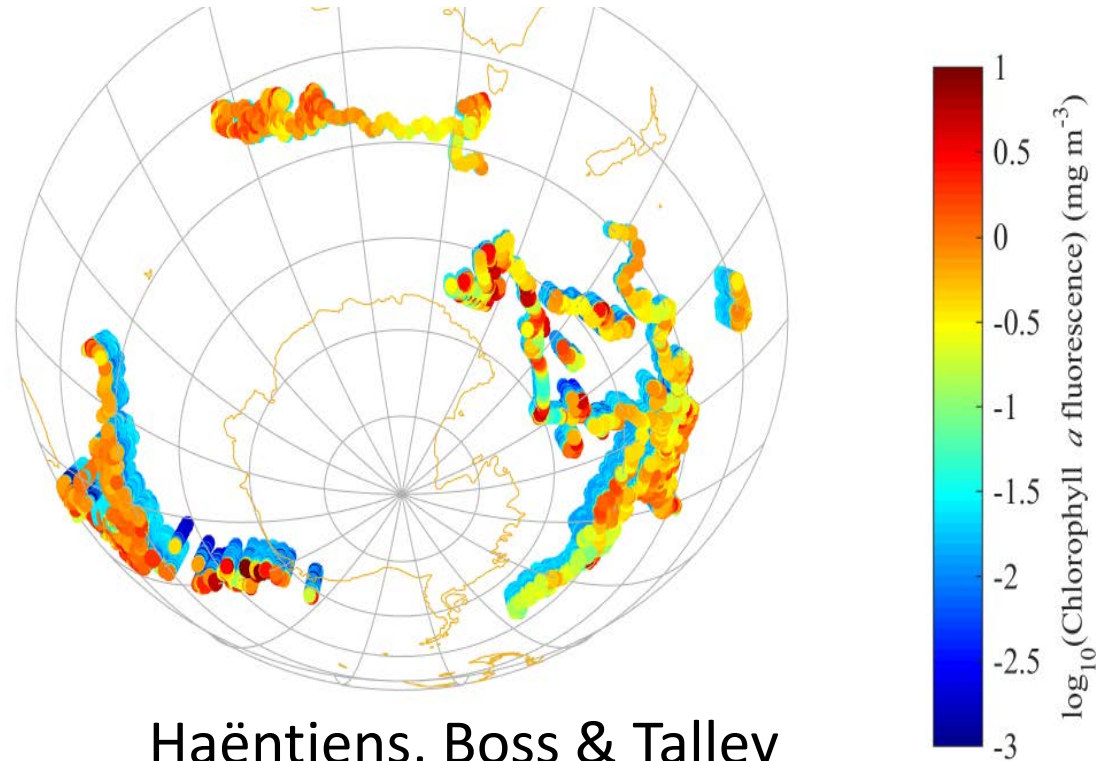


Revisiting Ocean Color Algorithms for Chlorophyll *a* and Particulate Organic Carbon in the Southern Ocean using Biogeochemical Floats



