

Global phytoplankton community composition from hyperspectral ocean color

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SEPTEMBER 24, 2025





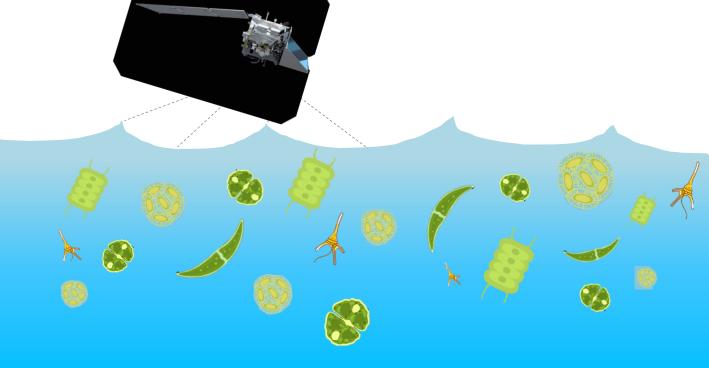






Goals for today

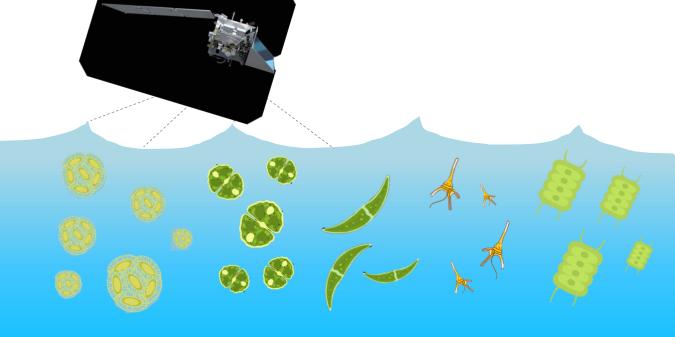
1) Model surface ocean phytoplankton pigments from hyperspectral $R_{rs}(\lambda)$



Goals for today

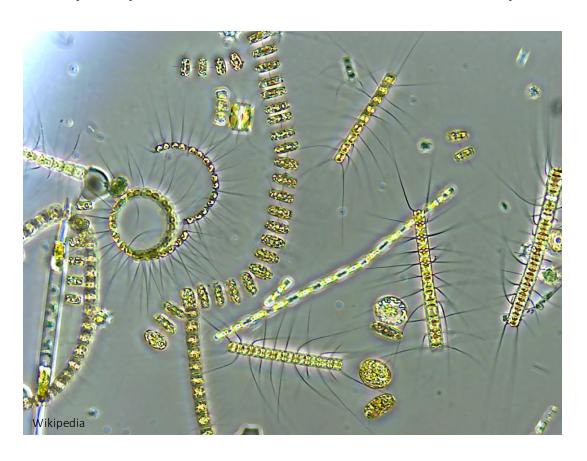
1) Model surface ocean phytoplankton pigments from hyperspectral $R_{rs}(\lambda)$

2) Compare the composition and distribution of communities derived from phytoplankton pigments and from hyperspectral $R_{rs}(\lambda)$



A fundamental challenge when describing phytoplankton communities

Phytoplankton are microscopic...



A fundamental challenge when describing phytoplankton communities

Phytoplankton are microscopic...

but can be seen from space



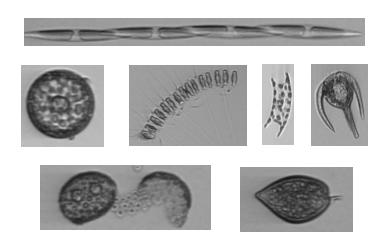


How do we measure phytoplankton community composition?

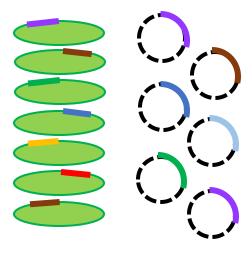
The "ideal" method is the one that is most effective for your system and question – understand the strengths and limitations.



microscopy



(imaging-in-) flow cytometry

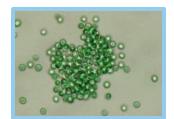


amplicon sequencing

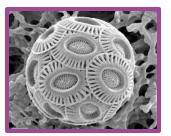
High Performance Liquid Chromatography pigments

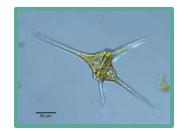
Phytoplankton have different pigments; some can be used as biomarkers to separate certain groups.





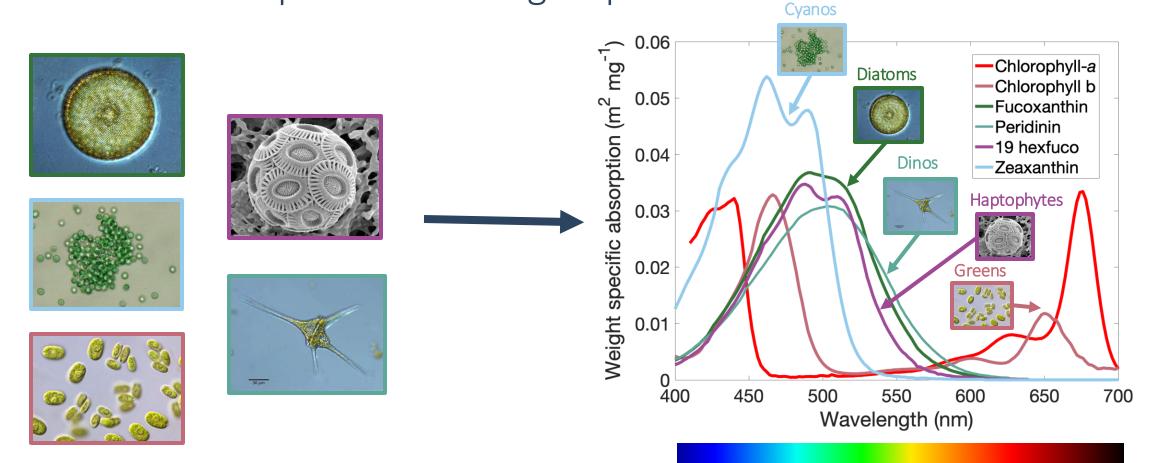




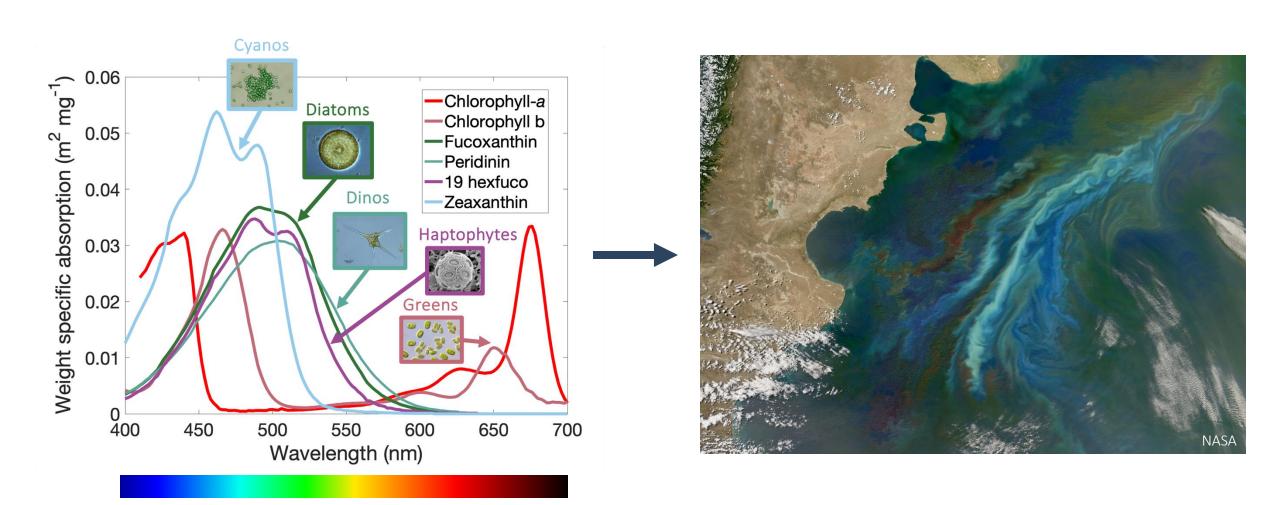


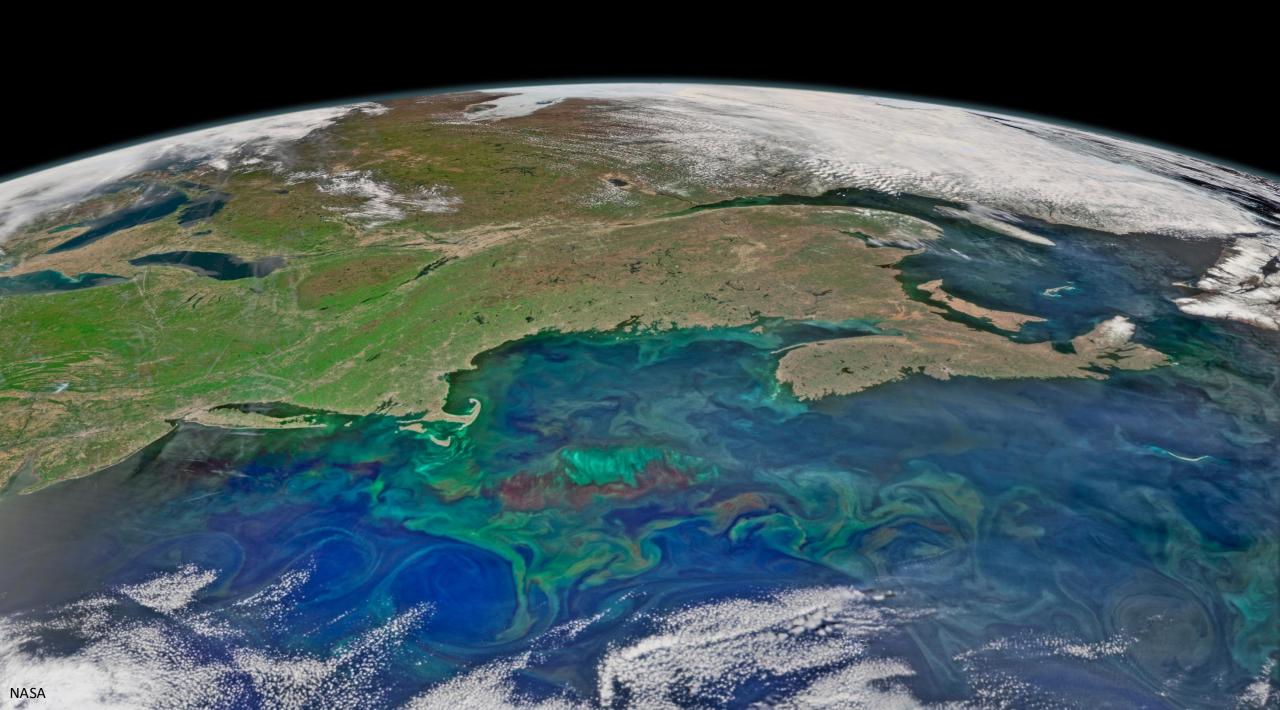
Phytoplankton pigments affect absorption

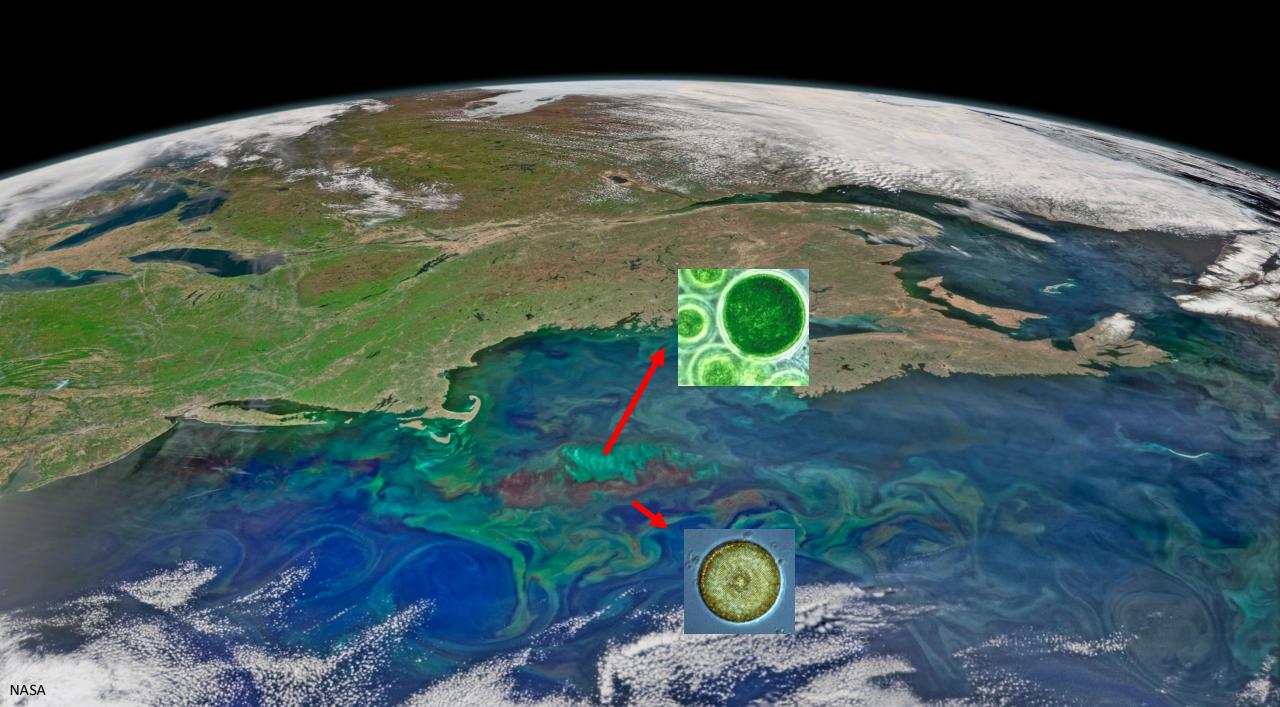
Phytoplankton have different pigments; some can be used as biomarkers to separate certain groups.

