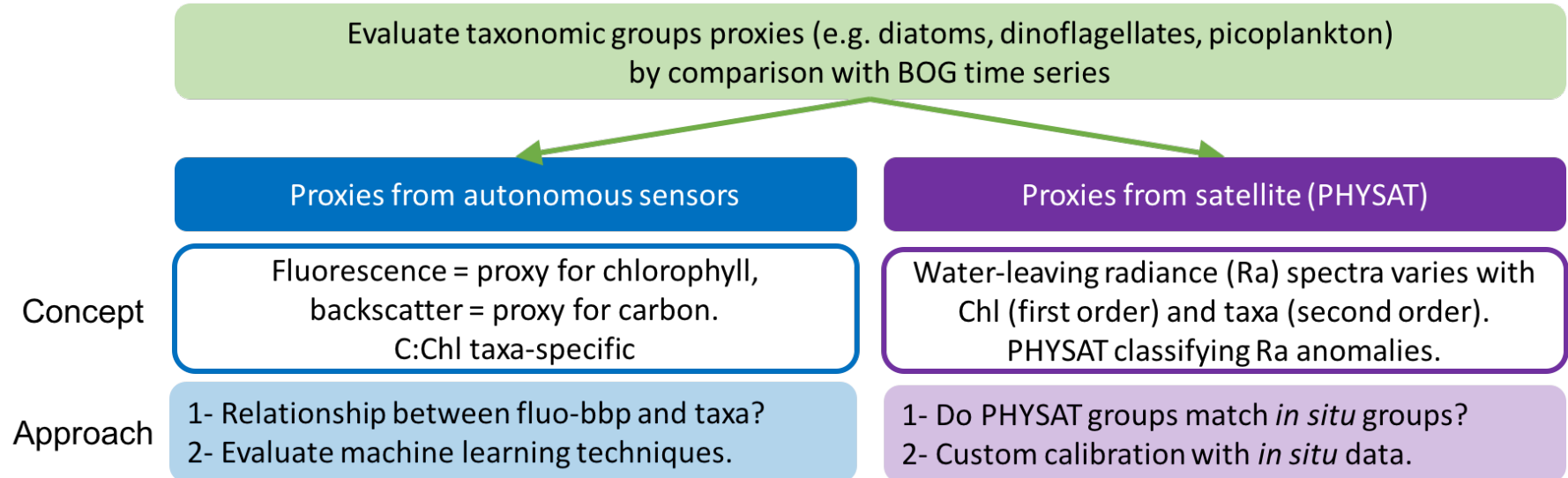


Plankton Proxies project



Proxies from autonomous sensors

Evaluate taxonomic groups proxies (e.g. diatoms, dinoflagellates, picoplankton)
by comparison with BOG time series

Proxies from autonomous sensors

Concept

Fluorescence = proxy for chlorophyll,
backscatter = proxy for carbon.
C:Chl taxa-specific

Approach

- 1- Relationship between fluo-bbp and taxa?
- 2- Evaluate machine learning techniques.

Proxies from autonomous sensors

Goal: investigate the presence/absence and robustness of a relationship between phytoplankton taxa and fluo-bbp, based on:

- (a) BOG phyto counts (carbon biomass per cell)
- (b) concurrent fluo / beam-c datasets from the CTD ($n \sim 1300$)
- (c) concurrent fluo / bbp datasets from the M1 mooring ($n \sim 100$)

Main conclusions:

- about 50% of BOG samples are mixed (no clear dominance by one group)
- when focusing on samples clearly dominated by one group, there is a relationship between taxa and Chl/POC (in situ)
- this relationship does not carry to fluo/beamc mostly because the POC/beamc relationship fails for dinos
- the mooring dataset (fluo/bbp) is more encouraging; however, the number of samples is too small to conclusively build a prediction model.

Proxies from autonomous sensors

Questions:

1) Is C:Chl taxa-specific?

(= relationship between Chl/POC and taxa from in situ samples)

YES

1pre) Are samples taxa-specific? (can samples be attributed to one taxon?)

only ~ 50%

2) Are fluo / beamc / bbp proxies for Chl / C?

(= relationship between fluo/Chl and Cproxy/POC, where Cproxy = beamc or bbp)

YES for fluo,

NOT REALLY for beamc,

NOT SURE for bbp

2pre) Is the mooring dataset reliable? (QC / interdeployment calibration)

QC / calibration difficult → small dataset

3) Can fluo / beamc or fluo / bbp be used to predict taxonomy?

MAYBE

(= relationship between fluo/Cproxy and taxa)

Concept

Proxies from autonomous sensors

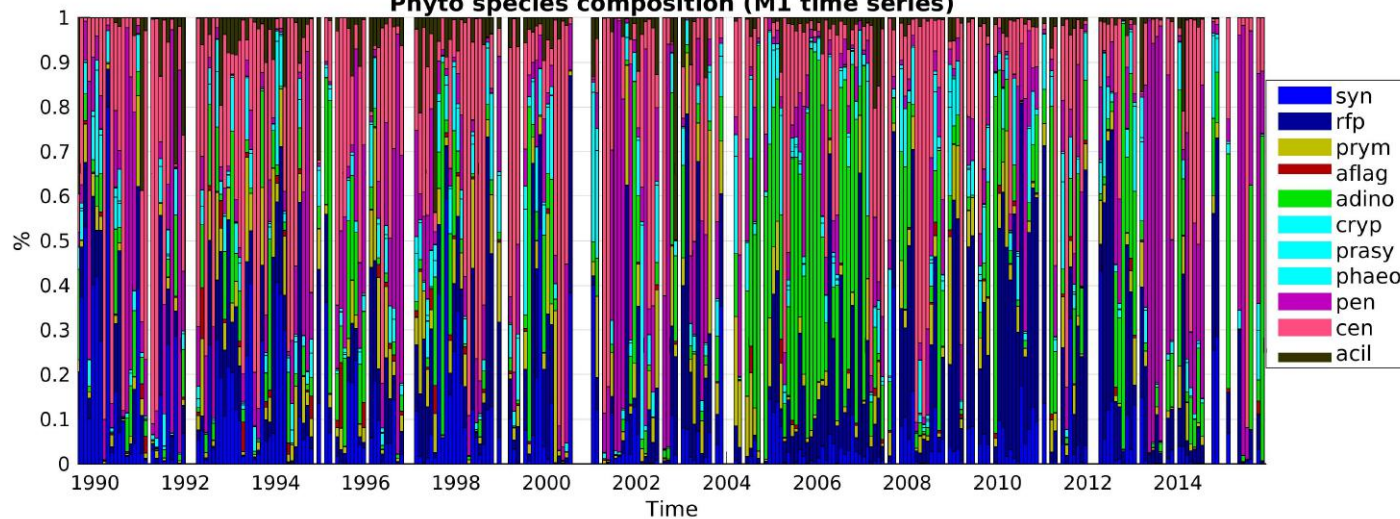
Fluorescence = proxy for chlorophyll,
backscatter = proxy for carbon.
C:Chl taxa-specific

Proxies from autonomous sensors

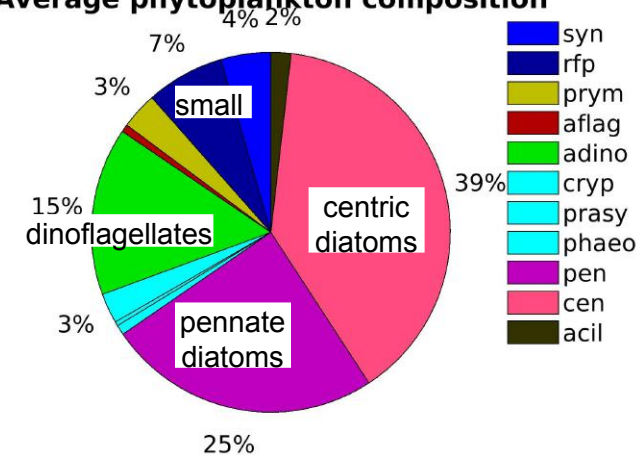
In situ dataset only
(Chl, POC, taxonomy)
1989-2015, N = 435

1pre) Phytoplankton community

Phyto species composition (M1 time series)



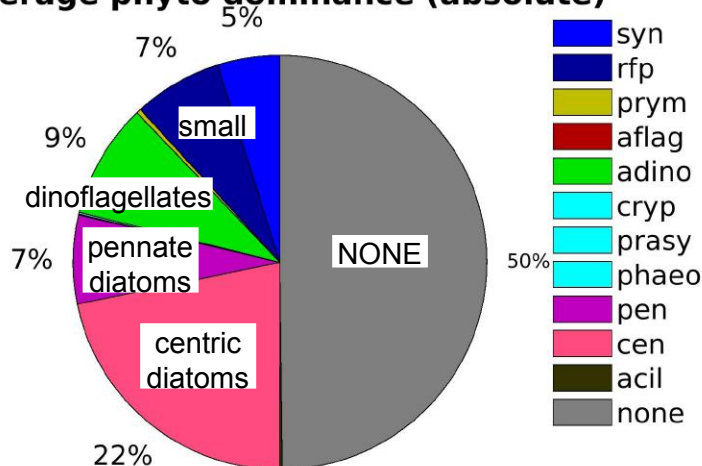
Average phytoplankton composition



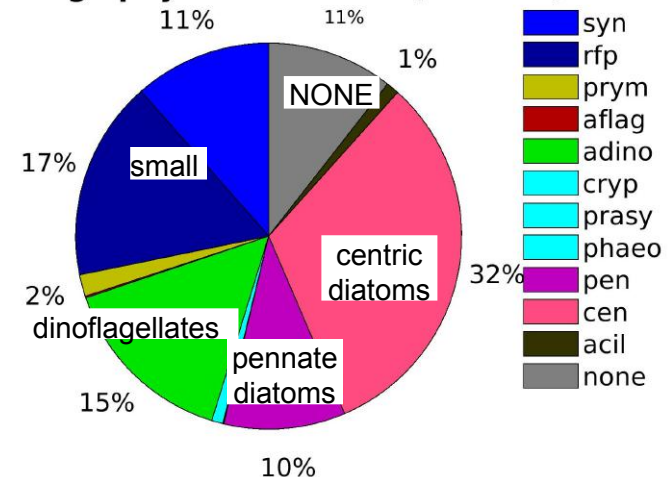
**Most samples are a mix of species,
not a lot of samples dominated by dinos.**

- *absolute dominance*: species $\geq 50\%$ of the biomass
- *relative dominance*: species is the most abundant and represents $> 30\%$ of the biomass

Average phyto dominance (absolute)



Average phyto dominance (relative)



1) Is C:Chl taxa specific?

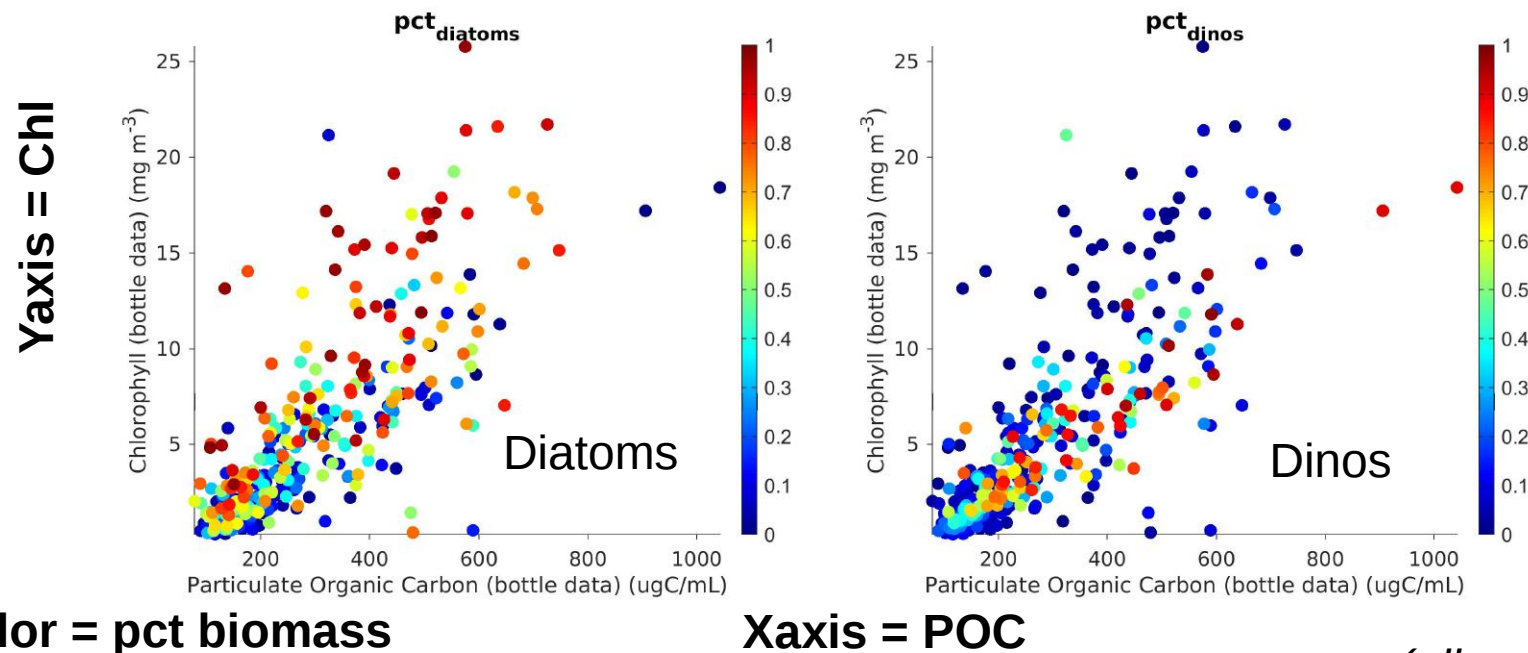
Looking at in situ (bottle) data only.

