



Remote sensing of highly concentrated phytoplankton blooms

**IOCCG SLS 2026
Pierre GERNEZ**

What causes a bloom? (open ocean)

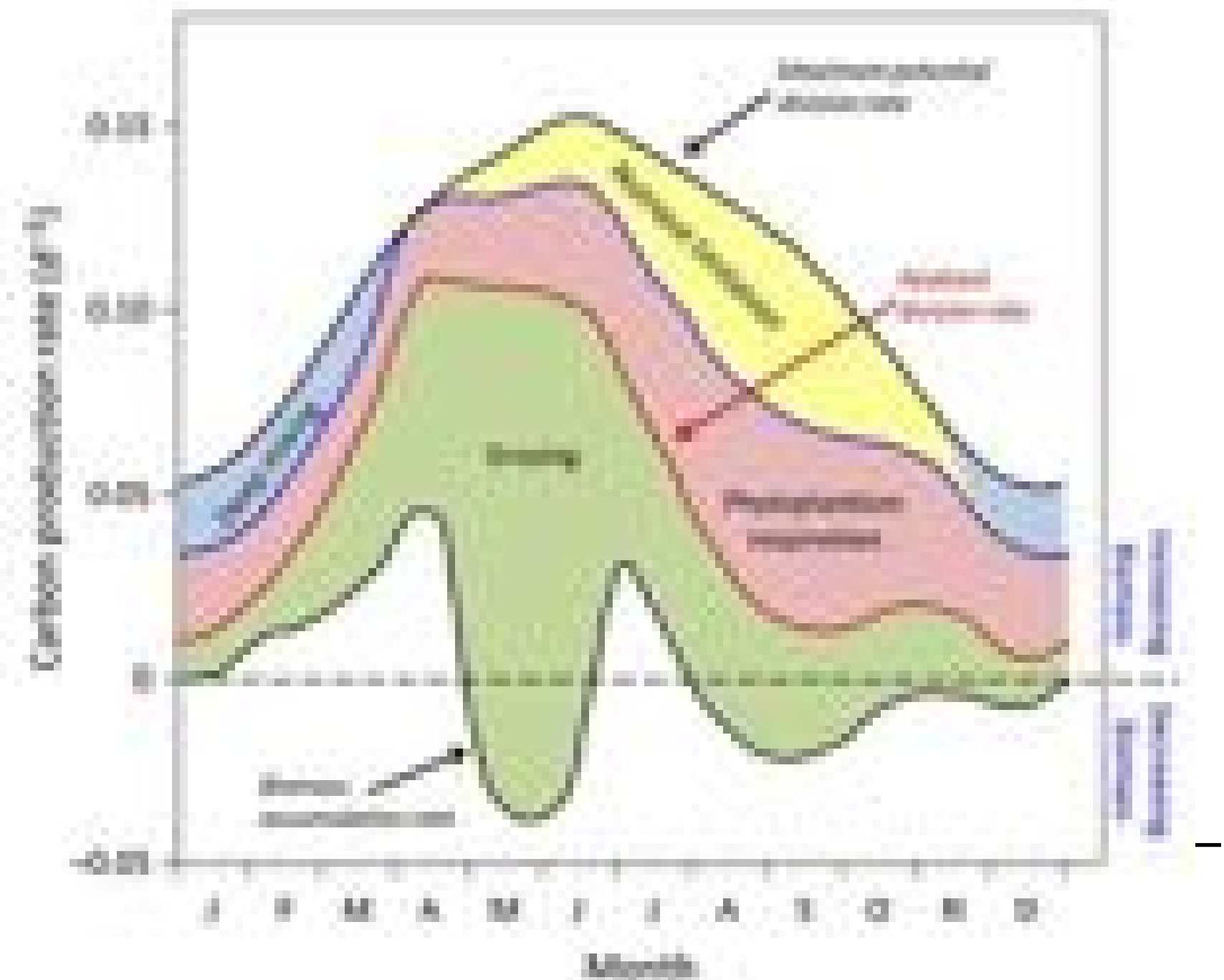


INVITED REVIEW Open Access

Student's tutorial on bloom hypotheses in the context of phytoplankton annual cycles

Michael J. Behrenfeld Emmanuel S. Boss

First published: 08 August 2017 | <https://doi.org/10.1111/gcb.13444>



What is a bloom? (coastal waters)



Article | [Free to Read](#)

What is a bloom? A commentary

[Theodore J. Smyda](#)

First published: 22 December 2003

Abstract

There are at least eight different modes and mechanisms by which harmful phytoplankton species can cause mortality, physiological impairment, or other negative in situ effects. Their distinction from nonharmful phytoplankton taxa is clearly warranted. Increasing use of bloom descriptors such as "exceptional," "unusual," "nuisance," and subjective reference to specific occurrences as blooms reveal widespread confusion concerning: what is a bloom?, how is it to be defined?, what is harmful?, and what distinguishes a harmful bloom from other blooms? Such efforts have been influenced by comparison with perceived spring diatom bloom characteristics. Examples, selected from among harmful bloom events, are presented to support the view that the classical focus on the spring diatom bloom has significant conceptual and operational biases as to what constitutes a bloom. This and the lax criteria used to define a bloom compromise