Haëntjens, Boss & Talley

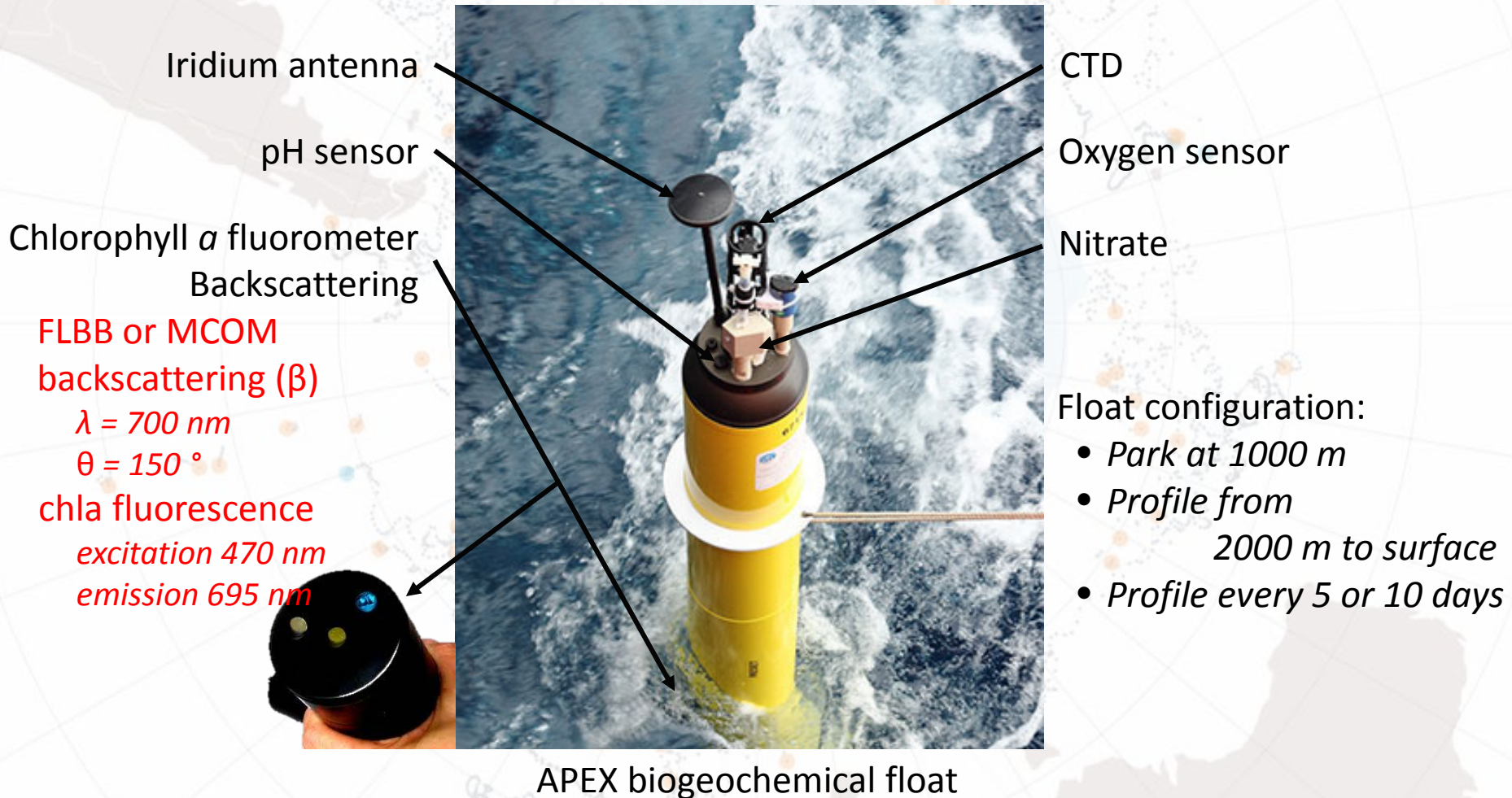


SOCCOM



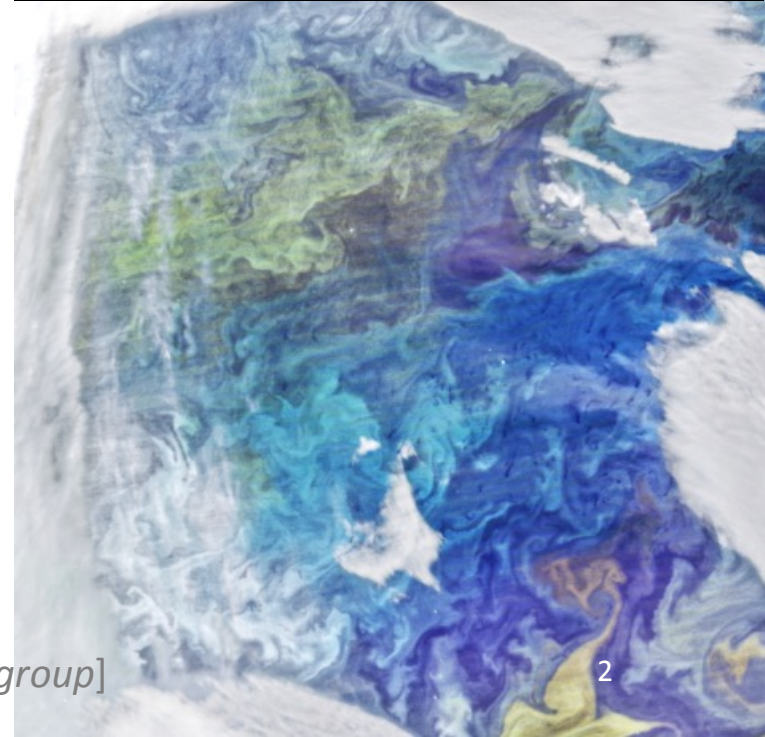
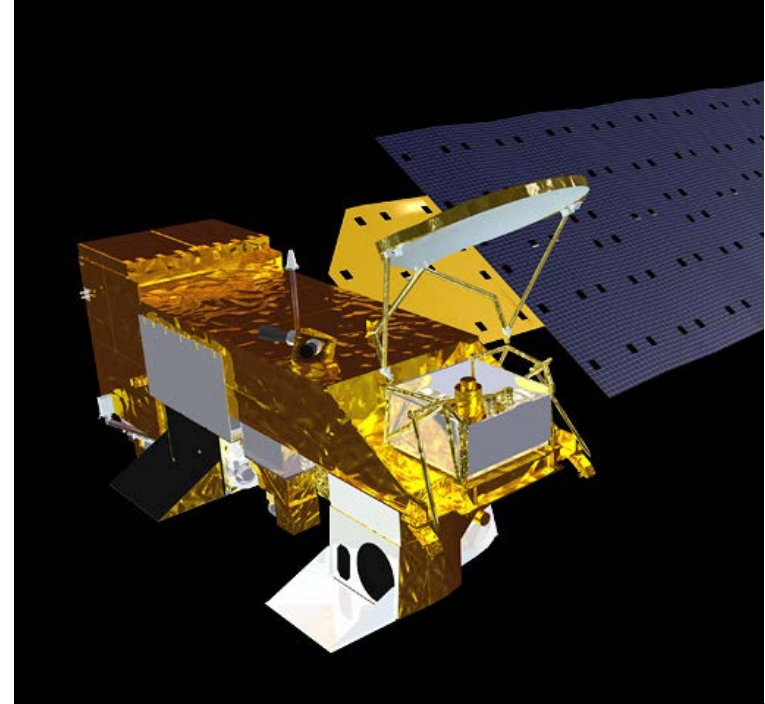
SOCCOM Profiling Floats