Chl: measured at the same time as plankton data

C: trickier. 2 measures:

- total biomass = sum(plankton biomass), very uncertain & noisy*.
- particulate organic carbon (**POC**, independent), might contain heterotrophs too.

Expected relationship: C:Chl is very high for picoplankton (low biomass anyways), high for dinos, low for diatoms.



No clear separation when considering all samples.

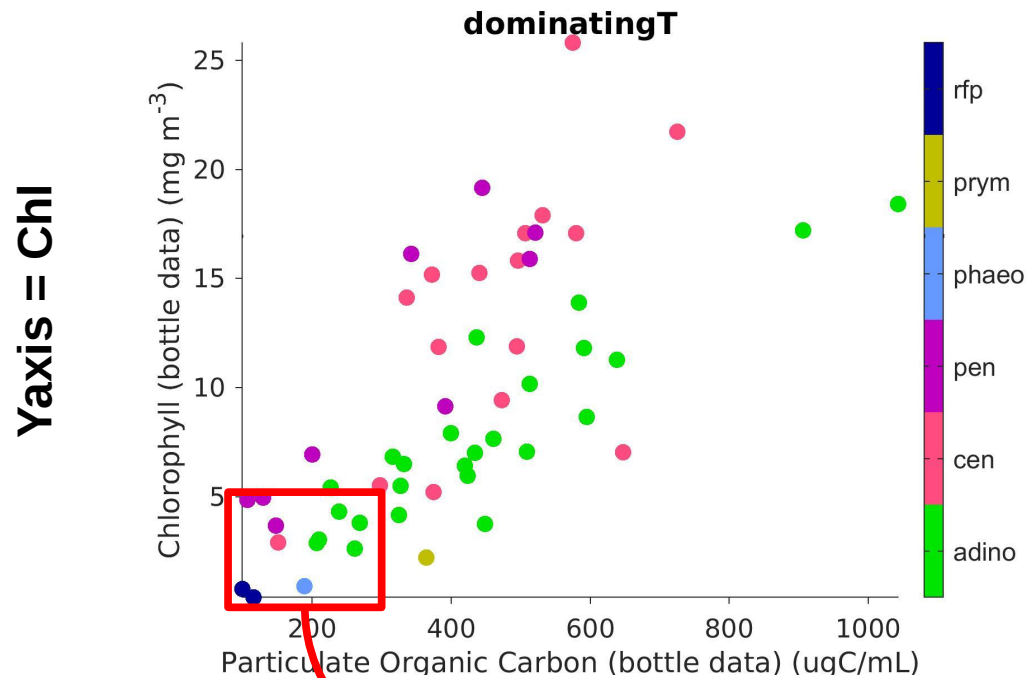
(all available BOG data)

*Both relationships with taxonomy and with beamc are worse for total biomass than for POC.

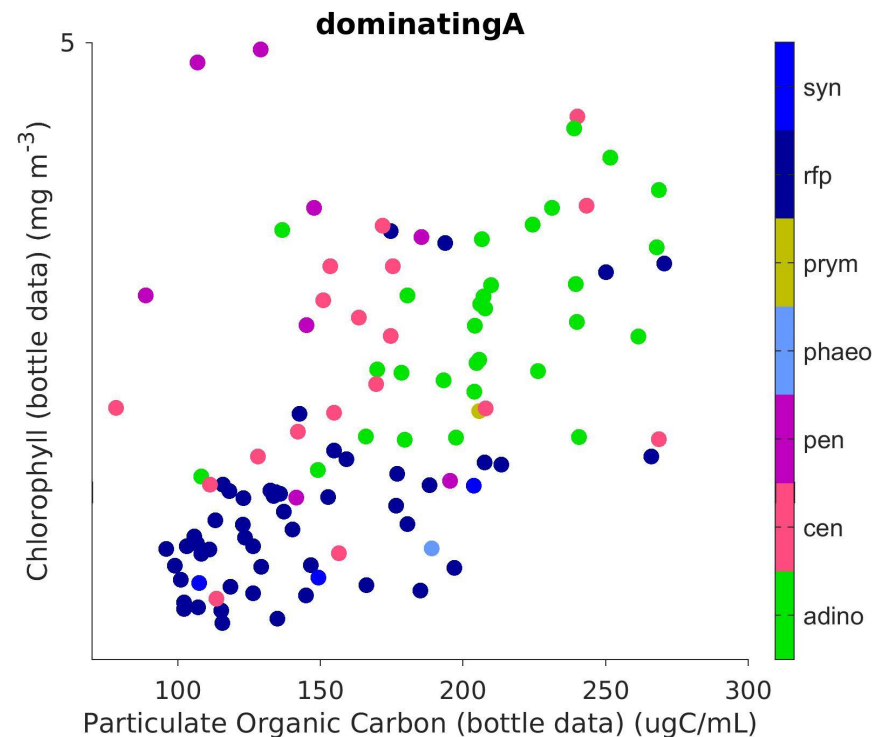
1) Is C:Chl taxa specific?

BUT samples strongly dominated by one taxon are separable in Chl/POC space.

Dominating = **above 80%**



Dominating = **above 50%**



zoom

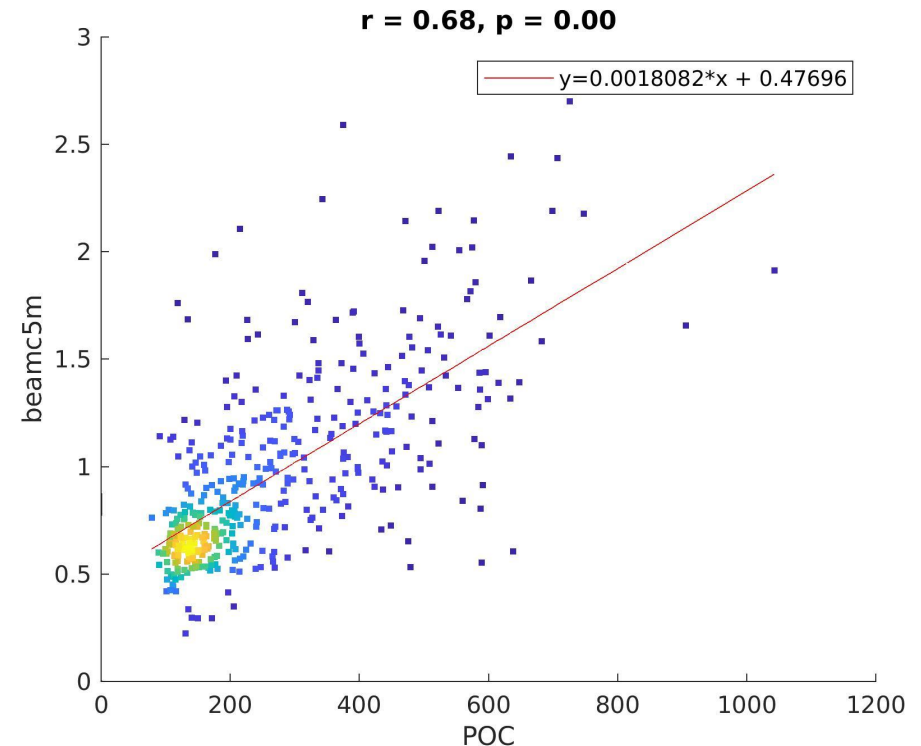
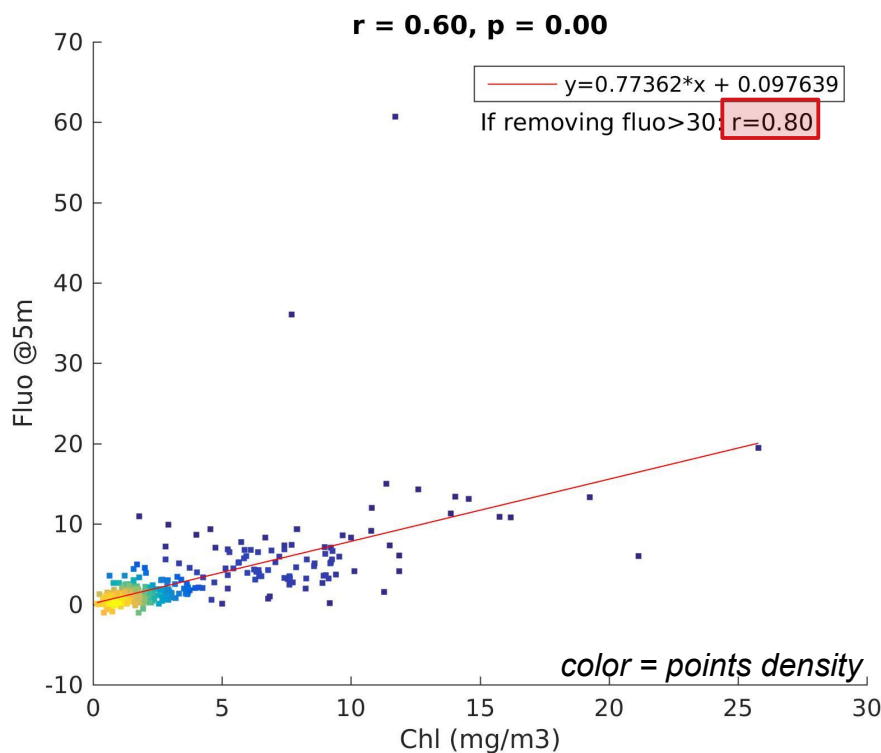
(only samples dominated by one group)