Highly concentrated phytoplankton blooms aka red tides

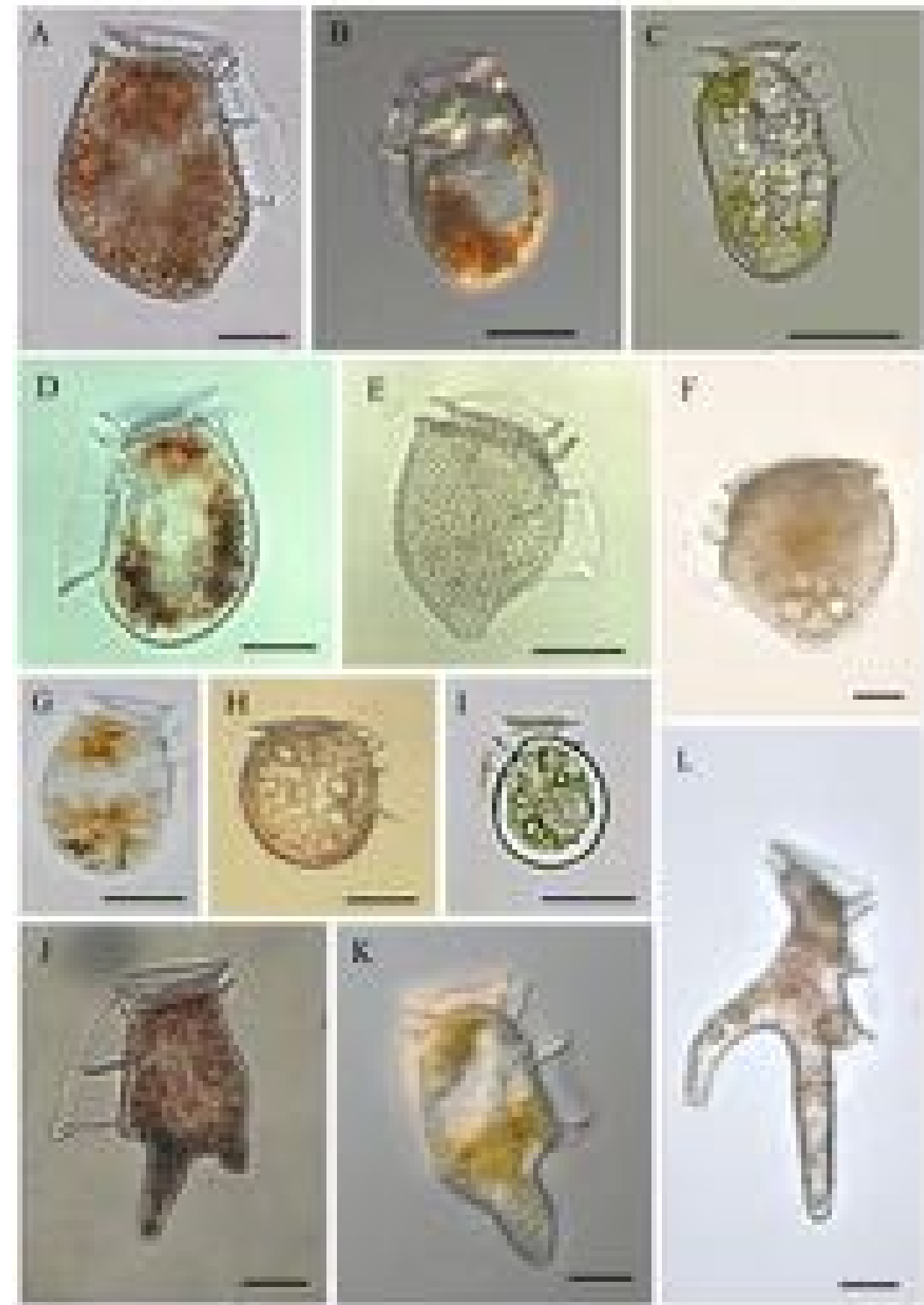
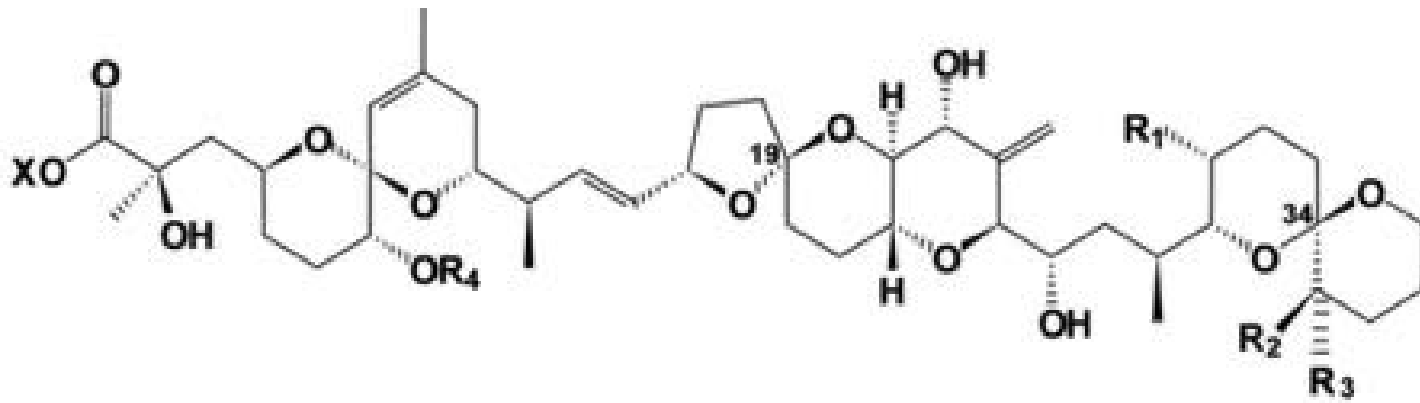
sometimes: harmful algal blooms (HABs)