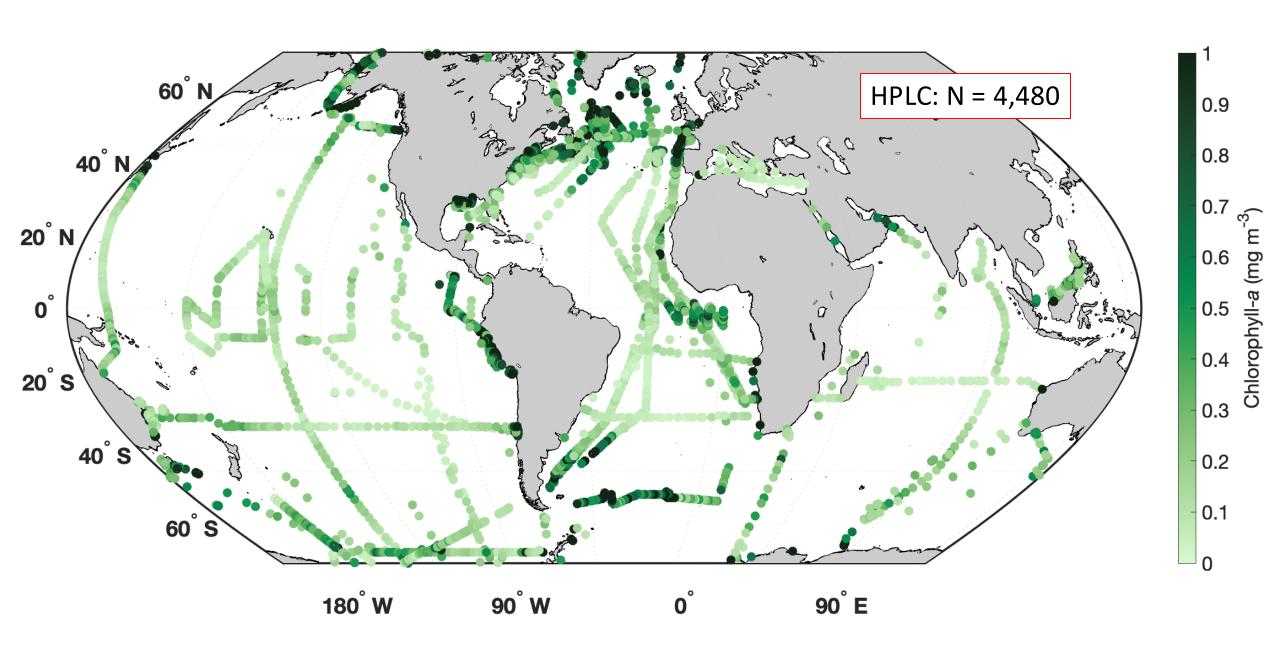


Pigments link phytoplankton and ocean color

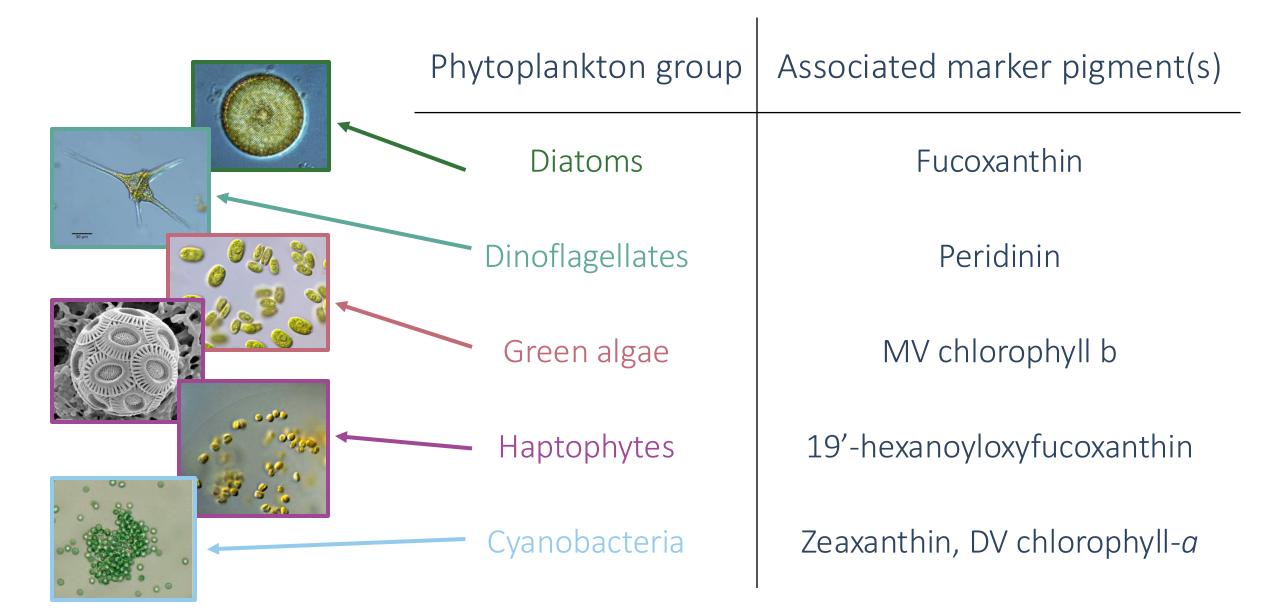


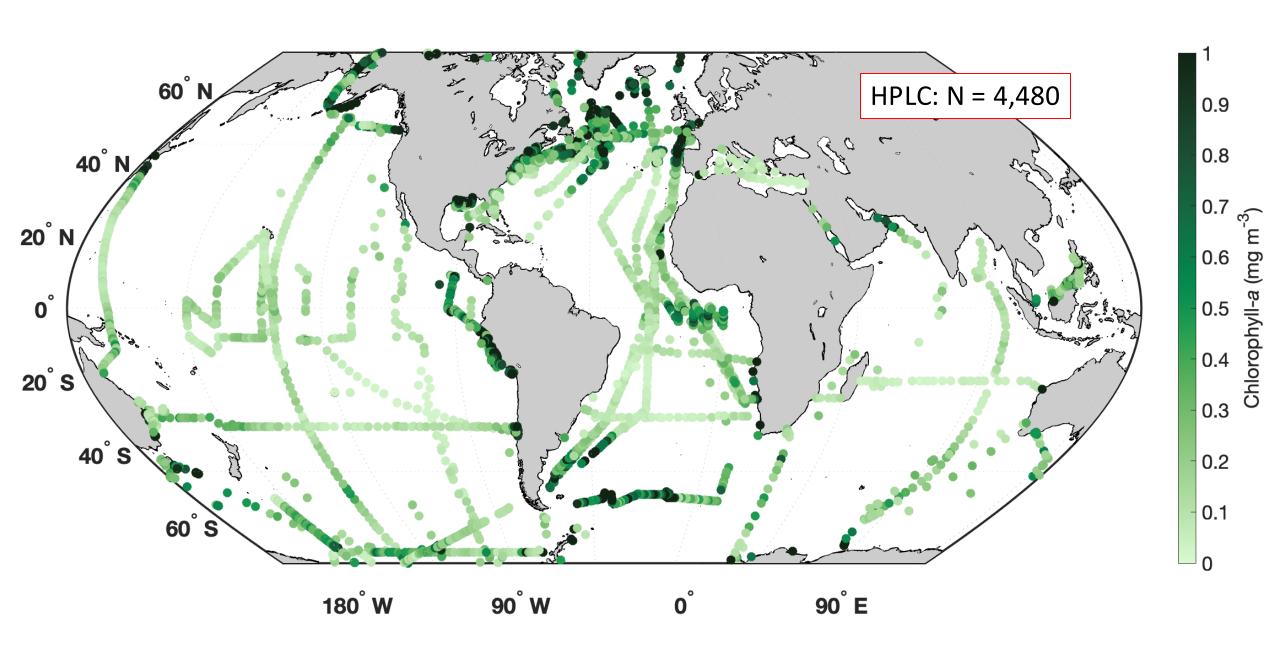






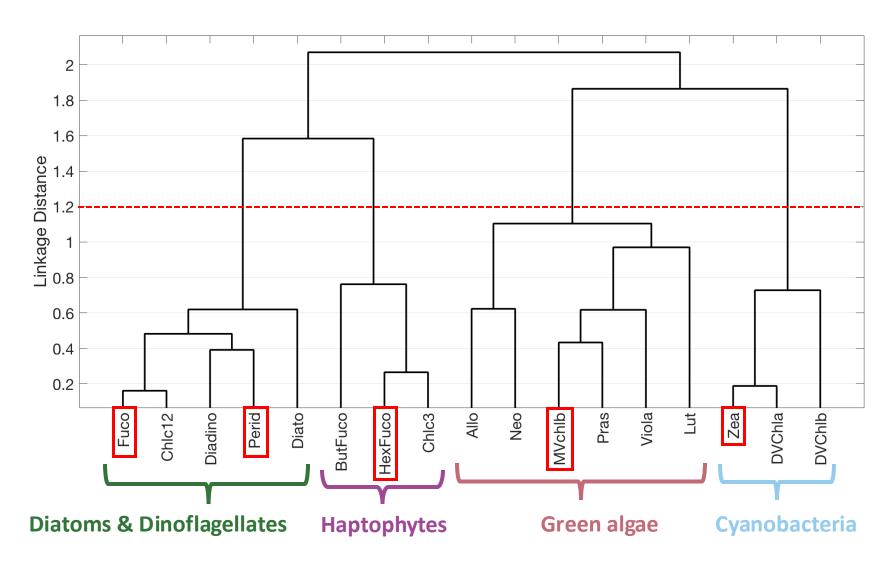
Phytoplankton pigments and taxonomy





Only 4 pigment-based groups can be separated globally

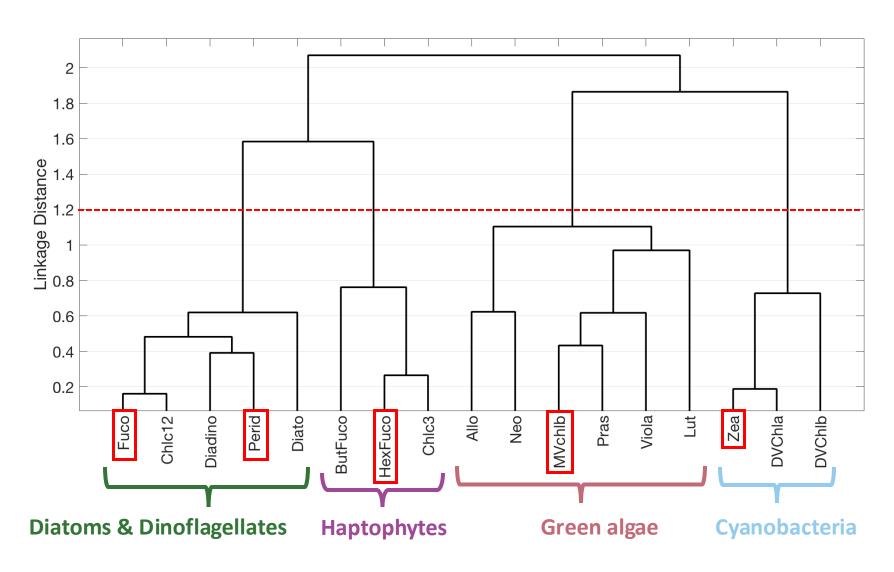
Using hierarchical cluster analysis to divide pigments based on the strength of correlations.