Proxies from autonomous sensors

CTD dataset (fluo, beamc)
vs taxonomy
1989-2015, N = 1228

2) Fluorescence / beam-c as proxies for Chl / C

Autonomous sensors @5 m

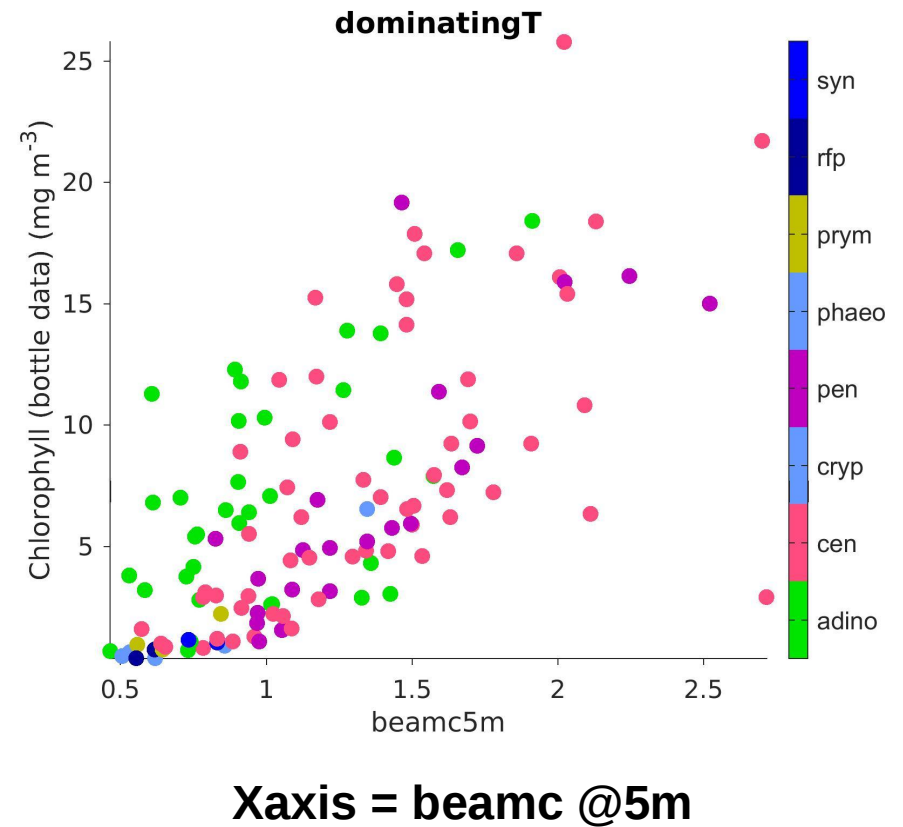
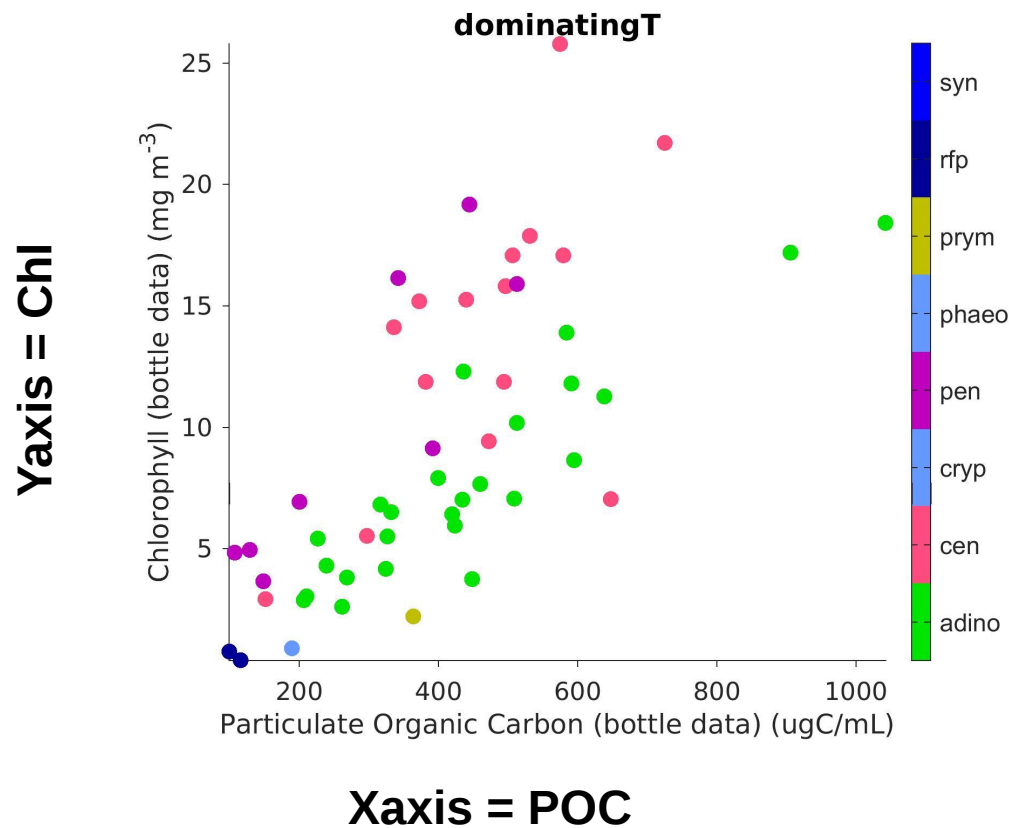


In situ surface bottle measurements

beamc is not a great proxy for biomass.

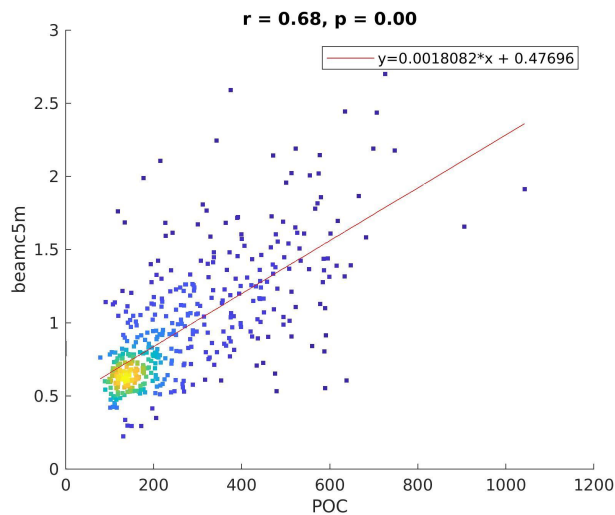
(A lot worse than published relationships in the open ocean, e.g., $r^2 = 0.92$ Graff et al 2015)

3) Taxonomy = $f^{\circ}(\text{Chl}, \text{beamc @5m})$

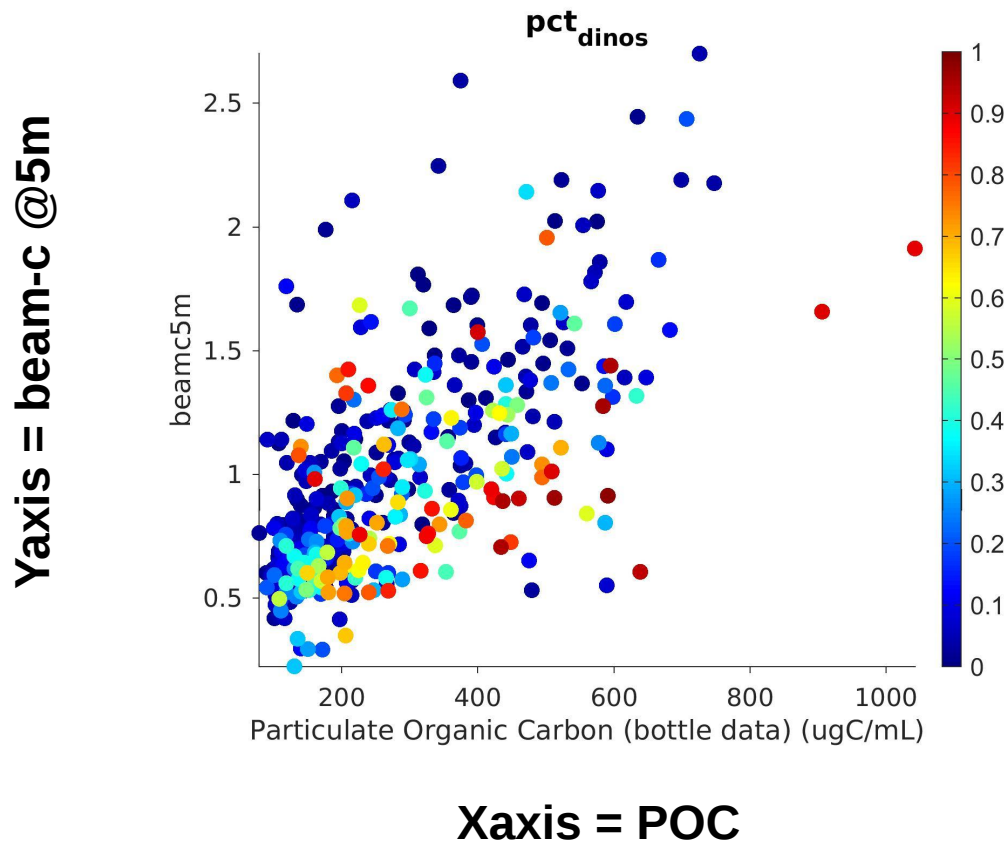


C:Chl higher for dinos – as expected
(same plot as slide 8)

beamc:Chl lower for dinos
opposite to expected relationship



Beam-c vs. POC



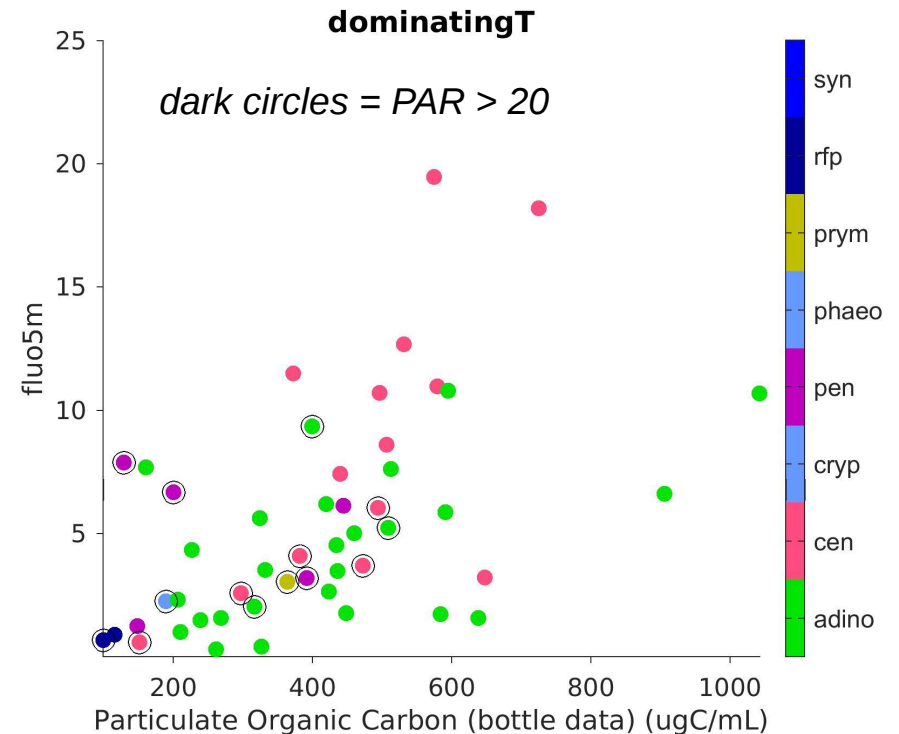
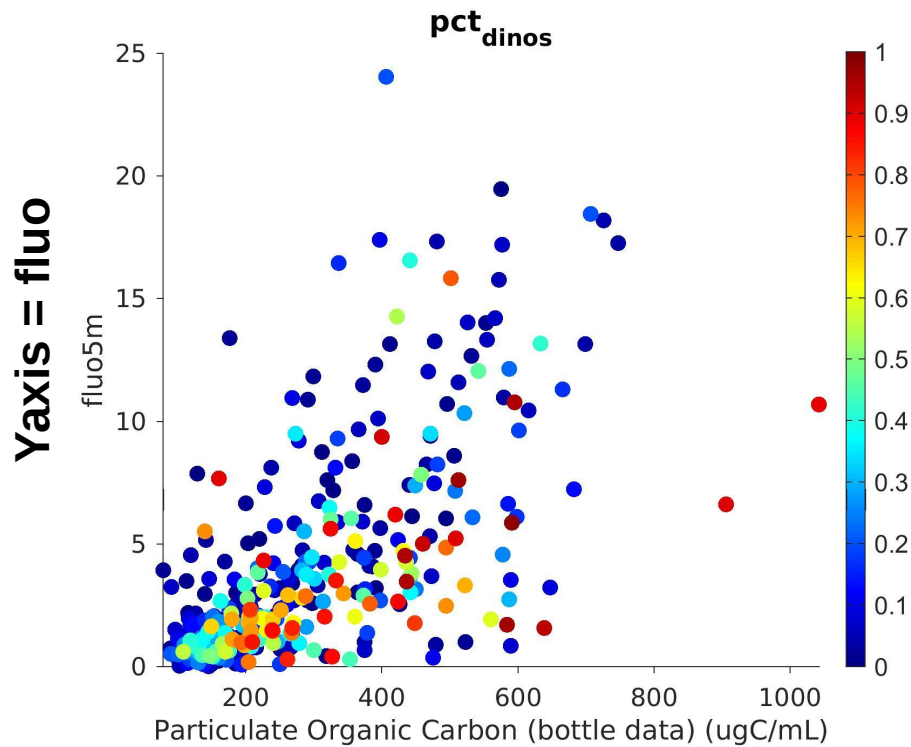
Correlation POC / beamc @5m

All samples	$r = 0.68$
pct_dinos > 0.5	$r = 0.50$
pct_dinos < 0.5	$r = 0.74$
pct_diatoms > 0.5	$r = 0.70$
pct_diatoms < 0.5	$r = 0.64$

→ relationship POC / beamc
breaks down for dinos.

