

GOCAD and the Remote Sensing of Dissolved Organic Carbon and CDOM in the Global Ocean.

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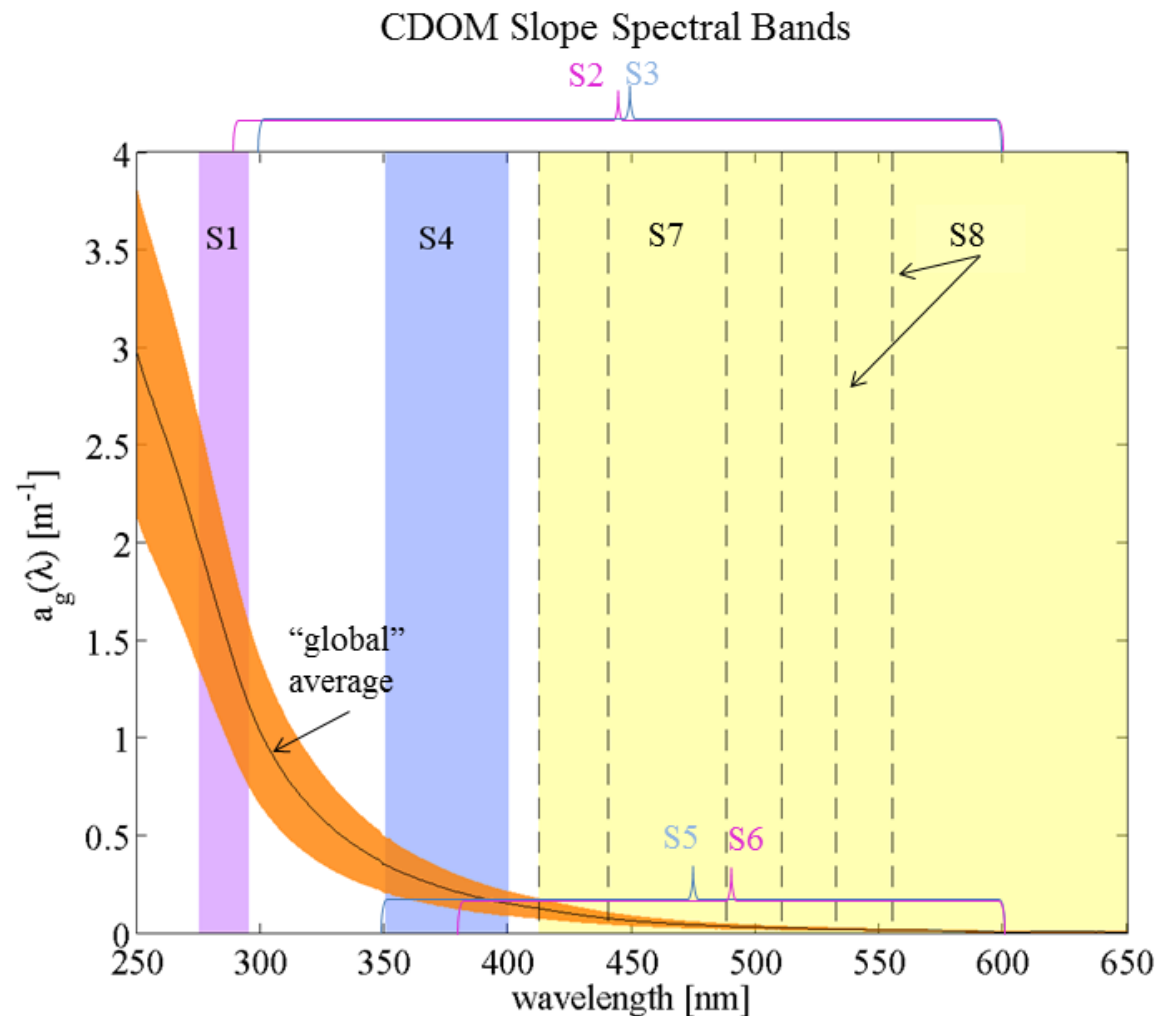
Funded by the Science of Terra and Aqua. Acknowledgements to Chris Proctor and Jeremy Werdell of NASA/OBPG, and all PIs who contributed CDOM & DOC data to SeaBASS.

DOC and Climate

- S. Arrhenius 1896: CO₂ from fossil fuel burning leads to greenhouse effect
- Siegenthaler and Sarmiento 1993: DOC comprises vast majority of marine organic carbon ~equivalent to atmospheric CO₂ pool
- Hedges 2002: Oxidation of just 1% of DOC would generate a flux of CO₂ into the atmosphere equal to all fossil fuel burned in a year
- Belanger et al. 2006: Photoproduction of CO₂ through oxidation of CDOM increased ~15% in the Arctic recent years due to the decrease in sea ice

DOC and Ocean Color; CDOM

$$a_g(\lambda) = a_g(\lambda_0) \exp(-S_g(\lambda - \lambda_0))$$



Spectral slopes reflect age, molecular weight, origin, photooxidation, etc.

NOMAD and the synthetic IOCCG datasets have been extremely useful to algorithm development, but for our purposes, a carbon-centric, ocean color database is required which extends into the UV, includes larger numbers of field stations (~40X more), and matches CDOM & DOC to satellite imagery as well as in situ radiometry.

Global Ocean Carbon Algorithm Database GOCAD

GOCAD

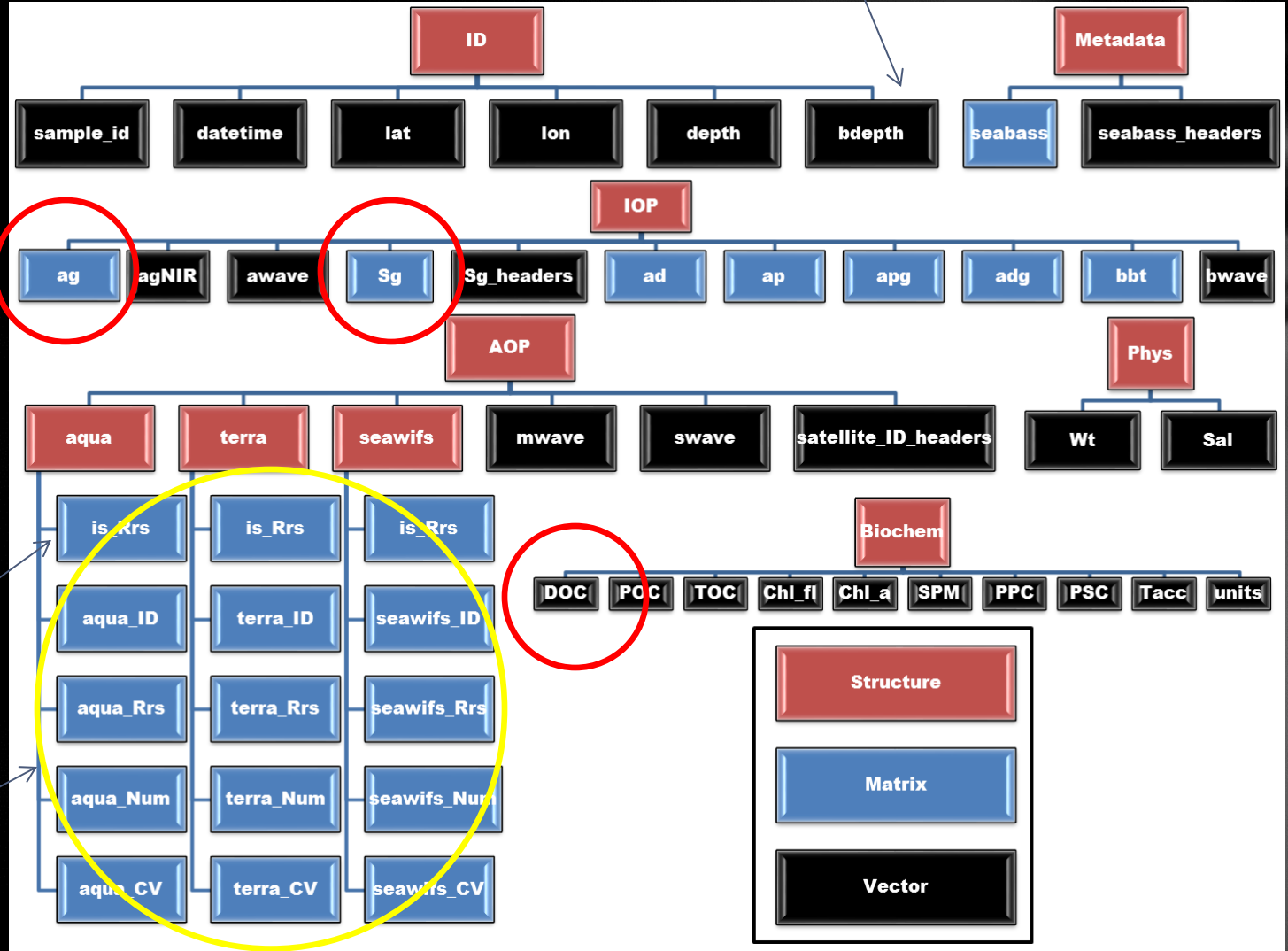
GEBCO 08 0.05

SeaBASS
Hansell/Carlson

5-10 m surface
average, 1 km
gridded,
hyperspectral,
Bailey &
Werdell 2006
(5x5, 8 hrs,
etc.)

RSR-weighted

OBPG L2



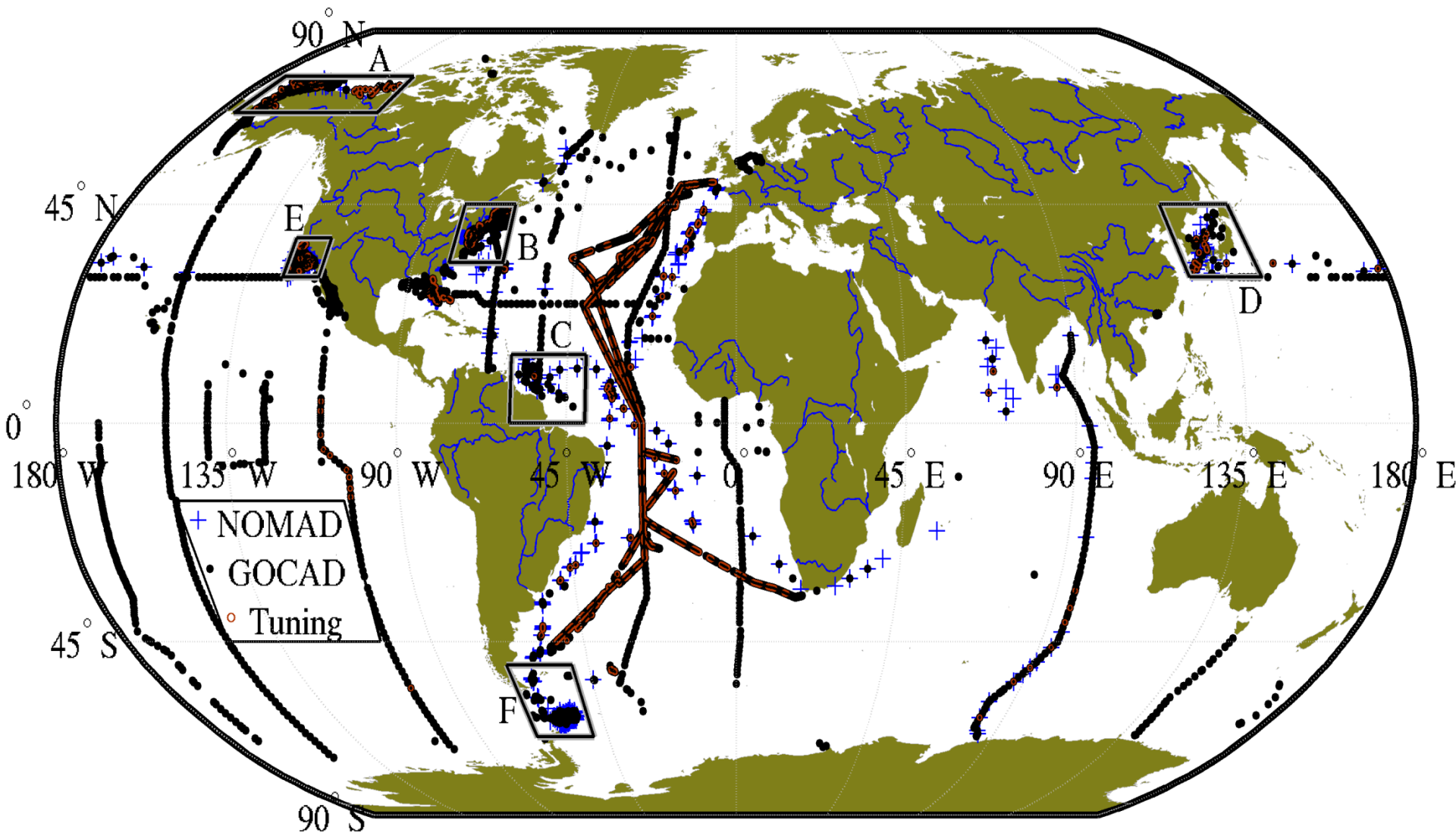
Rigorously quality controlled!

Table 2. Summary of field data collection campaigns.