Active floats
80 / 200



Ocean Color

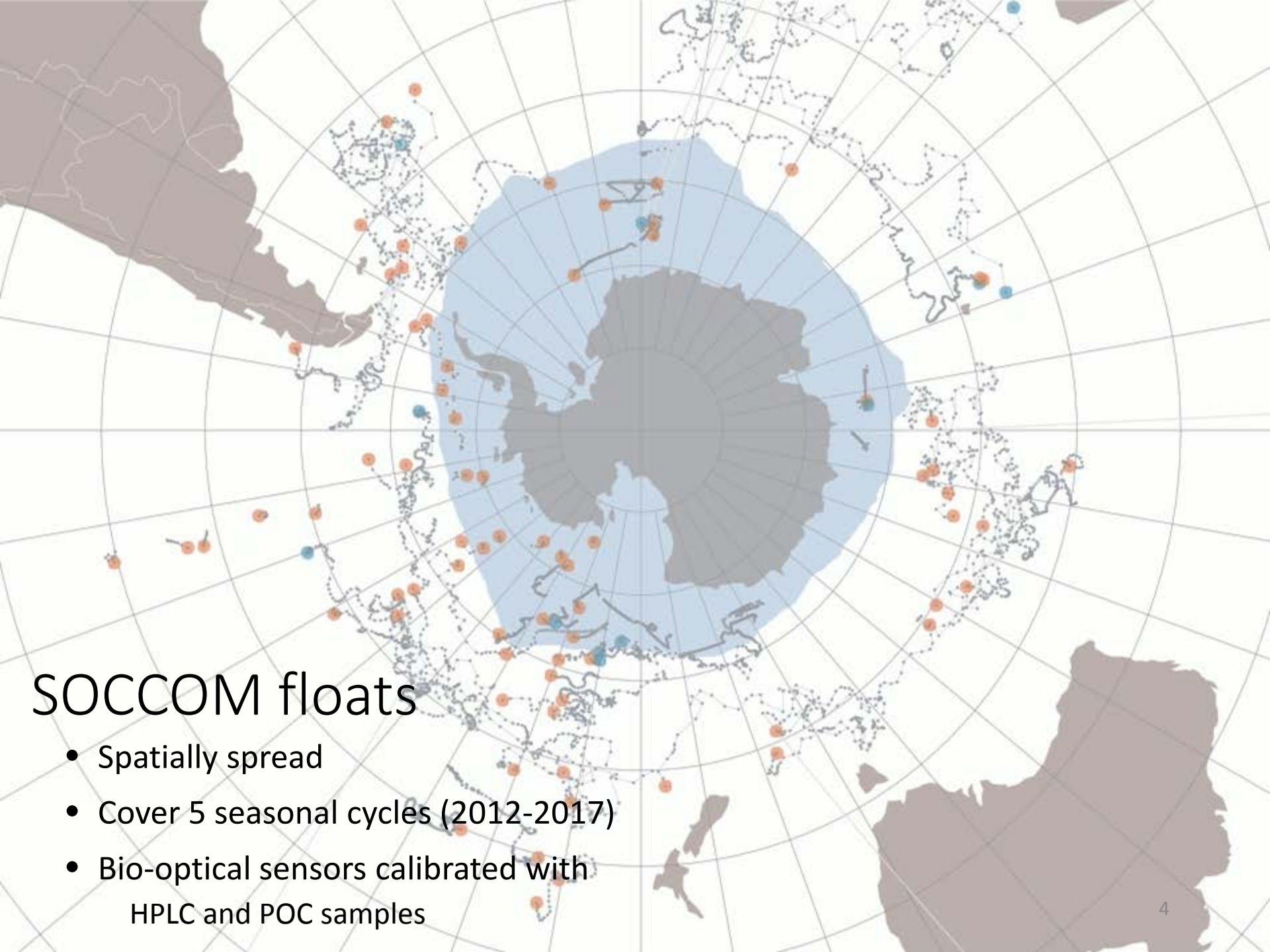
- VIIRS
Visible Infrared Imaging Radiometer Suite
- MODIS *Aqua*
Moderate Resolution Imaging Spectroradiometer
- Both satellites are:
 - polar orbiting
 - multispectral (visible and infrared detectors)
 - measure top of the atmosphere radiance
 - apply atmospheric correction
 - derive chl_a and POC from band ratios



Previous studies

Globally derived chl_a algorithm OCX validation in the Southern Ocean.

Bias	Dataset Type	Satellite(s)	Reference(s)
2.4	Fluorometrically extracted chl _a	CZCS	Mitchell and Holm-Hansen [1991] Sullivan et al. [1993]
~2	Fluorometrically extracted chl _a	SeaWiFS	Dierssen and Smith [2000]
~2	HPLC	ADEOS SeaWiFS MODIS	Mitchell and Kahru [2009] Kahru and Mitchell [2010]
~2	HPLC	MODIS	Guinet et al. [2013]
3-4	HPLC	MODIS VIIRS	Johnson et al. [2013]
1	HPLC	SeaWiFS	Marrari et al. [2006]



SOCCOM floats

- Spatially spread
- Cover 5 seasonal cycles (2012-2017)
- Bio-optical sensors calibrated with HPLC and POC samples

Calibrate chlorophyll *a* fluorometers

Fluorescence is a proxy for chlorophyll *a*

chl_a:fluorescence yield ratio is affected by: the light history, the species composition, and the physiology of the phytoplankton.



- calibrate all sensors against a "golden" sensor

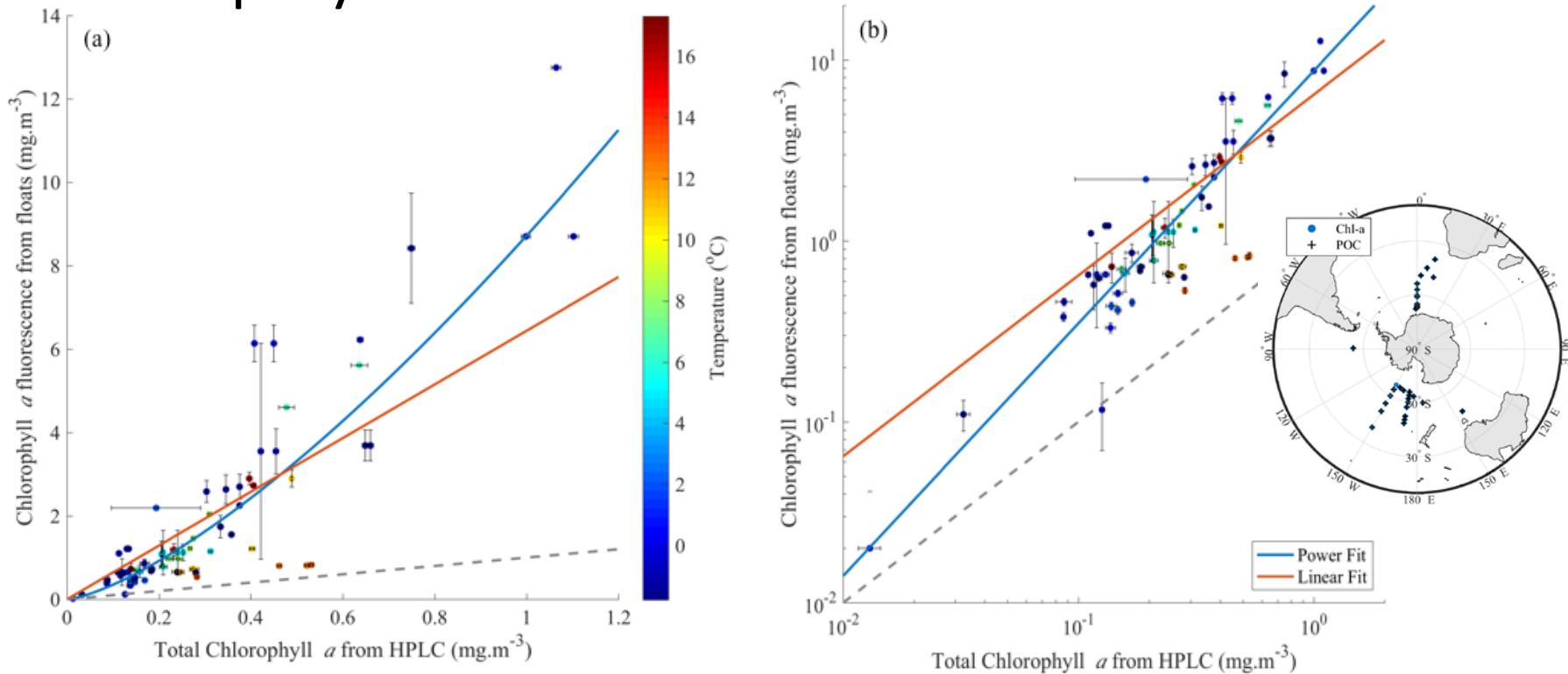
- correct for interference with fluorescent dissolved organic matter
- `find(min([chla]))` at depth and subtract it from all profiles
- similar correction as in *Xing et al.* [2016]