(beamc lower than
expected for dinos)

Fluo vs. POC



Xaxis = POC

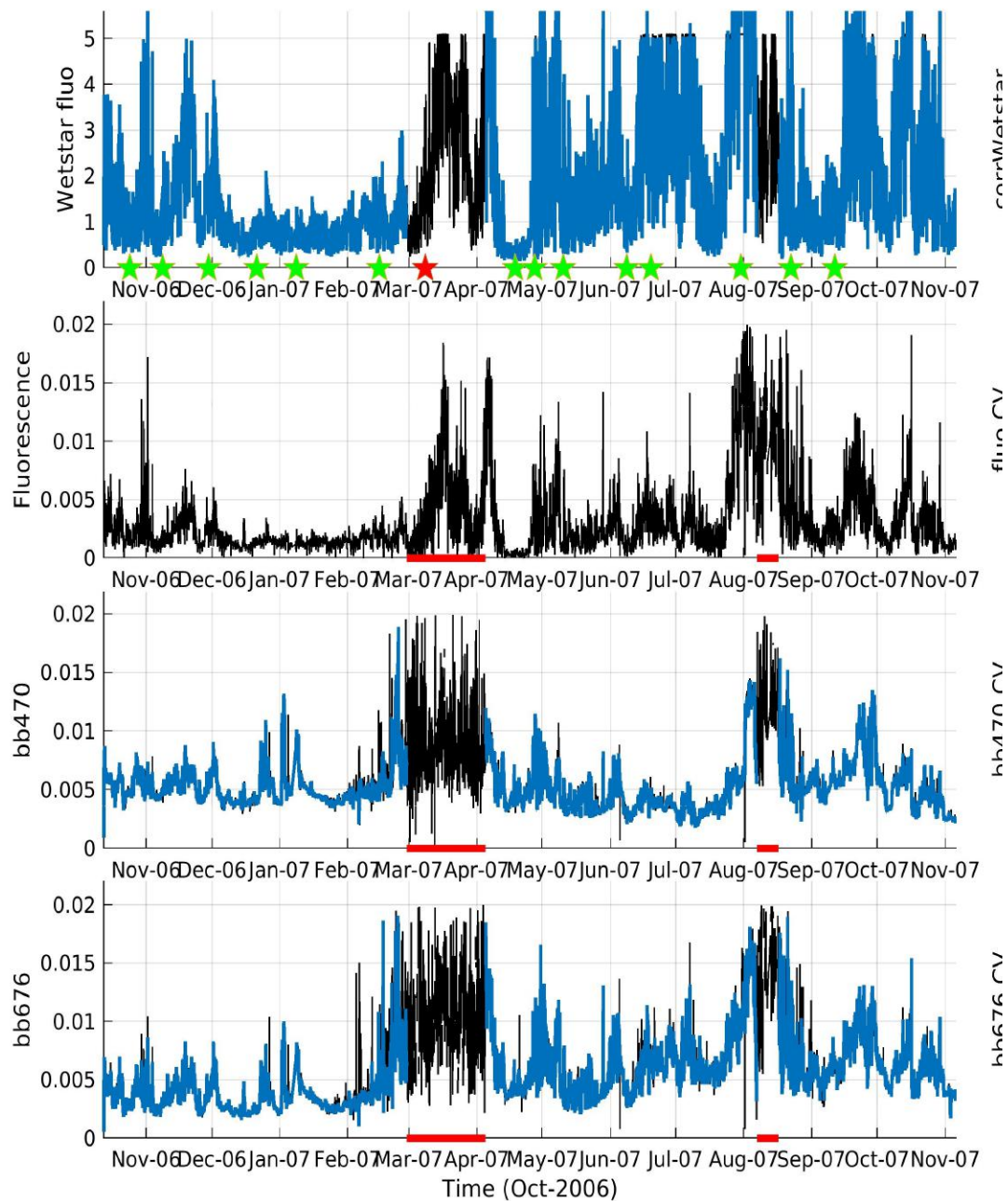
The relationship POC / fluo is (surprisingly) driven by taxonomy more than by quenching.
 $r(\text{Chl}, \text{fluo @5m}) = 0.78$

→ the real issue is with beamc.

Proxies from autonomous sensors

Mooring dataset (fluo, bbp470,
bbp676) vs taxonomy
2004-2014, N = 84 (after QC)

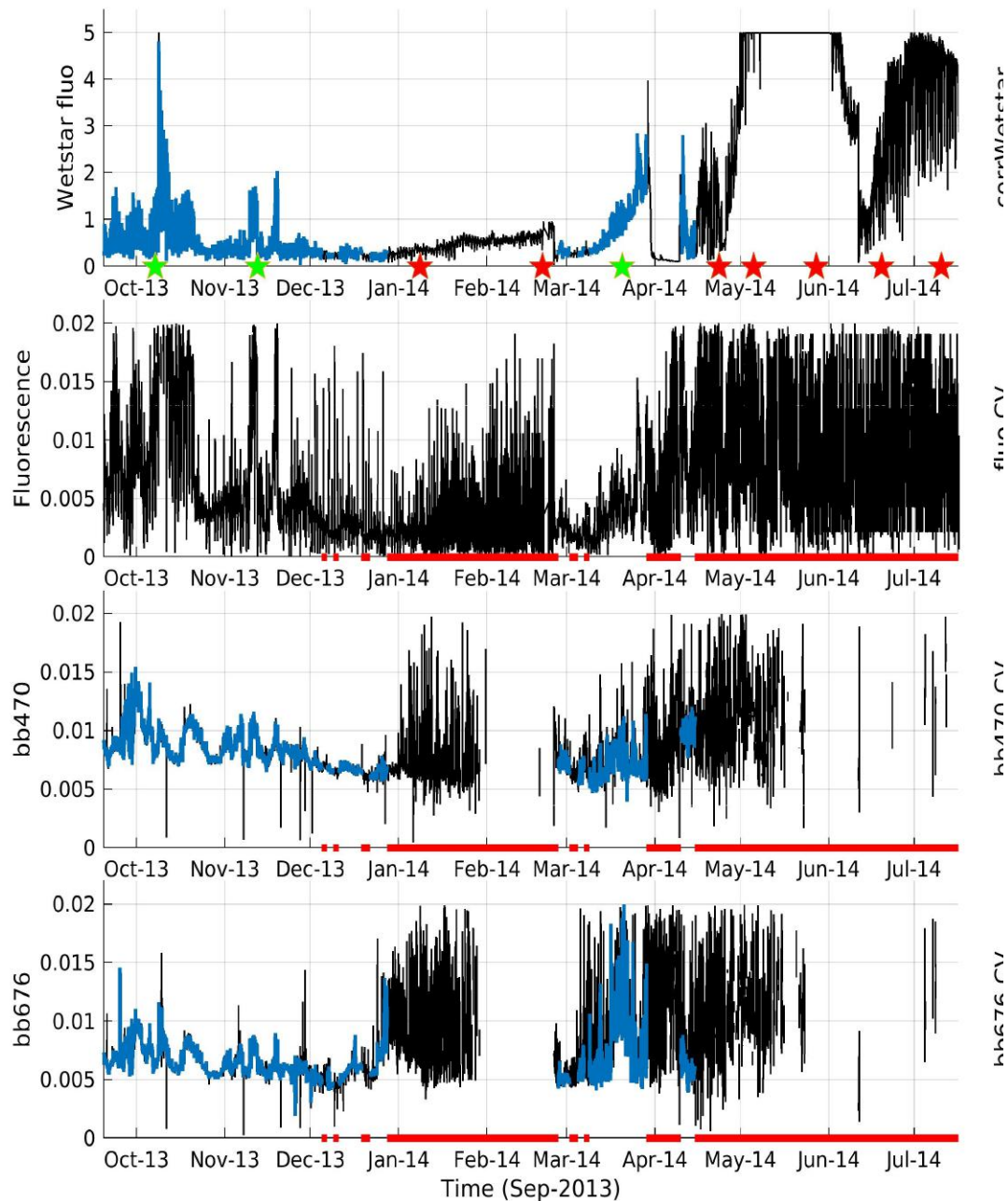
2pre) Mooring QC



Example of good deployment (200610).

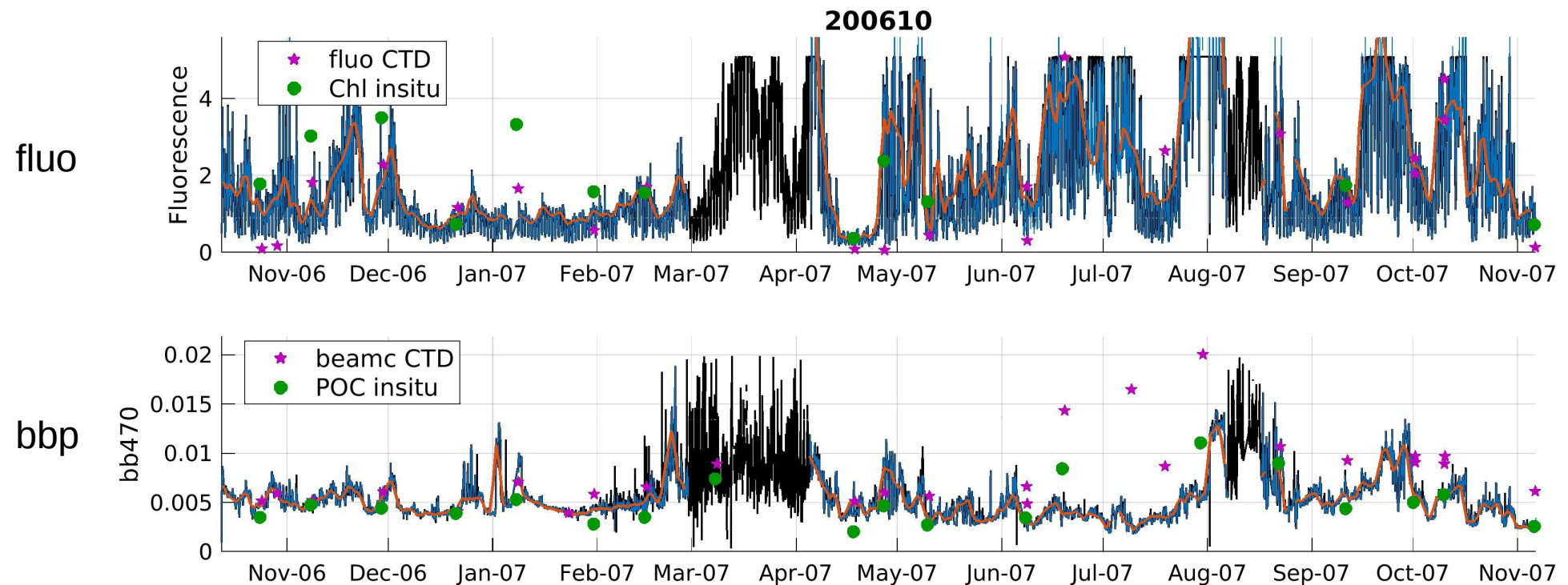
Stars = in situ taxa counts,
red = removed.

2pre) Mooring QC



Example of bad deployment (201309).
Stars = in situ taxa counts,
red = removed.

2pre) Mooring QC



Even for good deployments, fluo mooring / fluo CTD / Chl insitu (top) and bbp mooring / beamc CTD / POC in situ (bottom) do not agree very well.