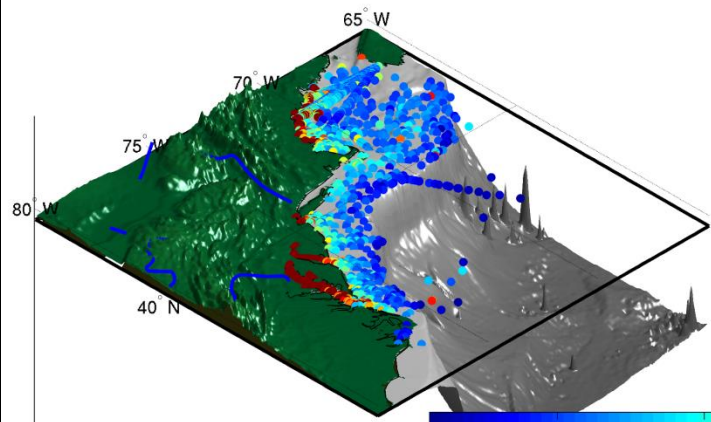
Experiment	Principal Investigators	Cruises	Numbers of Stations					Min. Lat	Max. Lat	Min. Lon	Max. Lon	Year(s)	
			CDOM	CDOM & IS*	CDOM & SAT**	DOC & IS*	DOC & SAT**						
CV1	A. Mannin	PenBaySurvey	C. Roesler	NSF-BWZ	G. Mitchell	2							
CV2	A. Mannin	San Diego Coastal Project	D. Stramski, M. Stramska	Benthic Ecol. from Space	H. Dierssen, R. Zimmerman	4							
CV3	A. Mannin	NASA Gulf of Maine	D. Phinney, D. Pi J. Brown	Kieber	Zimmerman								
CV4	A. Mannin			Photochemistry 03	H. Sosik	1							
CV5	A. Mannin			MVCO	H. Sosik	34							
CV6	A. Mannin	NOAA Gulf of Maine	J. Brown		J.R.V Zaneveld, W.S. Pegau	6							
D01	A. Mannin		D. Phinney, D. Pi	GOCAL		5							
D02	A. Mannin	Panama City Florida	J. Brown	PREPP	J. Chen	5							
D03	A. Mannin	Plumes and Blooms	D. Siegel	SAB Mapping	J. Nelson, A. Subramaniam	2							
D04	A. Mannin	CLIVAR A13.5 2010	D. Hansell	GEOTRACES	J. Chaves	1							
OCV1	A. Mannin	CLIVAR I05 2009	D. Hansell	COOA	J. Salisbury, D. Vandemark, C. Hunt	3							
OCV2	A. Mannin	CLIVAR I08S 2007	D. Hansell		J. Brock, A. Subramanian, K. Waters	1							
OCV3	A. Mannin	CLIVAR P02 2004	D. Hansell	NOAA CSC		1							
OCV5	A. Mannin	CLIVAR P16N 2006	D. Hansell	BOA	K. Carder	1							
PL6	A. Mannin	CLIVAR P16S 2005	D. Hansell	EcoHAB	K. Carder	19							
GOMECC	A. Mannin	CLIVAR P18 2007	D. Hansell	Okeechobee	K. Carder	1							
GEO-CAPE	A. Mannin	HLY-02-01	D. Hansell	Redtide	K. Carder	2							
MONTEREY BAY	B. Arnone	HLY-02-03	D. Hansell	TOTO	K. Carder	3							
WOCE P14S P15S	B. Tilbroo	SR03	D. Hansell		L.W. Harding Jr., M. Mallonee, A. Magnuson	6							
Carbon Transport MS		WOCE AR01 A05	D. Hansell	ACE-INC									
R.	C. Del Cas	ACE-ASIA	G. Mitchell, M. F		L.W. Harding Jr., M. Mallonee, A. Magnuson	11							
GasEx	C. Del Cas	AMLR	G. Mitchell, M. F										
Big Bend	C. Hu	Aerosols Index	G. Mitchell, M. F										
GEO-CAPE		CALCOFI	G. Mitchell, M. F										
CBODAQ	C. Hu	Sea_of_Japan	G. Mitchell, M. F										
GOM Oil Spill	C. Hu	Arc00	G. Cota										
		TAR07	G. Cota										
Catlin Arctic Survey	V. Hill	1	8	0	0	0	0	0	78.771	78.771	-104.721	-104.721	2011
AMT	W. Balch	5	23211	7703	1918	0	0	0	-47.27	49.716	-55.455	18.611	2005-2011
Gulf of Maine	W. Balch	33	8077	0	3251	0	0	0	42.683	44.058	-70.267	-66.172	2005-2008
Scotia Prince Ferry	W. Balch	47	11521	51	8397	0	0	0	43.604	43.798	-70.026	-66.164	2001-2004
2009 Oct Chesapeake	W.J. Rhea	1	13	10	4	0	0	0	38.136	39.062	-76.448	-76.229	2009
Totals		535	48574	8857	14568	1957	255	302	-70.999	78.771	-177.99	179.545	1984-2012

* IS is *in situ* $Rrs(\lambda)$, **SAT is satellite $Rrs(\lambda)$

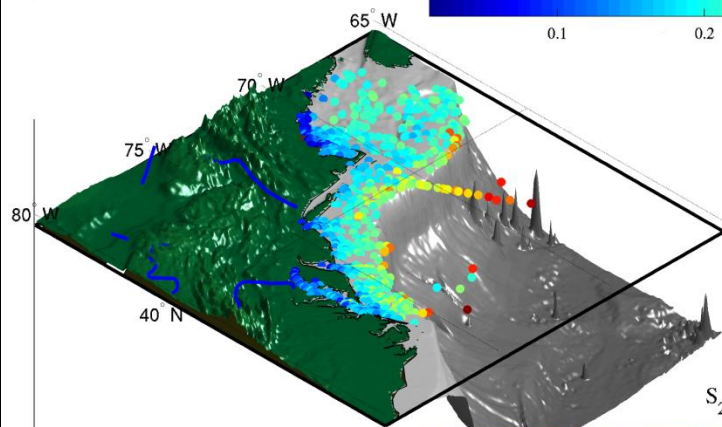
Field Stations with CDOM and/or DOC



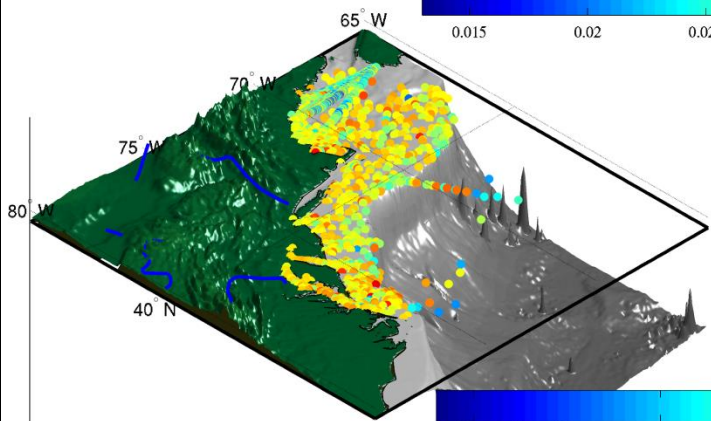
Northeast US



$a_g(412) [\text{m}^{-1}]$



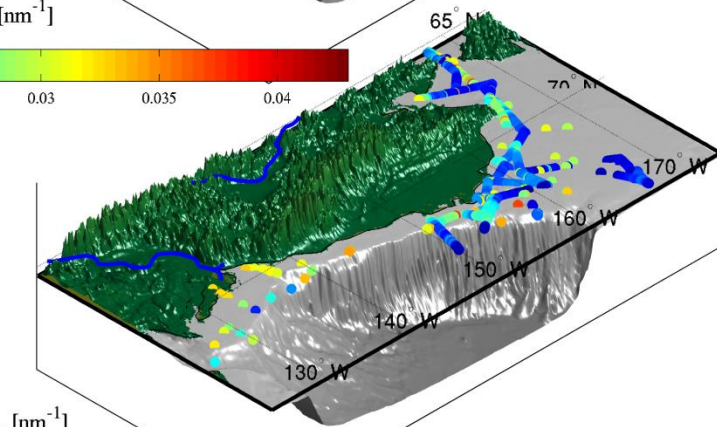
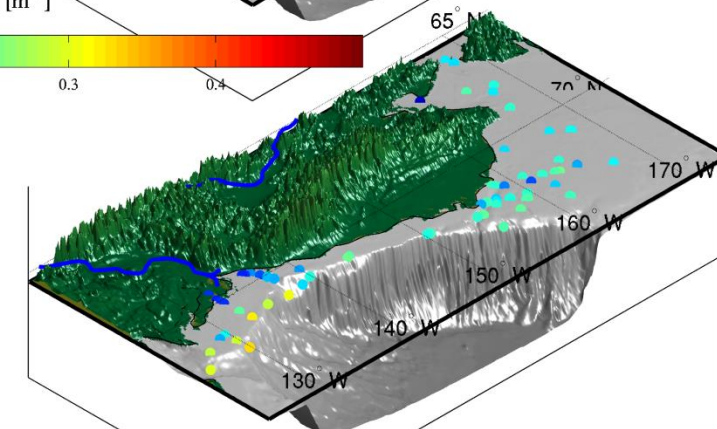
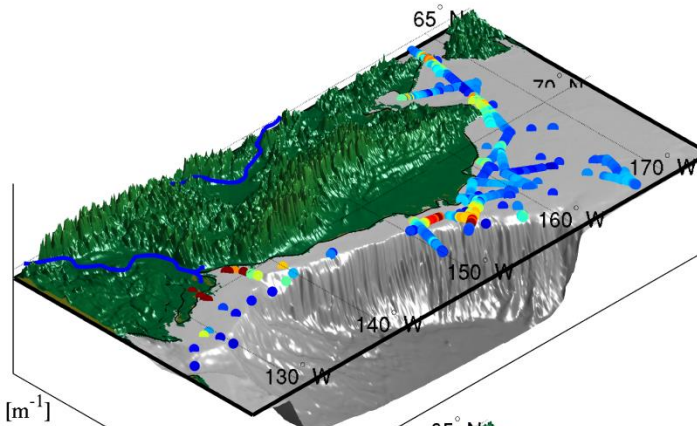
$S_{275-295} [\text{nm}^{-1}]$

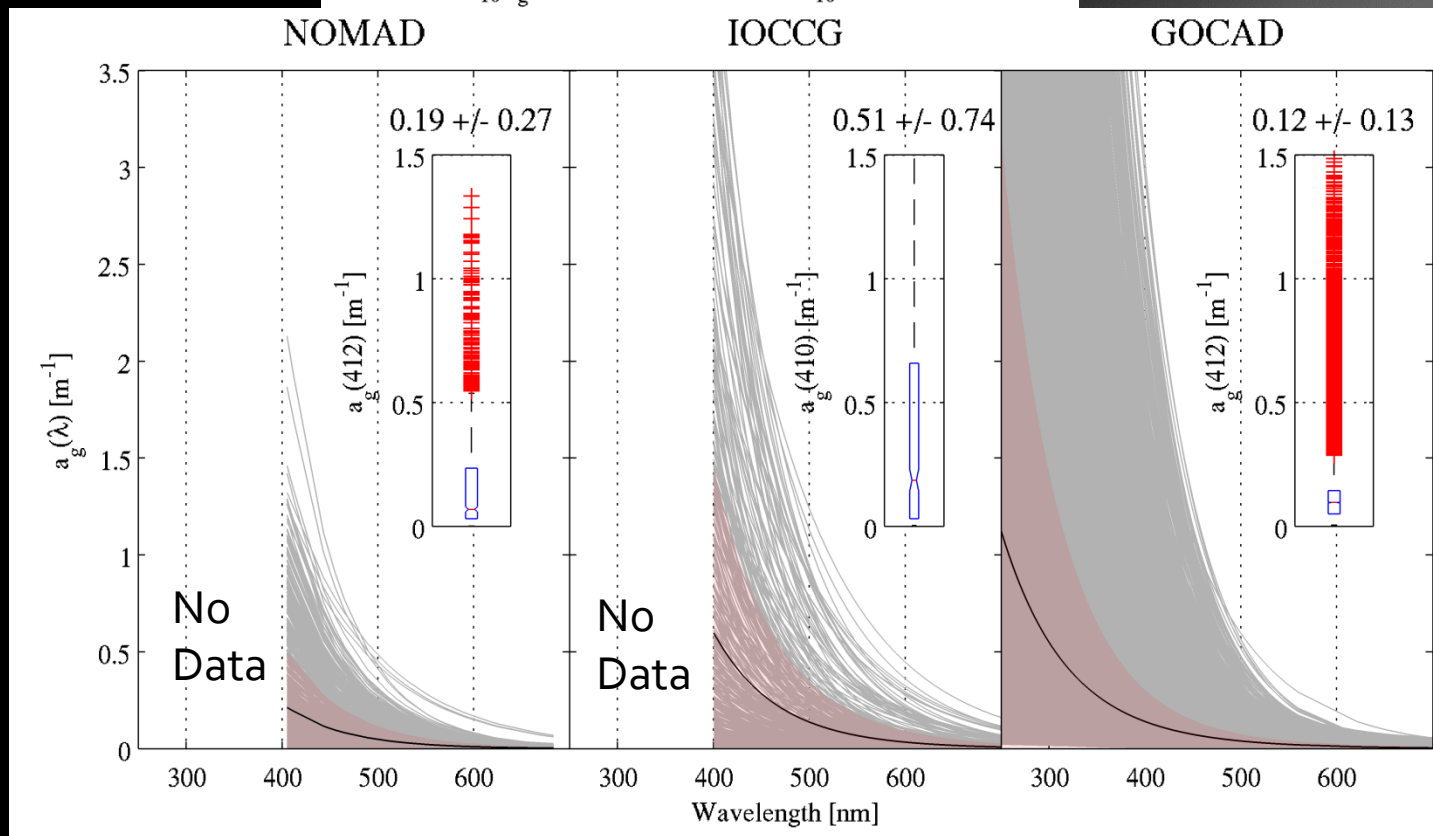
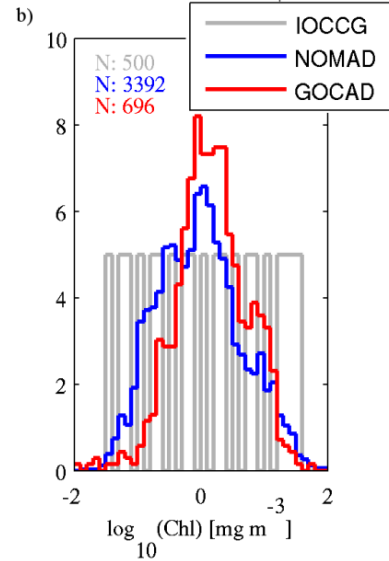
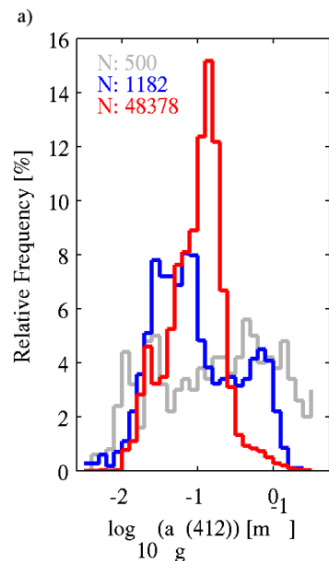