Only 4 pigment-based groups can be separated globally

Using hierarchical cluster analysis to divide pigments based on the strength of correlations.

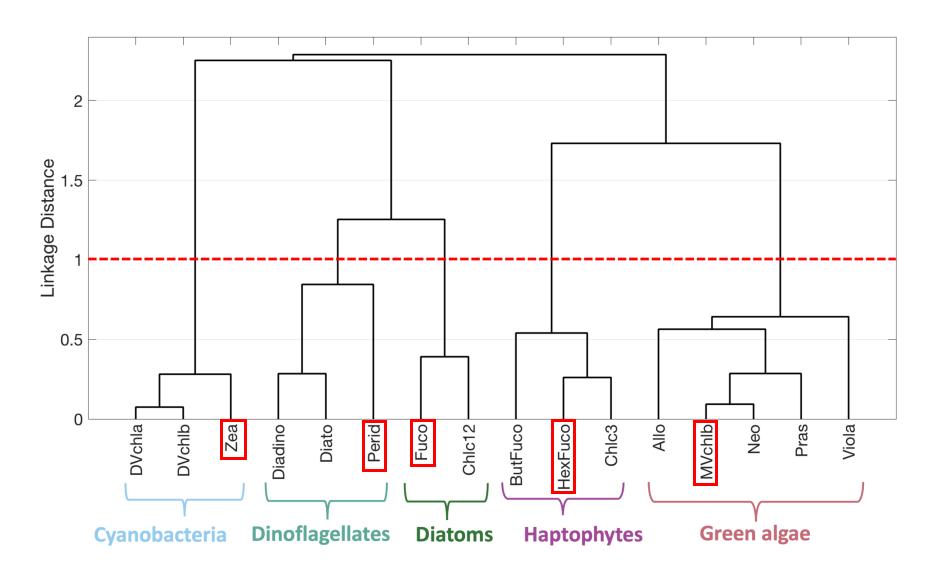
Pigments are limited in the number of groups they can separate.

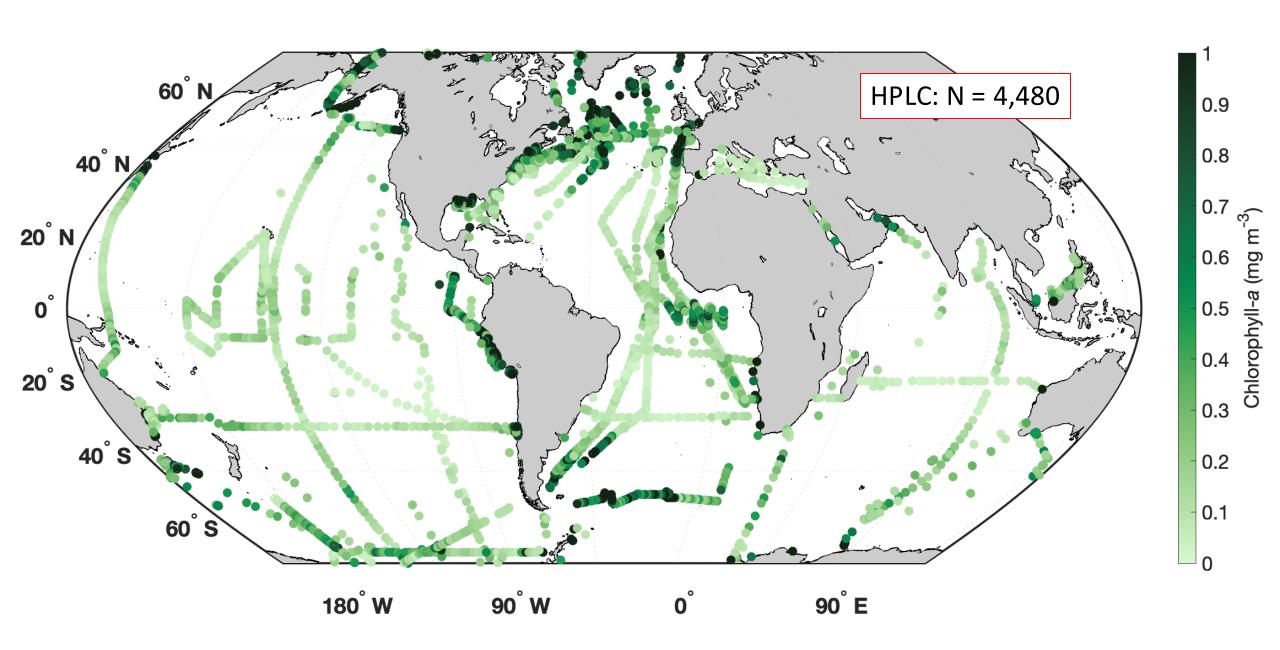


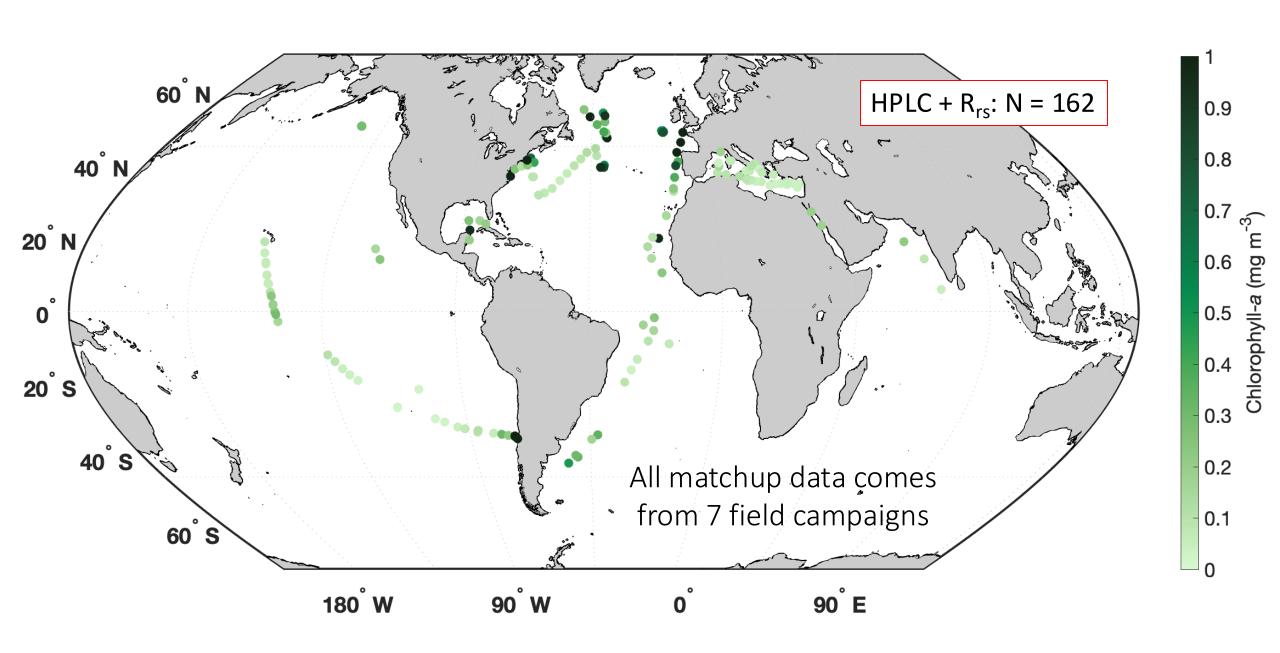
But 5 groups are separated in the western North Atlantic

Comparing global pigment clusters to a dataset collected over a seasonal cycle in the North Atlantic...

Five pigment-based groups separate from hierarchical cluster analysis on a regional scale.

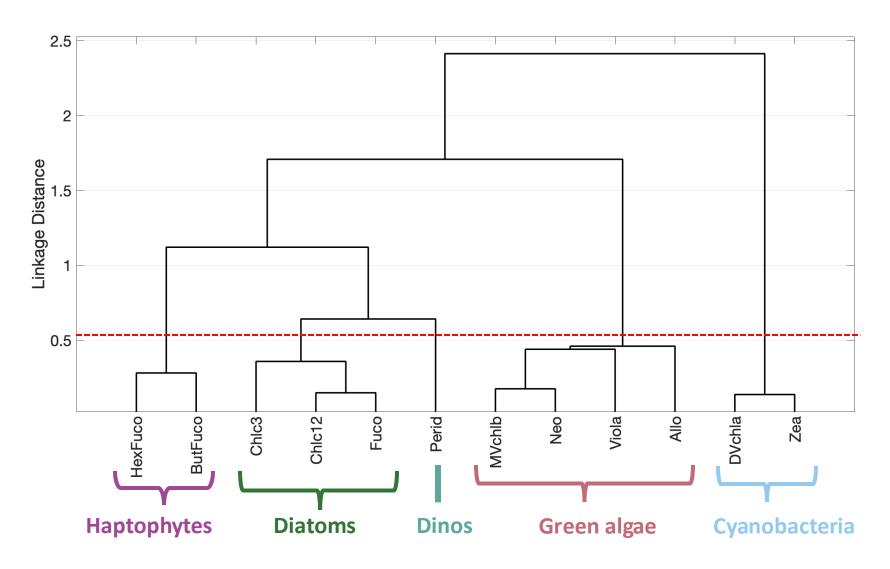


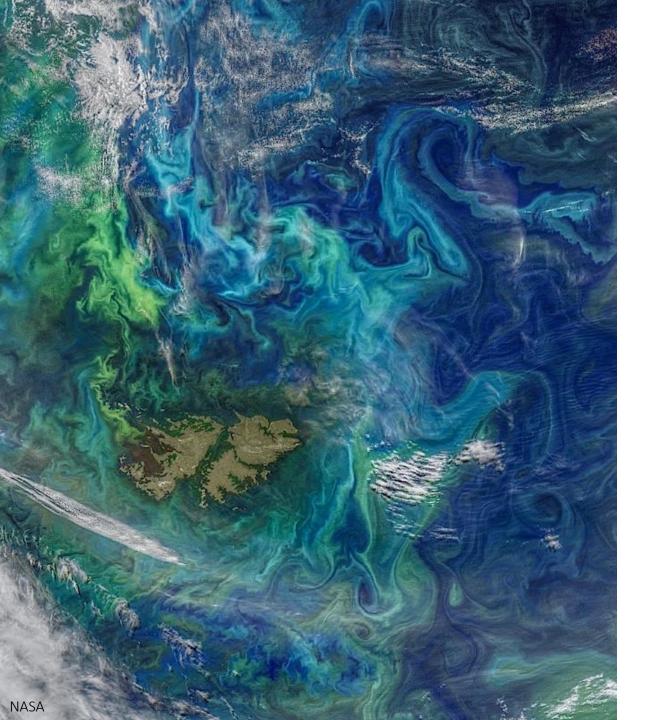




Max five pigment-based groups separate in this dataset

Paired global dataset of hyperspectral $R_{rs}(\lambda)$ and **HPLC** pigments can be separated into more groups than the global HPLC dataset.



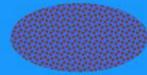


In-water data provide a baseline for ocean color modeling

Pigments can be used to separate a limited number of phytoplankton groups, depending on the dataset / ecosystem.

But need to test if the relationships between pigments and optics are as strong as we expect...