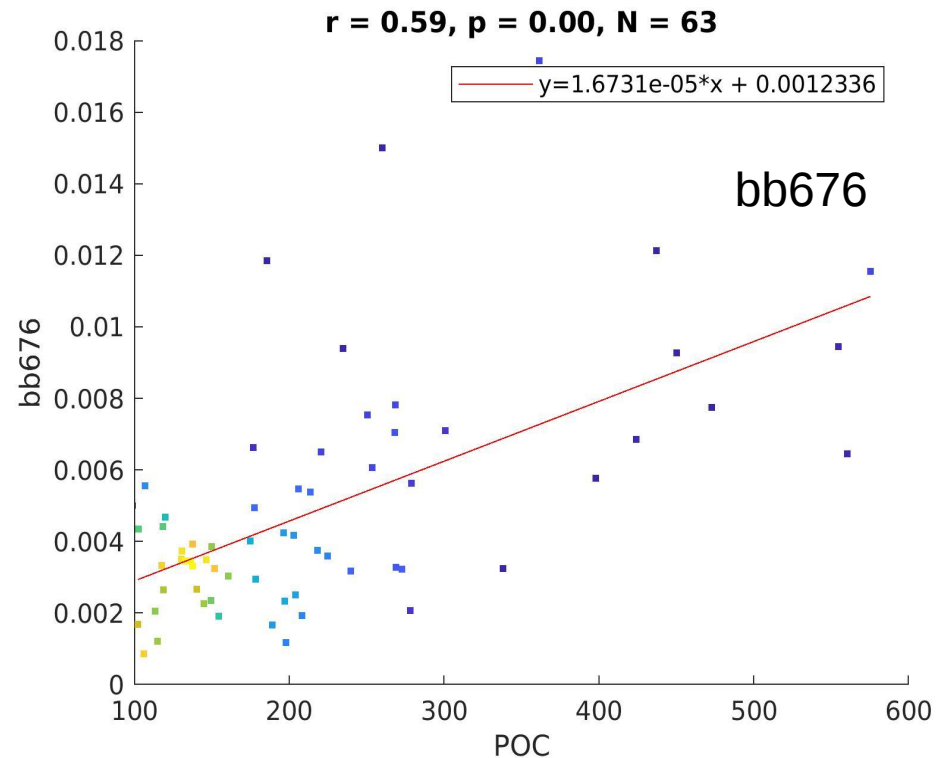
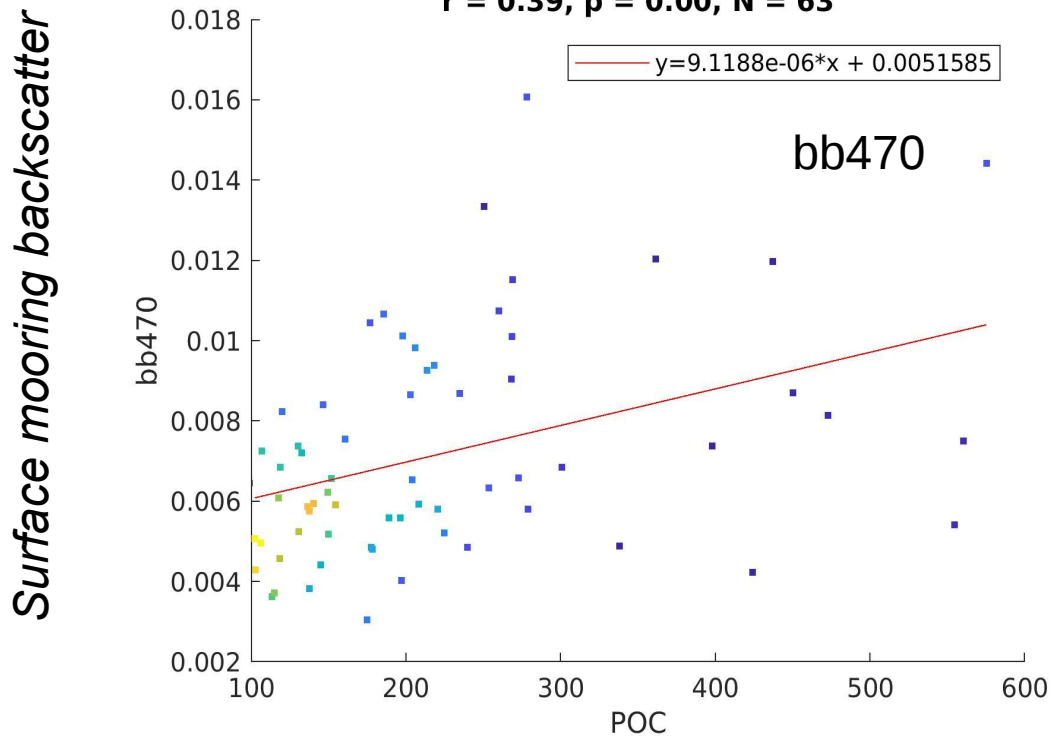
+ issue of fluorescence intercalibration between deployments.

Tested based on: (1) concurrent fluo CTD, (2) concurrent Chl insitu, (3) fluo/bb676 correlation. (2) gives the best results (used hereafter) but correlation is not always great.

→ choice of subdataset (5 deployments)

2) bbp as proxy for C

based on the entire mooring dataset (10 deployments) since bbp doesn't require calibration



In situ POC surface bottle measurements

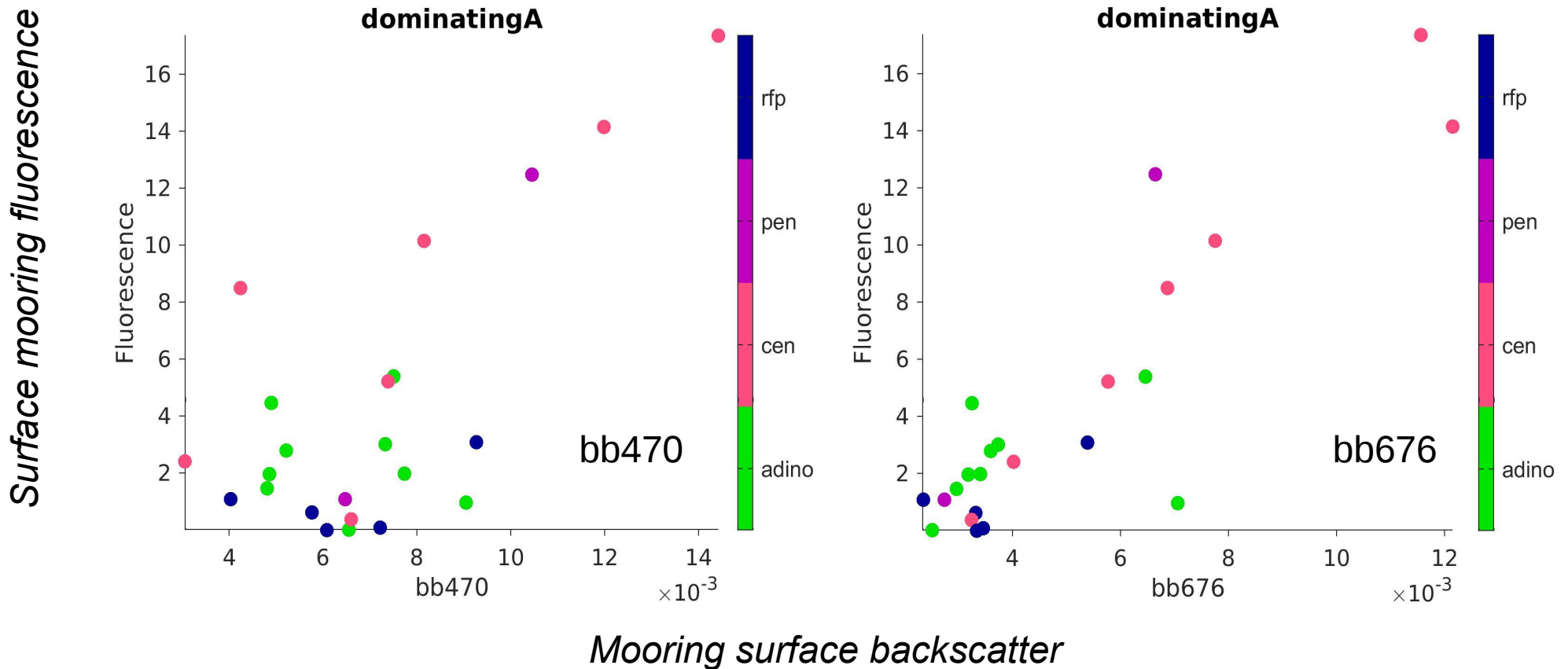
Correlations POC/bbp are not great...

(A lot worse than published relationships in the open ocean, e.g., $r^2 = 0.74$ for bb470 Graff et al 2015)

... but neither are correlations fluo/Chl **ie. there are not enough data points to conclude**

($r=0.54$ for Chl/fluor on the same dataset (no outliers), vs 0.80 on the CTD dataset when removing fluo>30)

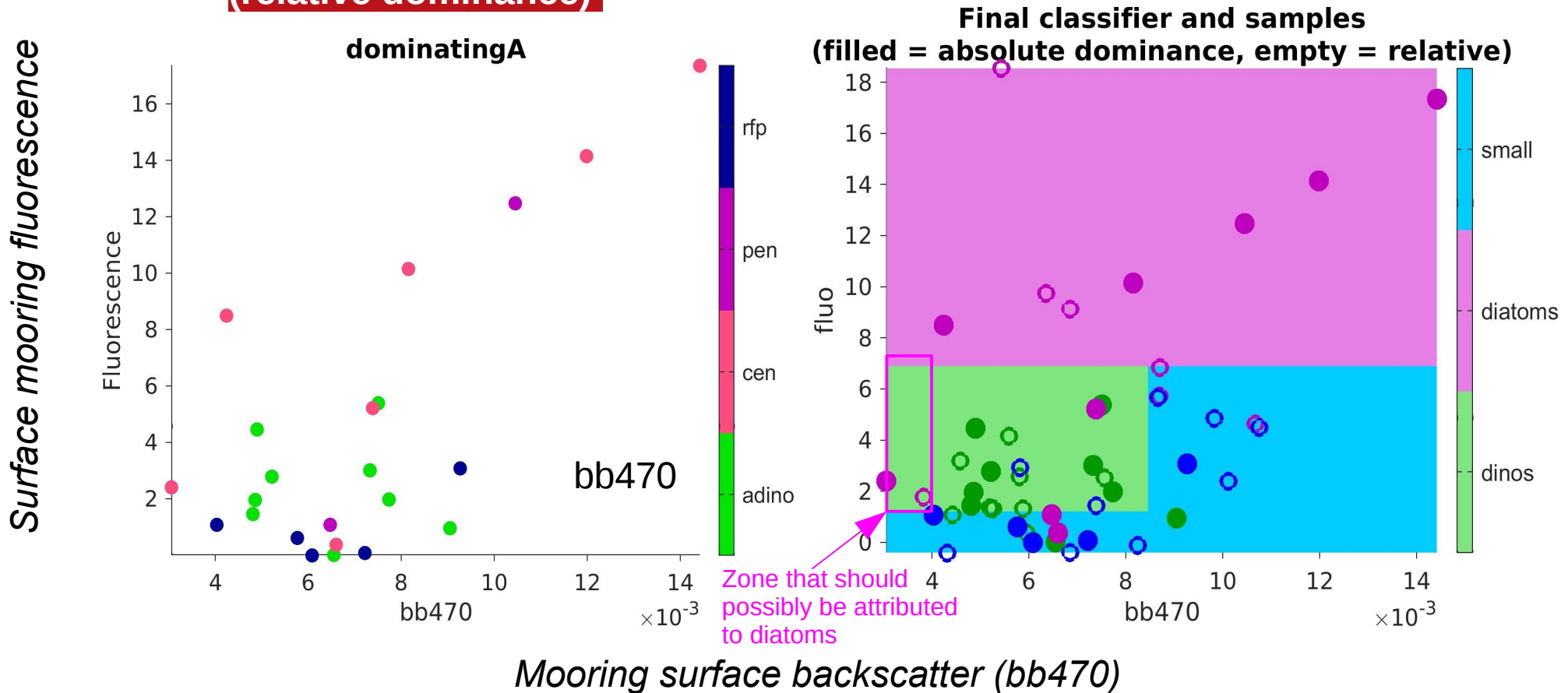
3) fluo/bbp as proxies for taxonomy



Nothing obvious... note that this is dominatingA (50%) and not dominatingT (80%) as with the CTD dataset (few samples, all dominatingT are diatoms)

3) fluo/bbp as proxies for taxonomy

... however **machine learning tree algorithm succeeds in predicting taxonomy (relative dominance)**



Algorithm trained on absolute dominance (> 50%, filled circles, N=23) and evaluated on relative dominance (open circles, N=26): 55% accuracy on training dataset, **65% on evaluation dataset** (only predictors = fluo & bb470).

Proxies from satellite (PHYSAT)

Evaluate taxonomic groups proxies (e.g. diatoms, dinoflagellates, picoplankton)
by comparison with BOG time series

Proxies from satellite (PHYSAT)

Concept

Water-leaving radiance (Ra) spectra varies with
Chl (first order) and taxa (second order).
PHYSAT classifying Ra anomalies.

