



1. SOCCOM observations

(L. Talley, K. Johnson, S. Riser, E. Boss, S. Gille, A. Dickson)

200 Biogeochemical Argo floats (10 day profiling to 2000 m) Temperature, salinity, pressure Oxygen, Nitrate, pH Bio-optics

What are the value of these data?

2. SOCCOM state estimation Biogeochemical Southern Ocean State Estimate (BSOSE) M. Mazloff, A. Verdy

What is a state estimate?



Year 4 (6



Biogeochemical constraints:



	2009	2010	2011	2012	2013	2014	2015	2016	2017
# of O ₂ profiles	2416	2142	1254	906	821	1322	1,574	1,550	2,395
# of pH profiles	0	0	0	B	0	309	564	1,126	1,624
# of NO ₃ profiles	0	201	80	116	247	463	625	1,293	2,057
Ship profiles	679	380	447	88	125	180	102	333	149
SOCAT5 pCO2	14,296	17,242	20,548	23,626	18,238	17,021	20,566	14,070	0







State estimation vs. Iterative methods (e.g. reanalysis)

For scientific understanding one prefers to adjusting the uncertain initial condition and "wind speed" allowing the trajectory to obey a *closed dynamical budget*. This has become known as "state estimation".



State estimation vs. Iterative methods: The Ocean Reanalyses Intercomparison Project (ORA-IP). Adapted from Balmaseda et al. 2015



The Biogeochemical Southern Ocean State Estimate (B-SOSE)

MIT general circulation model

+ BLING biogeochemistry (Galbraith et al, 2010)



observations: altimetry; sea surface temperature, sea ice concentration; profiling floats temperature, salinity, oxygen, nitrate, pH; bottle nutrients and carbon system; underway pCO₂; POC & CHL





We use the adjoint method (4d-Var) to iteratively solve our weighted least-squares problem: solve for model inputs to minimize distance between model and observations.



forward model with closed budgets $cost = ((obs_i - model_i)/error_i)^2$

adjust ICs and atmospheric state

produce estimate



adjoint model sensitivity of cost to control variables

ARIZON





BGC-Argo is achieving substantial in situ coverage. Now appropriate to construct a state estimate synthesizing all available observations and the laws of nature represented by a general circulation model

Goals

•Produce a gridded data set consistent with model physics and observations, increasing accessibility

 Provide quantitative baselines: Estimating the "normal" and a framework to understand the "normal"

2008-2012 solution publicly available: sose.ucsd.edu

Now producing a 2013 to present SOCCOM-era BSOSE









Validation against observations: holding the *model* accountable to the *data*











Comparisons with gridded products satellite data mapped products

Comparisons with in situ obs. float profiles shipboard measurements

Comparisons with estimated "quantities of interest"

air-sea CO₂ flux Drake Passage pCO₂ nutrient transport across 32°S

Validation online

http://sose.ucsd.edu/bsose_valid.html





Biogeochemical state
Gomparisons with climatology:

- air-sea CO2 flux with Takahashi climatology
- ALK, DIC, pH with GLODAPv2 climatology
- O2, NO3, PO4 with WOA13 climatology
- Comparisons with mapped products:
 - surface pCO2 with SOCATv5 & Landschutzer monthly maps
 - air-sea CO2 flux with Landschutzer monthly maps
 - chlorophyll with MODIS Aqua and VIIRs monthly maps
- Comparisons with in situ observations:
 - pCO2 with SOCATv5 underway
 - O2, pH, NO3, CHL, POC with SOCCOM floats profiles
 - ALK, O2, NO3 with GLODAPv2 profiles
 - O2 with non-SOCCOM BgcArgo profiles
 - O2 and CHL with shipboard profiles

Physical state

- Comparisons with mapped products:
 - SSH with Jason 1&2, and with AVISO maps
 - SST with MW SST maps
 - Ice concentration with NSIDC maps
 - T, S with Argo maps
- Comparisons with in situ observations:
 - T, S with Argo, MEOP, and shipboard profiles

pCO₂ in Drake Passage



B-SOSE is giving a "low-pass" estimate of normal



6

4

2

0

-2

-4

-6



B-SOSE progress report for SOCCOM 2018

What's working, what not, what's next?

- Multiyear BGC data assimilation via adjoint method at 1/3°.
- Closed budgets for BGC variables (DIC, ALK, O₂, NO₃, PO₄, Fe).
- SOCCOM float observations as constraints.
- Establishing a BLING and B-SOSE international user base.

not working:

SOCCOM

- CO₂ flux seasonal cycle out of phase with observations in some regions.
- In general: Biological activity weak. Surface pCO₂ high. pH low.
- Mid-lat. NCP low and Argentine Basin misfits large due to circulation
- Multigrid 4D-Var optimization at 1/6°. Some development needed.
- Some outliers still present in data constraints

next:

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 - Optical constraints and improved mixed layer fidelity for BGC.
 - Removal of remaining outliers
 - Reporting of misfit statistics to aid SOCCOM QC effort.
 - Publication of 1/6° B-SOSE for 2013-2017, dissemination via LAS.
 - Multigrid optimization at 1/6° for improved 2013-2018 B-SOSE.
 - Extend validation to include other quantities of interest (e.g. NCP).