- if `sun_elevation > 5°`
- `mean(chlaXing2012, chlaSackmann2008)`

FYI: No significant difference day-only or night-only data.

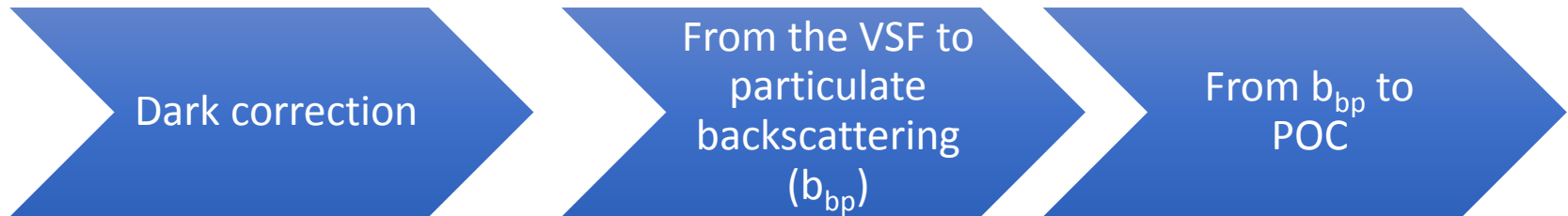
- compare first profile with [chl_a] from HPLC taken within 24 hours of deployment (Roesler et al., 2017)

Relationship between total chlorophyll *a* from HPLC and chlorophyll *a* fluorescence from floats



Relationship	<i>n</i>	<i>r</i> ²	<i>RMSD</i>	<i>RMSRD</i>
$chl_{a_{HPLC}} = 0.15(\pm 0.017) \times chl_{a_{float}}$	73	0.77	0.20	0.48
$chl_{a_{HPLC}} = 0.213(\pm 0.016) \times chl_{a_{float}}^{0.714(\pm 0.242)}$	73	0.80	0.12	0.37