- **Directly toxic**
- **Oxygen depletion during bloom decay**
- **Accumulation of ammonium inside cells**

Low-concentration HABs

Some species can produce high amount of toxins at low cell abundance

***Dinophysis* sp.** (~ 100 cells / L)
Diarrhetic Shellfish Poisoning (DSP) toxins
Okadaic acids (Reguera et al., 2014)



Several species of the *Dinophysis* genera are toxic

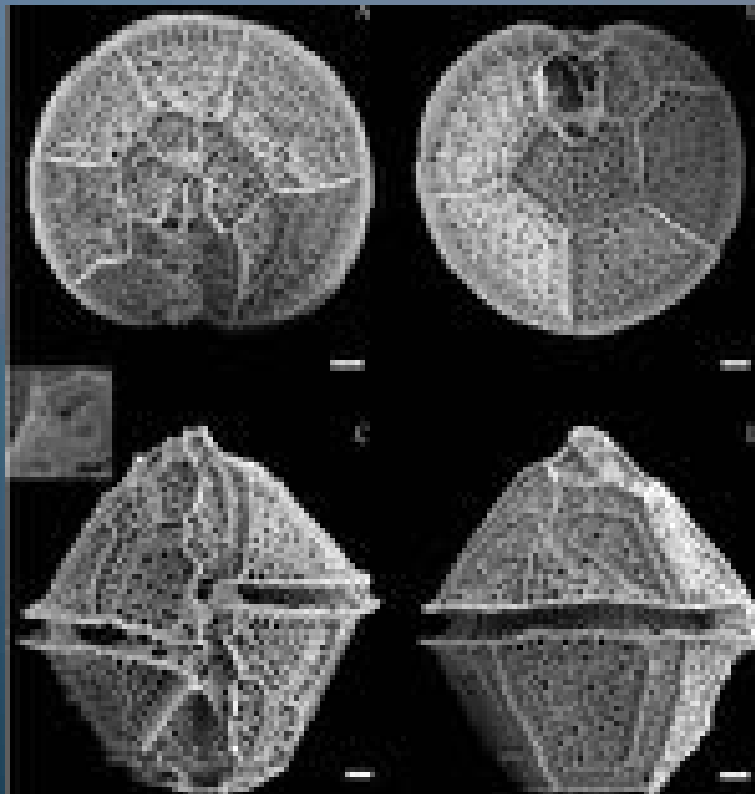
Example of yessotoxins (YTX) producers

Main YTX producers

Lingulodinium polyedra

Protoceratium reticulatum

Gonyaulax spinifera



Mertens et al., (2023)

L. polyedra cell concentration up to 3,600,000 cells/L...
... but no toxicity recorded in shellfish

YTX were produced by *P. ceratium* (low cell concentration)

269 known harmful phytoplankton species



214 known toxic and 55 non-toxic (but still harmful) species

- 55% dinoflagellates
- 20% diatoms
- 16% cyanobacteria

Be careful with taxonomy...



Lingulodinium polyedra ? Not any more! ***Lingulaulax polyedra***

★ <i>Lingulodinium polyedra</i> (F.Stein) J.D.Dodge, 1989	
AphiaID	233592 (urn:lsid:marinespecies.org:taxname:233592)
Classification	Biota > ★ Chromista (Kingdom) > ★ Harosa (Subkingdom) > ★ Alveolata (Infrakingdom) > ★ Myzozoa (Phylum) > ★ Dinozoa (Subphylum) > ★ Dinoflagellata (Infraphylum) > ★ Dinophyceae (Class) > ★ Peridiniophyceae (Subclass) > ★ Gonyaulacales (Order) > ★ Lingulodiniaceae (Family) > ★ <i>Lingulodinium</i> (Genus) > ★ <i>Lingulodinium polyedra</i> (Species)
Status	✖ unaccepted > superseded combination
Accepted name	★ <i>Lingulaulax polyedra</i> (F.Stein) M.J.Head, K.N.Mertens & R.A.Fensome, 2024