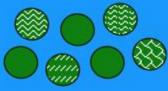




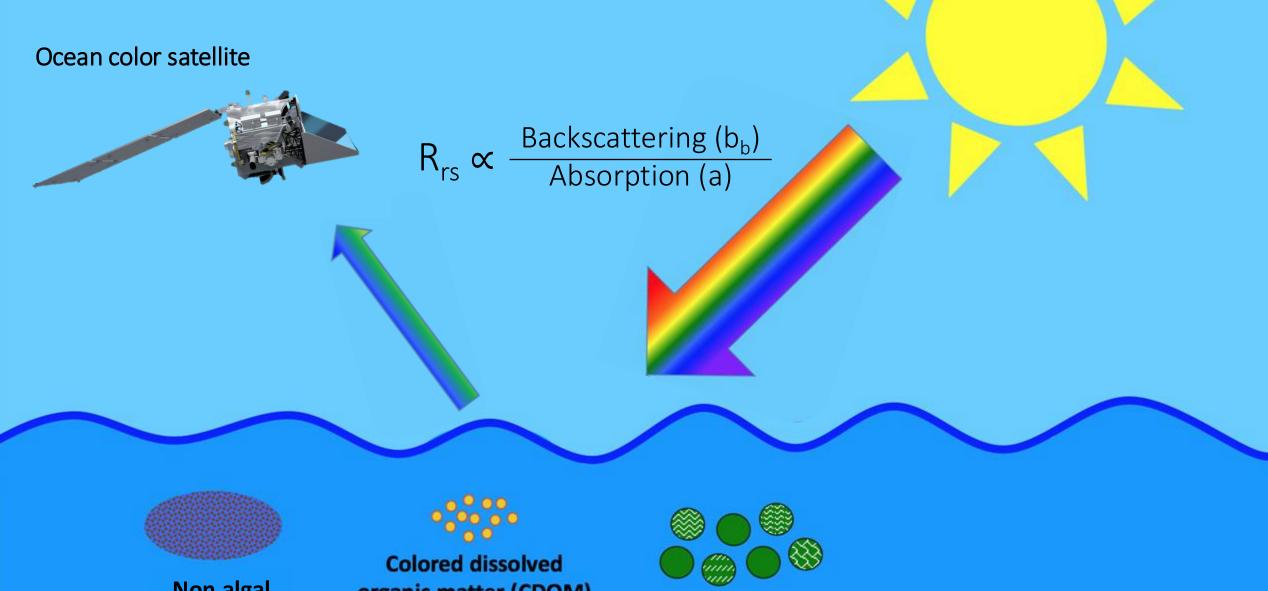
Non algal particles (NAP)



Colored dissolved organic matter (CDOM)



Phytoplankton



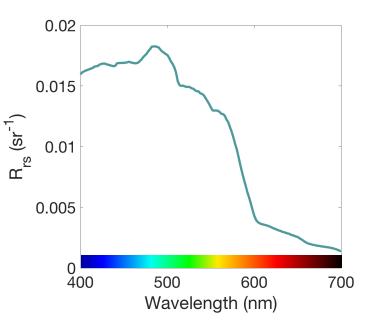
Non algal particles (NAP)





Phytoplankton

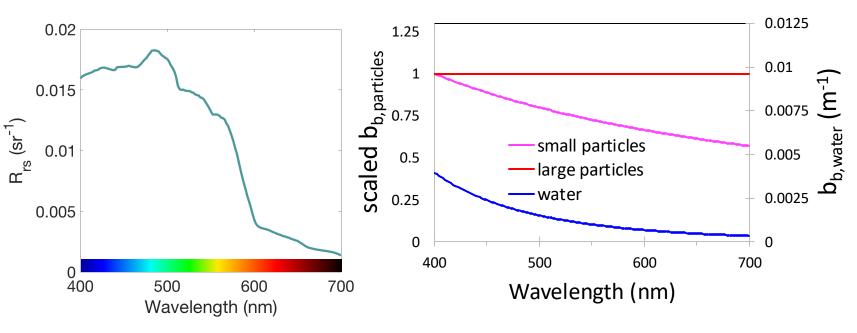
$$R_{rs}(\lambda) \approx \frac{b_b(\lambda)}{a(\lambda) + b_b(\lambda)}$$



AOP and IOP spectra from Roesler and Perry 1995

$$R_{rs}(\lambda) \approx \frac{b_b(\lambda)}{a(\lambda) + b_b(\lambda)}$$

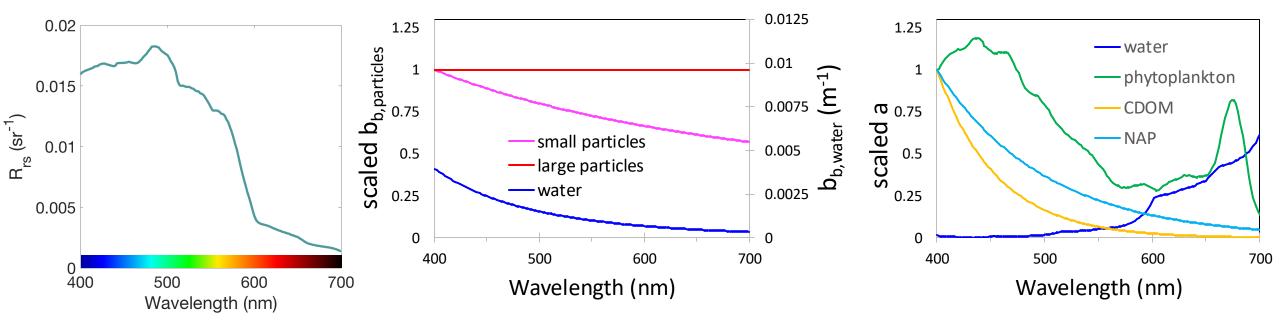
$$b_b = b_{b,water} + b_{b,large\ part.} + b_{b,small\ part.}$$



AOP and IOP spectra from Roesler and Perry 1995

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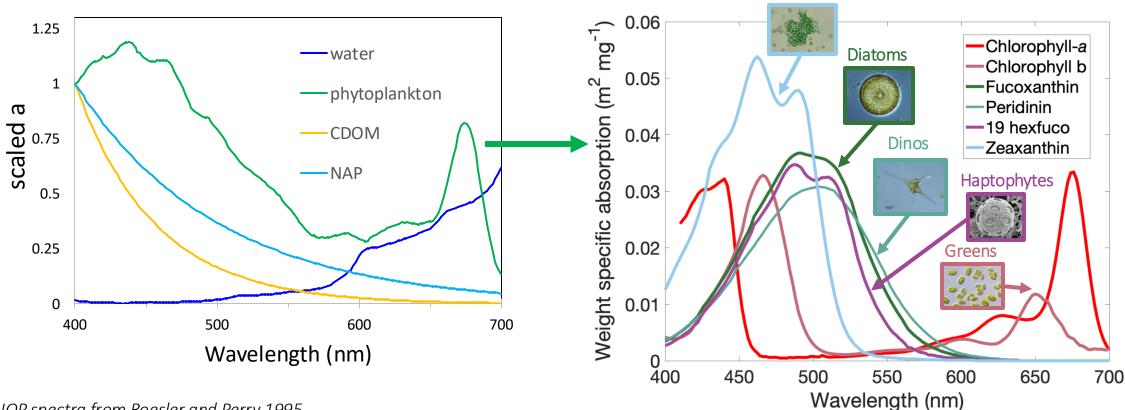


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Cyanos

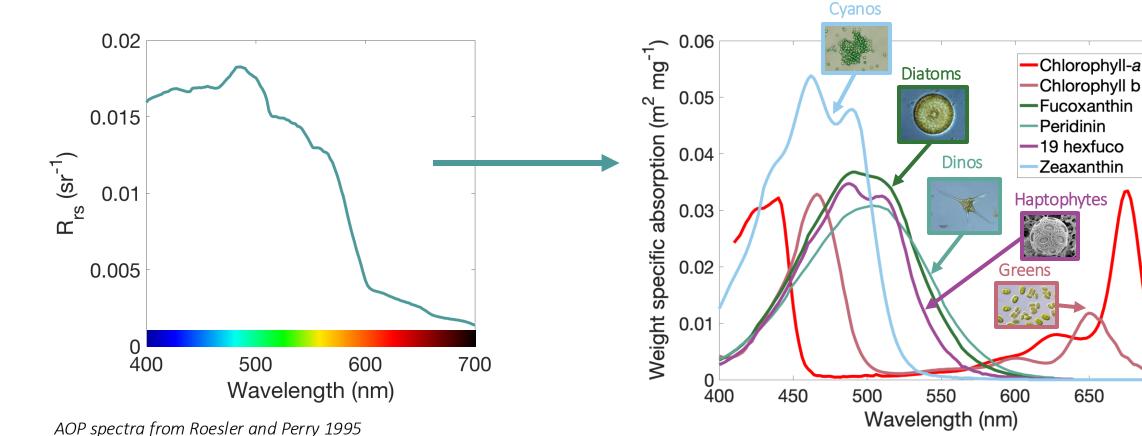


From reflectance to pigments via absorption

$$R_{rs}(\lambda) \approx \frac{b_b(\lambda)}{a(\lambda) + b_b(\lambda)}$$

$$R_{rs}(\lambda) \approx \frac{b_b(\lambda)}{a(\lambda) + b_b(\lambda)}$$
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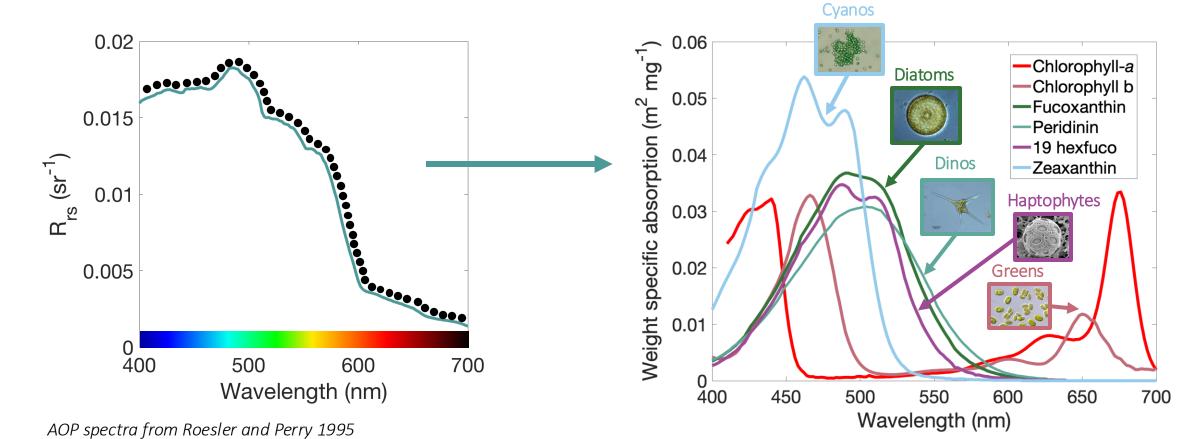
700



Maximize short (<50nm) spectral scale variability

$$R_{rs}(\lambda) \approx \frac{b_b(\lambda)}{a(\lambda) + b_b(\lambda)}$$

$$R_{rs}(\lambda) \approx \frac{b_b(\lambda)}{a(\lambda) + b_b(\lambda)}$$
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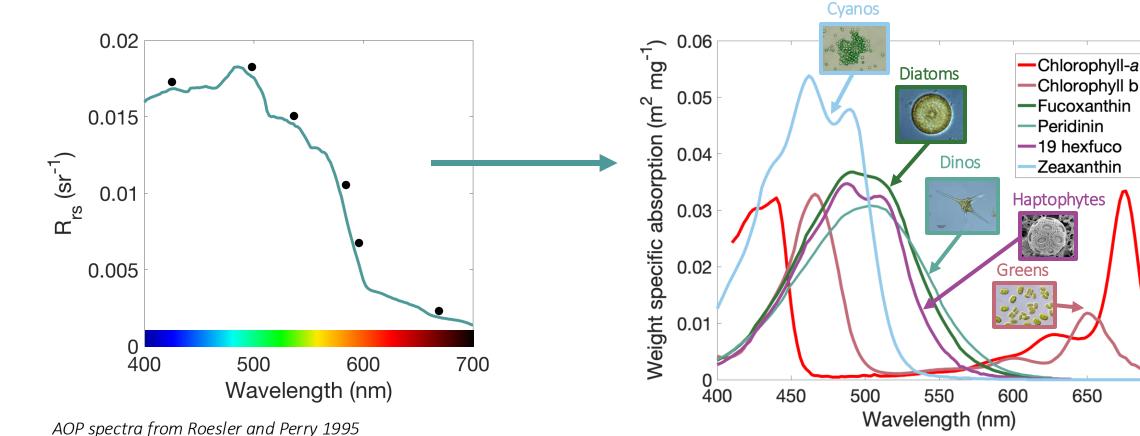