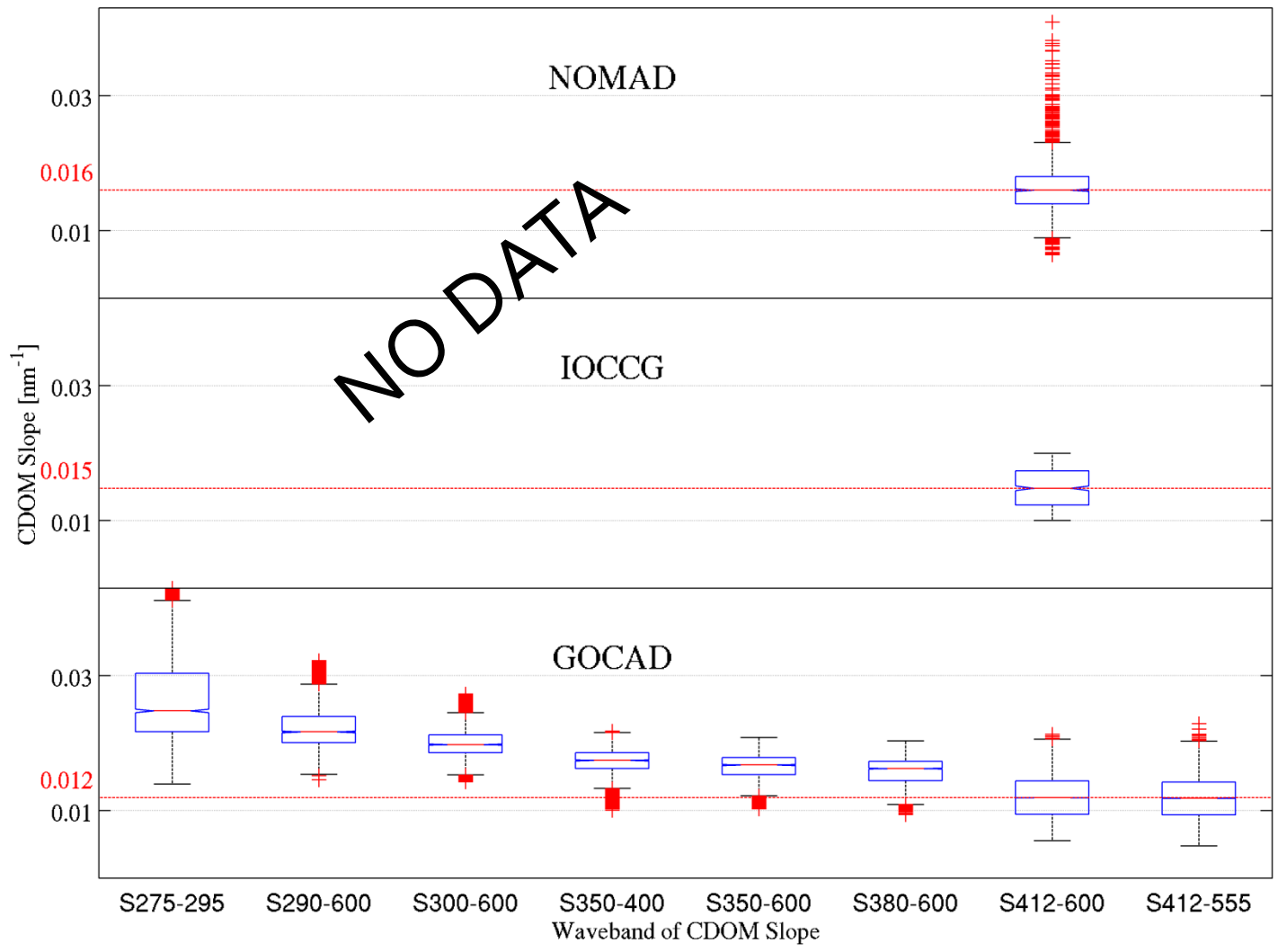
$S_{412-555} [\text{nm}^{-1}]$



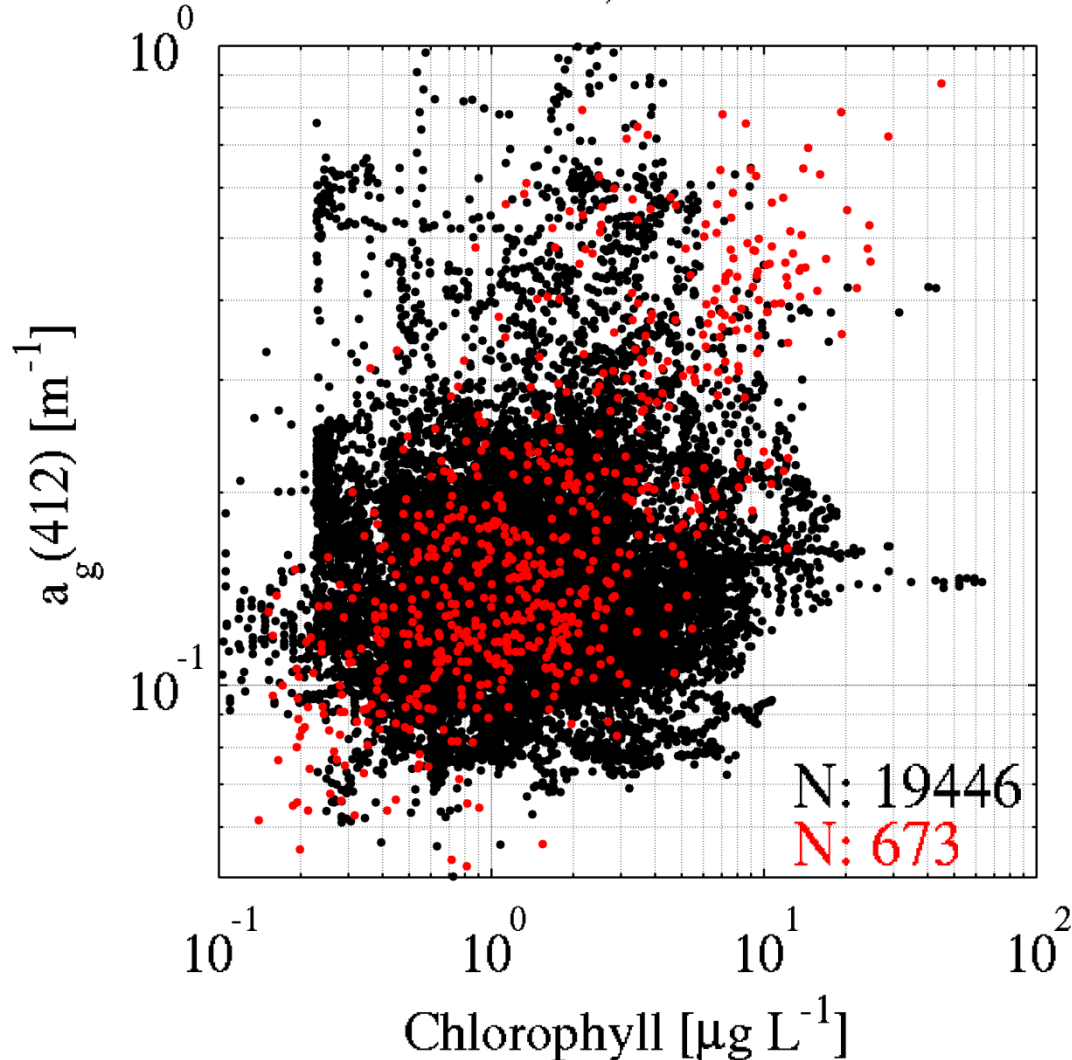
Chukchi/Beaufort Seas







Chl Fluor. r^2 : 0.00; Chl HPLC r^2 : 0.37



What/where *are* Case 1 waters, anyway?

Algorithm Development

- **Empirical Approaches**

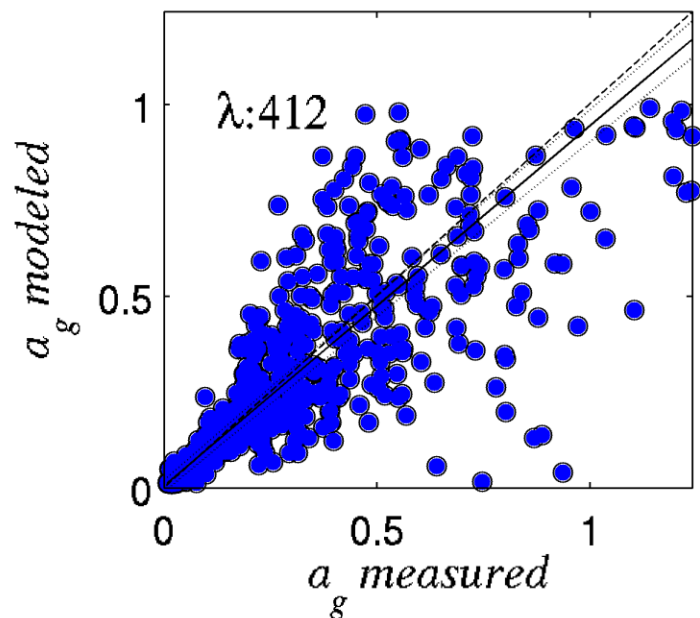
- Band ratio, one-phase exponential decay model (**EXP**)
- Multiple Linear Regression (**MLR**)
- Machine Learning
 - Random Forest Tree-Bagger (**RFTB**)

- **Semi-analytical Approaches** (*Hybridized here for CDOM, not CDM, but no UV, and no DOC)

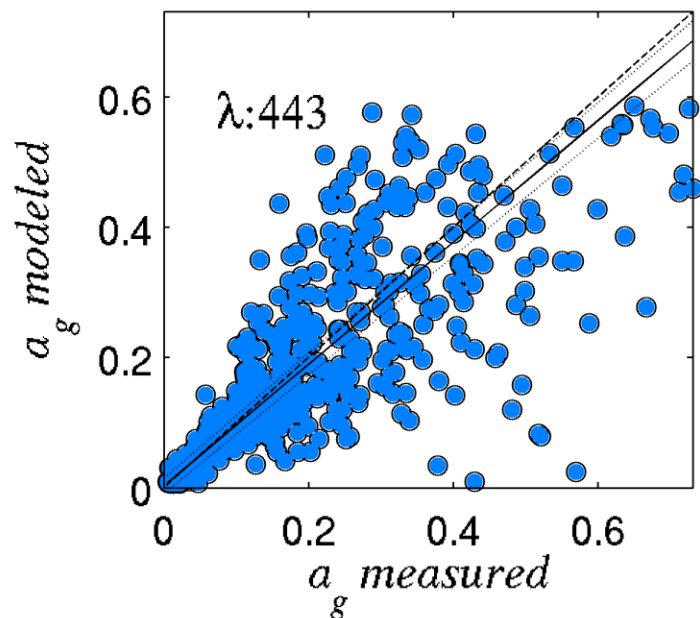
- Quasianalytical Algorithm (**QAA**)
- Generalized Inherent Optical Property (**GIOP**)
- *CDM is related to a_g (CDOM) and b_{bp} (NAP)

- All algorithms were tuned/optimized/trained on *in situ* reflectances and validated on satellite data from independent field stations.