★ <i>Lingulaulax polyedra</i> (F.Stein) M.J.Head, K.N.Mertens & R.A.Fensome, 2024	
AphiaID	1778370 (urn:lsid:marinespecies.org:taxname:1778370)
Classification	Biota > ★ Chromista (Kingdom) > ★ Harosa (Subkingdom) > ★ Alveolata (Infrakingdom) > ★ Myzozoa (Phylum) > ★ Dinozoa (Subphylum) > ★ Dinoflagellata (Infraphylum) > ★ Dinophyceae (Class) > ★ Peridiniophyceae (Subclass) > ★ Gonyaulacales (Order) > ★ Lingulodiniaceae (Family) > ★ <i>Lingulaulax</i> (Genus) > ★ <i>Lingulaulax polyedra</i> (Species)
Status	accepted
Rank	Species
Type taxon of	★ <i>Lingulaulax</i> M.J.Head, K.N.Mertens & R.A.Fensome, 2024
Parent	★ <i>Lingulaulax</i> M.J.Head, K.N.Mertens & R.A.Fensome, 2024
Original name	★ <i>Gonyaulax polyedra</i> F.Stein, 1983
Synonymised names	★ <i>Gonyaulax polyedra</i> F.Stein, 1983 - unaccepted > superseded combination ★ <i>Myrionosphæridium macrocephalum</i> Deflandre & Cockson, 1955 [- unaccepted > junior subjective synonym ★ <i>Lingulodinium polyedra</i> (F.Stein) J.D.Dodge, 1989 - unaccepted > superseded combination



Red tides

not always red

not always harmful

always extremely
concentrated



Simplified IOP model

L. polyedra

95% abundance microphyto

Optical variability dominated by phyto

$$a(\lambda) \sim a_w(\lambda) + a_{\text{phy}}(\lambda) + a_{\text{cdom}}(\lambda) + a_{\text{nap}}(\lambda)$$

Phytoplankton dominated by 1-2 species

$$a_{\text{phy}}(\lambda) \sim a_{\text{phy},1}(\lambda) + a_{\text{phy},2}(\lambda) + a_{\text{phy},3}(\lambda) \\ + a_{\text{phy},4}(\lambda) + a_{\text{phy},5}(\lambda) + \dots$$

Two main challenges

To what extent is it possible to identify the main bloom-forming species from space in the absence of *in situ* observations?

Do we know the optical properties of all bloom-forming species?

Spectral libraries



The USGS Spectral Library contains reflectance spectra, including samples of minerals, rocks, soils, physically constructed as well as mathematically computed mixtures, plants, vegetation communities, microorganisms, and man-made materials. The samples and spectra collected were assembled for the purpose of using spectral features for the remote detection of these and similar materials. The latest version is an update to the previously published version (2007) that is the leading database used by the global scientific community to identify materials by their "spectral fingerprints." The library is fundamental for Mineral Resources Program projects to characterize mineral resources with imaging spectrometer data.



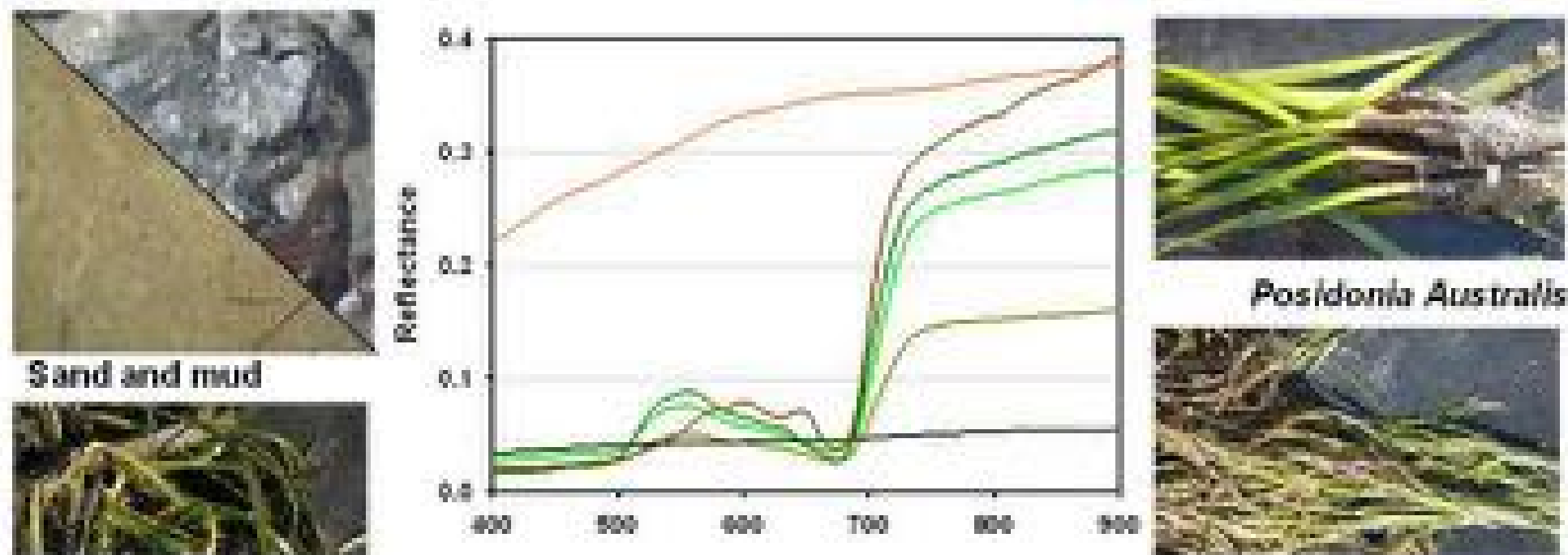
The National Spectral Database (NSD) houses best-practice spectral data collected by Australian remote sensing scientists from multiple locations and institutions across the country. As the collection grows the NSD will service continental scale research for the Earth observation community and support application such as the validation of satellite-based measurements of surface reflectance.