DIC budget for five regions of the Southern Ocean







Verdy and Mazloff, 2017. JGR

Dec. 2017 oxygen at 55 m. BGC-Argo observations shown with filled circles.

B-SOSE Iteration 121. RMS fit to obs: 19.08 µmol/kg

World Ocean Atlas 2013 climatology. RMS fit to sobs: 15.72 µmol/kg

SOCCOM

WOA 18% more consistent with obs.



50

0

100

150

200

250

300

350

June 2017 oxygen at 55 m. BGC-Argo observations shown with filled circles.

B-SOSE Iteration 121. RMS fit to obs: 14.68 µmol/kg

World Ocean Atlas 2013 climatology. RMS fit to sobs: 13.74 µmol/kg

SOCCOM

WOA 6% more consistent with obs.



50

0

100

150

200

250

300

350

Dec. 2017 nitrate at 55 m. BGC-Argo observations shown with filled circles.

B-SOSE Iteration 121. RMS fit to obs: 19.08 µmol/kg

World Ocean Atlas 2013 climatology. RMS fit to sobs: 15.72 µmol/kg

SOCCOM

WOA 33% more of consistent with obs.



June 2017 nitrate at 55 m. BGC-Argo observations shown with filled circles.

B-SOSE Iteration 121. RMS fit to obs: 3.4 µmol/kg

World Ocean Atlas 2013 climatology. RMS fit to solve obs: 3.9 µmol/kg

SOCCOM

B-SOSE 12% more consistent with obs.











Validation

Hold the *model* accountable to the *data*

AND ALSO

Hold the *data* accountable to the *model*

Feedback to QC effort. Now posting table of largest normalized misfits.

Normalized pH misfit





BGC-Argo is achieving substantial in situ coverage. Now appropriate to construct a state estimate synthesizing observations and the laws of nature represented by a model





SOCCOM

State estimate goals:

•Propagate observational information via model dynamics.

•Produce a gridded data set consistent with model physics and observations, increasing accessibility

 Provide quantitative baselines: Estimating the "normal" and a framework to understand the "normal"

sose.ucsd.edu

Input and output freely available

Code is well-document and on github

Verdy and Mazloff, 2017. JGR.

2018 OCB Workshop

Nitrate transport



Volume transport in neutral density, γ , layers for the (A) Atlantic, (B) Indian, and (C) Pacific sectors of the Southern Ocean in Sv. Nitrate transport in γ -layers is shown below (E, F, G) in Tmol/yr. The transport is quantified across 32°S into the Atlantic Ocean for the layer 25.99< γ < 27.76, into the Indian Ocean for the layer 25.99< γ < 27.55, and into the Pacific for the layer 25.18< γ <27.34 for (D) volume and (H) nitrate (layers denoted by a dashed black line).

The mean volume transport is 9.71 ± 5.07 Sv, 13.09 ± 6.77 Sv, and 7.51 ± 5.18 Sv for the Atlantic, Indian, and Pacific Oceans, where the \pm gives the standard deviation from the 3-day average time series. The nitrate volume transport for these three basins is 5.57 ± 2.94 Tmol/yr, 5.09 ± 3.06 Tmol/yr, and 1.78 ± 1.91 Tmol/yr.

Budget diagnostic software and examples

http://sose.ucsd.edu/budgets/

BGC budgets being presented for BSOSE iteration 105 (2008-2012) In development. Currently has: DIC, O2, NO3, and FE for top 150m and top 650m For five regions of the Southern Ocean we present:



The Southern Ocean DIC budget analyzed in a state estimate revealing the impact of components at various time and space scales. **Rosso et al., 2017**

Seasonal cycle averaged 26.6°-78°S =>

adv + dif + bio, air-sea, dilution = dDIC/dt



SOCCOM

Other B-SOSE related highlights

Extensive work on observing system design and mapping methods using the B-SOSE model, machine learning techniques, float data, and satellite data

- Correlation scales of heat and carbon content and their fluxes as first guess at observational density requirements.
 Mazloff et al., 2018
- Assessment of importance of sampling strategy and T,S covariance for NO₃ reconstruction. Liang et al., 2018
- Assessment of importance of covariance and sampling strategy for mapping O₂ at 100 m via "Random Forest" machine learning technique. Giglio et al., 2018
- Methodology developed to estimate where a float will be in 100 days based on 1/12° model. Validate estimate with actual trajectories. Wang et al., 2018

Correlation to OOI location









178⁰W

176⁰W



Changing biogeochemical budgets and the role of the ocean in climate and supporting marine resources

For B-SOSE results and validation: http://sose.ucsd.edu/

- Don't hesitate to request diagnostics or assistance with analysis!
- References:
- Verdy and Mazloff, 2017. A data assimilating model for estimating Southern Ocean biogeochemistry. JGR.
- Rosso, Mazloff, Verdy, and Talley, 2017.
 Space and time variability of the Southern Ocean carbon budget. *JGR*.
- Mazloff, Cornuelle, Gille, and Verdy, 2018. Correlation Lengths for Estimating the Large-Scale Carbon and Heat Content of the Southern Ocean. JGR.



O2 at 50m. Floats superimposed on B-SOSE