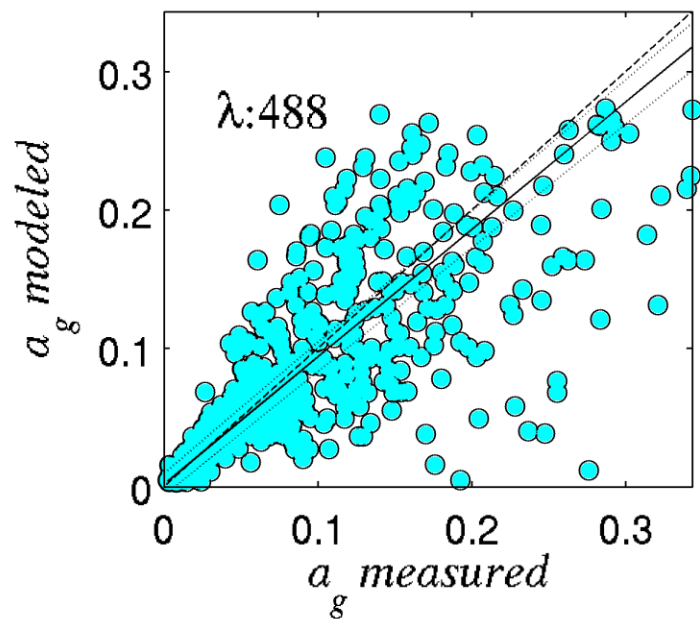
n:654 r^2 :0.66 MAPD:30.2 MSE:0.02 Bias:-0.01



n:654 r^2 :0.65 MAPD:31.3 MSE:0.01 Bias:-0.01

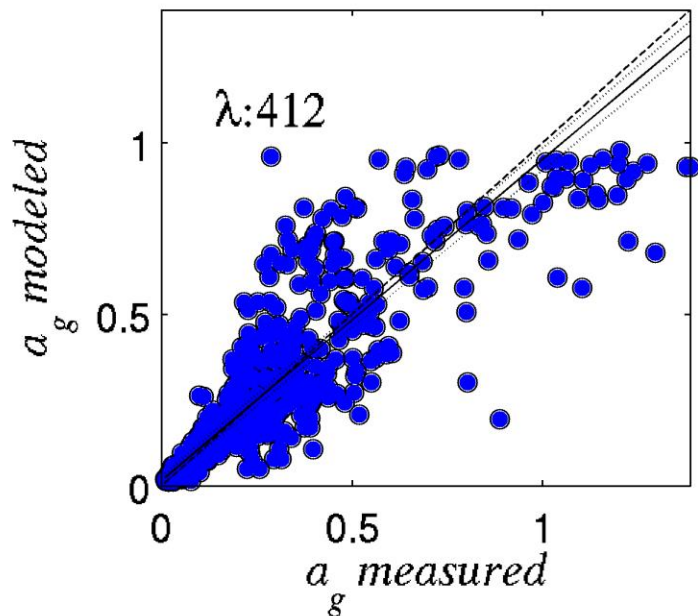


n:654 r^2 :0.63 MAPD:33.6 MSE:0 Bias:0

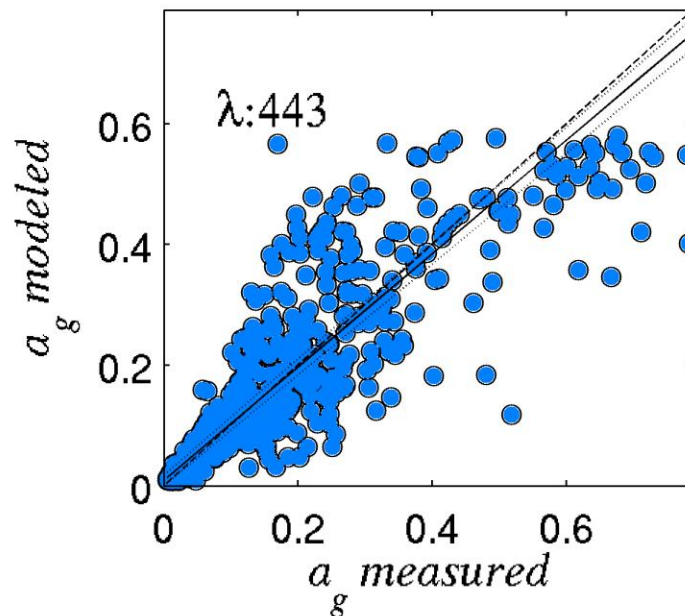


QAA hybrid

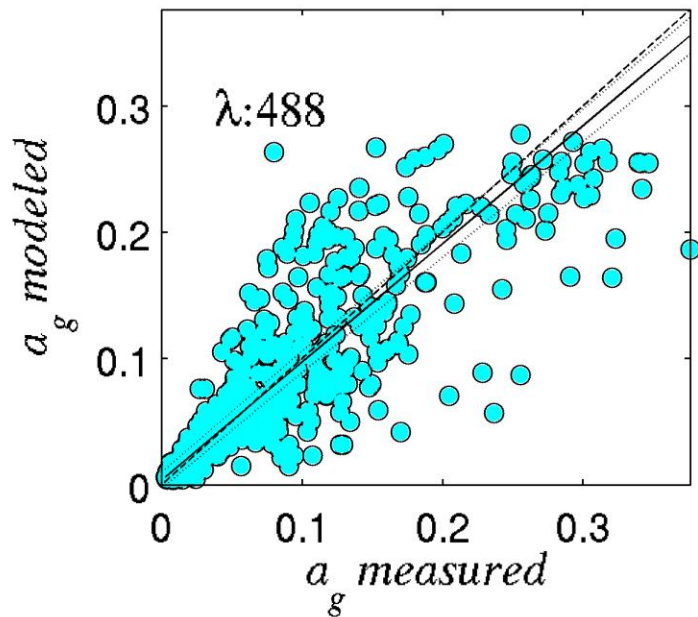
n:625 r^2 :0.77 MAPD:28.4 MSE:0.018 Bias:0



n:625 r^2 :0.76 MAPD:29.8 MSE:0.006 Bias:0



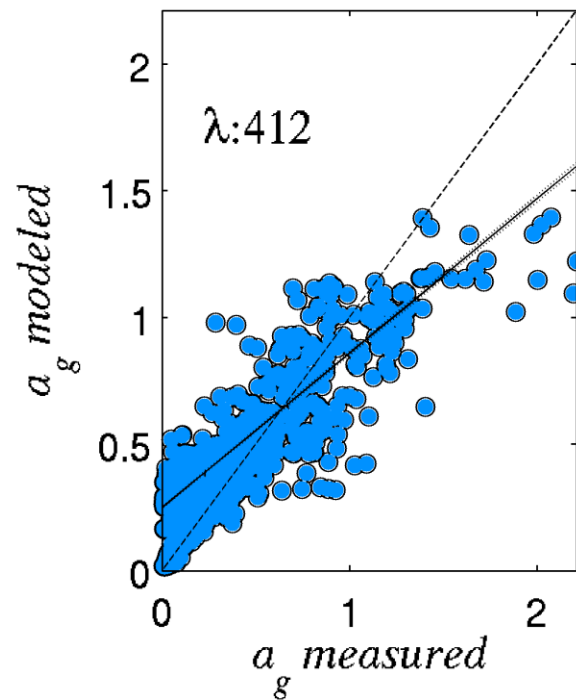
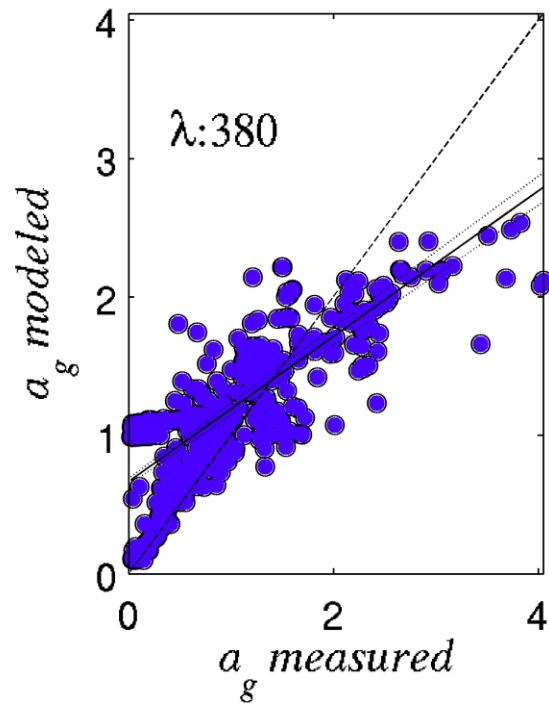
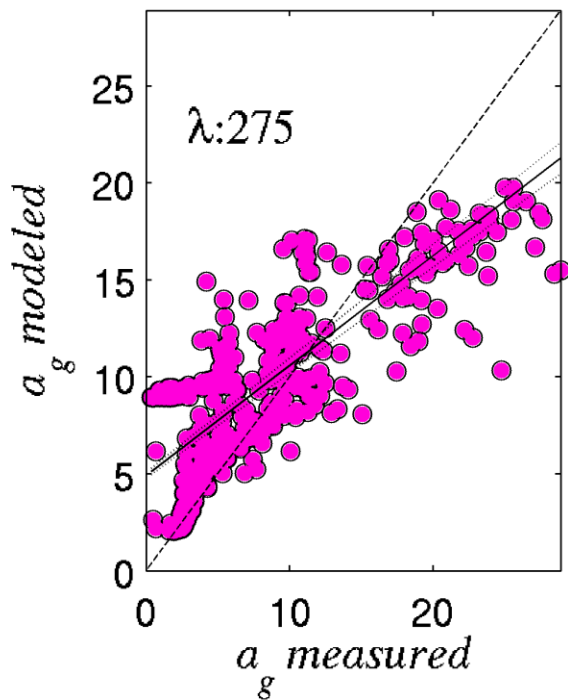
n:625 r^2 :0.74 MAPD:33 MSE:0.001 Bias:0



GIOP hybrid

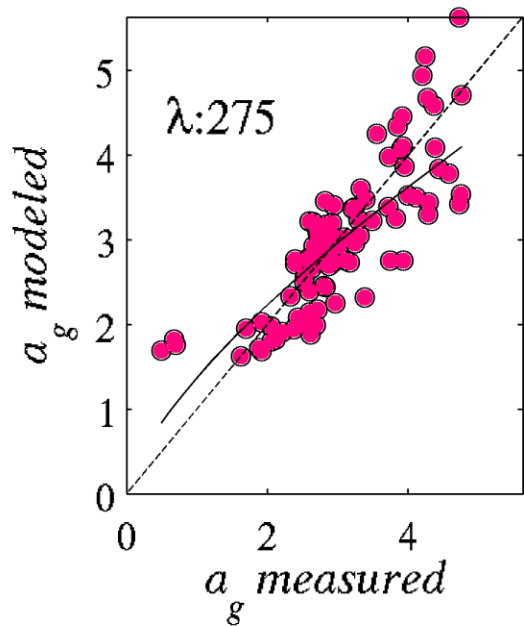
RFTB

n:431 r^2 :0.59 MAPD:161.5 MSE:20.61 Bias:1.48 n:542 r^2 :0.55 MAPD:436.2 MSE:0.38 Bias:0.29 n:8703 r^2 :0.56 MAPD:657.2 MSE:0.06 Bias:0.21

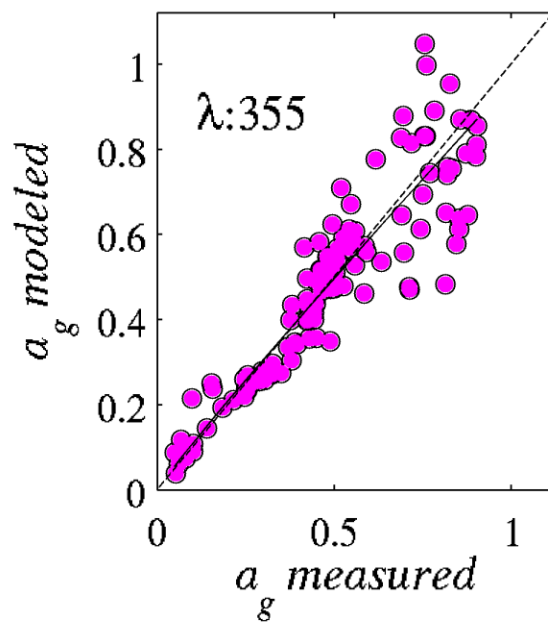


MLR

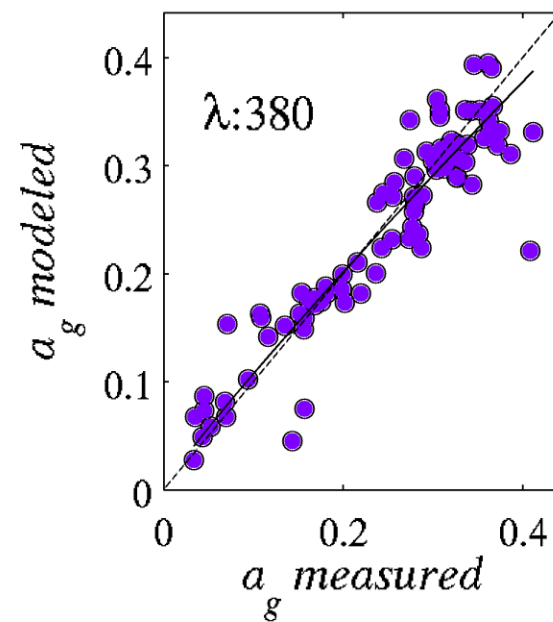
n:100 r^2 :0.58 MAPD:16.4 MSE:0.249 Bias:-0.04



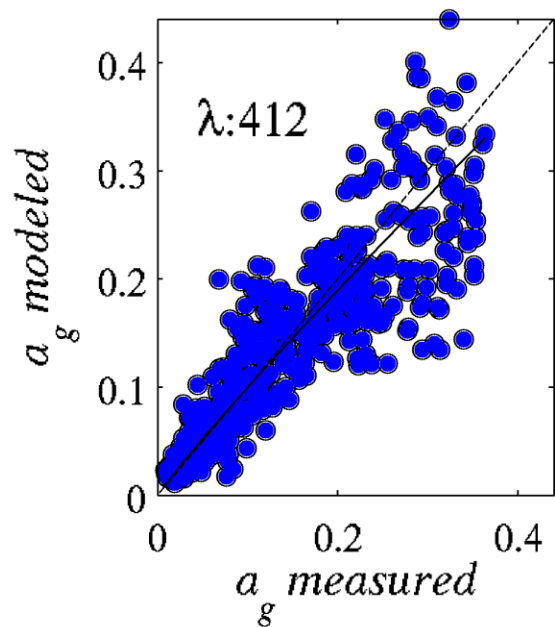
n:116 r^2 :0.91 MAPD:15.6 MSE:0.01 Bias:0



