

# Spanning the scales: Multi-platform approaches for integrated studies of biogeochemistry and physics



Craig Lee Applied Physics Laboratory, University of Washington

Luc Rainville, Andrey Shcherbina, Eric D'Asaro (APL-UW), Steve Riser (APL-UW), Melissa Omand (URI), Dan Rudnick (SIO), Ken Johnson (MBARI), Tom Farrar (WHOI)



## **Observational Challenges**



Episodic evolution (K. Fennel)





Small patch scales, 4D physics

### Synoptic 4D sampling

- Sampling speed vs. timescales of dynamics
- Tradeoffs... temporal resolution vs. spatial resolution

### Persistence

- Episodic events, multiple realizations
- Extended timescales

Representative spatial/temporal coverage

• Scaling from patches to basins, seasons to years

Recent autonomous technologies provide access scales that were previously difficult or impractical to sample.

## Match Approach, Technology to Spatial and Temporal Scales



## Match Approach, Technology to Spatial and Temporal Scales



## Many Platforms, Complementary Capabilities



## What are They Good For?

## Boundary Current Example...



- Area defined by regional model
- 28893 glider profiles
- 2482 Argo profiles
- Similar per-profile cost (\$20k/200 profiles, \$100k/1000 profiles), but usage differs
- Floats- distributed, mapping
- Gliders- concentrated, small scales, string gradients





Salinity Processes in the Upper Ocean Regional Study

SPURS

Salinity Processes in the Upper Ocean Regional Study

What are the physical processes responsible controlling the upper ocean salinity balance: air-sea interactions, mixing, oceanic transport, etc.

### SPURS programs involve coordinated field work, numerical models, and remotesensing:

Towed Surface Salinity Profiler, Asher et al., UW SPURS Data Management System, Bingham et al., UNC Multi-scale Modeling and Data Assimilation, Chao et al., RSS Near-surface Turbulence: Lagrangian Floats, D'Asaro et al., UW Toward a Salinity Budget (flux mooring), Farrar et al., WHOI Multiscale Autonomous Surveys, Fratantoni et al., WHOI Characteristics SSS Fluctuations, Gordon et al., LDEO Upper Ocean Salinity from Glider Surveys, Lee et al., UW Multi-Scale Modeling and Data Assimilation, Li et al., JPL Measurements of T, S, Wind Speed, and Rainfall (floats), Riser et al., UW

Spurs

Microstructure and Mixing, Schmitt et al., WHOI (NSF) SSS Drifters for SPURS, Centurioni et al., SIO (NOAA) Prawler Mooring, Kessler et al., NOAA/PMEL (NOAA) Sustained Ocean Observations, Goni et al., NOAA/AOML (NOAA).



PICO moorings: Kessler (NOAA) Rain lenses: Drushka, Asher, Jessup, Rainville (NSF)



## SPURS -

NASA

Salinity Processes in the Upper Ocean Regional Study





## SPURS - 2 Salinity Processes in the Upper Ocean Regional Study

#### NASA SPURS-2 Coordinated Drift

∔ APL Lagrangian float

- 🖉 WHOI Waveglider
- 💉 APL Sea Glider
- UW APEX Floats
- SIO Drifters
- AOML Drifters
- 🖉 R/V Lady Amber

NASA SPURS 2 APL/UW AShcherbina@apl.uw.edu

Data SIO, NOAA, U.S. Navy, NGA, GEBCO

Google earth

Shcherbina (APL-UW)

Salinity Processes in the Upper Ocean Regional Study

SPURS



Moorings \_\_\_\_\_\_
Drifters \_\_\_\_\_\_
Seagliders \_\_\_\_\_\_
Argo floats \_\_\_\_\_\_
UCTD

Decorrelation scales: 75 km and 5 days mapped as perturbations from remote sensing

Note: Size of data marker is scaled by when it was collected relative to the map time.