Past satellite resolution was insufficient...

$$R_{rs}(\lambda) \approx \frac{b_b(\lambda)}{a(\lambda) + b_b(\lambda)}$$

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 $b_b = b_{b,water} + b_{b,large\ part.} + b_{b,small\ part.}$ $a = a_{water} + a_{NAP} + a_{CDOM} + a_{ph}$

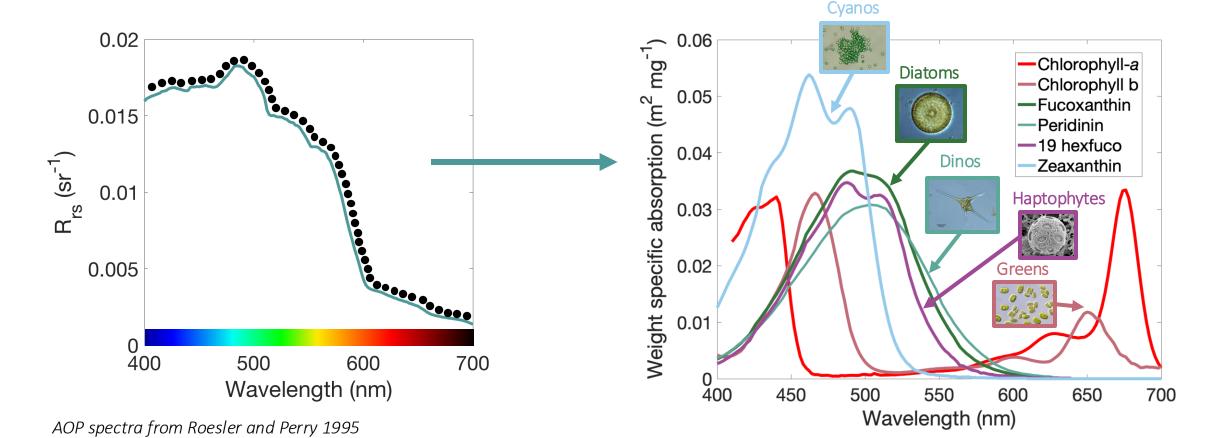
700



But PACE provides unprecedented resolution

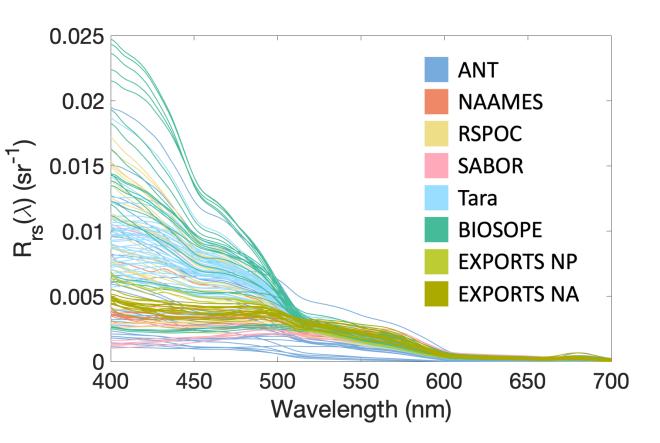
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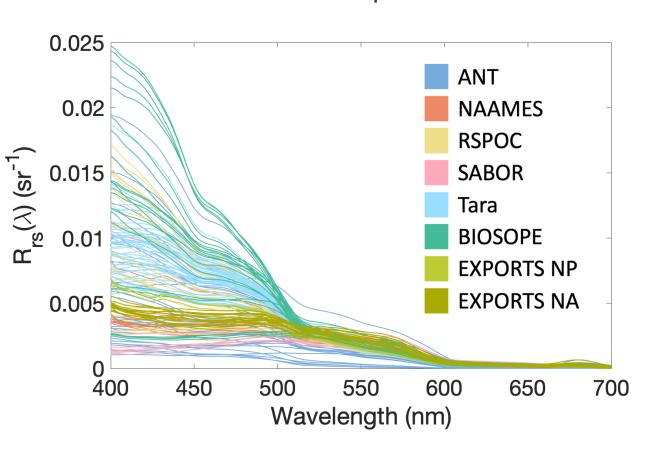
Implementing the SDP model

Measured spectra



Implementing the SDP model

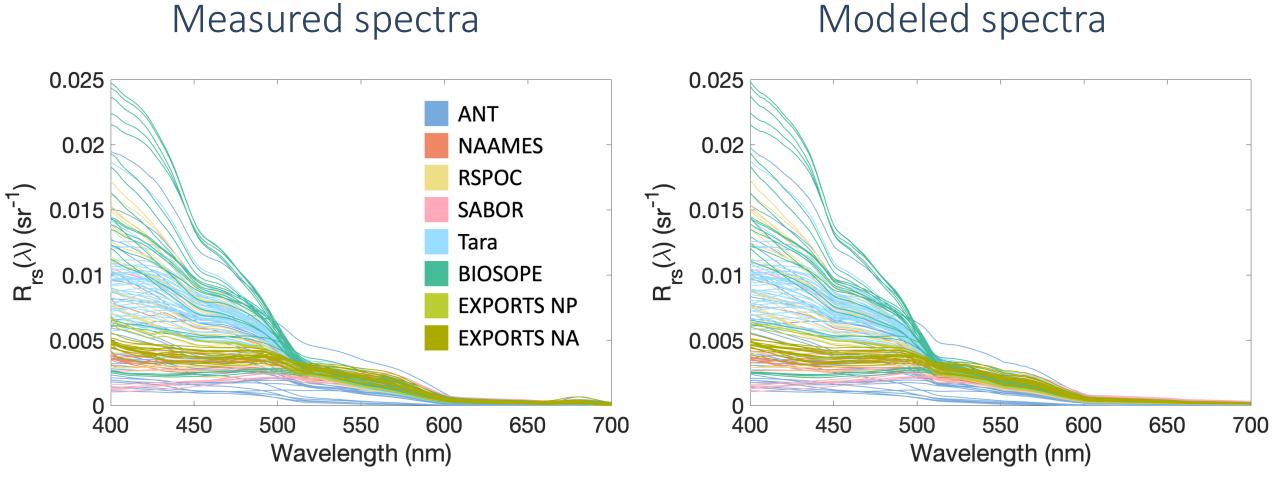
Measured spectra



Construct a generic hyperspectral model to reconstruct remote sensing reflectance:

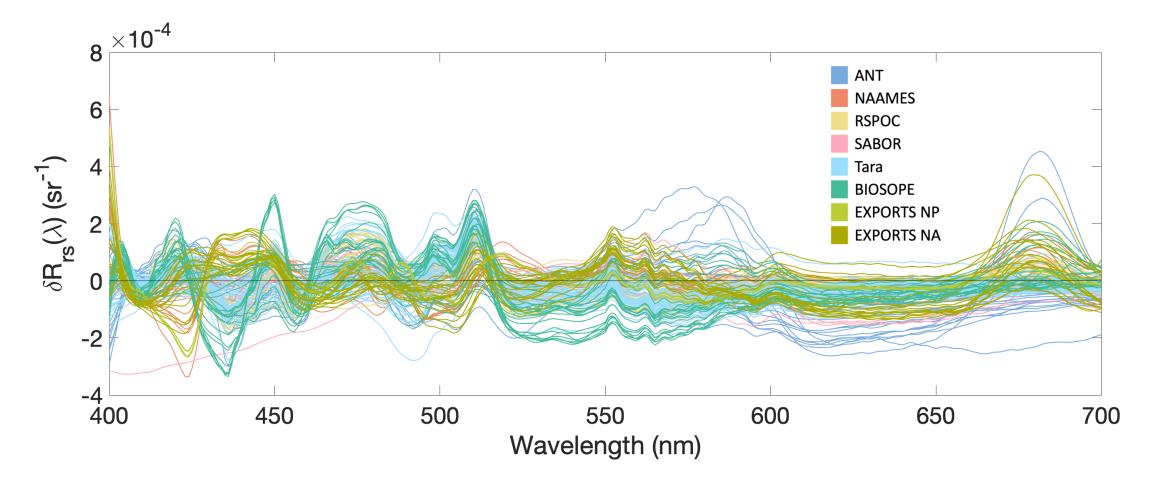
$$R_{rs,mod}(\lambda) = f(a,b_b)$$
 where $a=a_{ph}+a_{dg}+a_{water}$ and $b=b_{bp}+b_{bwater}$

Implementing the SDP model



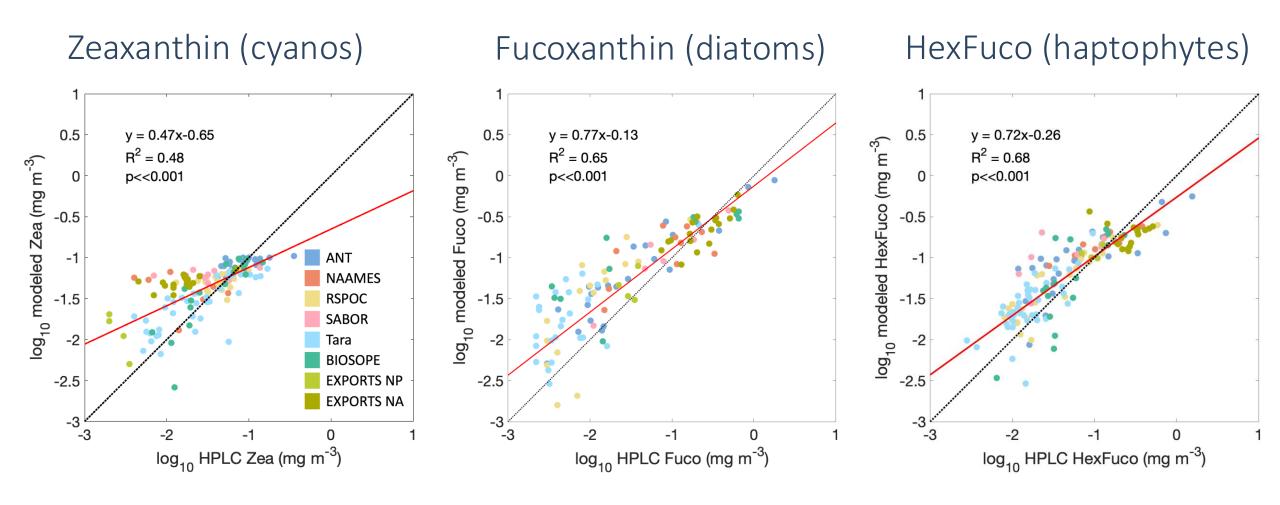
They should look identical if our assumptions were correct

R_{rs} residual $(\delta R_{rs}) = R_{rs,meas}(\lambda) - R_{rs,mod}(\lambda)$



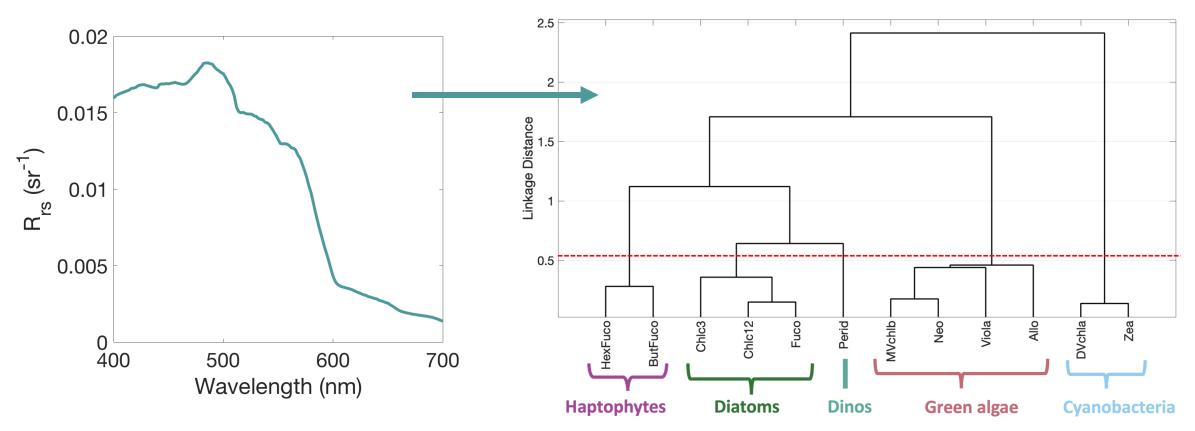
Use the reflectance residual (δR_{rs}) for further modeling...

