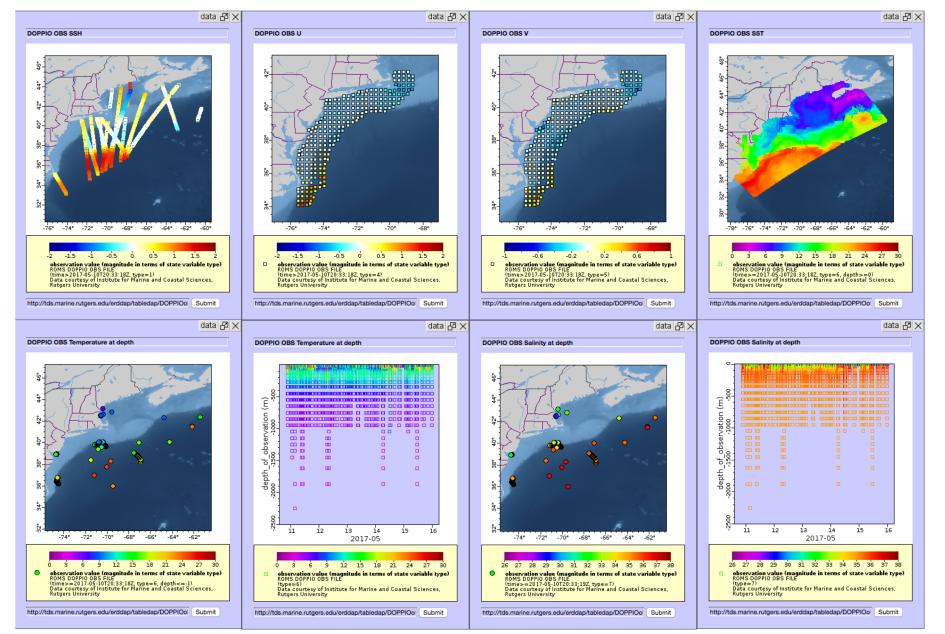
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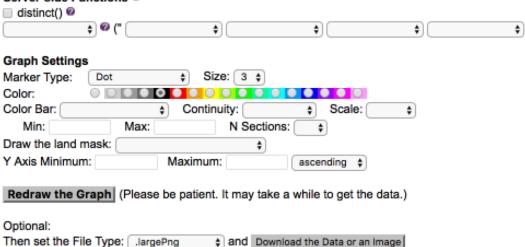
Institution: Rutgers University (Dataset ID: DOPPIO_REANALYSIS_OBS)

Range: Iongitude = -77.44505 to -59.690285°E, latitude = 32.23944 to 45.95889°N, depth = -4250.0 to 0.0, time = 2007-01-02T00:00:00Z to 2015-12-30T23:00:00Z Information: Summary @ | License @ | FGDC | ISO 19115 | Metadata | Background @ | Subset | Data Access Form

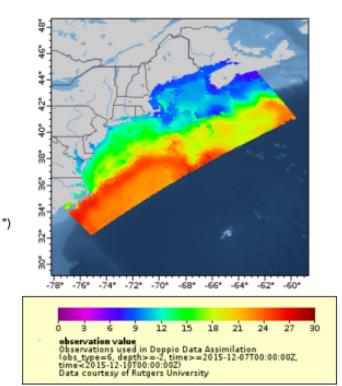
Graph Type:	markers	\$	
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Server-side Functions @



or view the URL: http://tds.marine.rutgers.edu/erddap/tabledap/DOPPIO_REANALYSIS_OI (Documentation / Bypass this form @) (File Type information) Click on the map to specify a new center point. Zoom: Out 8x Out 2x Out Data in in 2x in 8x Time range: 3 \$ day(s) \$



<u>ERDDAP</u> > <u>tabledap</u> > Make A Graph .

Dataset Title: Observations used in Doppio Data Assimilation

Rutgers University (Dataset ID: DOPPIO_REANALYSIS_OBS) Institution: longitude = -77.44505 to -59.690285°E, latitude = 32.23944 to 45.95889°N, depth = -4250.0 to 0.0, time = 2007-01-02T00:002 to 2015-12-30T23:00:00Z Range: Summary @ | License @ | FGDC | ISO 19115 | Metadata | Background & | Subset | Data Access Form Information:

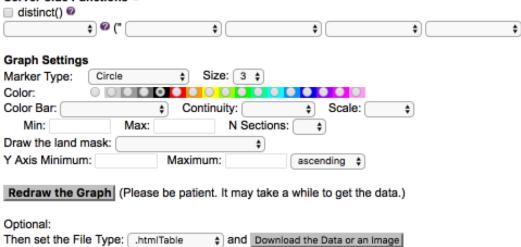


ERDDAP

Easier access to scientific data

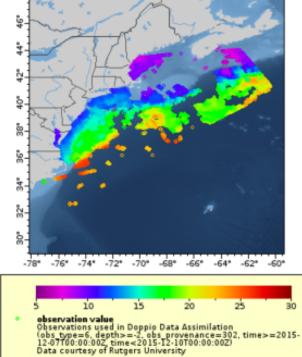
Constraints @	Optional Constraint #1 @	Optional Constraint #2 @	
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Server-side Functions @



or view the URL: http://tds.marine.rutgers.edu/erddap/tabledap/DOPPIO_REANALYSIS_OI (Documentation / Bypass this form @) (File Type information)





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Output goes to THREDDS server Forecast Model Run Collection* (FMRC) at <u>tds.marine.rutgers.edu/thredds</u>

*unidata.ucar.edu

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Catalog http://tds.marine.rutgers.edu/thredds/roms/doppio/catalog.html

Dataset	Size	Last Modified
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Averages/		
<u>History/</u>		
Floats /		