Approach

- 1- Do PHYSAT groups match *in situ* groups?
- 2- Custom calibration with *in situ* data.

Proxies from satellite (PHYSAT)

Goal: evaluate outputs from the PHYSAT algorithm, that predicts taxonomy from space (dominant species at each pixel).

- 1) PHYSAT only (spatio-temporal patterns)
- 2) PHYSAT in comparison with BOG data

Main conclusions:

- regional patterns look OK, noting that PHYSAT mostly finds nanoflagellates in winter
- there are a LOT of missing data nearshore, so that MB patterns are off
- comparison with BOG data is not conclusive.
- in general Nanos are likely overrepresented (ie some PHYSAT “nanos” are likely diatoms for instance)

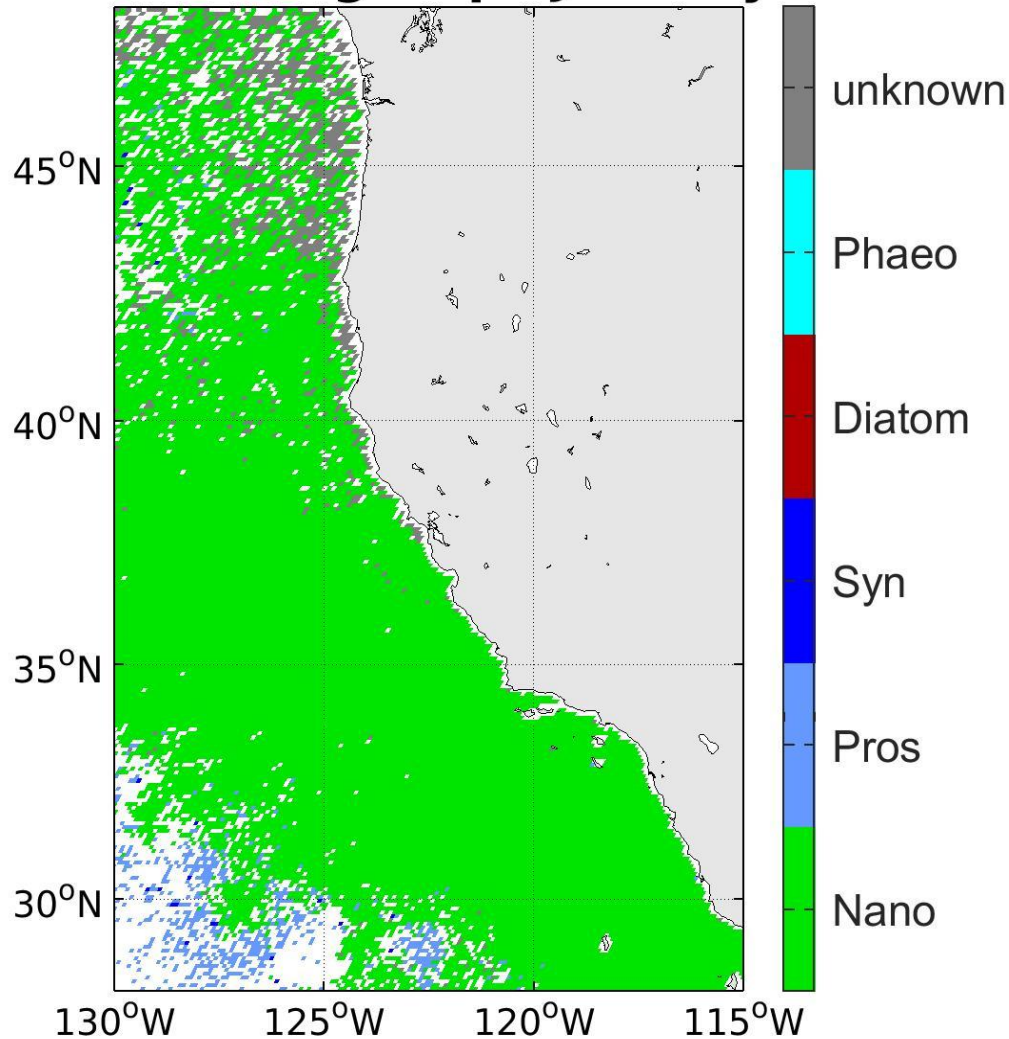
→ doesn't appear to work, at least locally (too many data gaps?)

Note that this is based on SeaWiFS only, not MODIS – also likely to work better regionally.

1) PHYSAT patterns

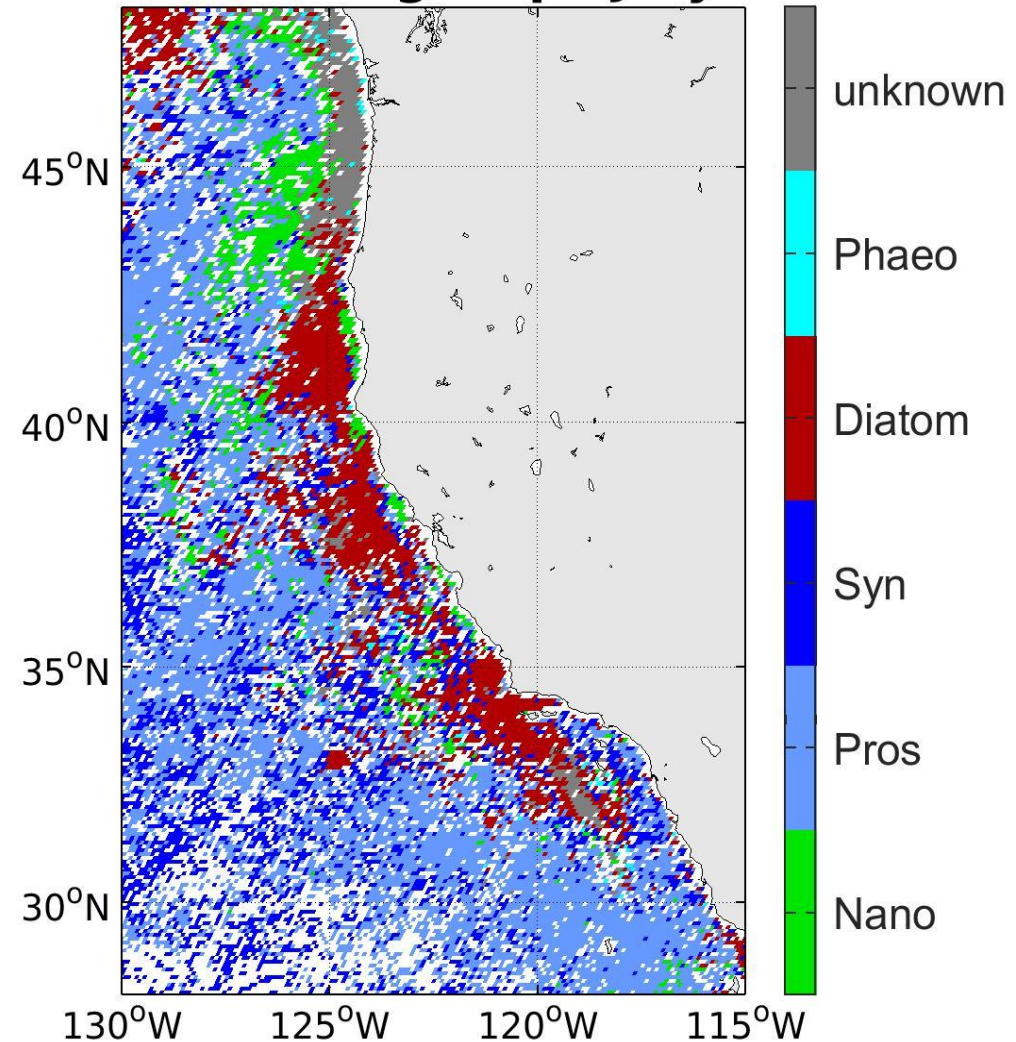
WINTER

PHYSAT groups January

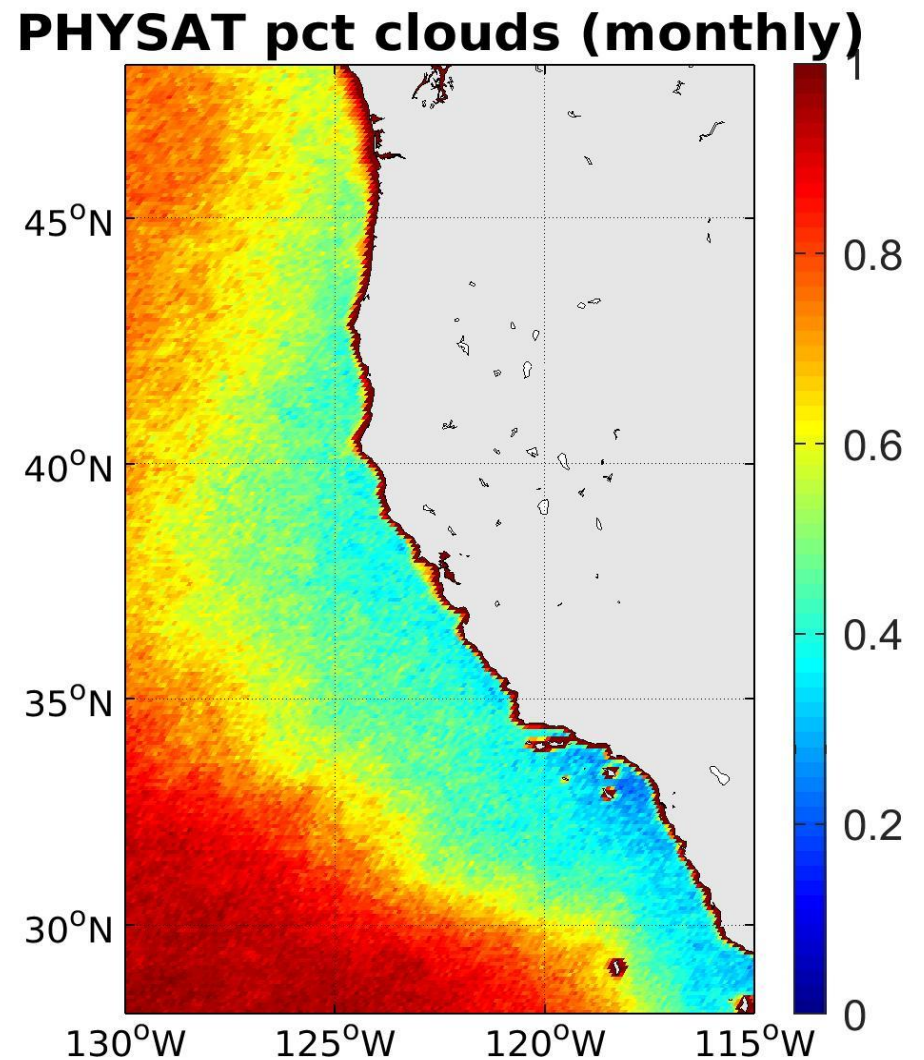
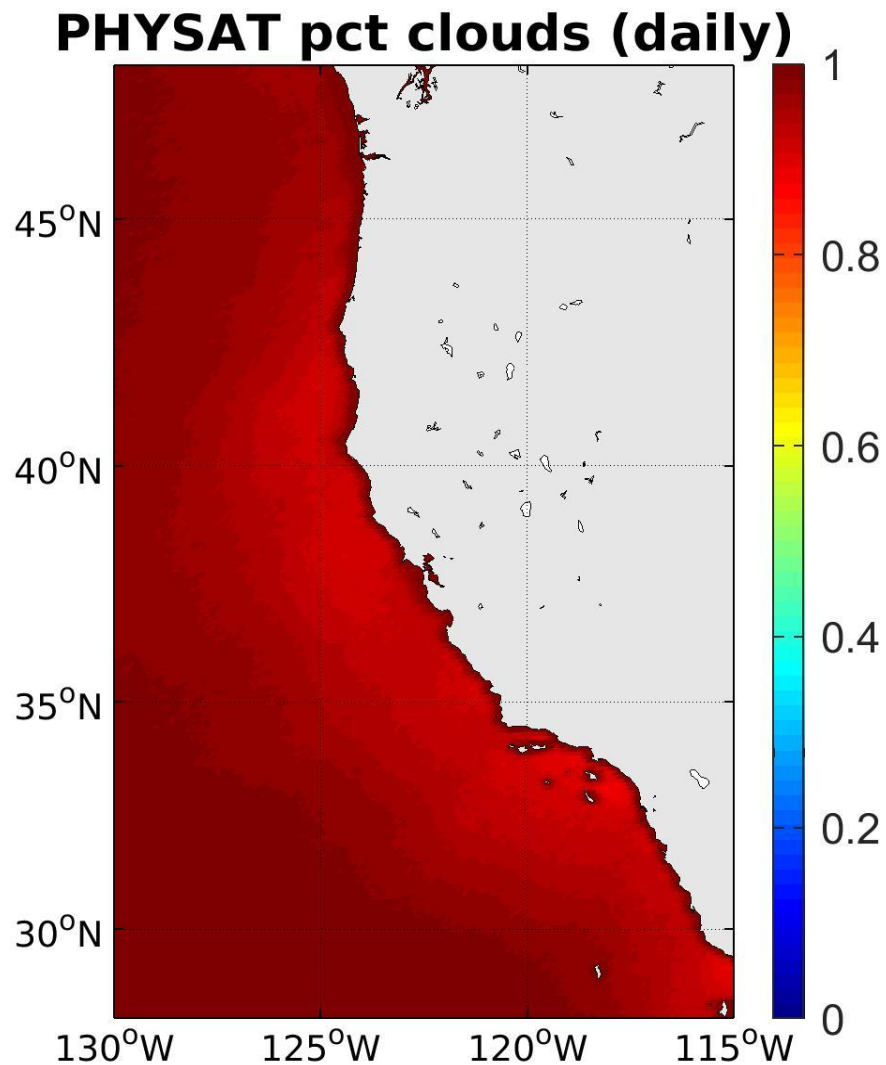