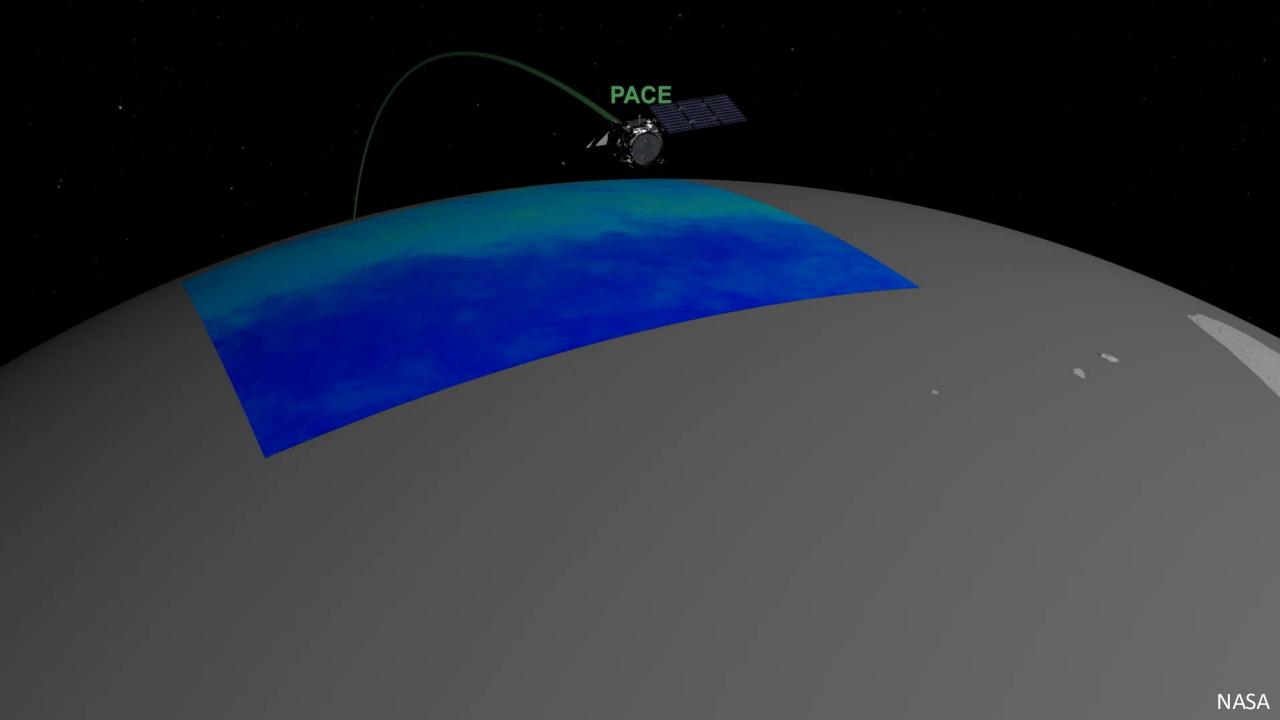
Modeled vs. measured pigments



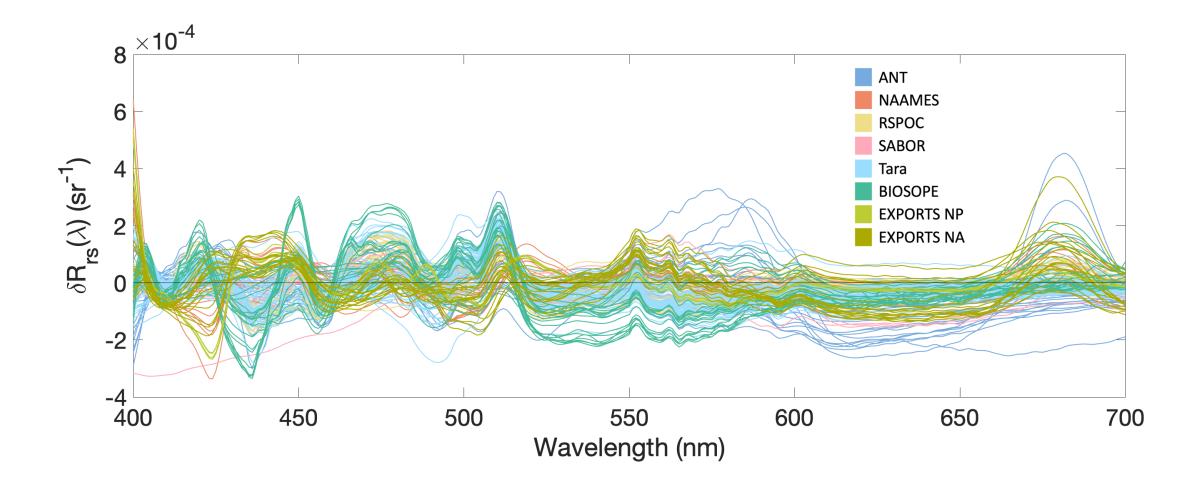
Going from ocean color to phytoplankton with PACE

Using the hyperspectral δR_{rs} data, we can model 13 phytoplankton pigment concentrations and distinct 5 phytoplankton groups.



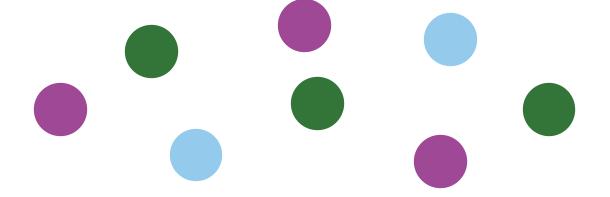


What else can we do with the R_{rs} residual (δR_{rs})?

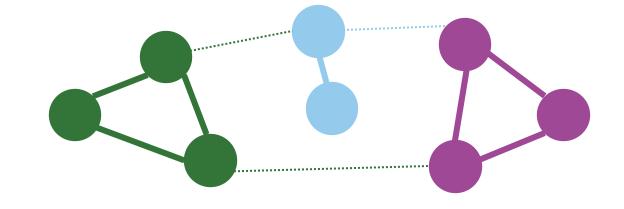


Network-based community detection analysis

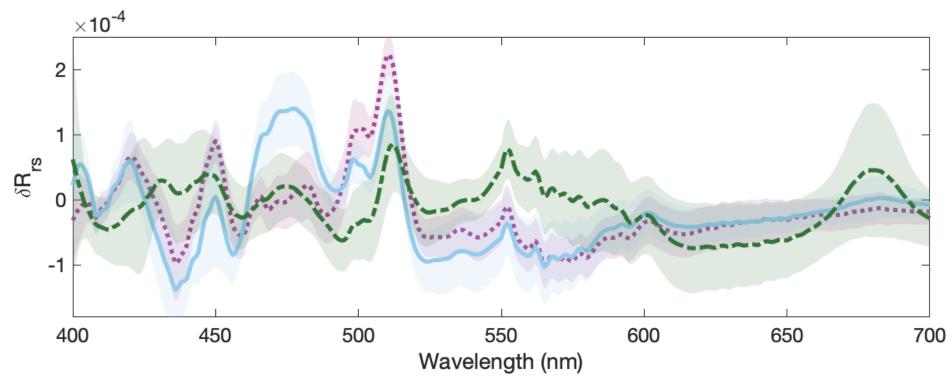
Assign each sample to a community based on its associated characteristics.



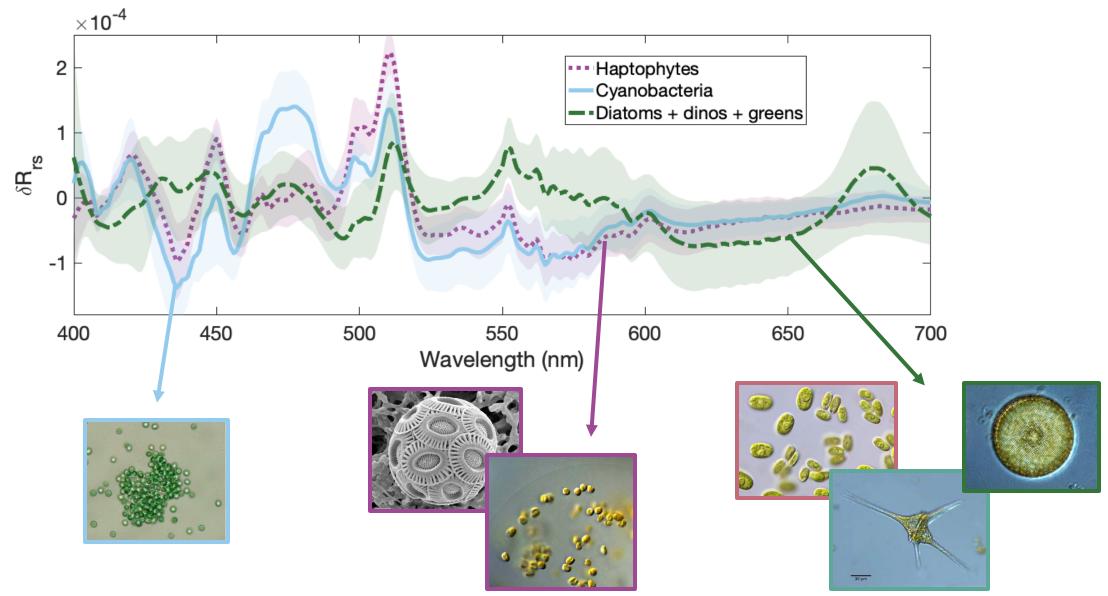
Form communities that maximize within-group connections and weaken between-group connections.



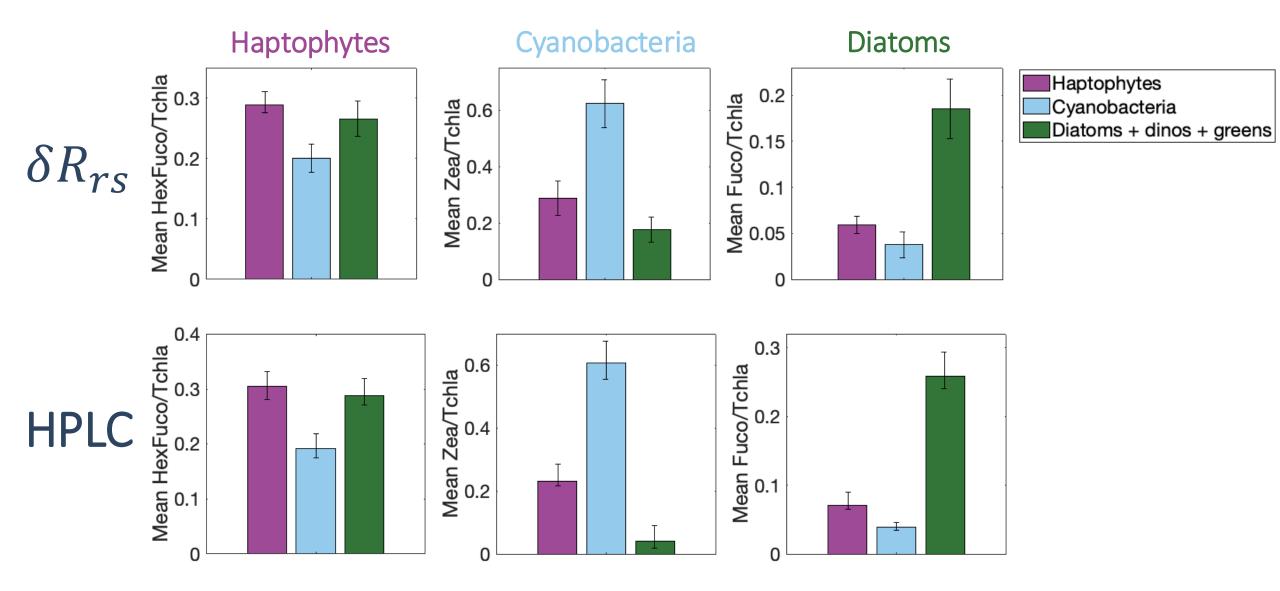
Community detection analysis: 3 δR_{rs} communities