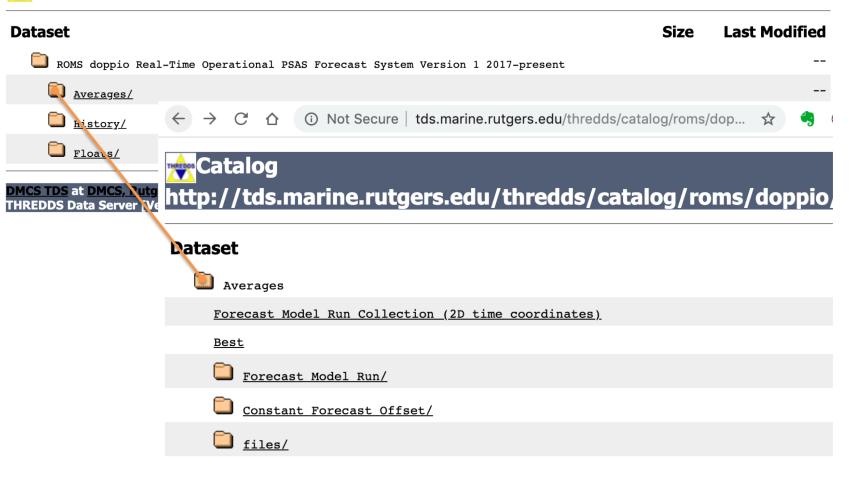
<u>DMCS TDS</u> at <u>DMCS, Rutgers University</u> see <u>Info</u> THREDDS Data Server [Version 4.6.8 - 2017-01-06T16:32:27-0700] <u>Documentation</u>

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*unidata.ucar.edu

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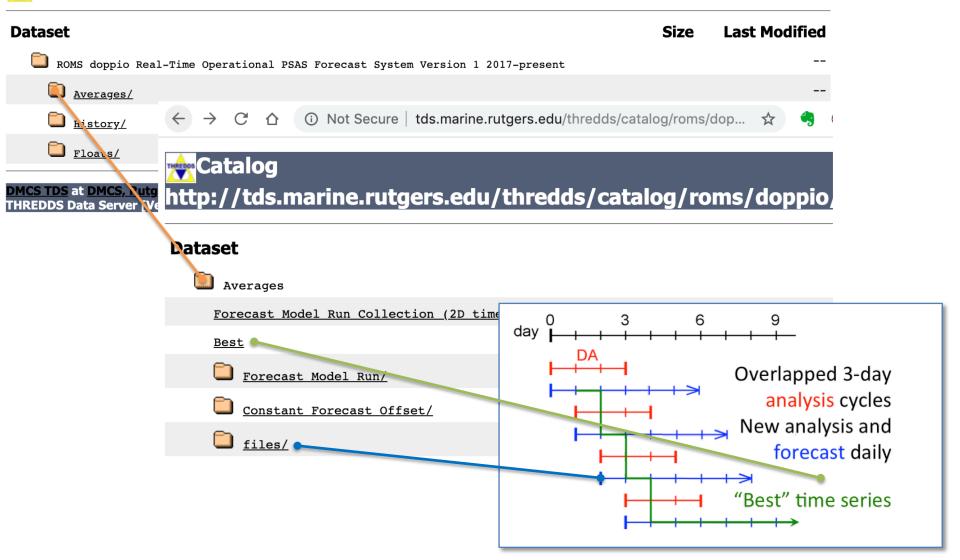


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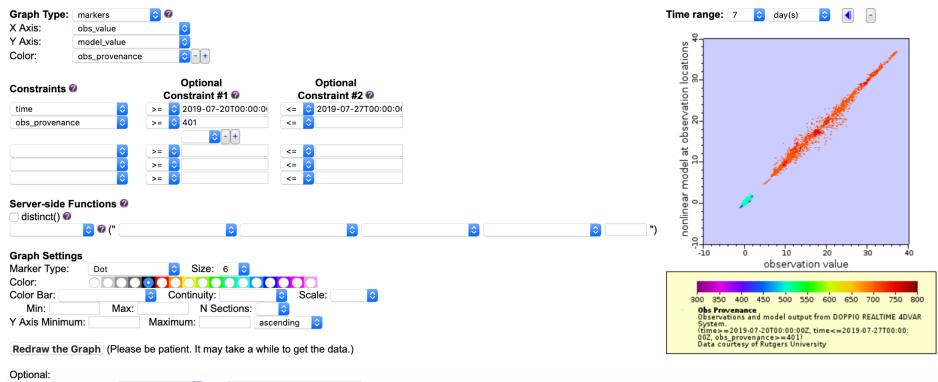


Dataset Title: Observations and model output from DOPPIO REALTIME 4DVAR System.

Institution: Rutgers University (Dataset ID: DOPPIO_REALTIME_MOD)

Range: longitude = -79.62816 to -59.690285°E, latitude = 32.23944 to 46.61133°N, depth = -3250.0 to 0.0, time = 2018-04-03T00:00:00Z to 2019-07-26T23:45:00Z

Information: Summary @ | License @ | FGDC | ISO 19115 | Metadata | Background & | Subset | Data Access Form