## SPURS

#### Salinity Processes in the Upper Ocean Regional Study

## Salinity and temperature budgets of the mixed layer

For the 3-week averages:

Salinity tendency mostly due to mesoscale advection.

Temperature balance has similar contributions from surface fluxes, entrainment, and advection.

Not perfect...

- Physics representation in terms
- Temporal and spatial scales

Farrar et al. (2015)



## SPURS

**Aquarius Ol** 

(7-day, 150 km)

days, 30 km

14 days, 70 km

distance [km]

-50

-150

-150

-100

-100

50

50

100

150

0000<sup>0</sup>

100

15

Feb 2016

#### Salinity Processes in the Upper Ocean Regional Study

150

#### Different scales emphasize different processes. 100 ×10<sup>-10</sup> 50 21-day running mean Salinity tendency (kg/kg/s) 0 0 0 5 0 0 0 -Ú -50 -100 -150 150 100 37.5 **ML** salinity tendency Feb Mar Apr May Jun Jul Aug Average over air-sea fluxes 50 lateral advection a time scale 37.45 7-day running mean 37.4 Salinity tendency (kg/kg/s) -50 Assume a 37.35 -100 spatial scale -150 150 100 -1.5 Oct Dec May Nov Jan Feb Mar Apr Jun Jul Aug 50 0.2 m/s 0 -0.2 1-h 200 km -50 $\langle U \rangle_{dt} \cdot dt$ 1-day 7-day - 14-day -100 100 -150 0 Sep May Jun Jul Oct Nov Dec Jan Feb Mar Aua

### BGC Observations: Leveraging Onto an Autonomous Sensor Network

#### **More Variables**

### <u>Ships</u>

**Discrete samples Pigment analysis** Phytoplankton POC absorption( $\lambda$ ) **Nutrients Community Structure** Rates Sensors CTD + velocity PAR (Ed)  $b_{bp}$ Chl fluor **Beam Attenuation** Oxygen many others... Many more...

### Moorings/IBOs

CTD + velocity Microstructure Nutrients (autoanalyzer) Zooplankton (image, acoustic) PAR (Ed) Spectral Irradiance b<sub>bp</sub> Chl, CDOM fluor Beam Attenuation Oxygen pH Meteorology Genomics

#### More Measurements

## Floats & Gliders

CTD + velocity Microstructure Nitrate (SUNA) Zooplankton (acoustic) PAR (Ed) Spectral Irradiance b<sub>bp</sub> Chl, CDOM fluor Beam Attenuation Oxygen pH

Calibration- interoperability between platforms Proxies- biogeochemical/biological variables from autonomous sensors

## Coupled Physical-Biogeochemical Processes- NAB08



#### Craig M. Lee, Eric A. D'Asaro, Mary Jane Perry, Katja Fennel

Matthew Alkire, Witold Bagniewski, Nathan Briggs, Ivona Cetinic, David Checkley, Amanda Gray, Kritstinn Gudmundsson, Jan Kaiser, Emily Kallin, Richard Lampitt, Amala Mahadevan, Patrick Martin, Nicole Poulton, Eric Rehm, Katherine Richardson, Ryan Rykaczewski, Tatiana Rynearson, Brandon Sackmann, Michael Sauer, Michael Sieracki, Toby Westberry

### Spring Phytoplankton Bloom Initiation and Evolution in the Subpolar North Atlantic

- Timing deploy before the bloom.
- Persistence measurements before, during, after bloom.
- Float Lagrangian frame.
- Seagliders spatial context.
- Real-time data adaptive sampling.
- Proxy sensors for carbon-cycle components.
- Ship-based sampling calibration, inform interpretation.
- Aggressive calibration efforts (lab, deployment, process & recovery).
- Satellite remote sensing ocean color, SST, Aviso SSH, NCEP winds.
- Models ecosystem, productivity, submesoscale circulation.