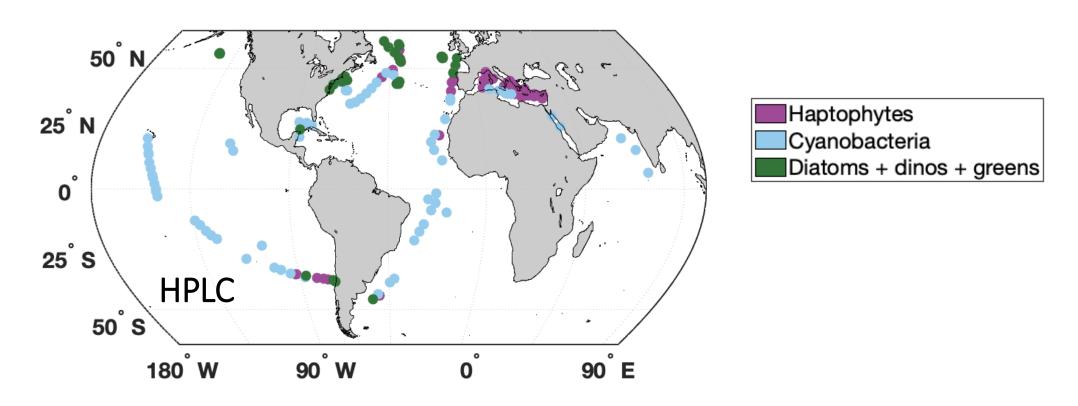
Community detection analysis: 3 δR_{rs} communities



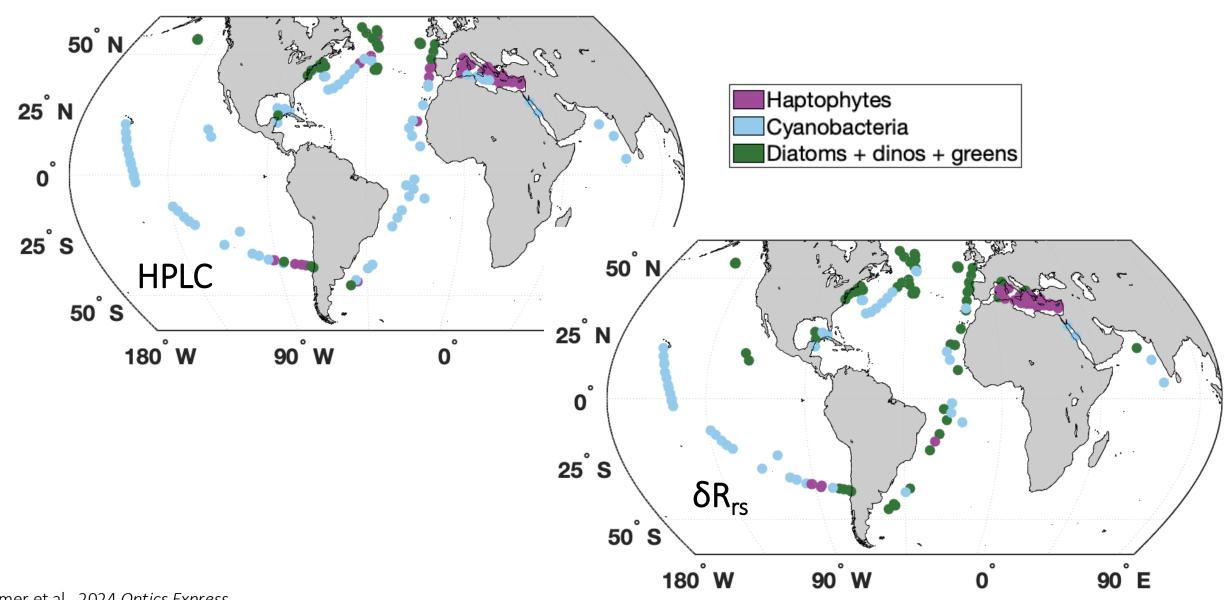
Three communities also separate from HPLC pigments

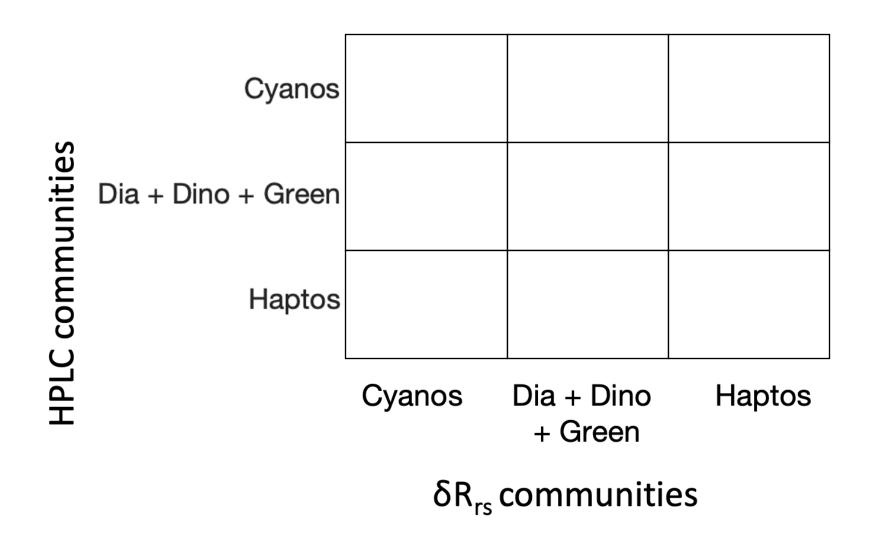


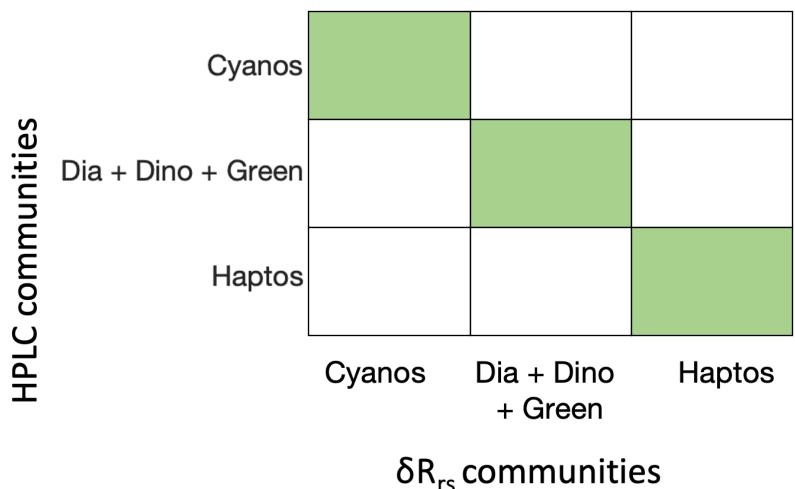
Global distribution of the three communities

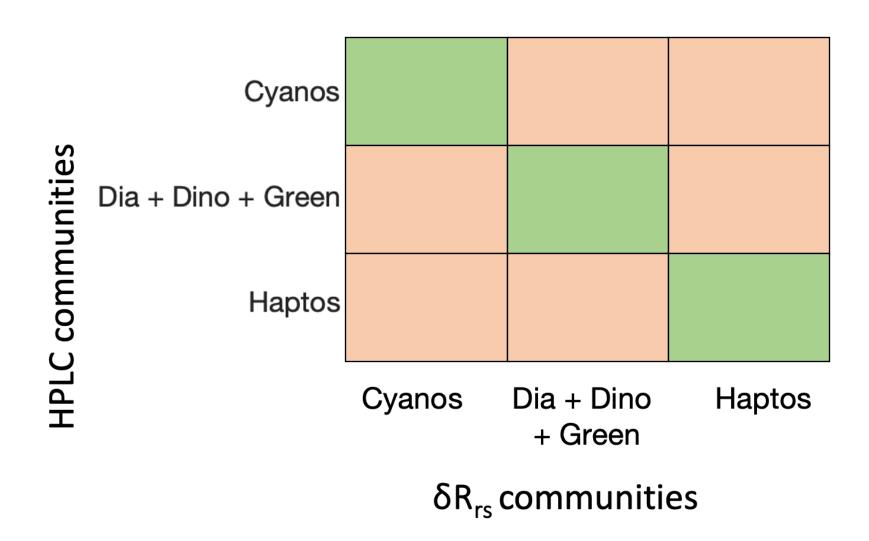


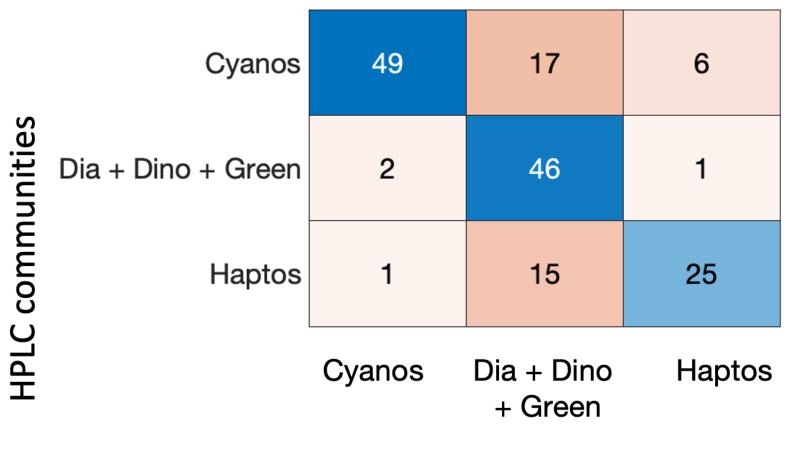
Global distribution of the three communities







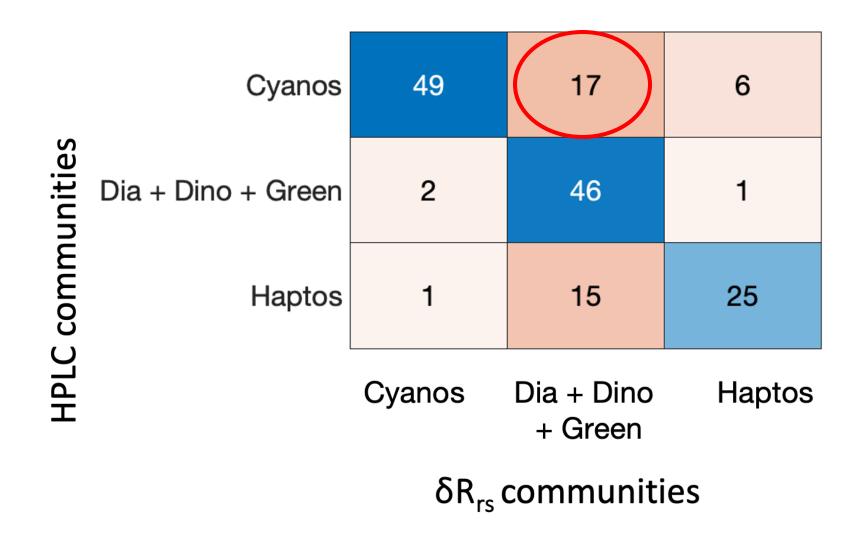




74% of samples correctly assigned (120 of 162)

 δR_{rs} communities

Can we use other methods to investigate "confused" samples?