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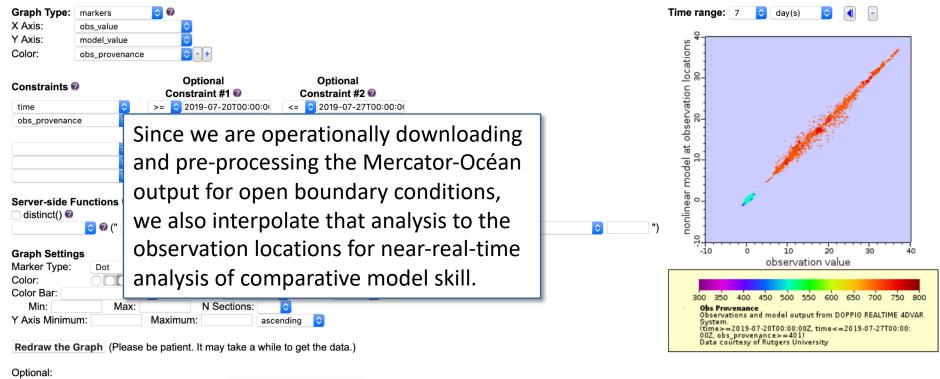


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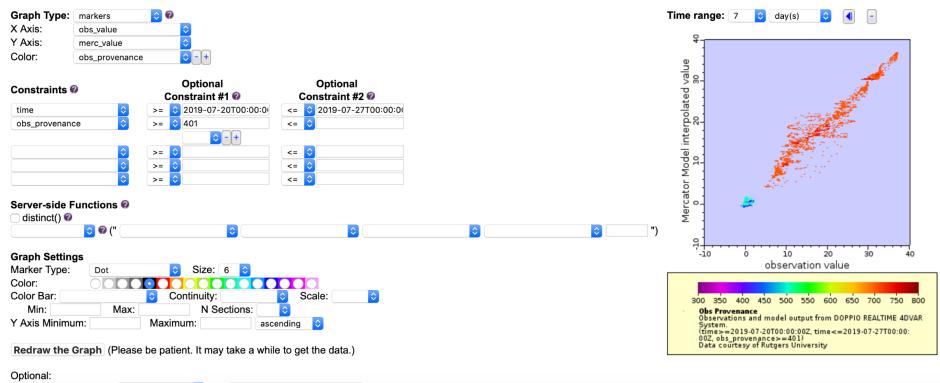


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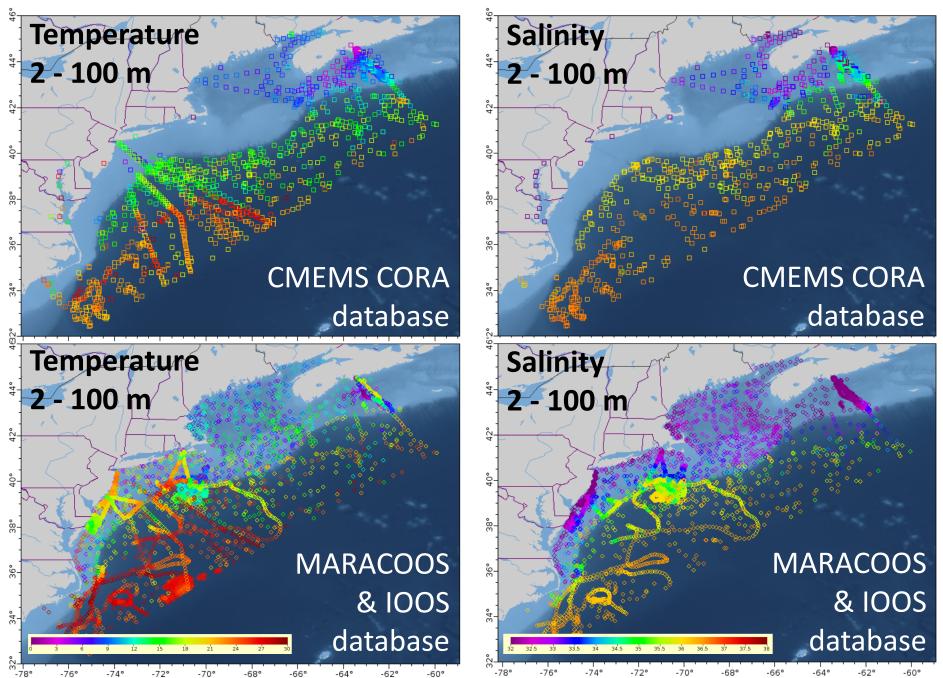
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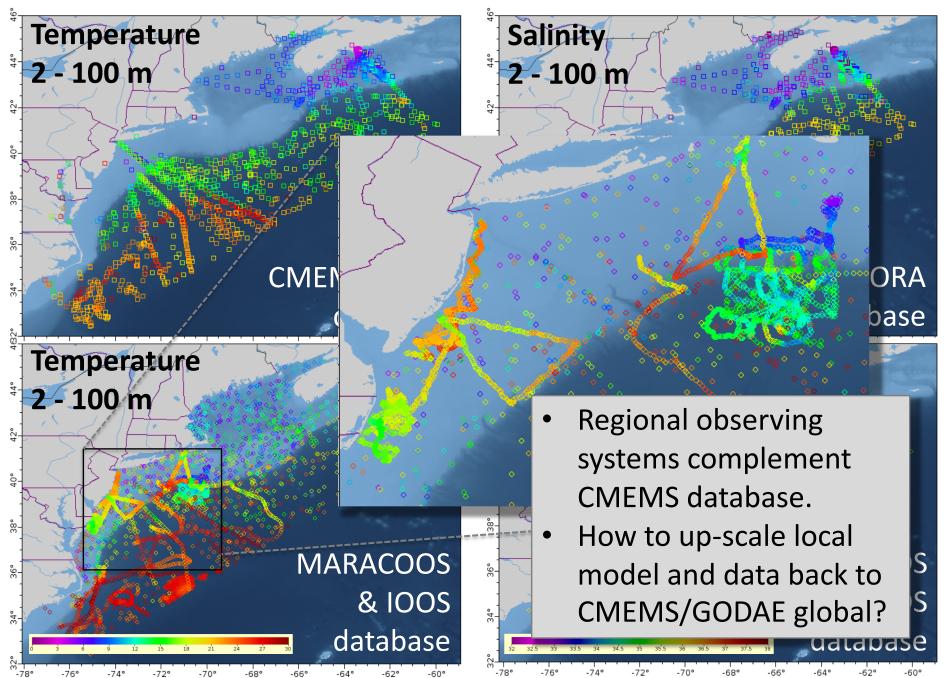