The National Spectral Database targets as diverse as soils, vegetation, waterbodies, coral, macro to micro algae, seagrasses and land surfaces.



Spectral libraries for shallow water RS

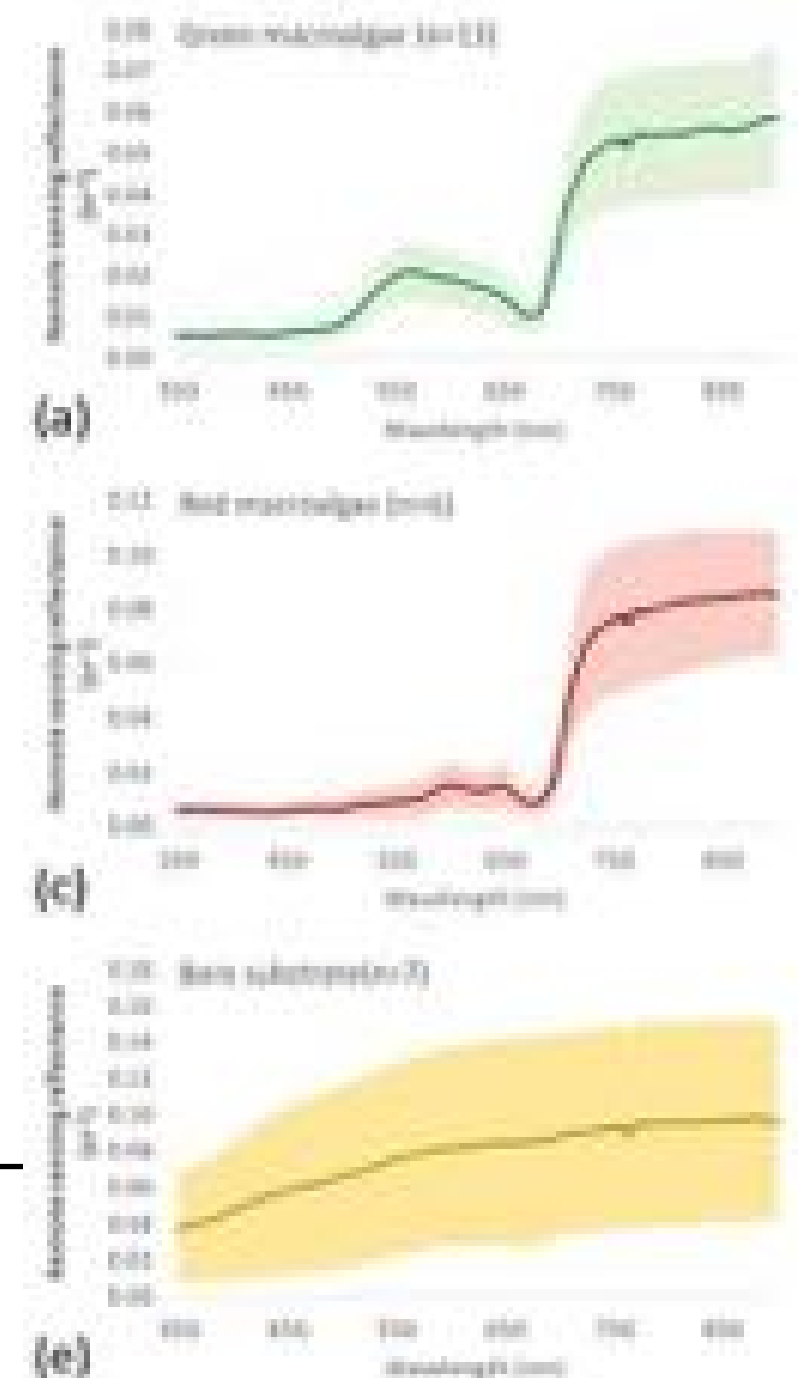
Roelfsema, Christiaan M; Phinn, Stuart R; Joyce, Karen (2016): Spectral reflectance library of algal, seagrass and substrate types in Moreton Bay, Australia [dataset]. PANGAEA, [doi https://doi.org/10.1594/PANGAEA.864310](https://doi.org/10.1594/PANGAEA.864310)