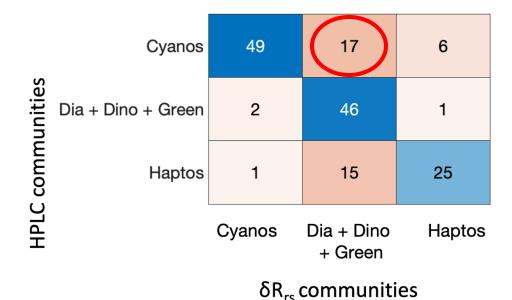
Methods with higher taxonomic resolution correct community assignment

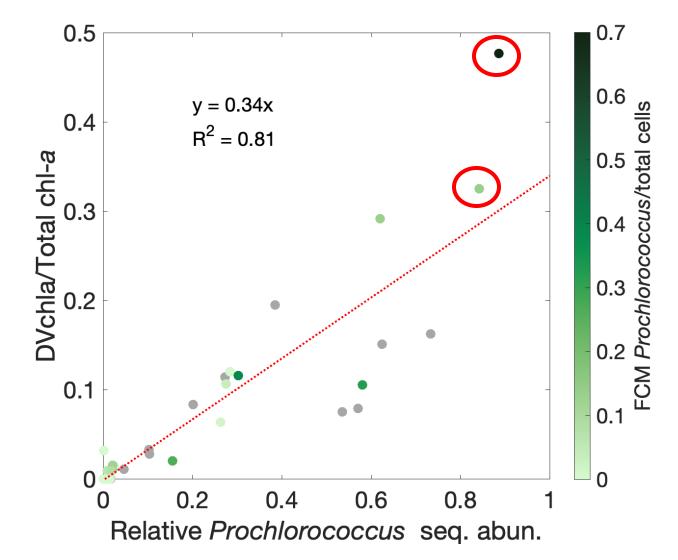
Kramer et al., 2024 Limnology & Oceanography: Methods

 δR_{rs} community: diatoms + dinos + greens

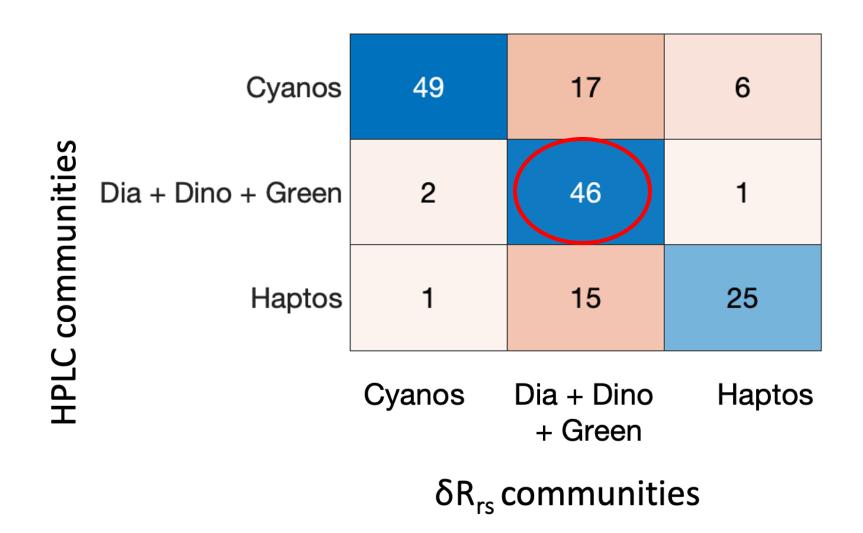
HPLC community: cyanos

16S rRNA + FCM: Prochlorococcus





Can we use other methods to investigate "confused" samples?

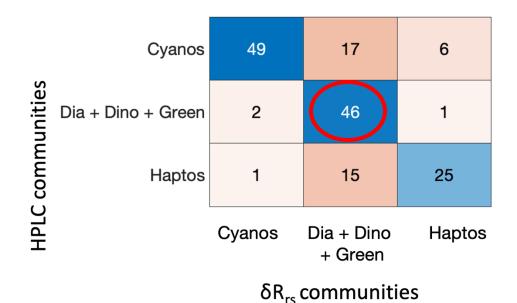


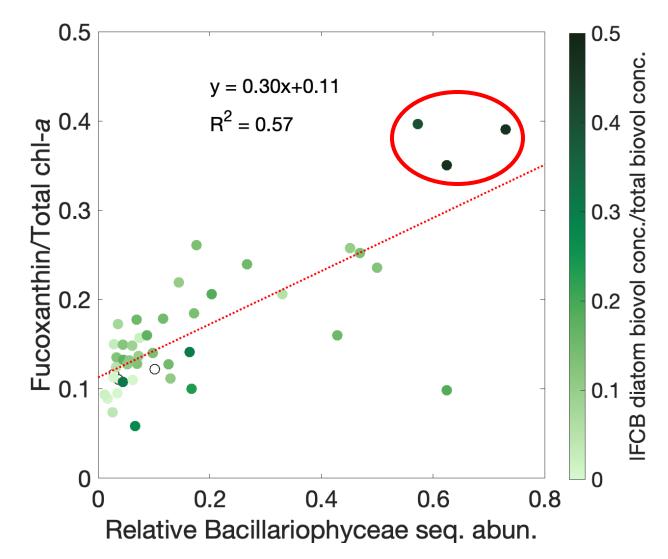
Methods with higher taxonomic resolution confirm community assignment

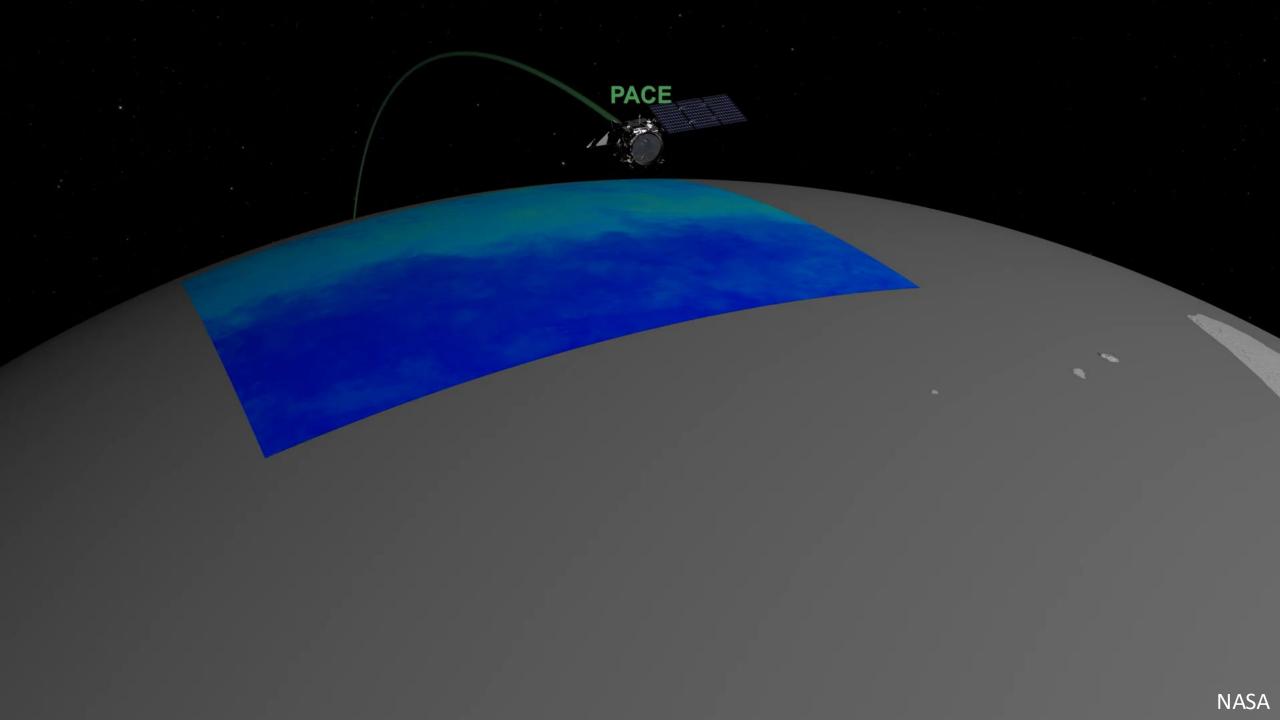
Kramer et al., 2024 Limnology & Oceanography: Methods

HPLC + δR_{rs} communities: diatoms + dinos + green algae

18S rRNA + IFCB: diatoms



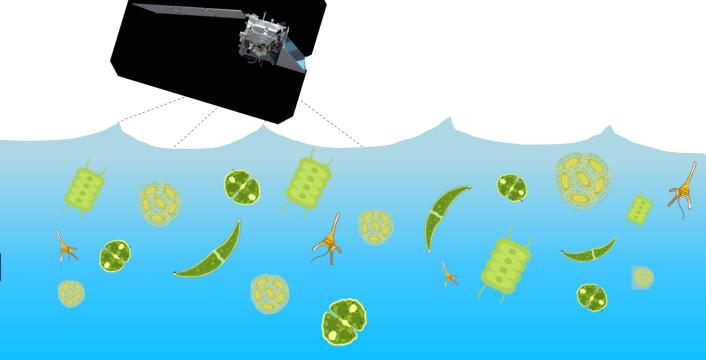




Next steps

SDP model is currently being implemented for PACE to model phytoplankton pigments

2) $\delta R_{rs}(\lambda)$ spectra will be available as a product from SDP and PACE: compare variability in space and time, compare communities from PVST HPLC.



Thanks and acknowledgements

Coauthors on work shown here: Dave Siegel, Jason Graff, Stéphane Maritorena, Jeremy Werdell, Ivona Cetinić, Dylan Catlett, Colleen Durkin, Erin Jones, Nicola Paul, Meg Estapa, Tatiana Rynearson, Alyson Santoro, Sebastian Sudek.

All researchers, technicians, captains, and crew who contributed to data collection, preparation, analysis, and submission (particular thanks to Ali Chase, Emmanuel Boss, Nils Haëntjens, Brian VerWey, Pat Kelly, Sean O'Neill, Melissa Omand, Ken Buesseler, and Elaine Steinberg).

The EXPORTS, NAAMES, and PACE science teams; the Carbon Flux Ecology lab at MBARI; BU Department of Earth & Environment.

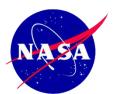






Funding sources for work shown here: National Defense Science and Engineering Graduate Fellowship, NASA Ocean Biology and Biogeochemistry, Simons Foundation Postdoctoral Fellowship in Marine Microbial Ecology, David and Lucile Packard Foundation.







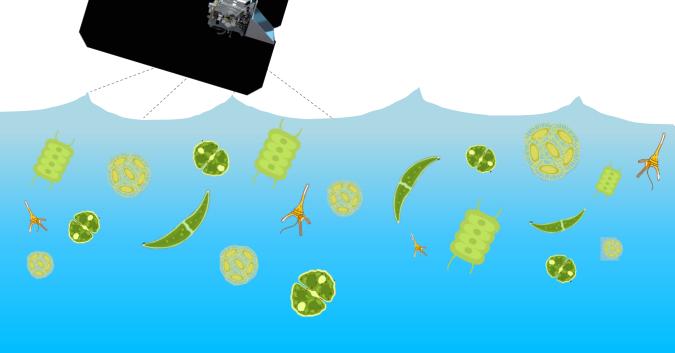


Questions?

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Phytoplankton pigments can be successfully modeled from short spectral scale variability in hyperspectral $R_{rs}(\lambda)$.

Communities from $\delta R_{rs}(\lambda)$ and from HPLC identify the same 3 pigment-based phytoplankton groups with good (74%) agreement across paired global samples.



Aquatic Cathartic documentary



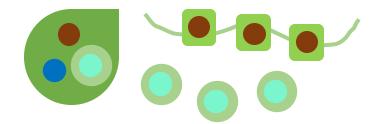
Migrations magazine



EXTRA SLIDES

Four major challenges for pigment-based PCC

1. Many phytoplankton groups share pigments, either due to evolution or mixotrophy.



3. Pigment concentration and composition can be affected by the environment (light, nutrients).



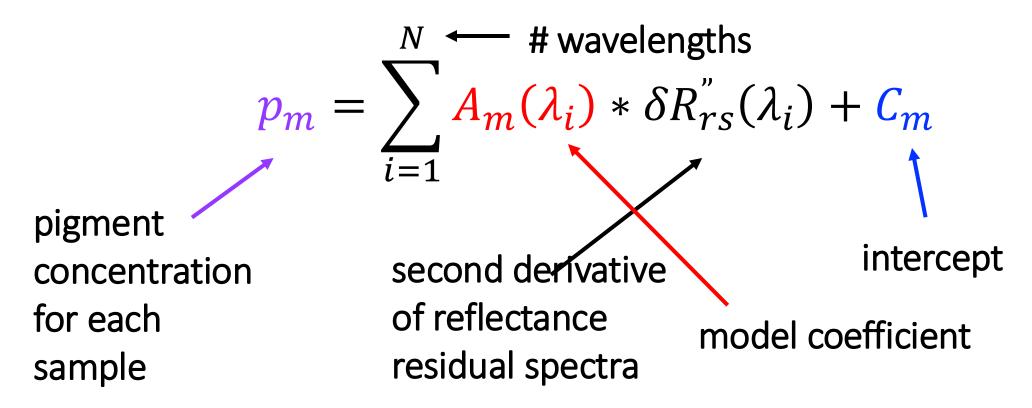
2. Within the same broad phytoplankton group, there can be variations in pigment composition.



4. Correlations between pigments and other methods may happen due to co-occurrence in nature.



Principal components regression model:



Repeat for all pigments. 100-fold cross validation used to test model. Use 75% of dataset for model training, 25% for testing.