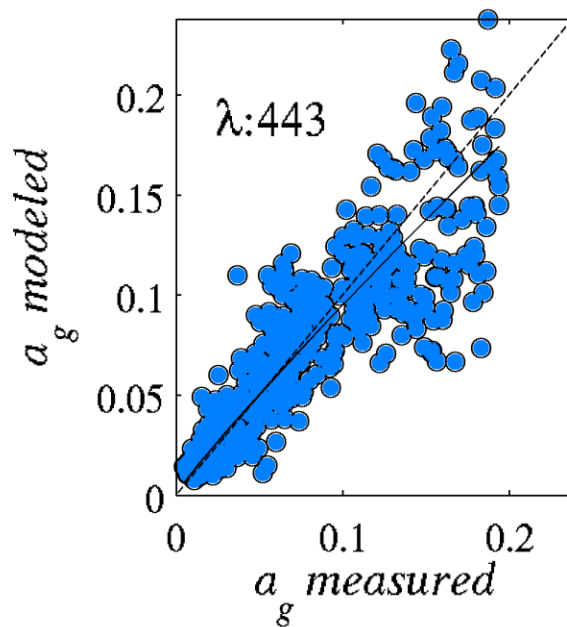
n:86 r^2 :0.84 MAPD:15.3 MSE:0.002 Bias:0



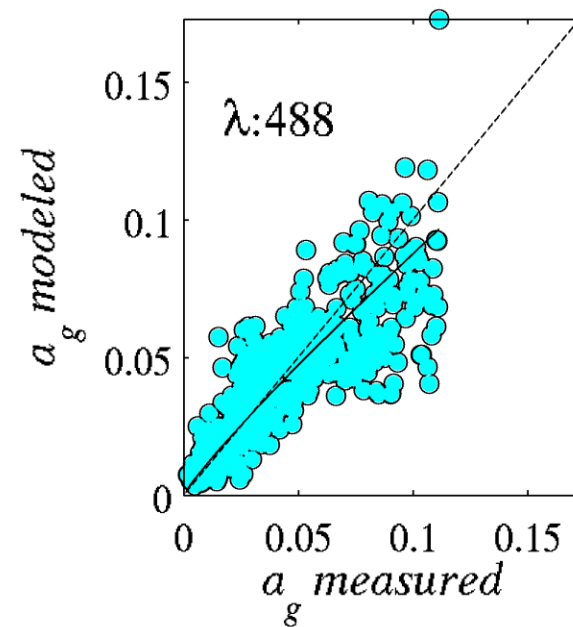
n:483 r^2 :0.87 MAPD:28.2 MSE:0.002 Bias:-0.01



n:462 r^2 :0.85 MAPD:28.6 MSE:0.001 Bias:0



n:490 r^2 :0.82 MAPD:33.7 MSE:0 Bias:0



What



Aqu
Sea Surface

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C07051

MANNINO ET AL.: SATELLITE VALIDATION OF DOC AND CDOM

C07051

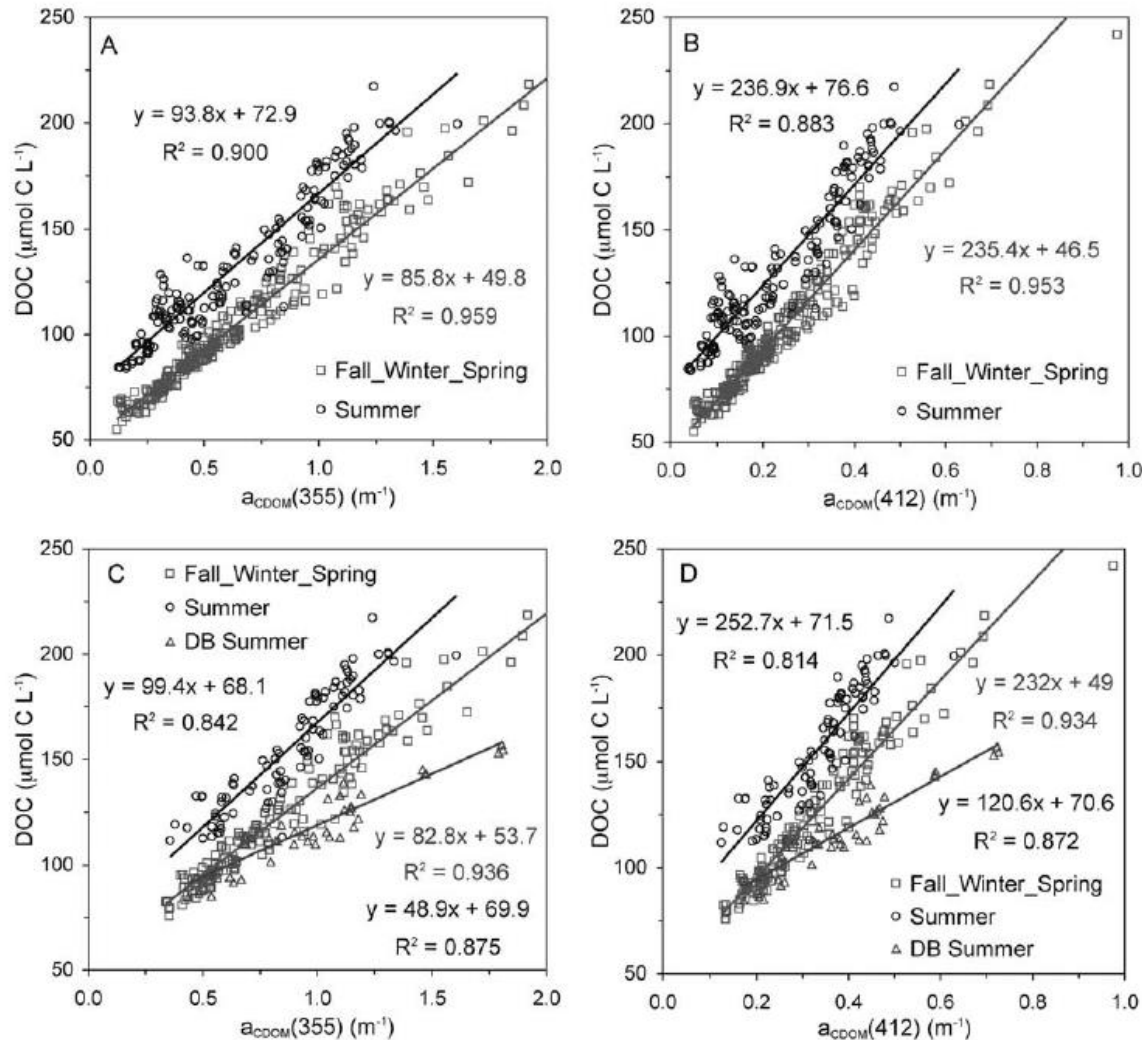
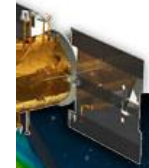
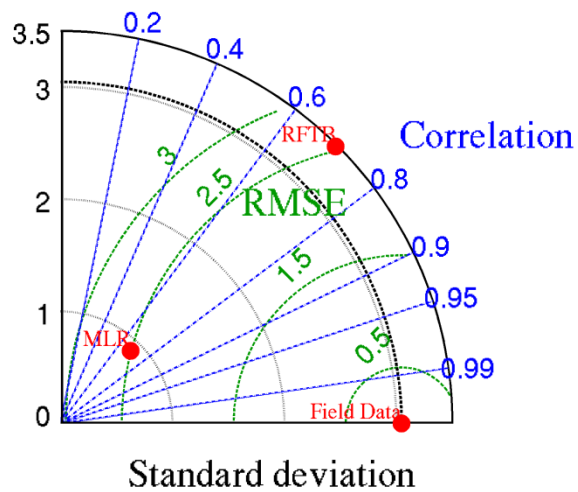
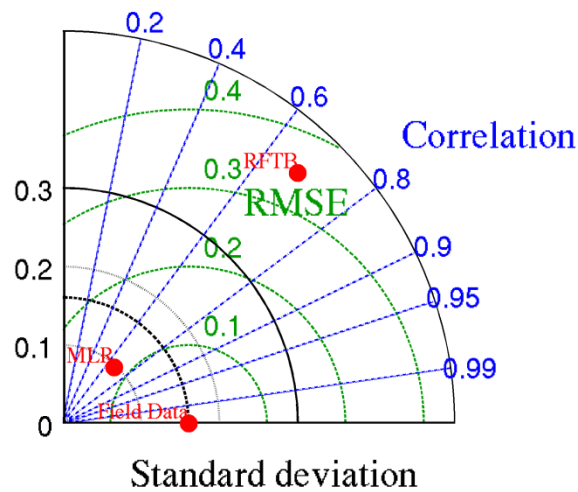
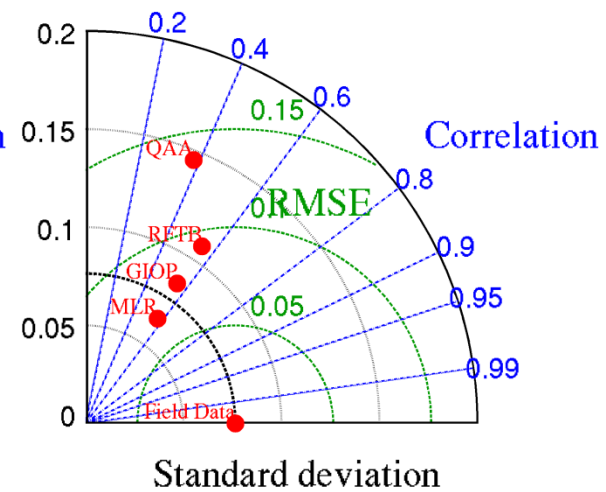
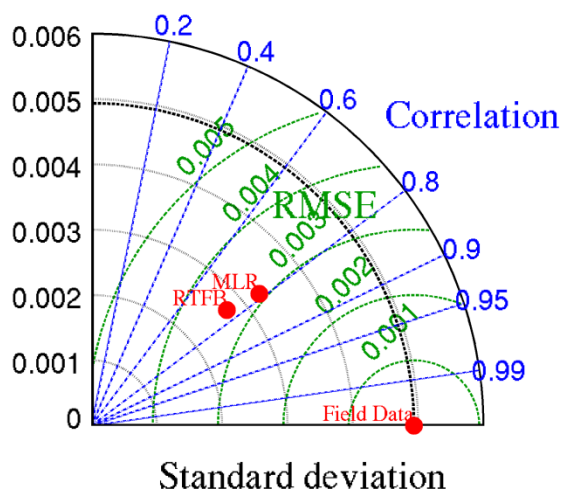
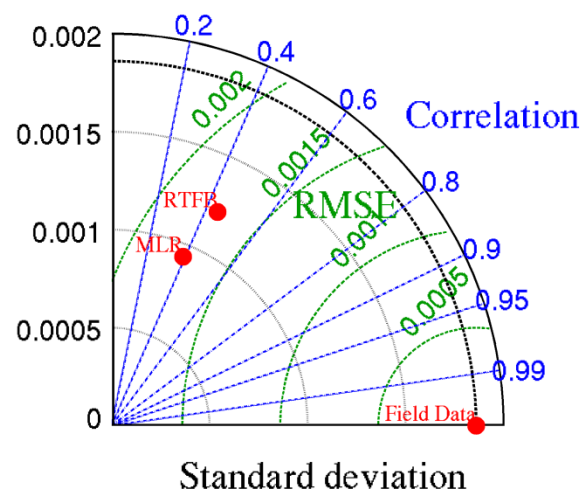
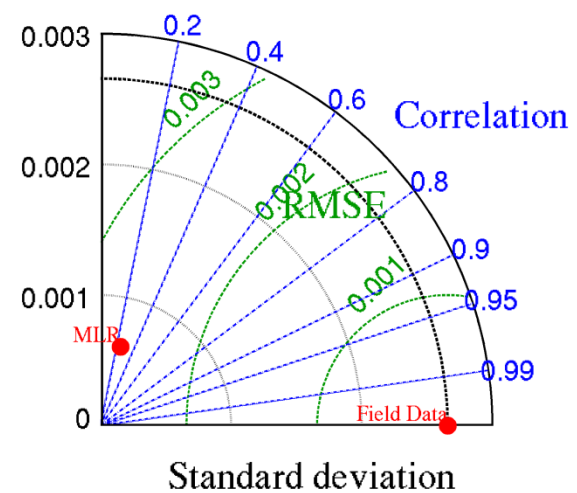


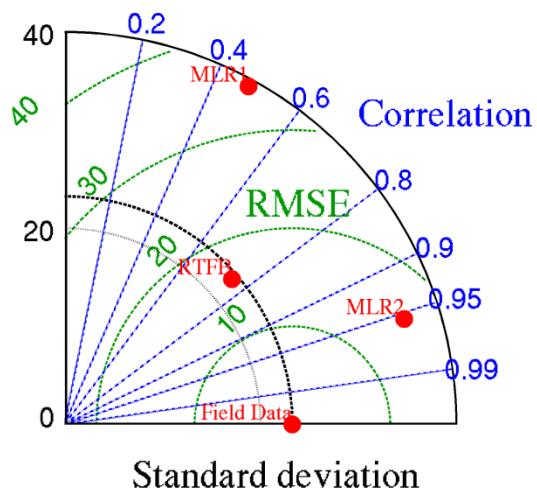
Figure 2. Seasonal relationships of dissolved organic carbon (DOC) and (A) $a_{\text{CDOM}}(355)$ (chromophoric dissolved organic matter absorption coefficient at 355 nm) or (B) $a_{\text{CDOM}}(412)$ within the MAB and between DOC and (C) $a_{\text{CDOM}}(355)$ or (D) $a_{\text{CDOM}}(412)$ in the Chesapeake Bay mouth and plume region for the 2004–2006 research cruises and Delaware Bay mouth and plume region for the summer 2005–2006 cruises (DB Summer). Data shown for Fall_Winter_Spring (October–May) include measurements from all depths sampled, but only the top 2 depths for the Chesapeake Bay mouth transect (depicted in Figure 1A). Summer (June–September) data only include the top 2 depths sampled (surface mixed layer) and exclude the Delaware Bay mouth and plume stations (open triangles in Figure 1).