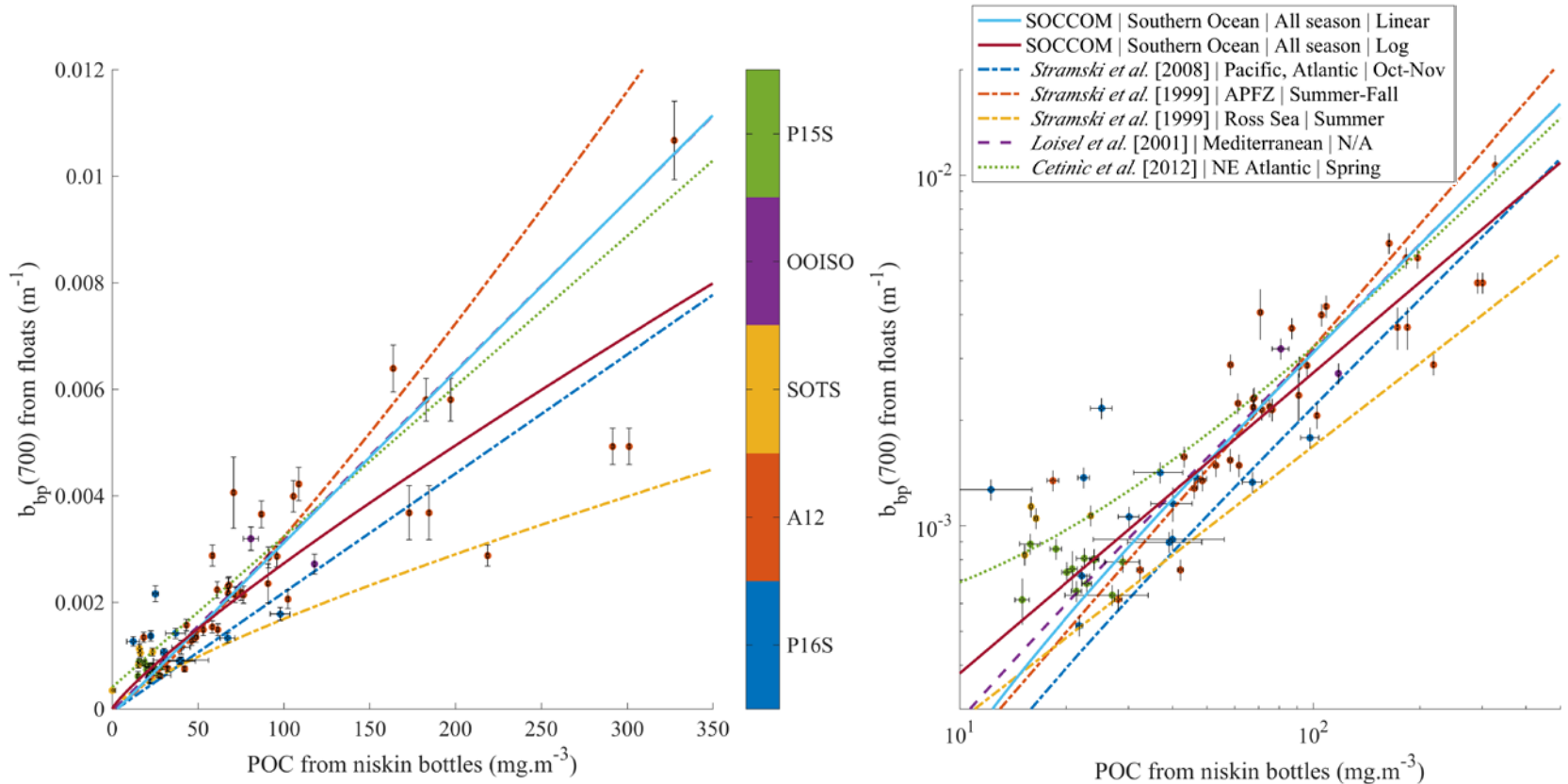
From backscattering to POC



- if available, use pre-deployment dark
- The VSF of seawater (β_{sw}) is estimated with *Zhang et al.*, [2009]
- The particulate conversion factor (χ_p) is estimated with *Sullivan et al.*, [2013]
- Build a relationship specific to our dataset

$$b_{bp} = 2 \times \pi \times \chi_p(\theta) \times (\beta(\theta) - \beta_{sw}(\theta))$$

Relationship between POC and particulate backscattering at 700 nm



Relationship	n	r^2	$RMSD$	$RMSRD$
$POC = 3.12 \times 10^4 (\pm 2.47 \times 10^3) \times b_{bp}(700) + 3.04 (\pm 6.78)$	67	0.76	35	0.47
$POC = 9.776 \times 10^4 (\pm 1.90 \times 10^4) \times b_{bp}(700)^{1.166 (\pm 0.173)}$	67	0.88	40	0.59

Deriving chl_a and POC from R_{rs}

Download images from NASA Ocean Color Level 2, reprocessing R2014.0

Product	Sensor	Algorithm	Equations
chl _a	MODIS	OCI ^a	Merged OC3M band ratio algorithm with color index (CI) of <i>Hu et al.</i> [2012]
chl _a	MODIS	J13 ^b	$R_{sw} = \log_{10} \left(\frac{\max(R_{rs}(443), R_{rs}(488))}{R_{rs}(547)} \right)$ $chl_{aJ13} = 10^{0.6994 - 2.0384R_{sw} - 0.4656R_{sw}^2 + 0.4337R_{sw}^3}$
chl _a	MODIS	SPGANTv4 ^c	$L_{wn}(\lambda) = \frac{F_0(\lambda) \times R_{rs}(\lambda)}{\pi}$ $R = \log_{10} \left(\frac{\max(L_{wn}(443), L_{wn}(488))}{L_{wn}(547)} \right)$ $chl_{aSPGANTv4} = 10^{0.5514 - 2.2434R + 0.0746R^2 - 0.0095R^3 - 0.7790R^4}$ blended with OC3M [<i>Kahru and Mitchell</i> , 2010]
chl _a	VIIRS	OCI ^a	Merged OC3V band ratio algorithm with color index (CI) of <i>Hu et al.</i> [2012]
chl _a	VIIRS	J13 ^b	$R_{sw} = \log_{10} \left(\frac{\max(R_{rs}(410), R_{rs}(443), R_{rs}(486))}{R_{rs}(551)} \right)$ $chl_{aJ13} = 10^{0.6736 - 2.0714R_{sw} - 0.4939R_{sw}^2 + 0.4756R_{sw}^3}$
POC	MODIS	S08 ^d	$POC = 203 \times \left(\frac{R_{rs}(443)}{R_{rs}(547)} \right)^{-1.034}$
POC	VIIRS	S08 ^d	$POC = 203 \times \left(\frac{R_{rs}(443)}{R_{rs}(551)} \right)^{-1.034}$

^a*Hu et al.* [2012]

^b*Johnson et al.* [2013]

^c*Kahru and Mitchell* [2010]; *Mitchell and Kahru* [2009]

^d*Stramski et al.* [2008]

Matching up

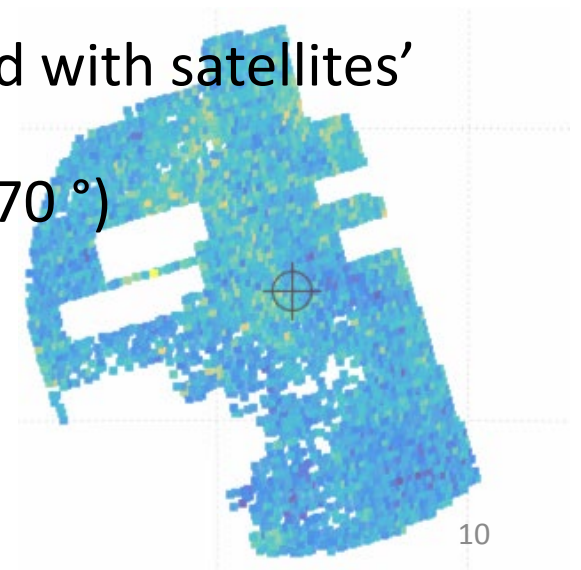
Maximize the quality of the samples:

- narrow time window (+/- 3 hours)
- mean 5x5 pixel box centered on float profile
- good atmospheric correction (mask level 2 flags)
- Weighted vertical integration.