SUMMER

PHYSAT groups July



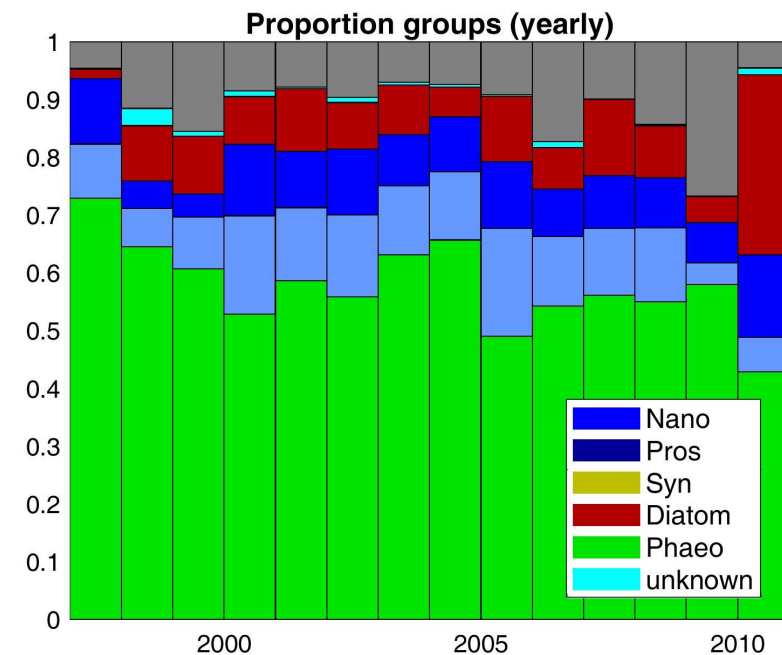
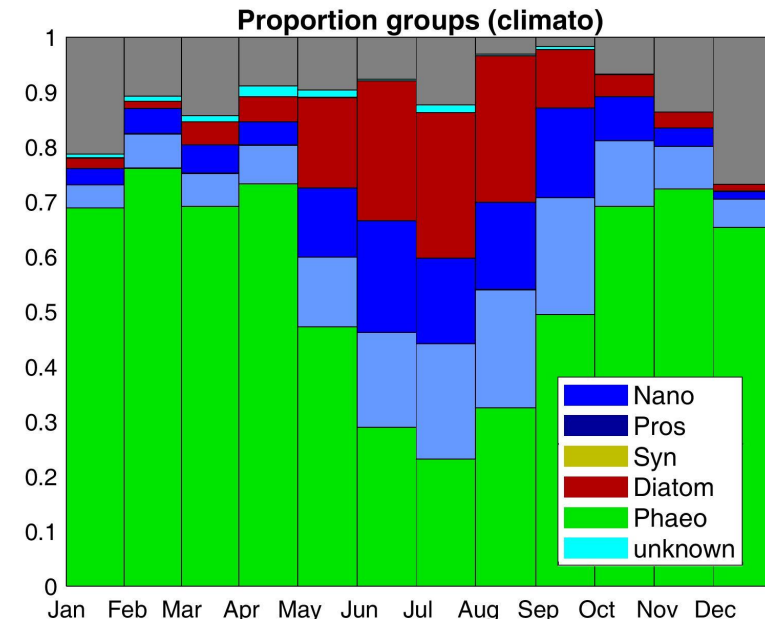
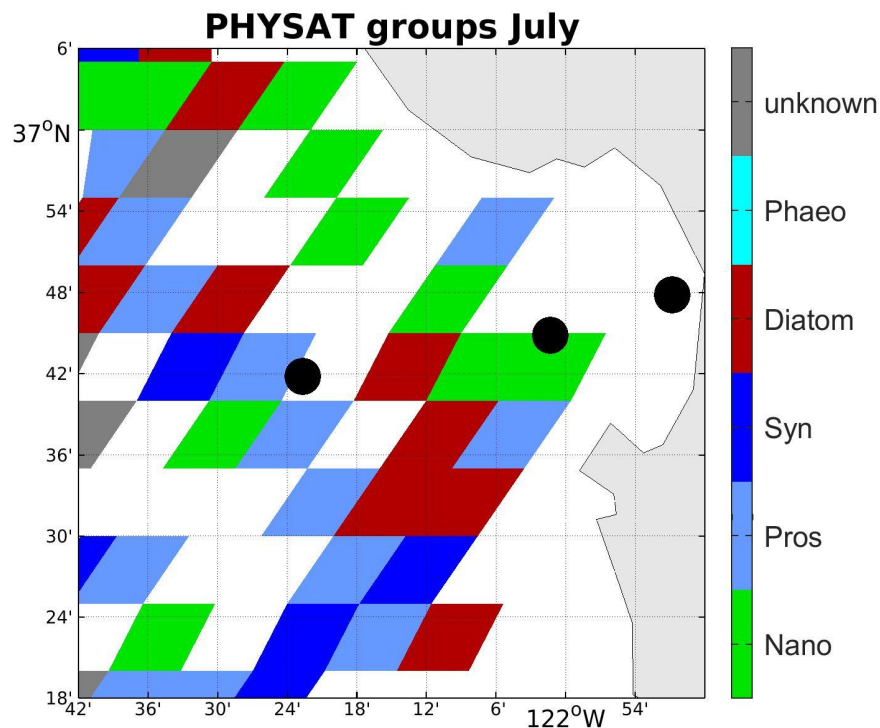
1) PHYSAT patterns: data gaps



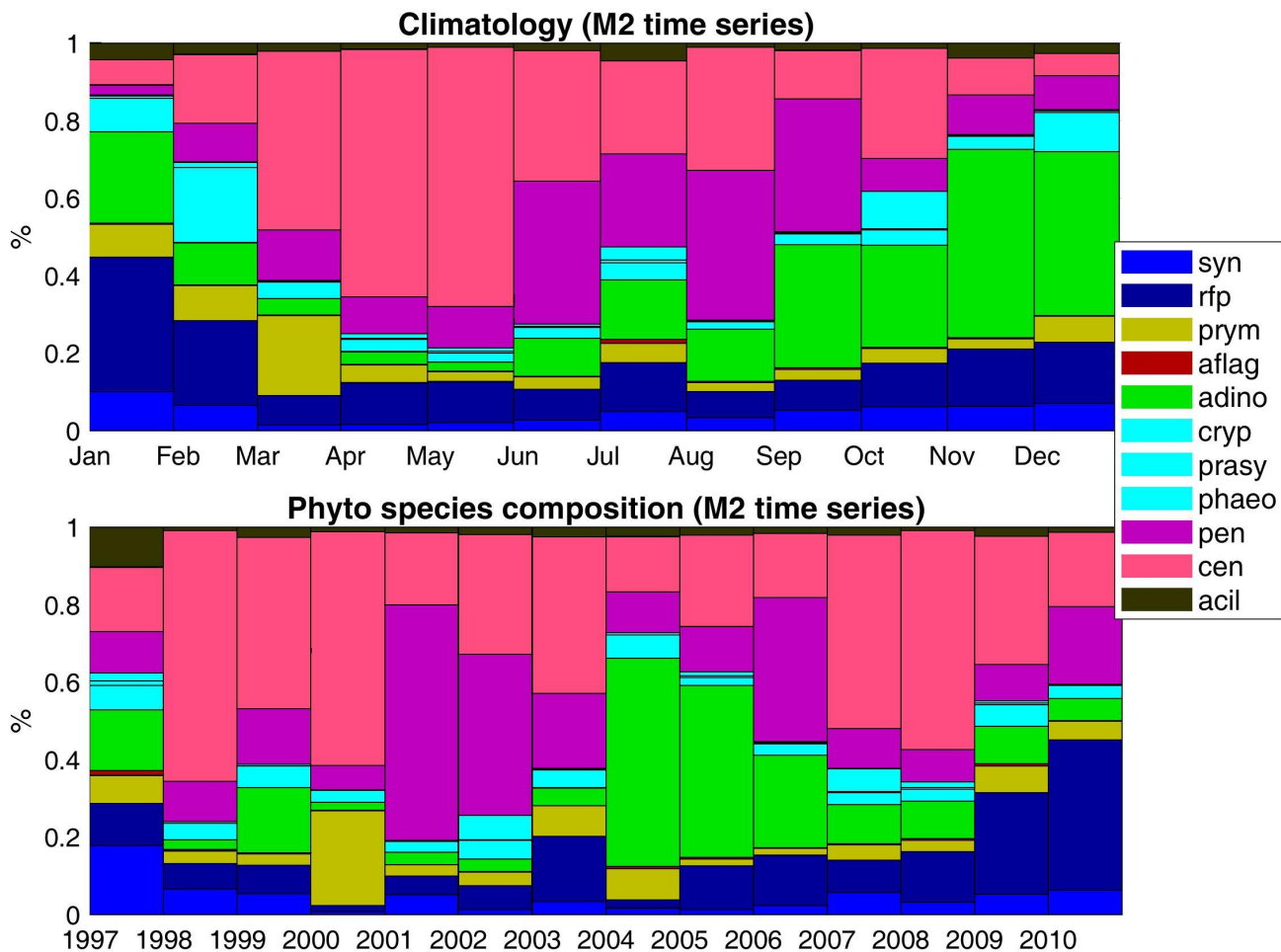
1) PHYSAT patterns: Monterey Bay

Gaps due to:

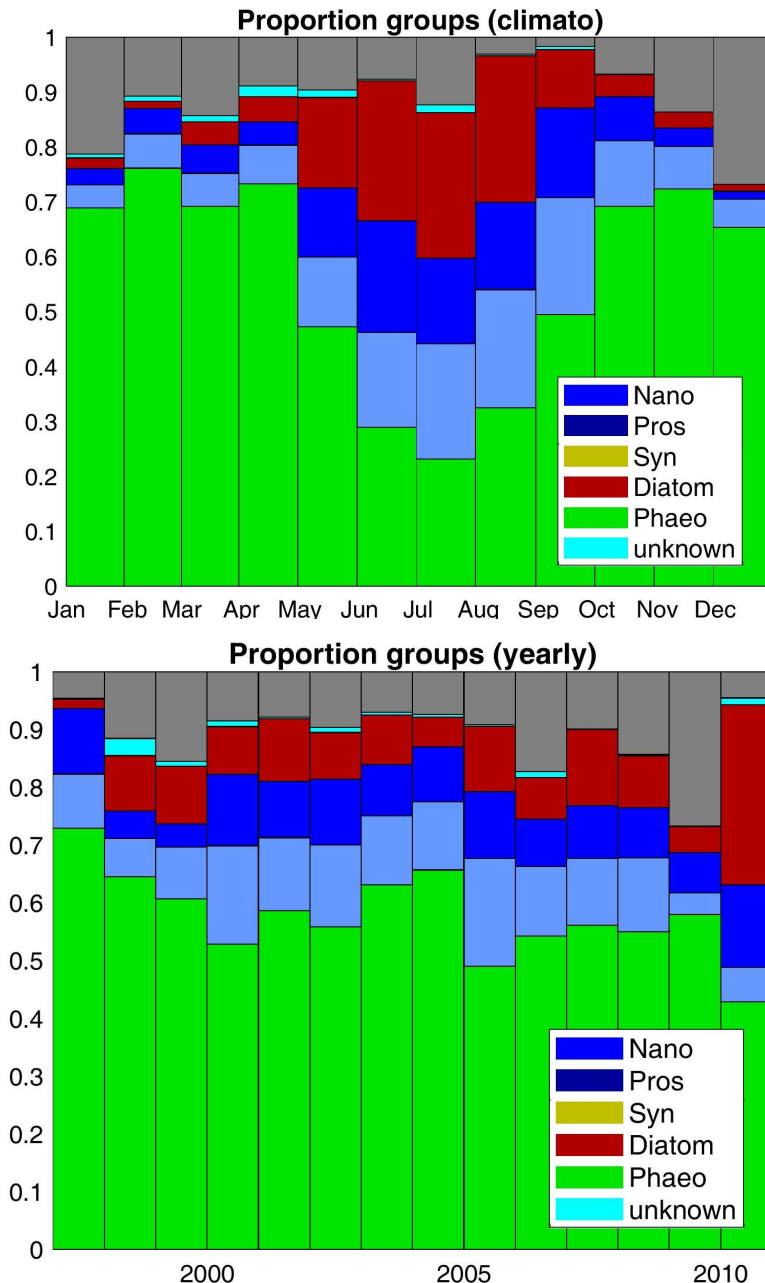
- clouds
- PHYSAT algorithm not identifying dominance
- no dominance for a given pixel over the years! (eg same number of 2 groups)