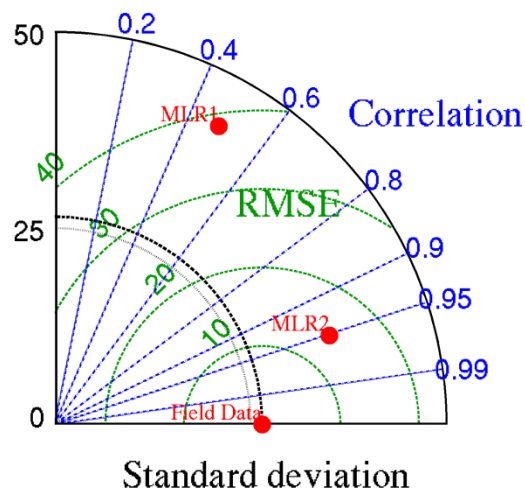
Algorithm Validation

$a_g(275)$  $a_g(380)$  $a_g(412)$  $S_g(275)$  $S_g(300)$  $S_g(412)$ 

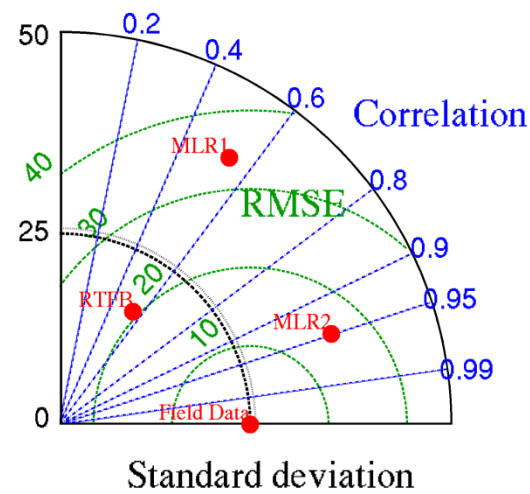
DOC(Aqua)



DOC(Terra)

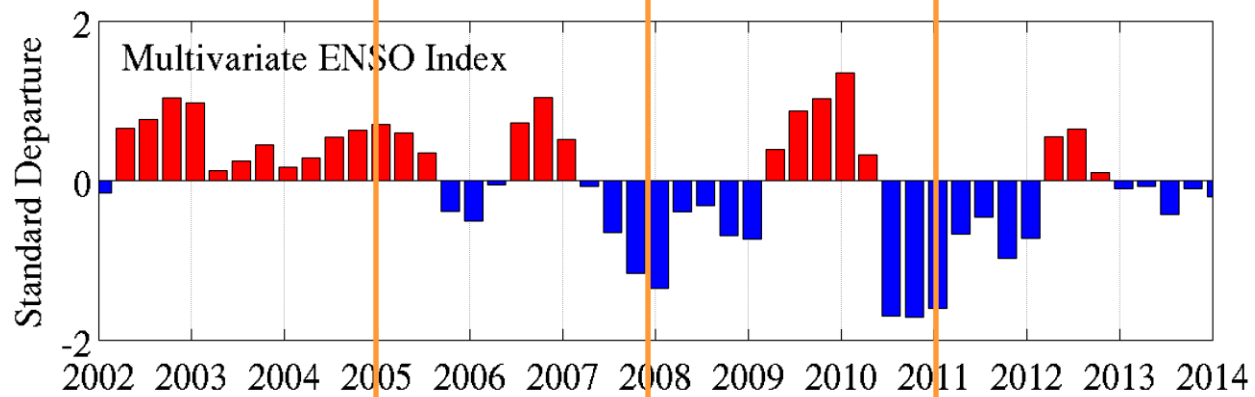
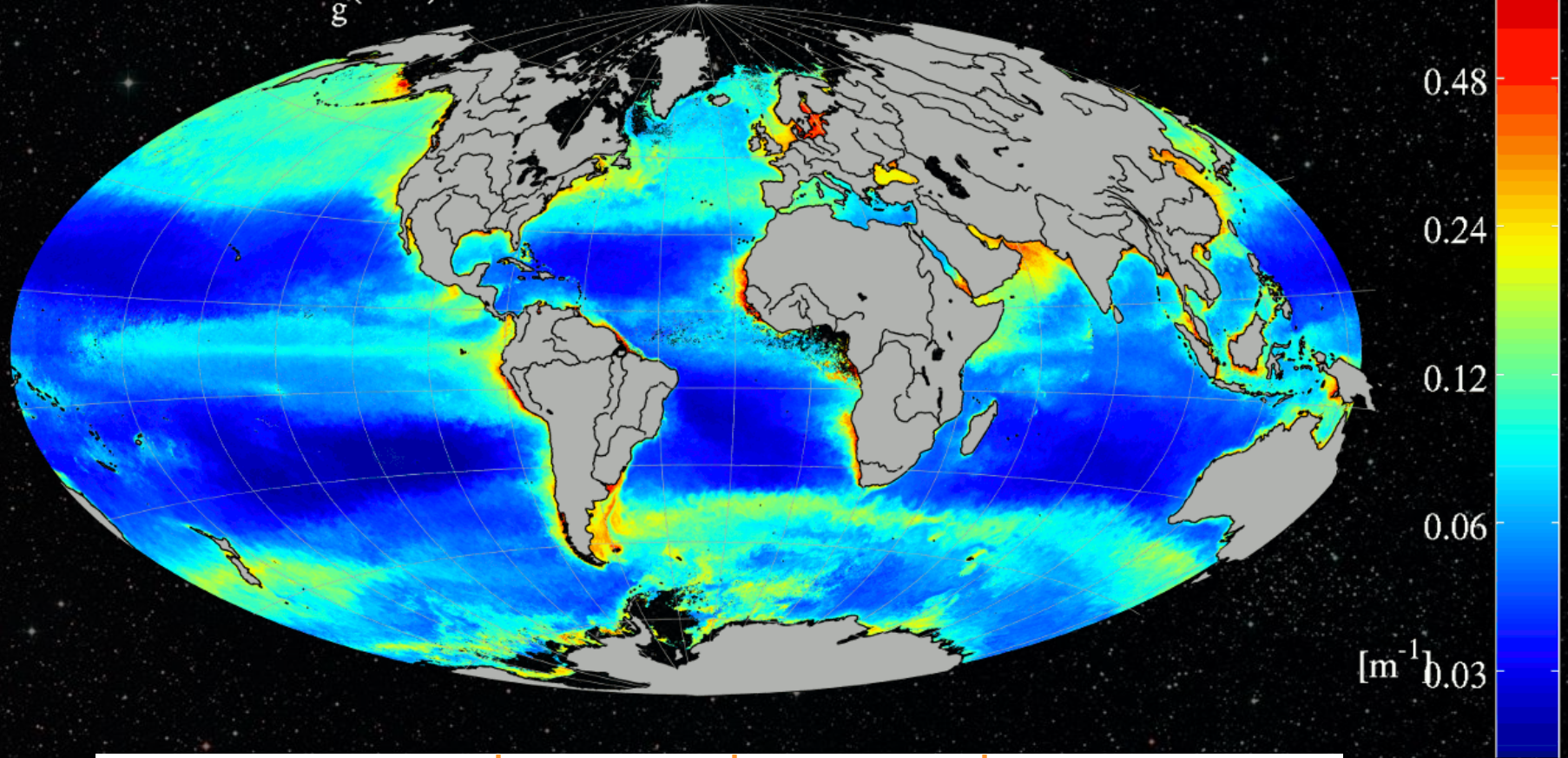


DOC(SeaWiFS)

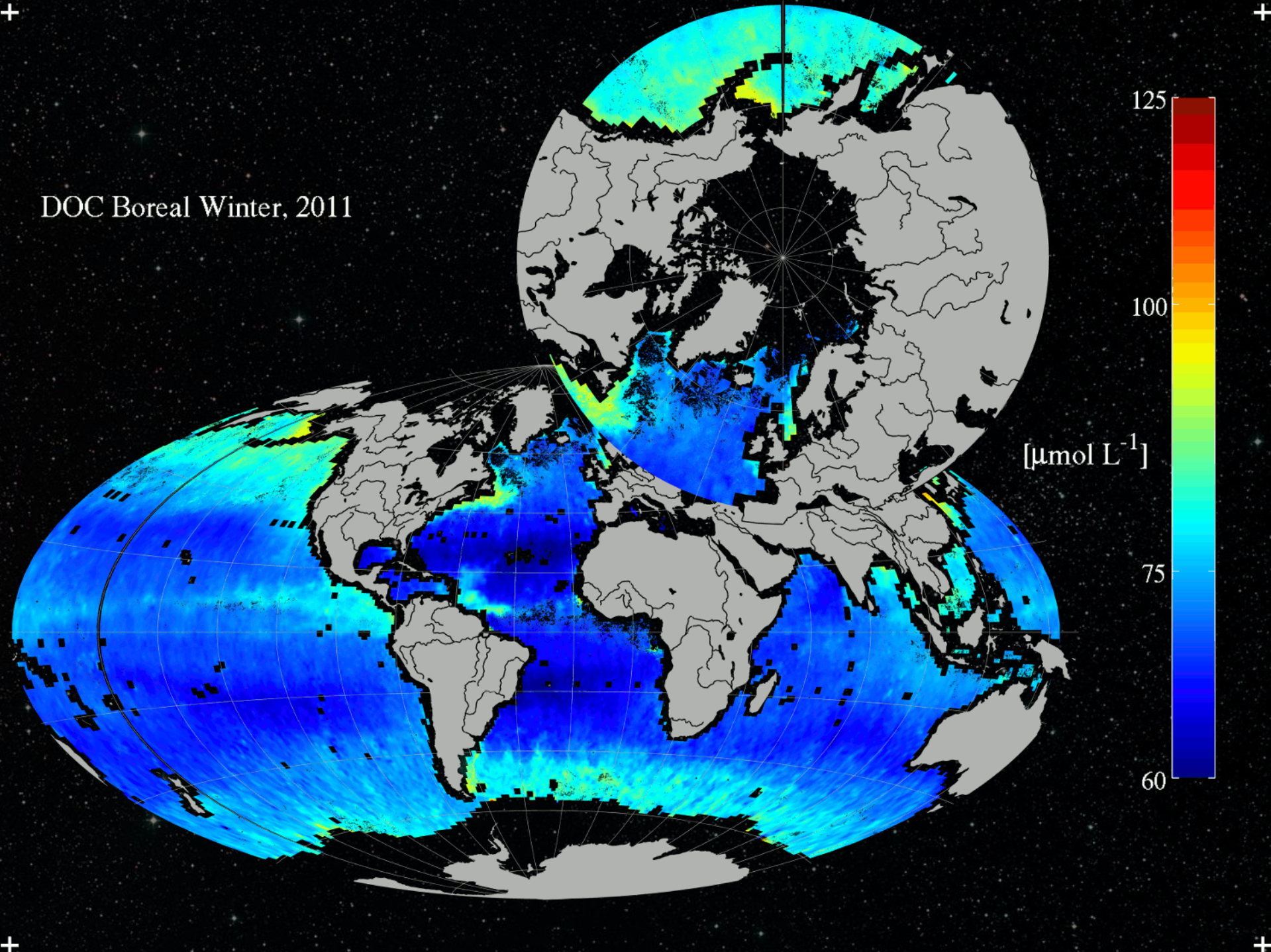


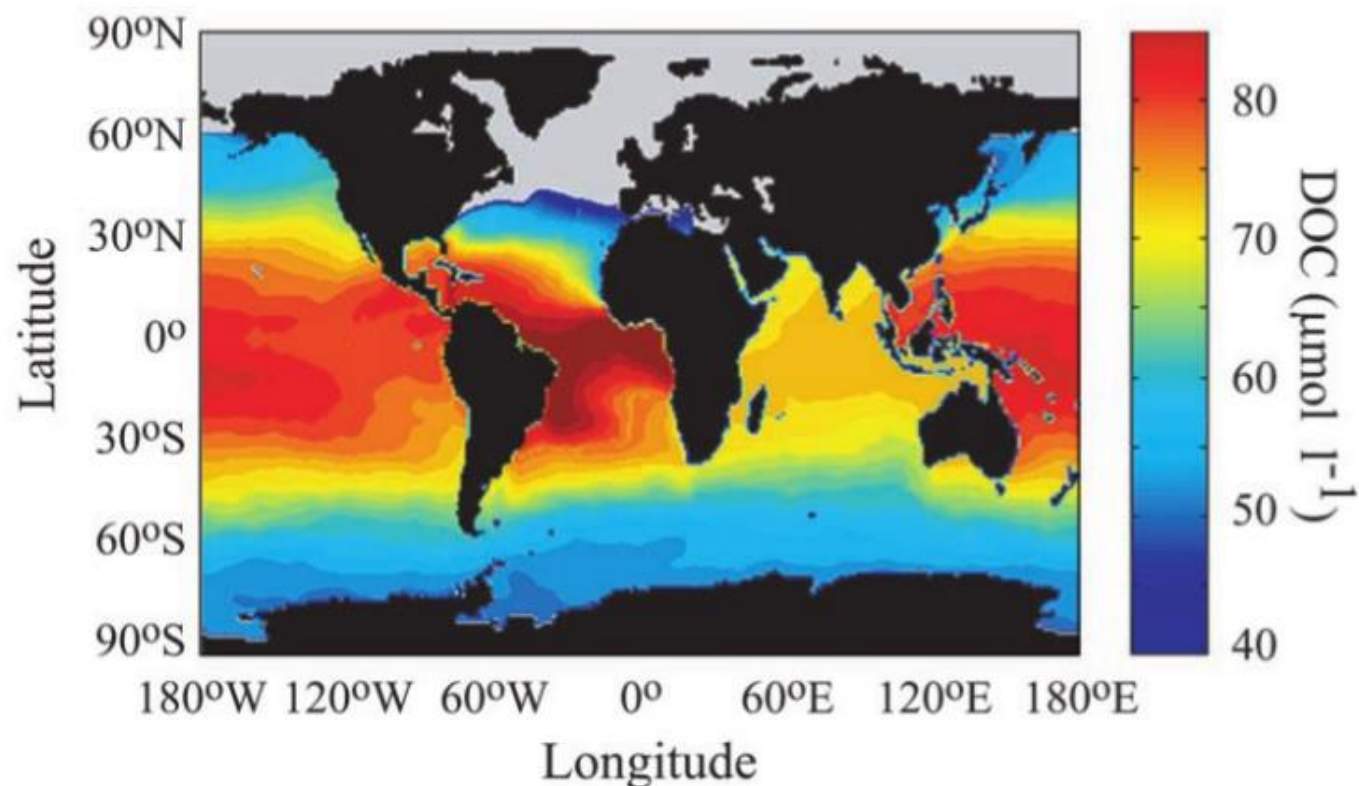
Algorithm Application

a_g (380) Boreal Winter, 2002 2003 2004




DOC Boreal Winter, 2011





JOURNAL OF GEOPHYSICAL RESEARCH, VOL. 107, NO. C12, 3228, doi:10.1029/2001JC000965, 2002

Figure 9. Climatological DOC distribution from a regression analysis based upon wintertime SST values. Concentrations of DOC are in units of $\mu\text{mol L}^{-1}$. The regression models used are presented in Table 2, and further details may be found in the text. 