➤ Very few matchups with MODIS

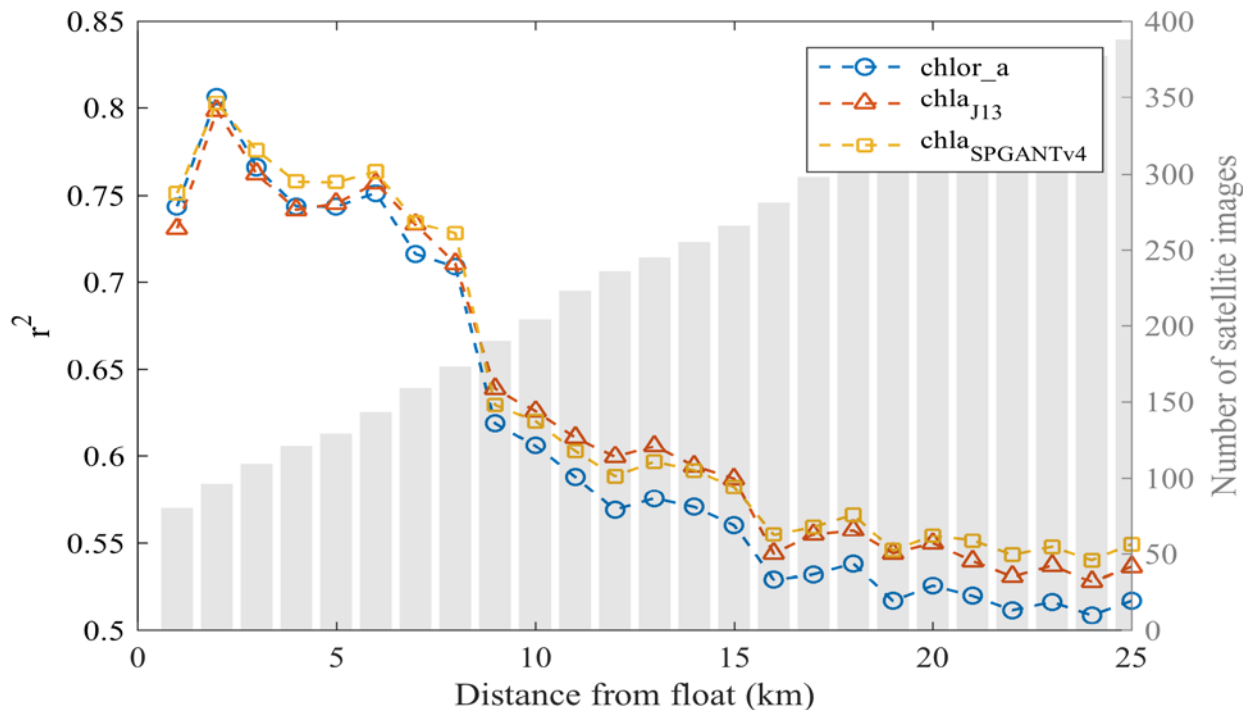
Explained by:

- floats' surface time is not synchronized with satellites' overpasses
- polar night, high solar zenith angle ($> 70^\circ$)
- high cloud coverage all year long

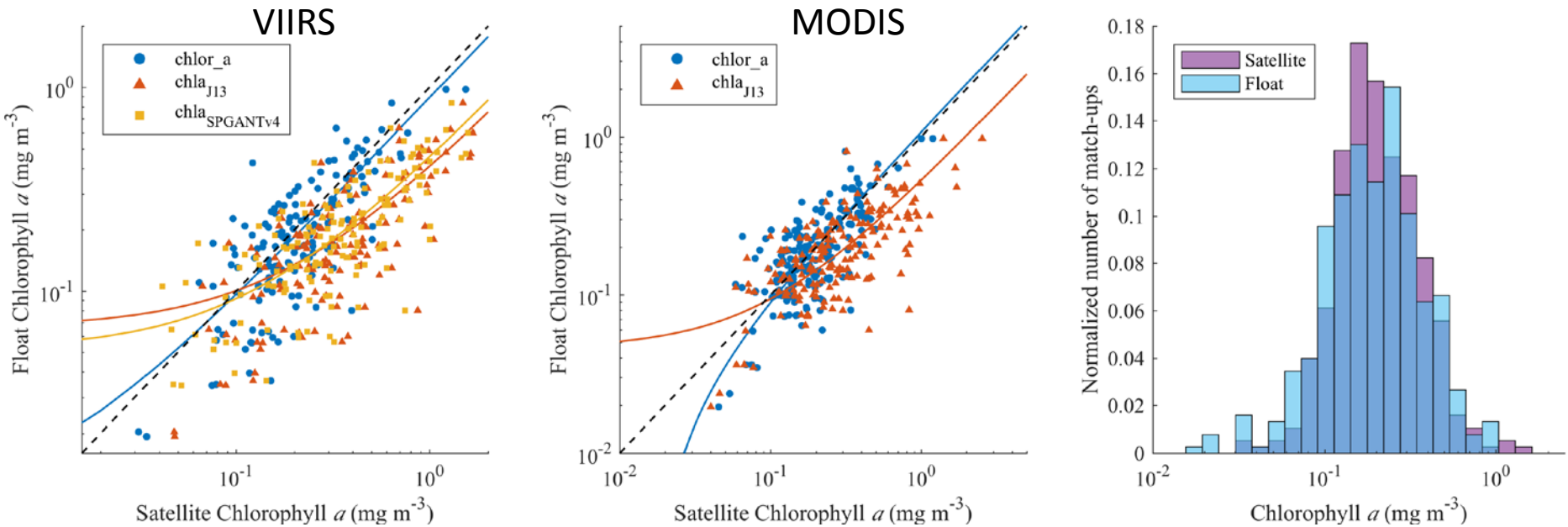


Maximize Number of Matchups

- Widening spatial (8 km) and temporal (24 hr) window
 - increase the number of matchups
 - at the possible cost of quality