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Subsurface in situ ocean observations for the northeast U.S. in 2015



Subsurface in situ ocean observations for the northeast U.S. in 2015



Evaluate the influence of individual observations on the analysis

$$\mathbf{x}_{a} = \mathbf{x}_{b} + \tilde{\mathbf{K}}(\mathbf{y} - H(\mathbf{x}_{b}))$$

analysis = prior + gain x innovation $\mathbf{d} = \mathbf{y} - H(\mathbf{x}_{b})$

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Define some scalar functional of model state: $I(\mathbf{x})$
e.g. transport through a section:
Change due to 4DVAR: $\Delta I = I(\mathbf{x}_{a}) - I(\mathbf{x}_{b})$
from 1st order Taylor $= I(\mathbf{x}_{b} + \tilde{\mathbf{K}}\mathbf{d}) - I(\mathbf{x}_{b})$
 $from 1^{st}$ order Taylor $= I(\mathbf{x}_{b} + \tilde{\mathbf{K}}\mathbf{d}) - I(\mathbf{x}_{b})$
 $= \mathbf{d}^{T}\tilde{\mathbf{K}}^{T}(\partial I / \partial \mathbf{x})|_{\mathbf{x}_{b}}$
innovation T • sensitivity of I mapped in
to observation space

Evaluate the influence of individual observations on the analysis

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from 1st order Taylor series expansion $= I(\mathbf{x}_{b} + \tilde{\mathbf{K}}\mathbf{d}) - I(\mathbf{x}_{b})$
 $\simeq \mathbf{d}^{T} [\tilde{\mathbf{K}}^{T} (\partial I / \partial \mathbf{x})|_{\mathbf{x}_{b}}$

 $\Delta I = \mathbf{d}^{\mathrm{T}} \mathbf{g} = \sum_{i=1}^{N} d_i g_i$ Each element of g is *uniquely* associated with each of the *N* observations.

We will group by observation type the elements of $\Delta I = \sum_{i=1}^{N} d_i g_i$ to examine the respective impact of SSH, SST, etc.

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Cross-Shelf Exchange Circulation Metrics *Volume transport and heat flux*

Cross-shelf volume transport:

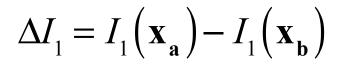
$$I_1 = \frac{1}{\tau} \int_0^\tau \int_{S-h}^\zeta \int_{-h}^\zeta u_n \, dz \, ds \, dt$$

Cross-shelf heat transport:

$$I_2 = \frac{\rho c_p}{\tau} \int_{0}^{\tau} \int_{S-h}^{\zeta} (u_n - \overline{u}_n) (T - \overline{T}) dz ds dt$$

Historical context:

Linder and Gawarkiewicz (1998) Chen and He (2014) Zhang et al. (2015) Along-shelf and vertical section following the 200 m isobath and passing through **OOI Pioneer Array** *Cross-shelf volume transport increments*