Global distribution and dynamics of colored dissolved and detrital organic materials

D. A. Siegel,¹ S. Maritorena, and N. B. Nelson

Institute for Computational Earth System Science, University of California, Santa Barbara, Santa Barbara, California, USA

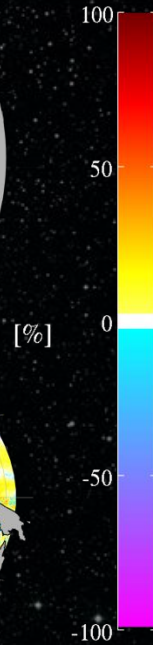
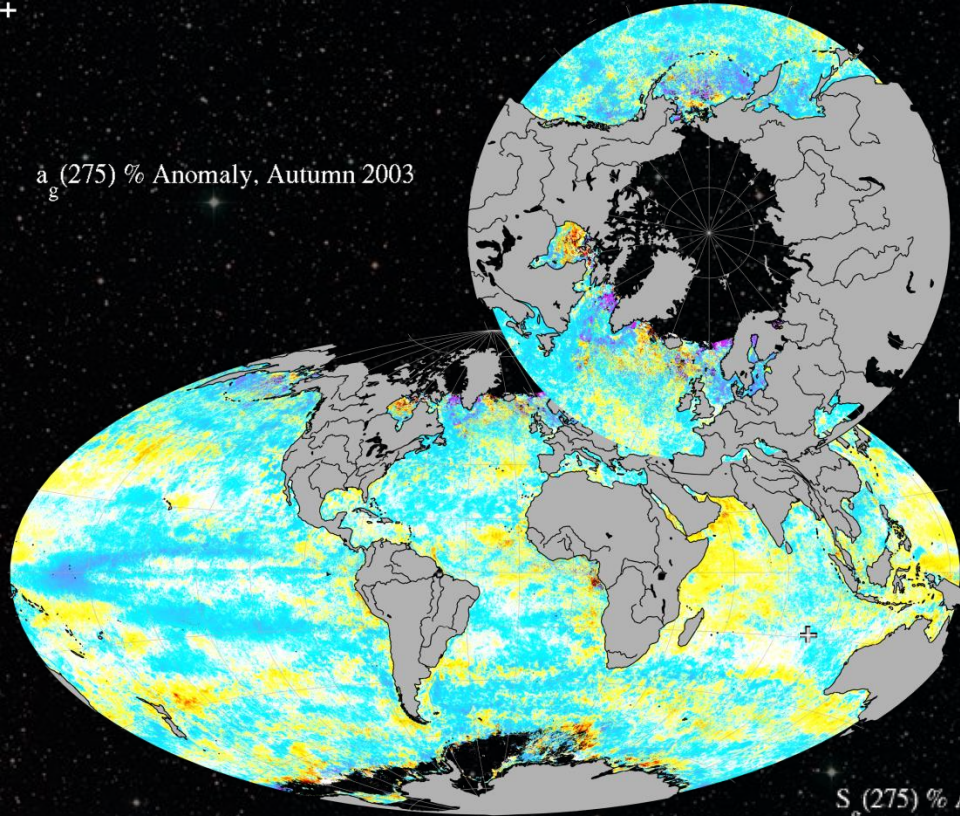
D. A. Hansell

Division of Marine and Atmospheric Chemistry, Rosenstiel School of Marine and Atmospheric Science, University of Miami, Miami, Florida, USA

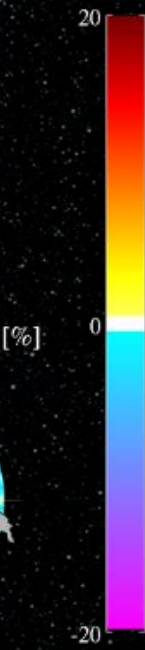
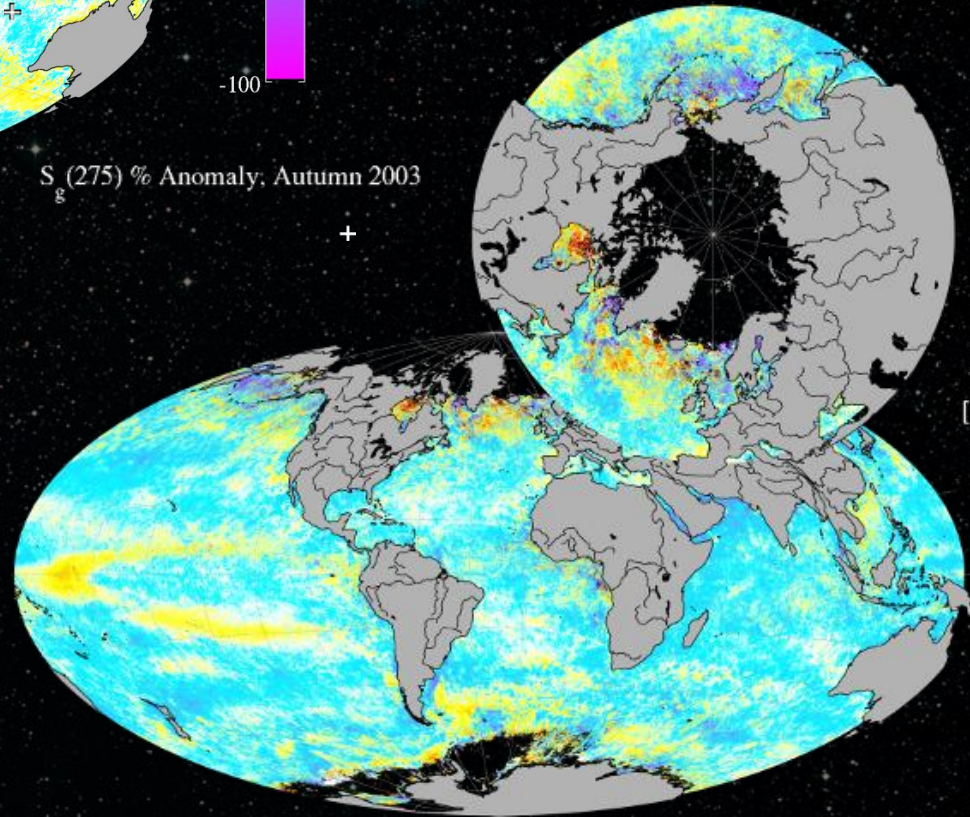
M. Lorenzi-Kayser

Institute for Computational Earth System Science, University of California, Santa Barbara, Santa Barbara, California, USA

\dot{a}_g (275) % Anomaly, Autumn 2003

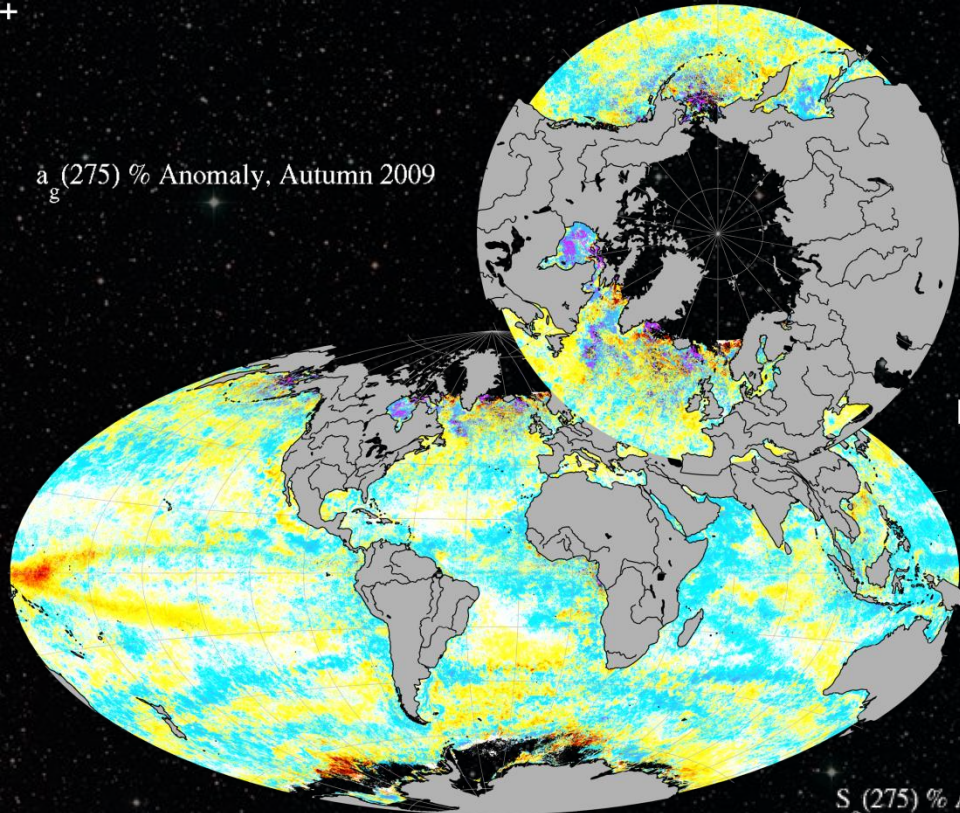


S_g (275) % Anomaly, Autumn 2003



ANOMALY =
YEARBLOCK - AQUA-ERA-AVERAGE

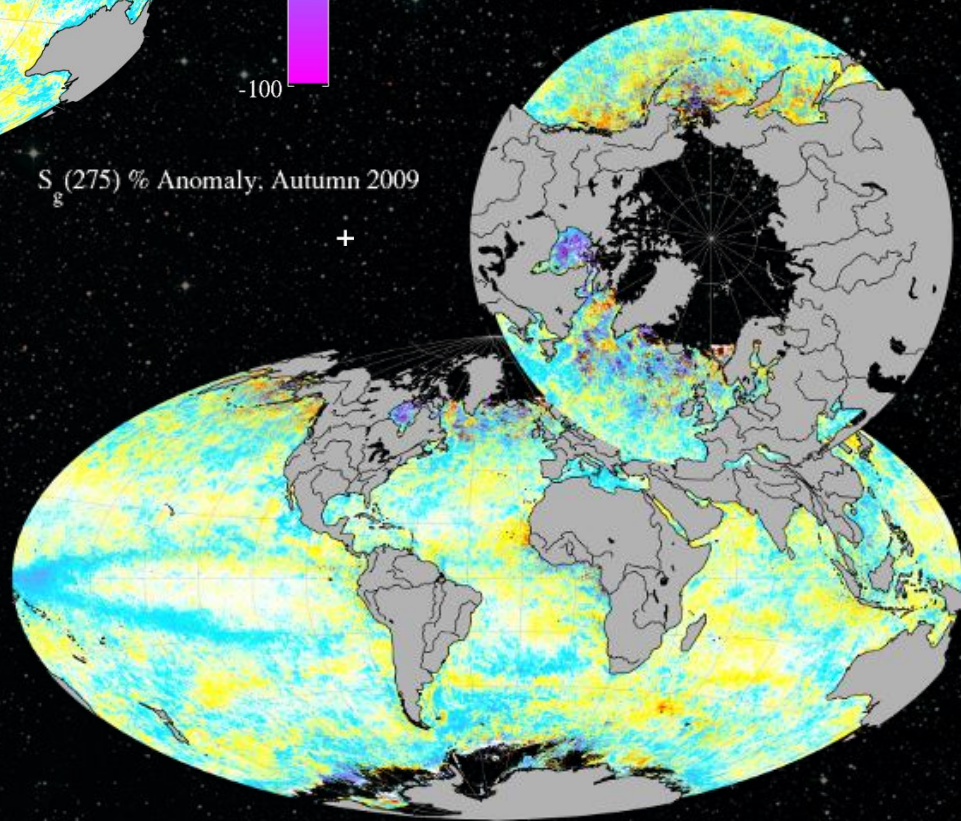
\dot{a}_g (275) % Anomaly, Autumn 2009



[%]