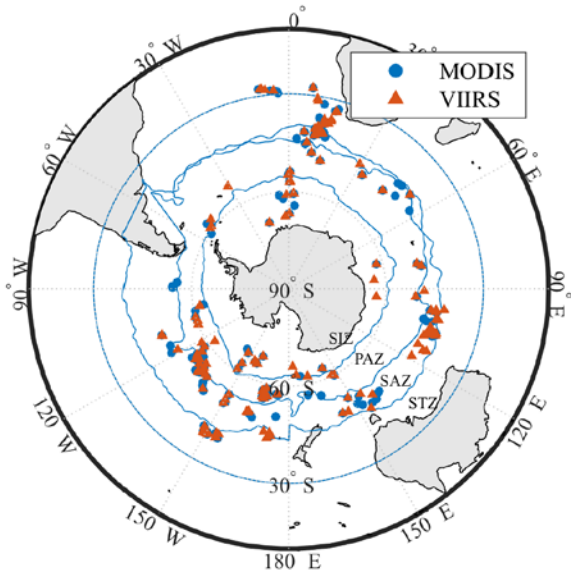


Comparison with Remote Sensing

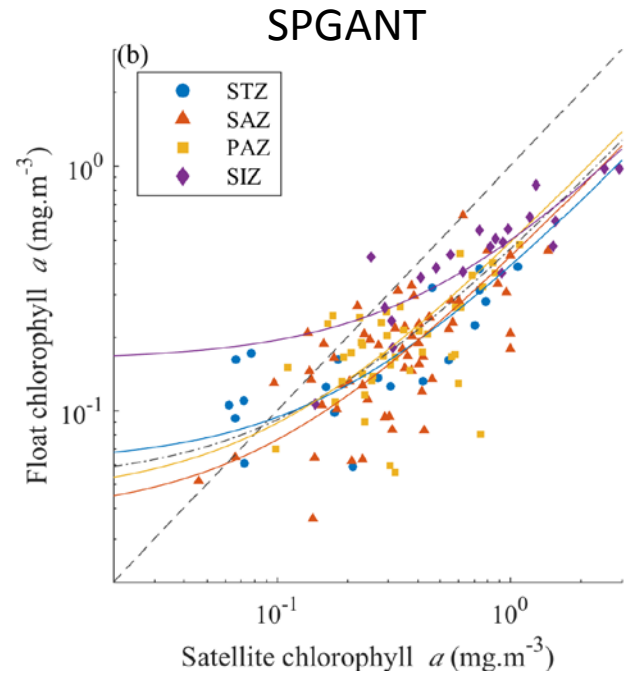
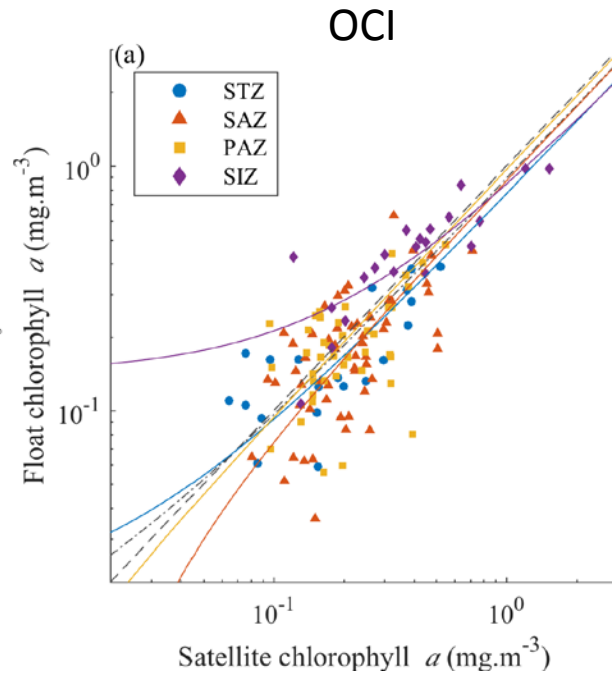


Sensor	Algorithm	Slope	Offset	n	r^2	$RMSD$	$RMSURD$
		Linear	Linear			Linear	Linear
MODIS	OCI	0.88(±0.04)	0.01(±0.01)	173	0.67	0.11	0.45
MODIS	J13	0.35(±0.02)	0.07(±0.01)	173	0.66	0.43	0.74
MODIS	SPGANTv4	0.41(±0.02)	0.05(±0.01)	173	0.70	0.35	0.68
VIIRS	OCI	1.09(±0.05)	-0.02(±0.01)	203	0.61	0.10	0.44
VIIRS	J13	0.49(±0.03)	0.05(±0.01)	203	0.49	0.28	0.62
both	OCI	0.97(±0.03)	-0.00(±0.01)	376	0.63	0.10	0.45

Regional variability:



SIZ: Sea Ice Zone
PAZ: Polar Antarctic Zone
SAZ: Subantarctic Zone
STZ: Subtropical Zone

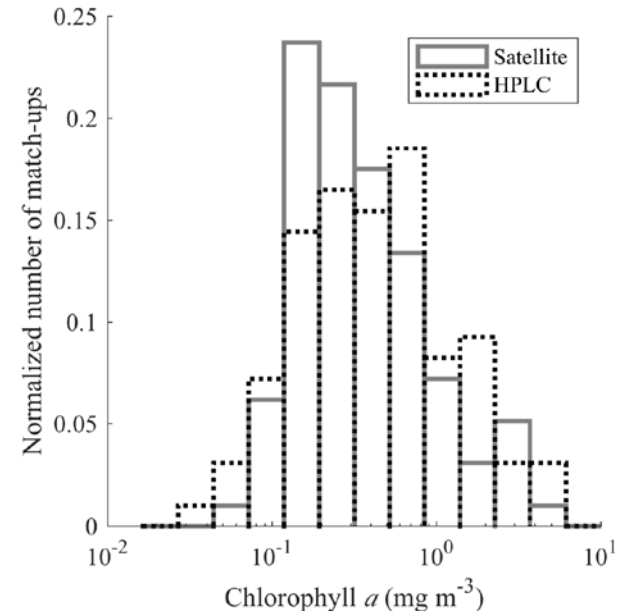
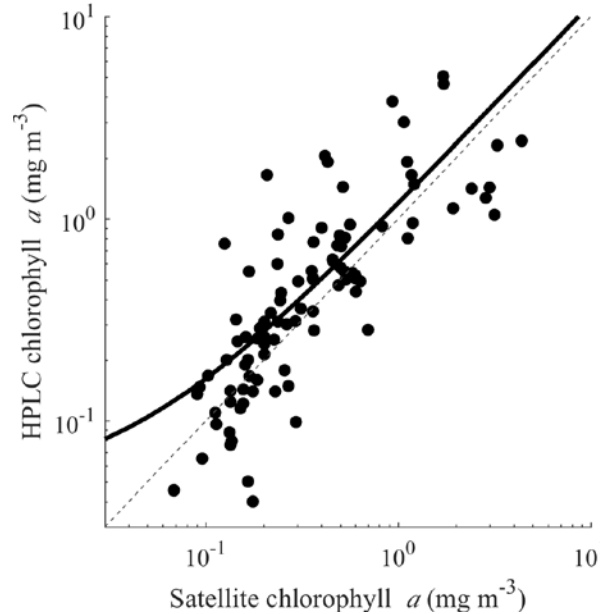
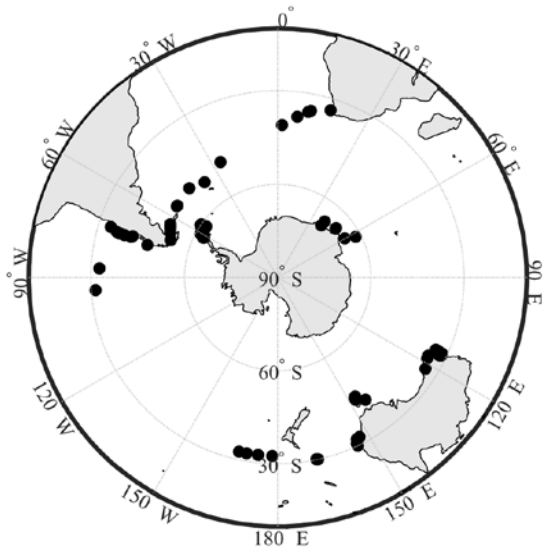