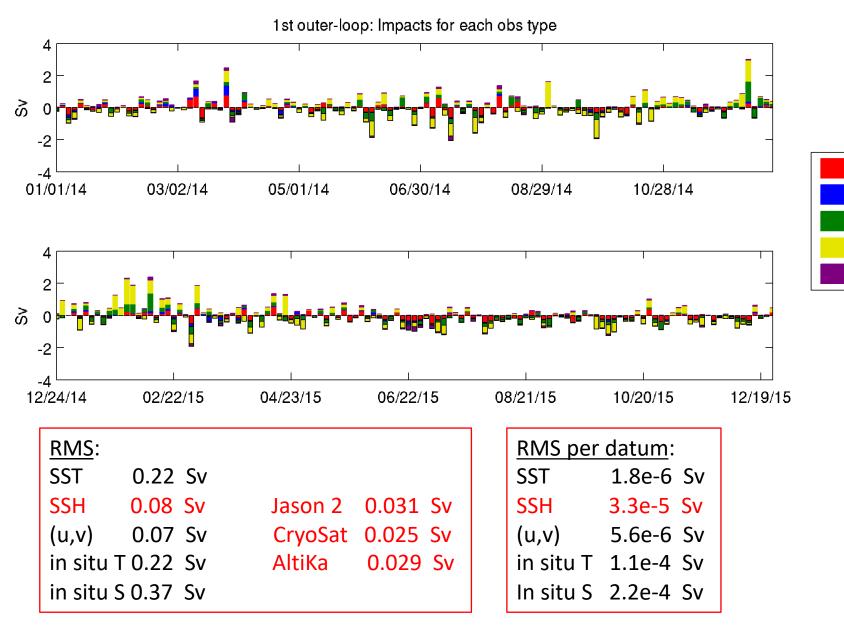
SST

SSH

т

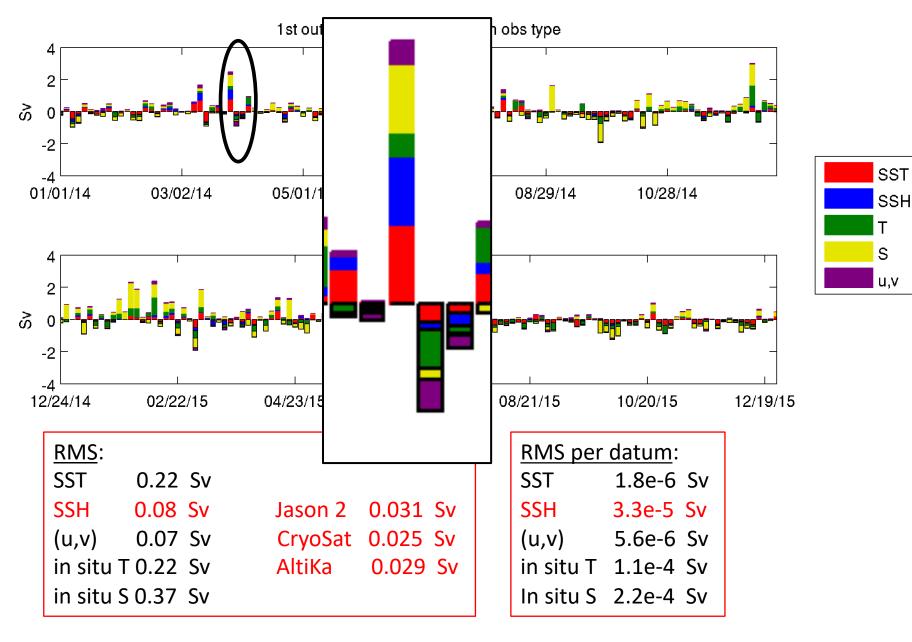
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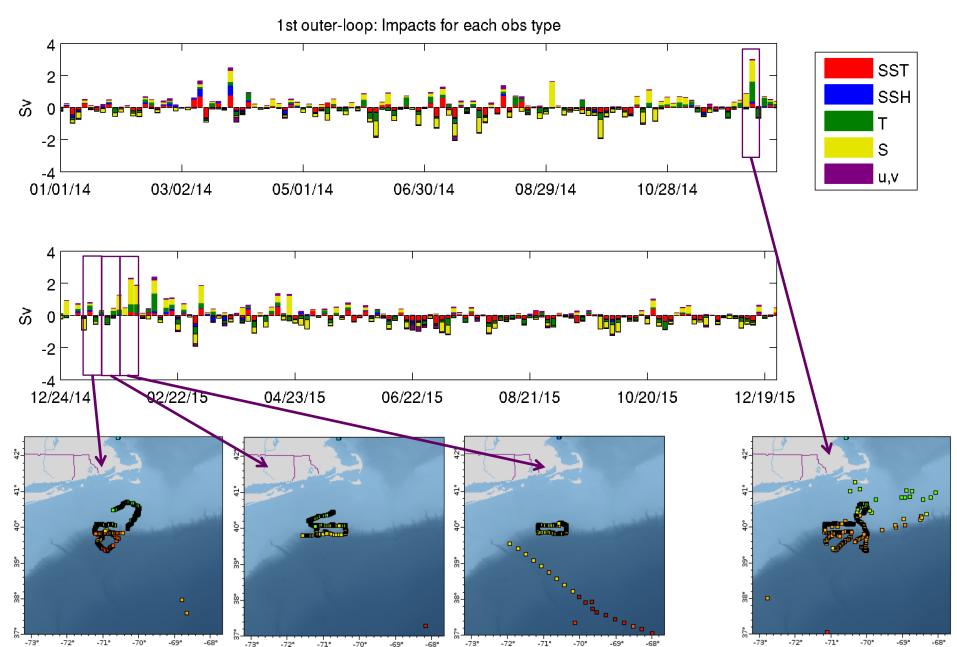


Cross-shelf volume transport increments

 $\Delta I_1 = I_1(\mathbf{x}_a) - I_1(\mathbf{x}_b)$

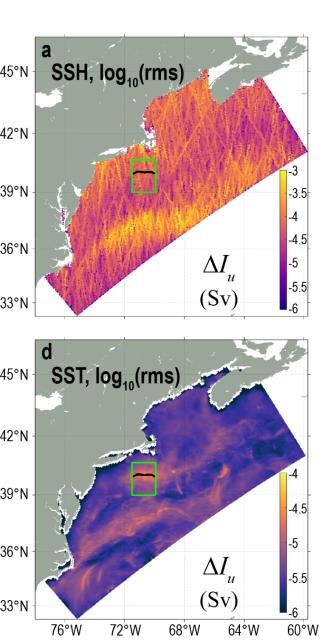


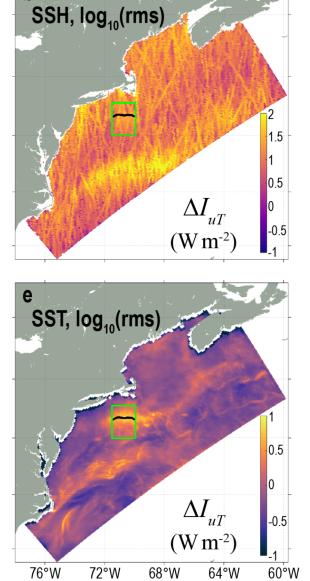
Cross-shelf volume transport increments $\Delta I_1 = I_1(\mathbf{x}_a) - I_1(\mathbf{x}_b)$