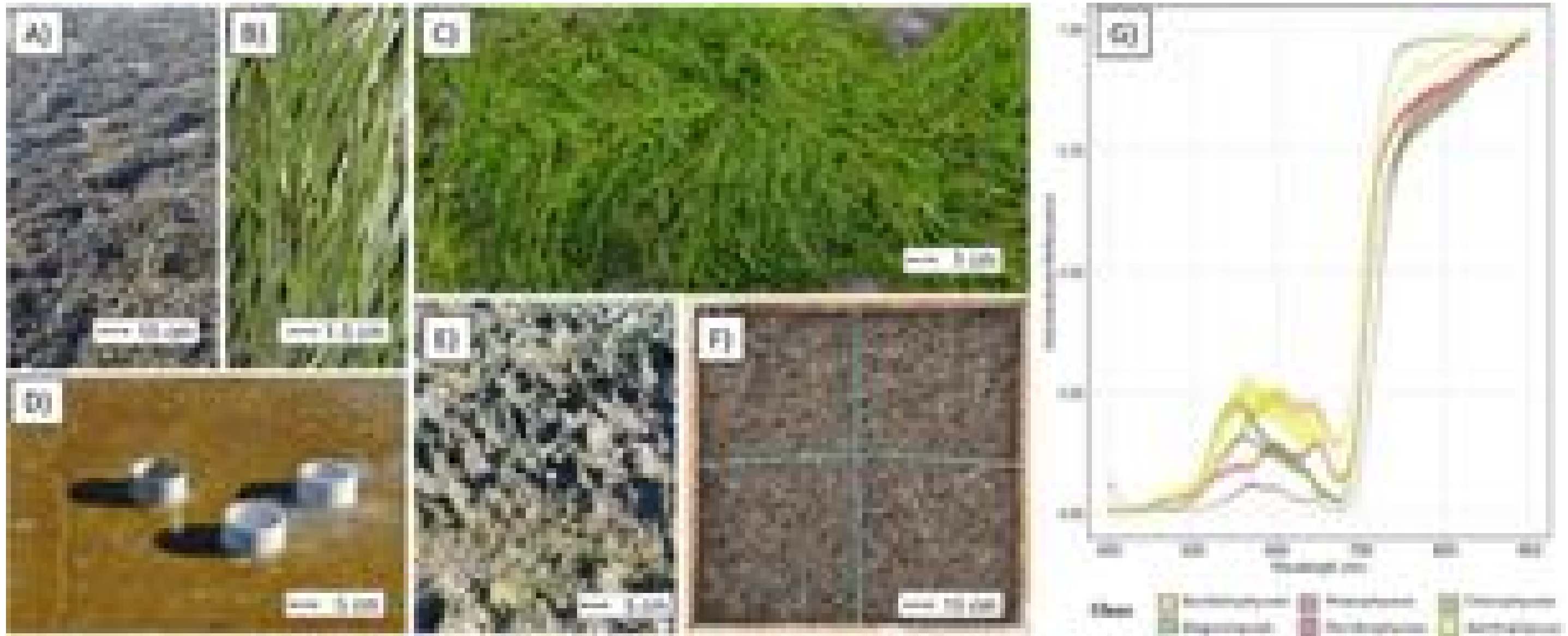
Earth System Science Data

Hyperspectral library of submerged aquatic vegetation and benthic substrates in the Baltic Sea

Ele Vahtmäe ✉, Laura Argus, Kaire Toming, Martin Ligi, and Tiit Kutser



Spectral library for intertidal vegetation



Davies et al., (2023) and Davies et al., (under review)

Phytoplankton spectral library??

Very complex as it requires to

- document the IOPs and AOPs of many phytoplankton species
- take into account the effects of changes in phytoplankton concentration
- take into account all the other optical constituents

Several approaches

- lab measurements of IOPs using cultures, combined with $R_{rs}(\lambda)$ simulations
- in situ measurements of IOPs and AOPs during very intense blooms
- satellite measurements of $R_{rs}(\lambda)$ during very intense blooms