## POC Proxies: $c_p$ and $b_{bp}$ ( >300 samples)



- Dual POC proxies.
- Empirical relationship derived from 296 (c<sub>p</sub>) & 321 (b<sub>bp</sub>) - POC pairs.

#### Redundancy

- Regions where c<sub>p</sub> and b<sub>bp</sub> POC records are identical provide confidence.
- Regions where c<sub>p</sub> and b<sub>bp</sub> POC records differ provide diagnostic.

## **Net Community Production and Export**

Matthew Alkire APL/UW



- = Decrease in  $NO_3 \times C:N$  Redfield
- = Increase in  $O_2 + O_2$  loss to atmosphere x O:C (PQ)
- = Increase in POC Carbon Export + [increase in DOC]



## Export Maps from Glider-based Measurements and Optical Proxies





- Map POC, Chl/bbp (community index)
- Combine to map export





## Growth Rates from Diel Cycles Melissa Omand (URI)

SCHMIDT



POC<sub>cp</sub> was calculated from a linear fit between filtered POC samples and Cp on the CTD Rosette (ie. *Claustre et al. 1999, Cetinic et al. 2012*).

## Growth Rates from Diel Cycles Melissa Omand (URI)





Growth rates calculated at each depth versus the average PAR measured over daylight hours at that depth

Resembles lab-based photosynthesisirradiance curves

- Take vertical average (10 m bin)
- Linearly fit night-time POC
- Specific growth rate is calculated from the extrapolated POC difference = 2.03 mgC(mgChl)<sup>-1</sup>h<sup>-1</sup>
- Compute growth rate at multiple depths



# Southern Ocean Carbon and Climate Observations and Modeling

Goals:

- Quantify and understand the role of the Southern Ocean (everything south of 30°S) in carbon cycling, acidification, and nutrient cycling on seasonal, interannual, and longer time scales.
- Develop the scientific basis for projecting the contribution of the Southern Ocean to the future trajectory of carbon, acidification, and nutrient cycling.





## SOCCOM Strategy

~200 profiling floats over 6 years with pH,  $NO_3^-$ ,  $O_2$ , biooptics

SOF

120%



Southern Ocean State Estimate model to get 4D fluxes

Improved coupled
 climate model (GFDL)
 predictions of Southern
 Ocean role in carbon
 and climate

180<sup>o</sup>W

00

SOCCOM Floats in White Pre-SOCCOM Floats in Yellow Non-op. Floats in Cyan

20-May-2017

1000

120°W













#### Table 1. Profiles to depth > 900 m.

Parameter	Ship Profiles	<b>BGC-Argo Profiles</b>	BGC-Argo	90°	120°	150°	-180°	-150°	-120°	-90°	-60°	-30°	0°
	per year	per year	/Ship	cal Argo	Sensor Types								March 2017
	(2001-2010)	(2016)			Latest location of operational floats (data distributed within the last 30 day:						days)		
					· c	Operational Floats	s (275)	Suspended partic Downwelling irrac	cles (132) diance (44)	<ul> <li>Nitrate (96)</li> <li>Chlorophyll a (132)</li> <li>Oxygen (257)</li> </ul>			
Oxygen	1730	11332	6.5					pri (10)					
Nitrate	1231	3835	3.1									Generated by www	.jcommops.org, 10/04/2017
pH direct	460	1862	4.0										
pH (TA/DIC)	540		3.4										
Source	US NODC	Argo GDAC											

## Summary

- Creative, multi-platform approaches offer great power.
- Match target spatial and temporal scales to system of platforms.
- Leverage autonomous sensors through BGC proxies.
- Understand process-level physics, biology and biogeochemistry.
- Scale up process knowledge + observations + remote sensing + models.
- Ship-based observations critical for calibration, proxy building. An application for long-range AUVs?
- New logistical approaches- flexible vessels, over-the-horizon deployments to access remote/dangerous locations, ...
- Expendable platforms. Aim for low-cost/long-endurance, easy logistics.
- How do we define expendable? Gliders and floats have similar cost per profile...