Maximal spatial variability off slope with regions $\sim 20\%$.

Temporal variability: Not significant.

Independent dataset

*all the data available on February 1, 2017, south of 30 °S, on SeaBASS
6242 HPLC samples from 1682 profiles between Oct 1995 and April 2011*



Out of the 659 matchups only 97 respected the criteria defined earlier
MODIS OCI matchups

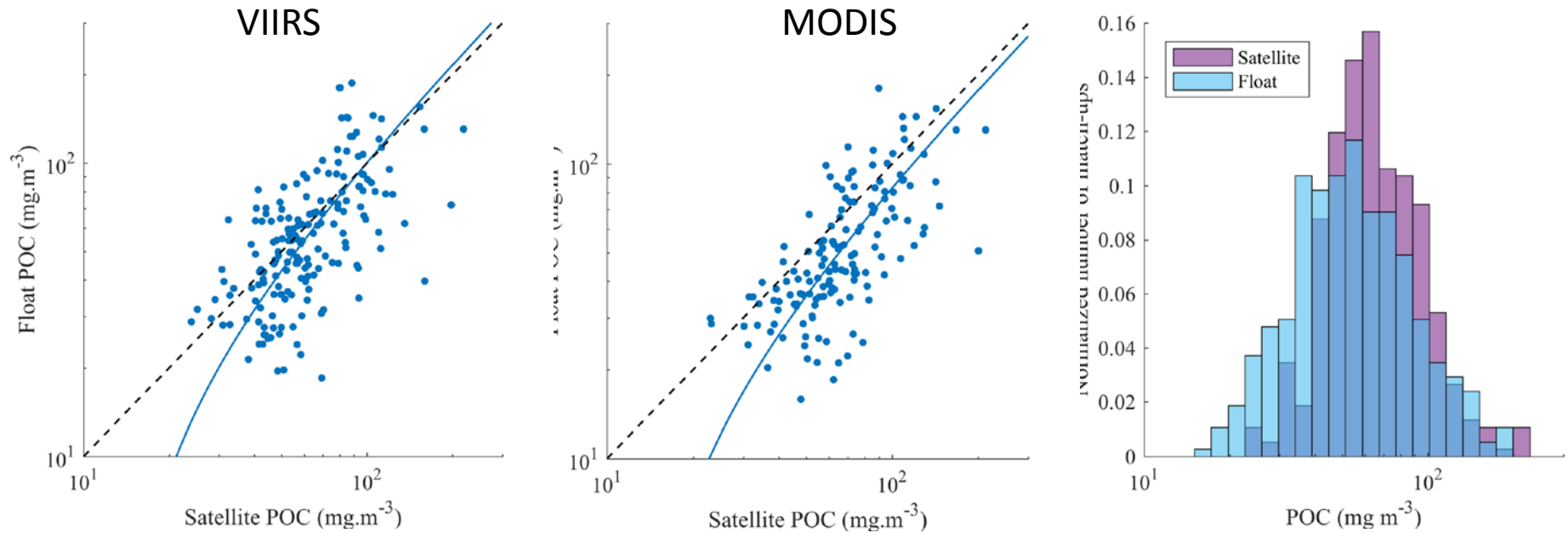
$$\text{chl}a_{\text{HPLC}} = 1.15(\pm 0.11) \times \text{chl}a_{\text{sat}} + 0.05(\pm 0.11)$$

$$\text{RMSD} = 0.78 \text{ mg m}^{-3}$$

$$\text{RMSURD} = 0.59$$

$$N = 97$$

Comparison with Remote Sensing



Sensor	Algorithm	Slope	Offset	n	r^2	$RMSD$	$RMSURD$
		Linear	Linear				
MODIS	POC	0.94(± 0.06)	-11.47(± 5.08)	173	0.37	31.46	0.44
VIIRS	POC	1.15(± 0.07)	-14.30(± 5.41)	203	0.34	28.60	0.38
both	POC	1.05(± 0.05)	-12.85(± 3.79)	376	0.33	29.95	0.41

Conclusion

- OCI performs well in the Southern Ocean
 - bias in our dataset of 9 % for VIIRS and 12 % for MODIS
- POC agrees well with float products
 - relatively low prediction capability

In addition, autonomous floats can be used as a third dimension (depth) to complement remote sensing in the Southern Ocean (4D)

Thank you !

Questions ?

Previous studies

- Under-estimation of global chl *a* algorithm:
 - Species composition
 - Physiology
 - Particulate Composition
- Spatially limited:
 - > 95 % in narrow corridor south of Tasmania [*Johnson et al., 2013*]
 - Primarily in the Scotia Sea [*Mitchell and Kahru, 2009*]
- Seasonally limited:
 - Primarily austral summer

Regional Trends with SeaBASS Dataset