In situ measurements of phytoplankton bio-optical properties during red tides

Very challenging to sample

Unpredictable events, difficult to anticipate

Patchy events, short duration

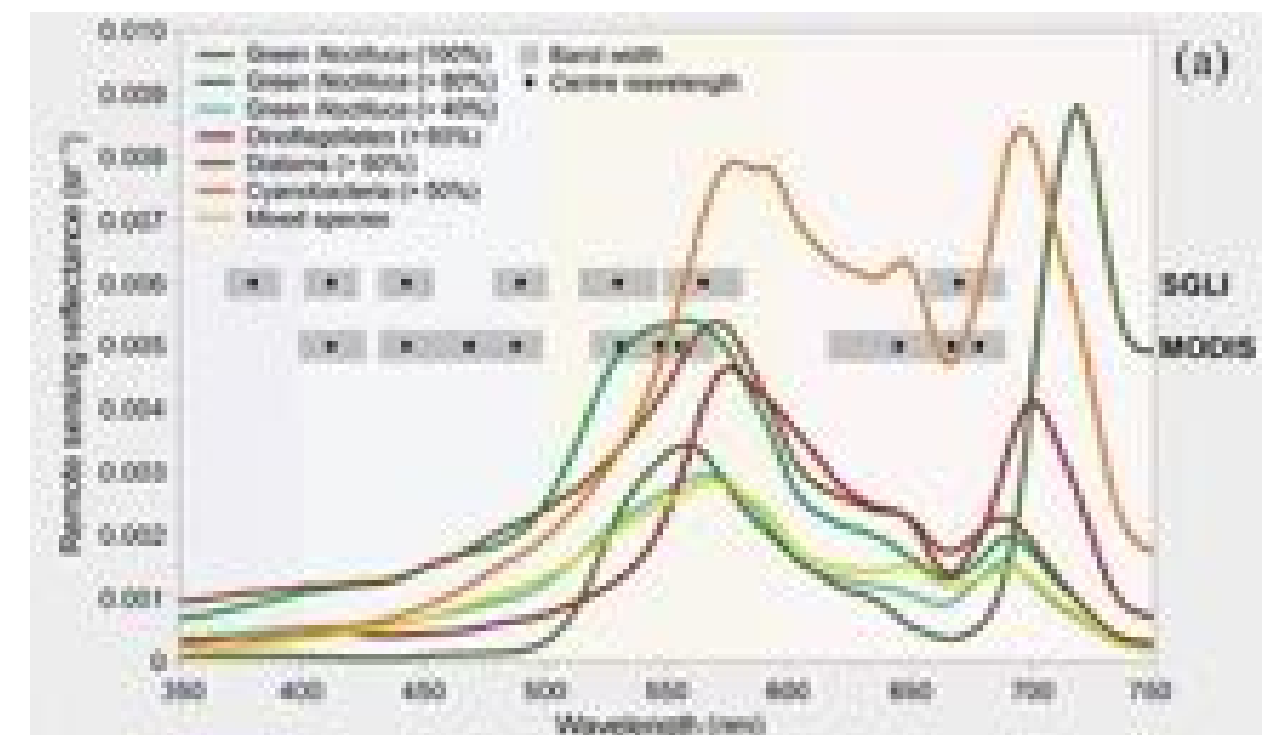
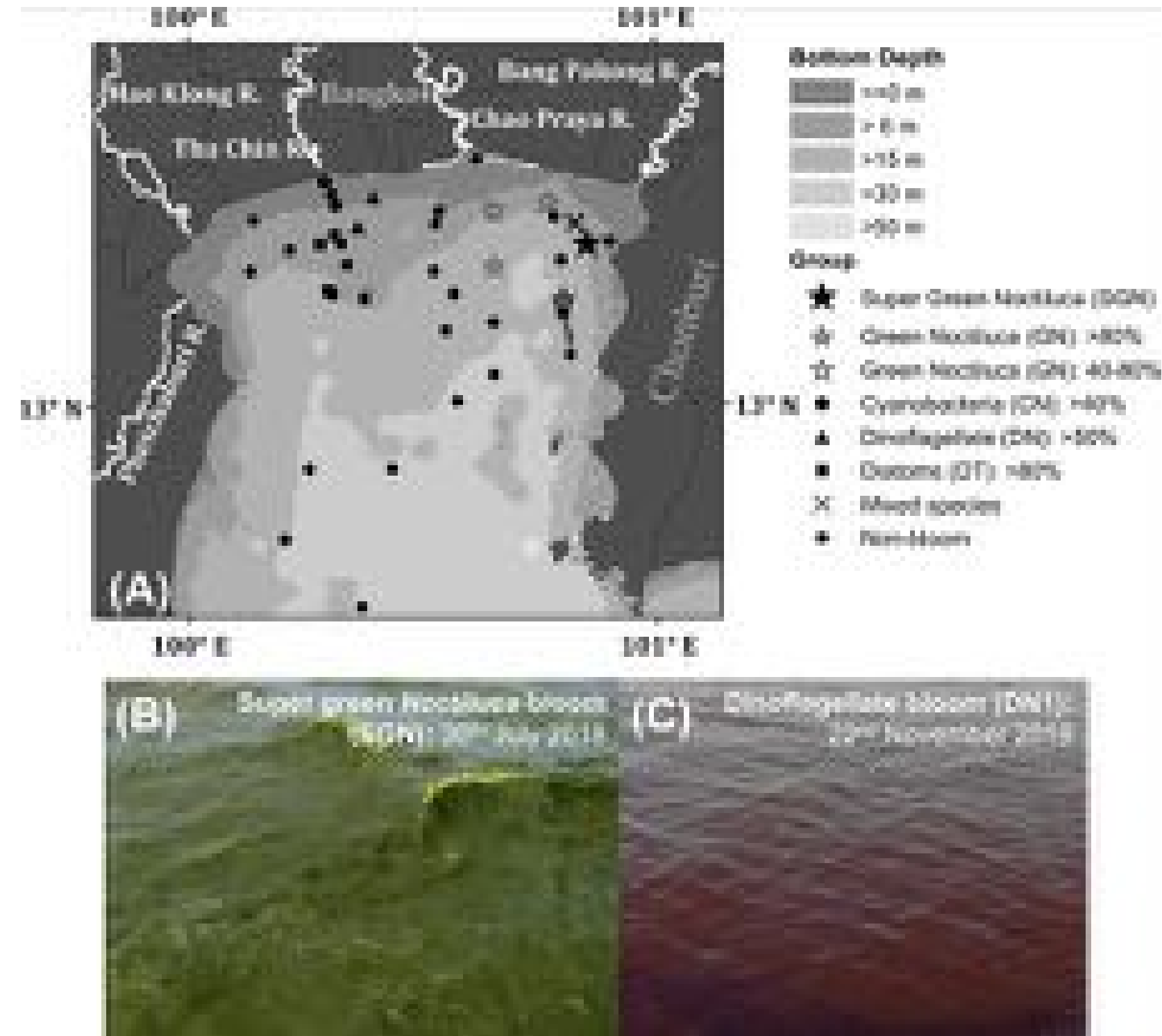
Large ships not ideal to sample relatively small patches