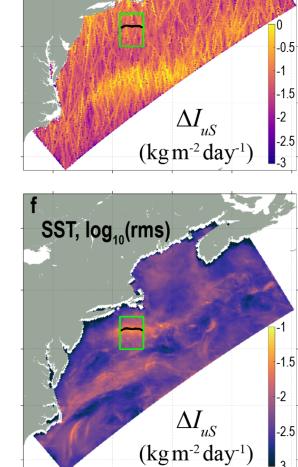


Observation impact

b







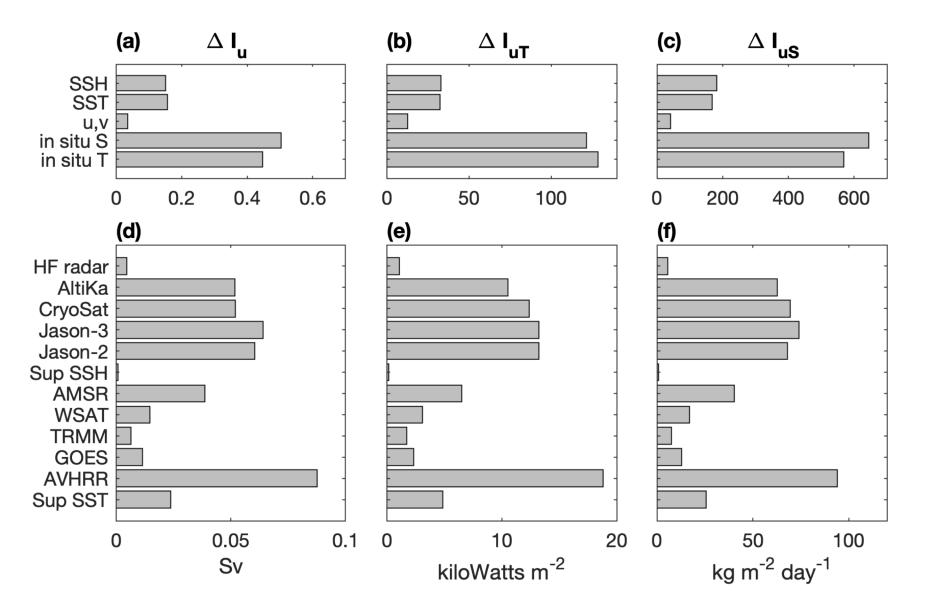
С

SSH, log₁₀(rms)

 (Kg m< day</th>
 J

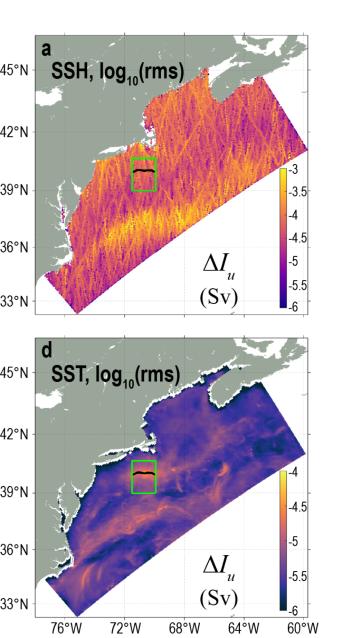
 76°W
 72°W
 68°W
 64°W
 60°W

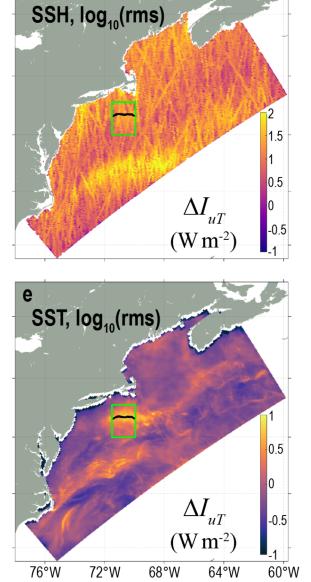
Observation impact

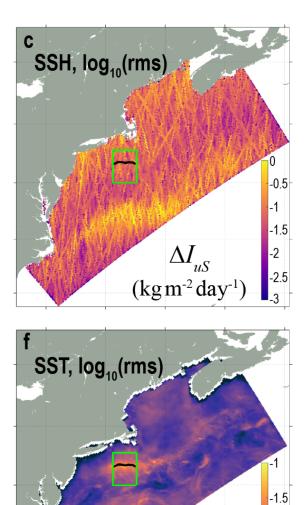


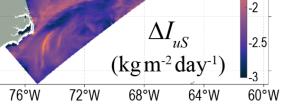
Observation impact

b









Doppio 4D-Var observation and background error assumptions have been driven by practical demands – needs further analysis and improvement

Type & platform	Source	Sampling rate and resolution	Super-obs averaging ¹	Obs error
AVHRR IR SST	MARACOOS.org & NOAA Coastwatch	4 passes per day, 1 km	3 hr	σ _b
GOES IR SST	NOAA Coastwatch	hourly, 6 km	3 hr	$2\sigma_b$
AMSR2, TRMM, and WindSat microwave SST	NASA JPL PODAAC	daily, 15 km		$1.25\sigma_b$
SSH: Jason, AltiKa, CryoSat	RADS, TU Delft	~ 1 pass daily, ~ 4 km		0.04 m
in situ T, S: NDBC buoys, Argo floats, XBT, surface drifters	Met Office En4.2	$variable^2$	$_{ m levels}^2$	$0.25\sigma_b\sigma_o/\sigma_{ m max}^3$
Surface currents: CODAR HF-radar	MARACOOS.org	hourly, 1 km	1 hr, 24 km	$0.5\sigma_b$
Glider T, S: MARACOOS	IOOS Glider DAC	variable ²	standard levels ²	$0.25\sigma_b\sigma/\sigma_{ m max}^3$
In situ T,S: GoM ⁵ buoys	NERACOOS.org ⁴	hourly, 10 buoys	1 hr	σ_b
In situ u,v: GoM ⁵ buoys	NERACOOS.org ⁴	hourly, 9 buoys ¹	1 hr	0.5σ _b

Table 1: A summary of the observational data assimilated into ROMS during 2014-2017, the procedure for forming super observations, and the observation errors assigned to each observation type. 1: All data that were sampled at a horizontal resolution higher than that of the model were formed into super observations at the resolution of the ROMS grid unless otherwise indicated. 2: Profile data were binned in the vertical using the WOD atlas standard depths (Boyer et al.) [2009). 3: Here σ is the standard deviation of all observations that fall within a vertical bin (see comment 1) and σ_{max} is the maximum value of all σ in a vertical profile. 4: NERACOOS = North East Regional Association Coastal Ocean Observing System. 5: GoM=Gulf of Maine.