S_g (275) % Anomaly, Autumn 2009

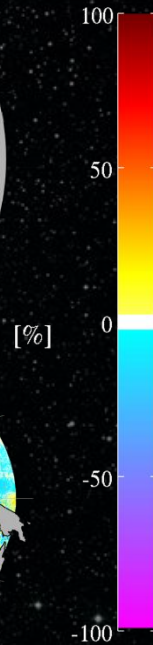
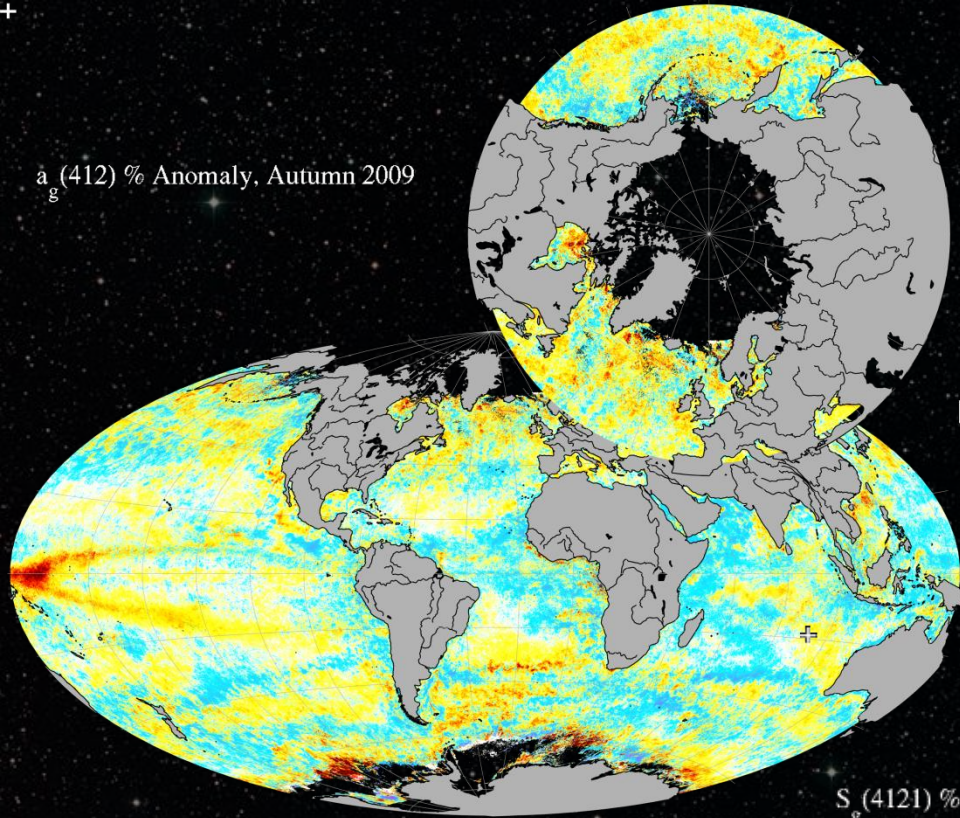


[%]

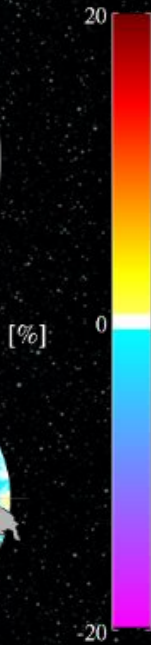
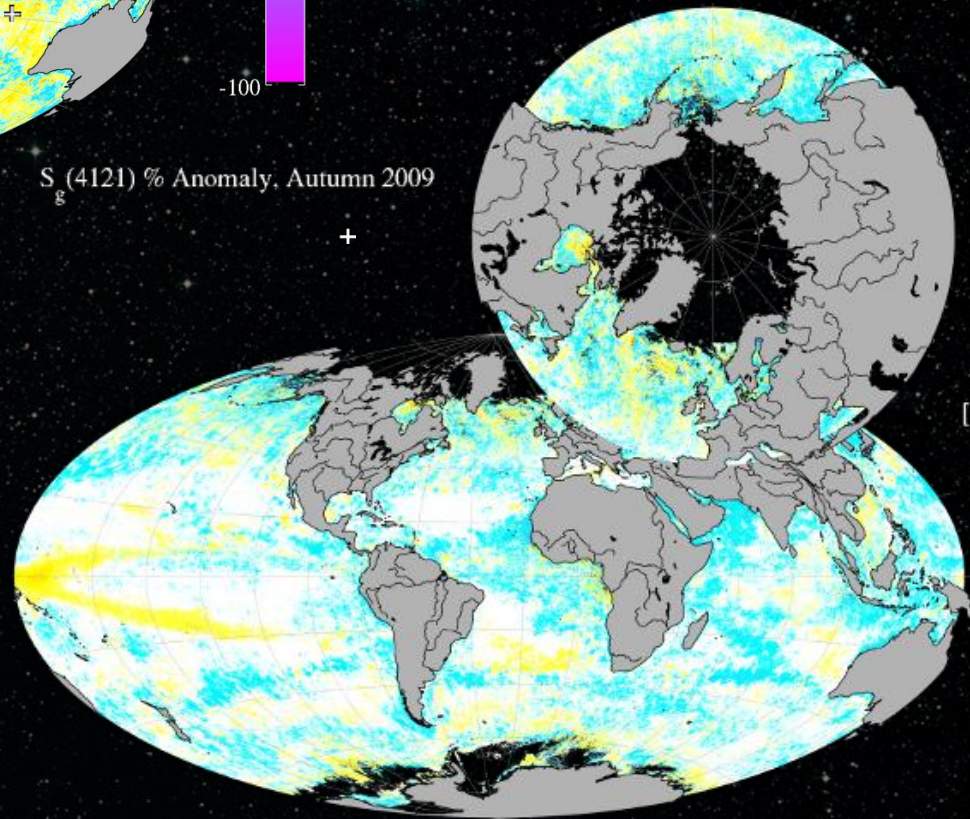


ANOMALY =
YEARBLOCK - AQUA-ERA-AVERAGE

\dot{a}_g (412) % Anomaly. Autumn 2009



S_g (4121) % Anomaly. Autumn 2009



ANOMALY =
YEARBLOCK - AQUA-ERA-AVERAGE

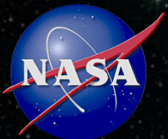
Summary

Remote sensing of CDOM, spectral slope, and DOC are important to understanding marine carbon budgets and the underwater light field

GOCAD is a global, carbon-centric, algorithm development database including hyperspectral data extending into the UV with ~50k field stations

Several algorithms performed reasonably well retrieving $a_g(\lambda)$, $S_g(\lambda)$, and DOC, but MLR proved the most versatile and easy to implement on satellite imagery

- Combining ocean color imagery with SSS from Aquarius vastly improved retrievals of DOC

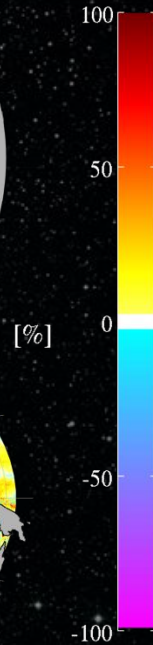
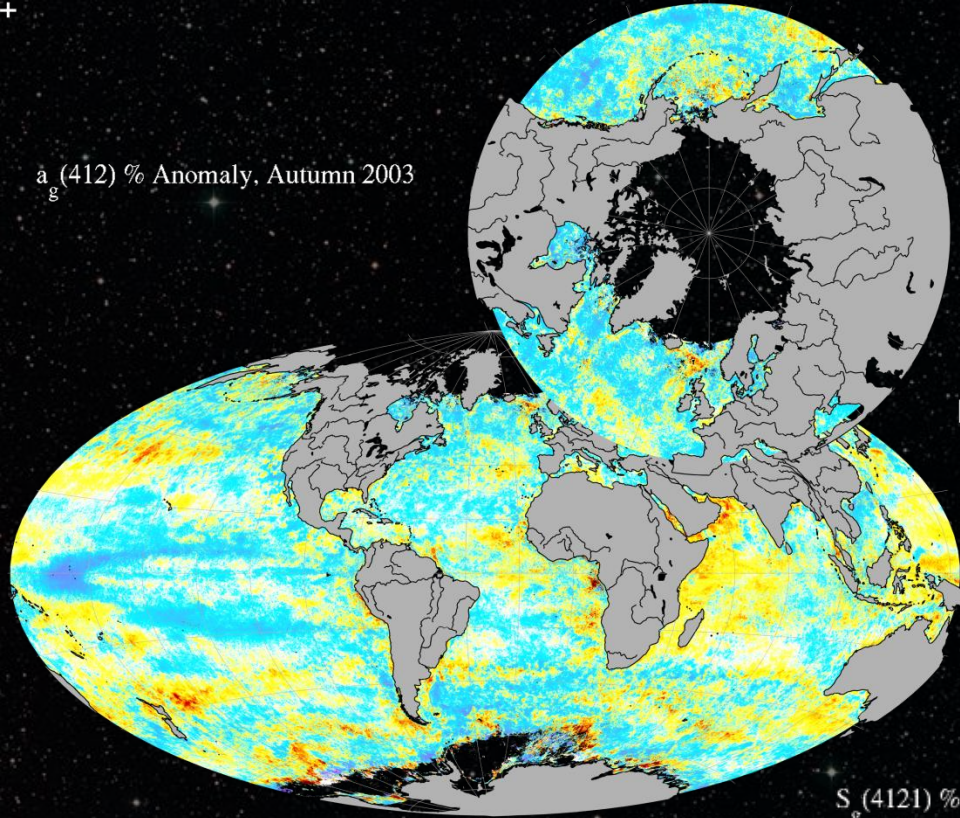


Aurin, Dirk A., Mannino, A., Larry, D., *Remote sensing of CDOM and dissolved organic carbon in the global ocean*, in prep.

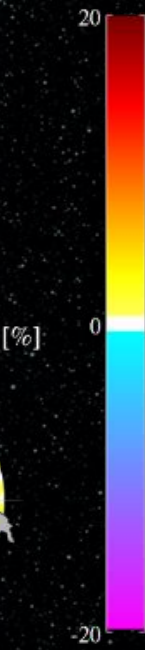
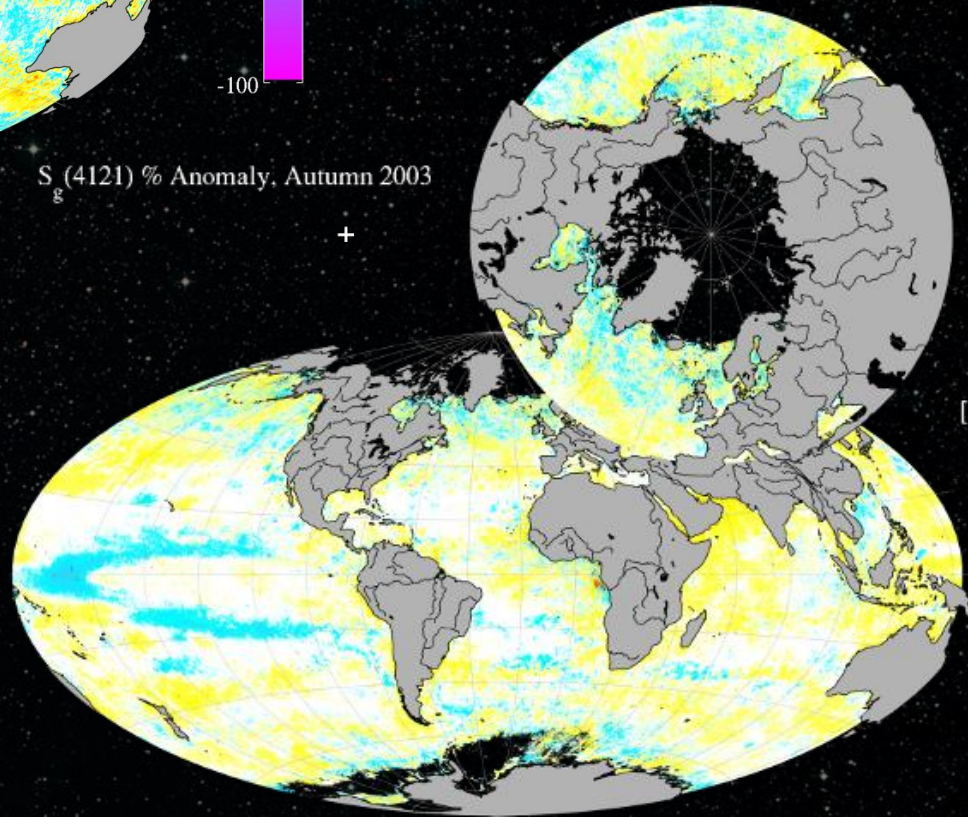
A slide projector is mounted on a boat's deck, silhouetted against a vibrant sunset sky. The sun is low on the horizon, casting a golden glow across the water and sky, with scattered clouds catching the light. The text "EXTRA SLIDES" is overlaid in a blue, 3D-style font with a reflection effect on the water's surface.

EXTRA SLIDES

\dot{a}_g (412) % Anomaly. Autumn 2003



S_g (4121) % Anomaly. Autumn 2003



ANOMALY =
YEARBLOCK - AQUA-ERA-AVERAGE

CDOM

- Strongly absorbs in the blue and UV
 - Limits light for photosynthesis
 - Shades cells from UV damage
 - Leads to surface heating/stratification
- Together with chlorophyll, CDOM dominates the blue-green ratio of sea-surface reflectance
 - Increases uncertainty in band-ratio algorithms for Chl
 - S_g important to semi-analytical retrievals, particularly those that separate CDOM from non-algal particulate absorption

CDOM Remote Sensing

2001 Kahru et. alia (regional)

2002 D'Sa et al. (regional)

2003 Johannsen et al. (regional)

2008 Mannino et al. (regional)

2009 Morel & Gentili (CDOM *index*)

2011 Zhu et al. (regional)

2011 Tiwari et al. (NOMAD)

2011 Shanmugan et al. (NOMAD)

2011 Matthews (review)

2012 Odermatt (review)

2012 Brando et al. (regional)

2012 Tilstone et al. (regional)

2013 Doug et al. (regional, NOMAD)

2013 Tahrani et al. (regional)

2013 Swan et al. (CLIVAR, n=127)

2013 Matsuoka et al. (regional)

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