1) Comparison with BOG



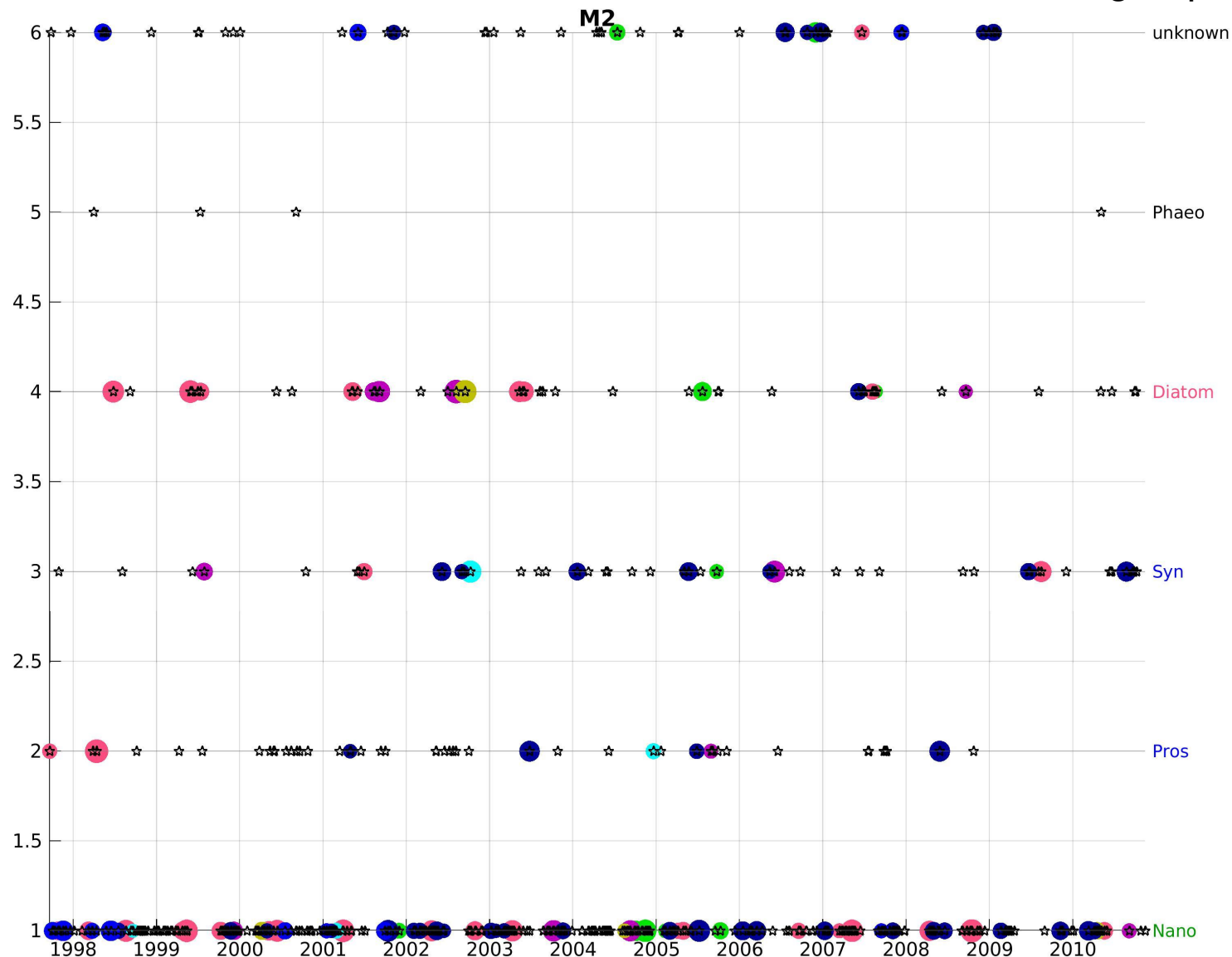
M2 = station most offshore, likely most representative of PHYSAT conditions, restricted to 1997-2010



2) Comparison with BOG

pixel closest to M2
(less clouds than M1),
PHYSAT data within 7 days

Stars = PHYSAT data, colors = matching BOG sample



PHYSAT
group

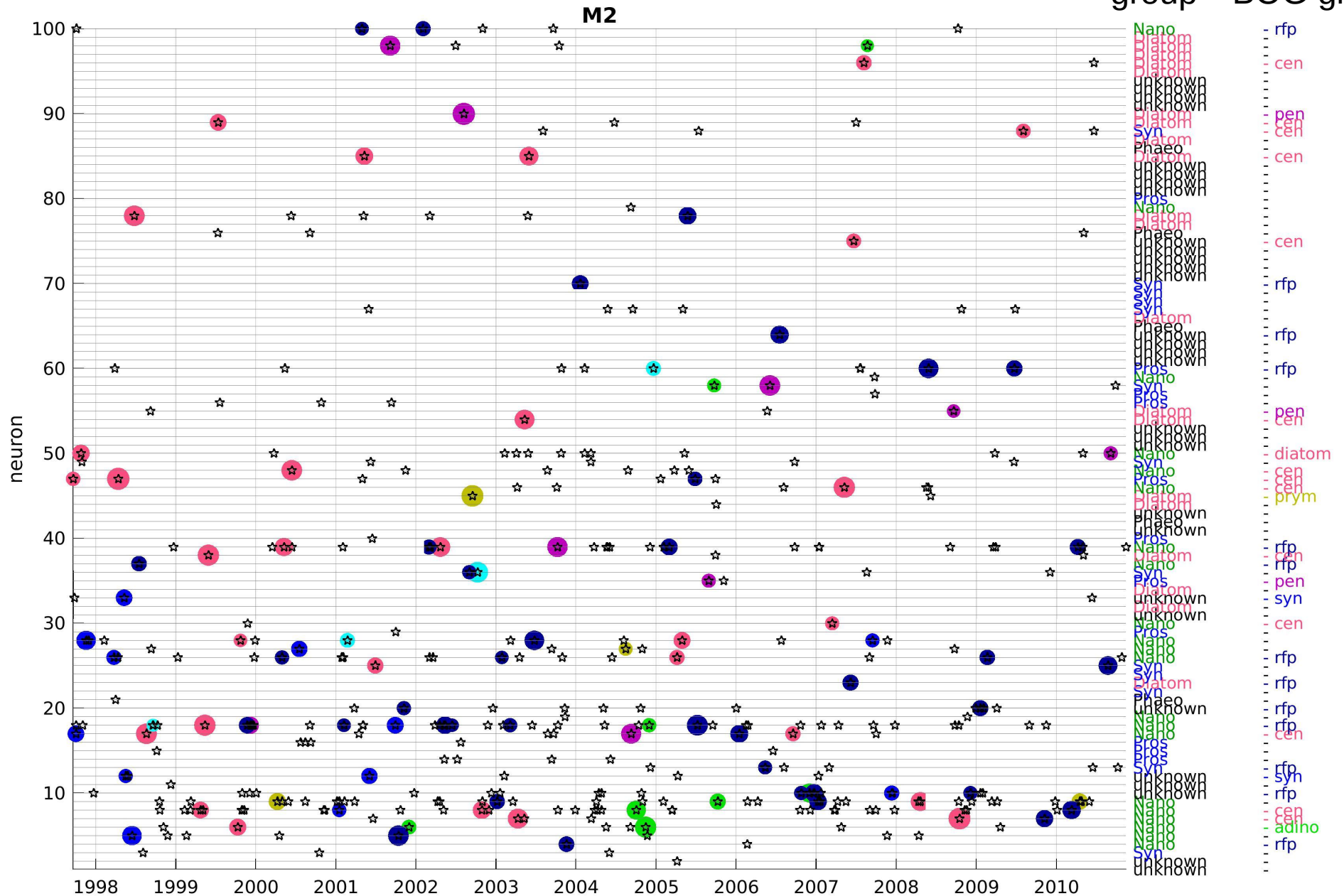
dominant
BOG group

not
too
bad

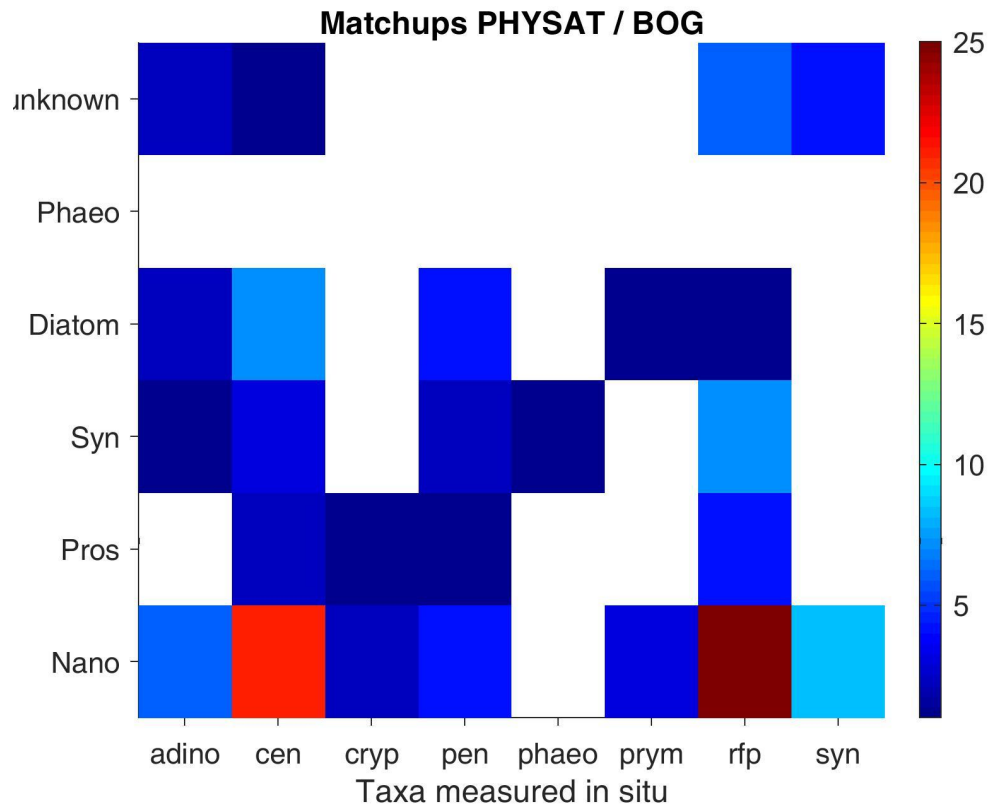
but most
PHYSAT
pixels
are nano

pixel closest to M2
(less clouds), PHYSAT
data within 7 days

PHYSAT dominant
group BOG group



2) Comparison with BOG



→ not very conclusive.

For instance, PHYSAT “diatoms” do seem to pick up mostly diatoms; however, most diatoms are labelled “Nano” by PHYSAT.