State variable	Horizontal decorrelation	Background quality	
	scale (km)	control parameter α	
SSH	40	5	
Velocity	40	3	
Temperature	40	6	
Salinity	15	6	

Table 2: A summary of the decorrelation scales assumed for background errors in each state variable. The vertical decorrelation length scale for all state variables was chosen to be 10 m. The parameter α used for the background quality control rejection criteria is also indicated.

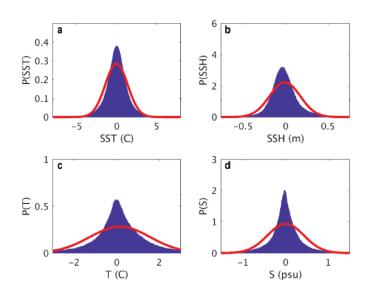
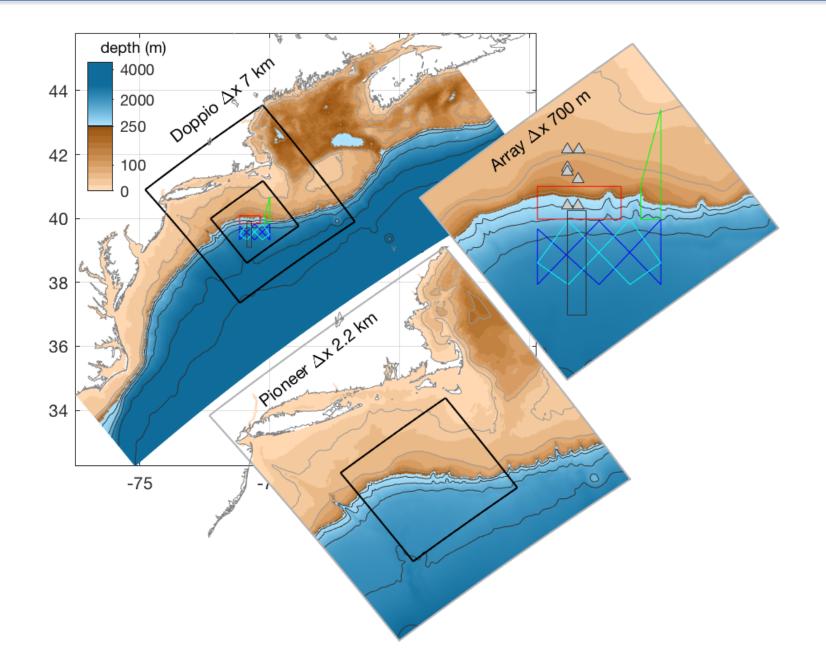
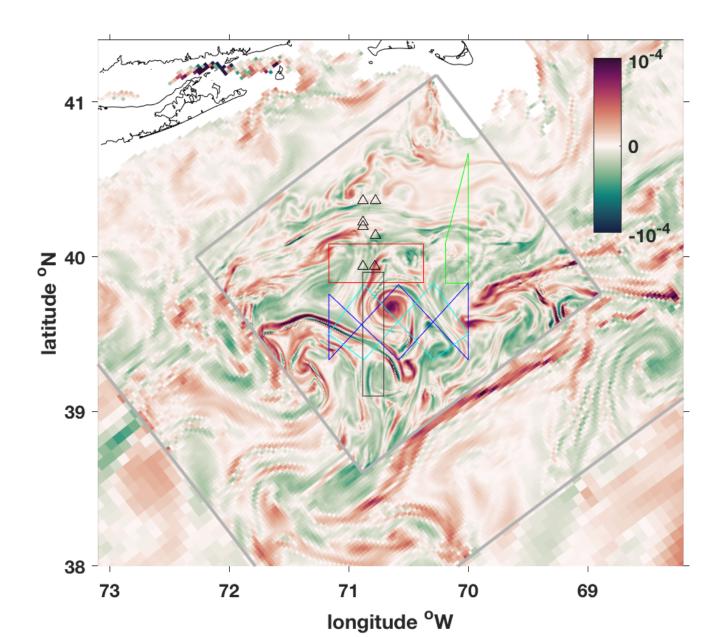


Figure 3: Probability density functions (pdfs) of the 4D-Var increments in (a) SST, (b) SSH, (c) *in situ* temperature, and (d) *in situ* salinity computed from all 4D-Var cycles spanning the 2014-2017 period.

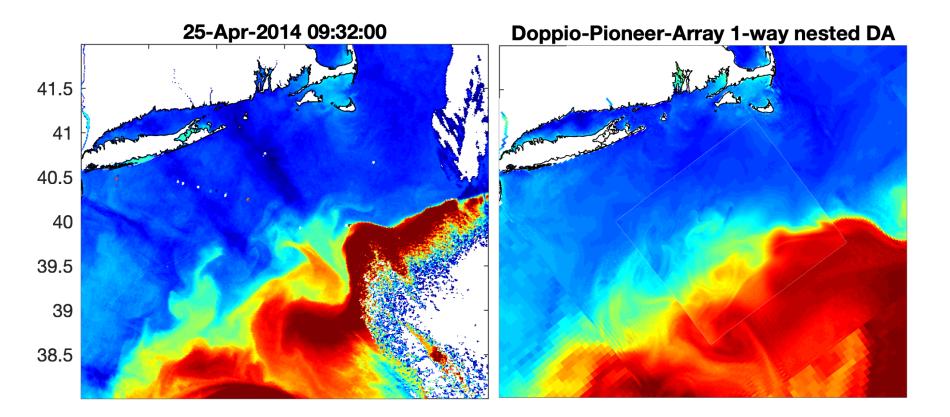
Nested modeling for local grid refinement



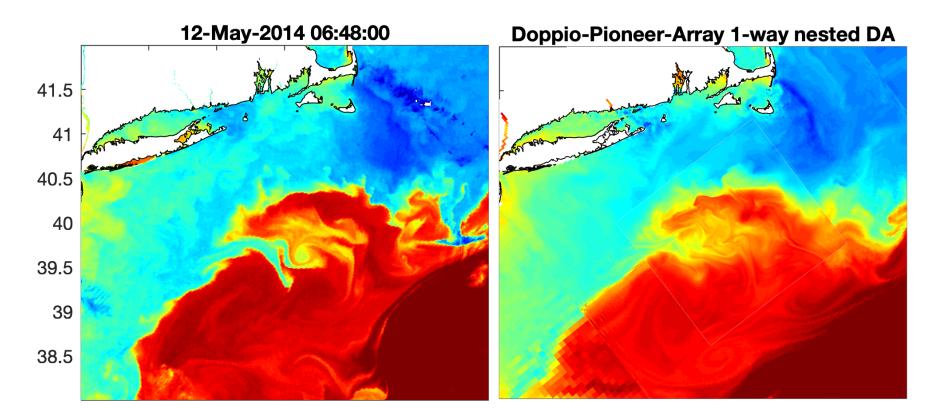
Doppio <-> Pioneer <-> Array 2-way nesting at all levels – free running forward model



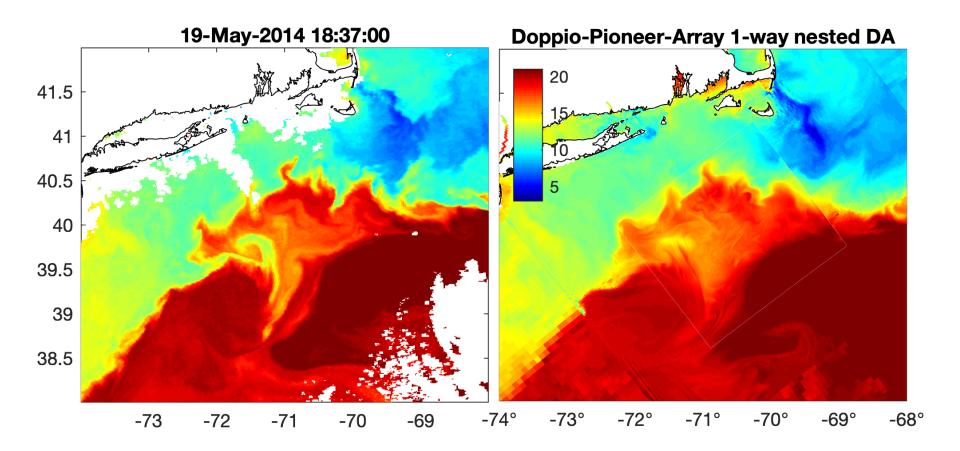
Doppio -> Pioneer -> Array Successive 1-way downscaling with data assimilation in each grid



Doppio -> Pioneer -> Array Successive 1-way downscaling with data assimilation in each grid



Doppio -> Pioneer -> Array Successive 1-way downscaling with data assimilation in each grid



Zhang, W. G., and G. G. Gawarkiewicz (2015), Dynamics of the direct intrusion of Gulf Stream ring water onto the Mid-Atlantic Bight shelf, Geophys. Res. Lett., 42, doi:10.1002/2015GL065530.

Summary (1/2)

Rutgers MARACOOS ROMS forecast system with 4DVAR DA uses all available data from a modern coastal ocean observing system

- Satellite SSH and SST, HF-radar, gliders, Argo, GTS XBT/CTD
- More and diverse data is better
- Pre-processing for QC; tides in altimetry; binning to independent obs.
- OBC bias removal essential: use mean state from 4DVAR-based climatology
- Data ingest and output exploits web services (OPeNDAP/THREDDS) and interoperability of data conventions (CDM, CF-conventions, NetCDF/HDF)

Useful skill for real-time applications

4 days for temperature and salinity; 1-2 days for velocity

Output to THREDDS/FMRC Forecast Model Run Collection

 Surface currents to U.S. Coast Guard; Bottom temperatures to NOAA Fisheries ...

Rutgers

2019 ROMS 4D-Var Training Workshop NOAA Center for Weather and Climate Prediction (NCWCP), July 29 to August 2, 2019

Summary (2/2)

Observation Impact Analysis

quantifies the influence of data on user-defined circulation metrics

- Impact assessment can be extended into the forecast interval
- We are working toward:
 - Real-time monitoring of IOOS observing elements (for QA/QC, etc.)
 - Observing system evaluation (e.g. long-term impact of individual radar sites in the network)
 - Adaptive sampling for moveable assets (e.g. OOI Pioneer gliders)

myroms.org Development Path

- 2-way nested forecasts (NOAA)
- Hybrid EnKF/4D-Var using DART to advance background covariance B (ONR)
- Coupling using ESMF/NUOPC (NOPP):
 ROMS <-> CICE <-> WRF/COAMPS <-> SWAN/WW3 and <u>hybrid model *nesting*</u>
- 2-way nested 4D-Var (NSF OOI Pioneer)
- Physics+bio (optical/IOP) joint assimilation

Rutgers

2019 ROMS 4D-Var Training Workshop NOAA Center for Weather and Climate Prediction (NCWCP), July 29 to August 2, 2019