More in situ data needed!

IOPs and AOPs

pigments, cytometry, microscope observations and DNA (taxonomy)

Luang-on et al., (2023, 2024)



Satellite approaches

In search of floating algae and other organisms in global oceans and lakes

Lin Qi ^{a b c}, Chuanmin Hu ^c  , Karlis Mikelsons ^{d e}, Menghua Wang ^d, Veronica Lance ^{d f}, Shaojie Sun ^{a b c}, Brian B. Barnes ^c, Jun Zhao ^{a b},
Dimitry Van der Zande ^g

Floating algae observed with VIIRS and S3/OLCI

- Macroalgae: several *Sargassum* and *Ulva* species
- Cyanobacteria scums: *Trichodesmium*, *Microcystis*, *Nodularia*
- Microalgae: red and green *Noctiluca*
- Other floating stuff: pumice raft, oil slick, jellyfish blooms?, shrimp eggs?

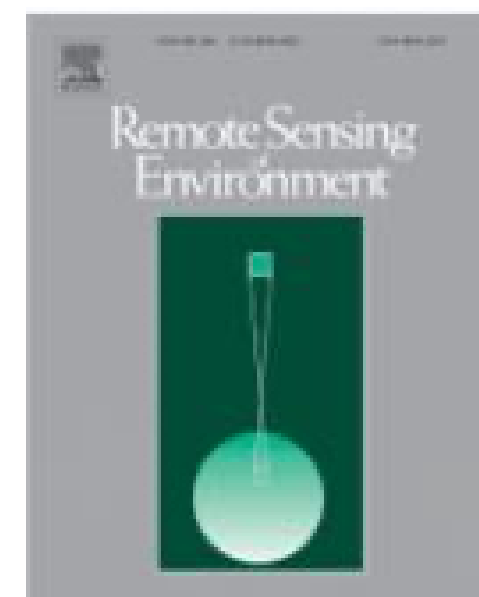
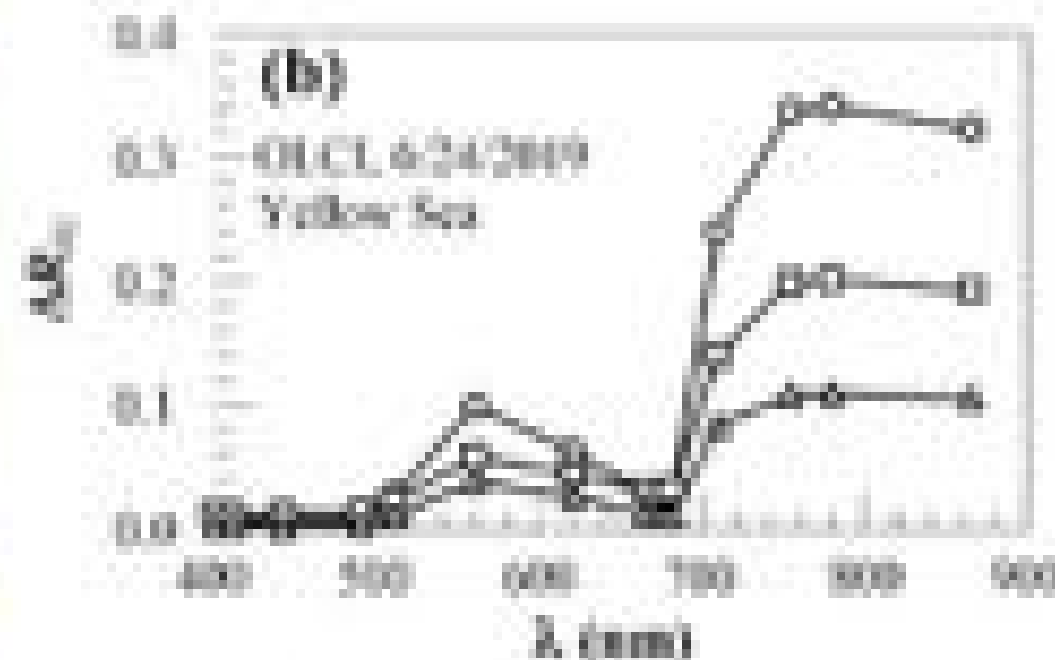
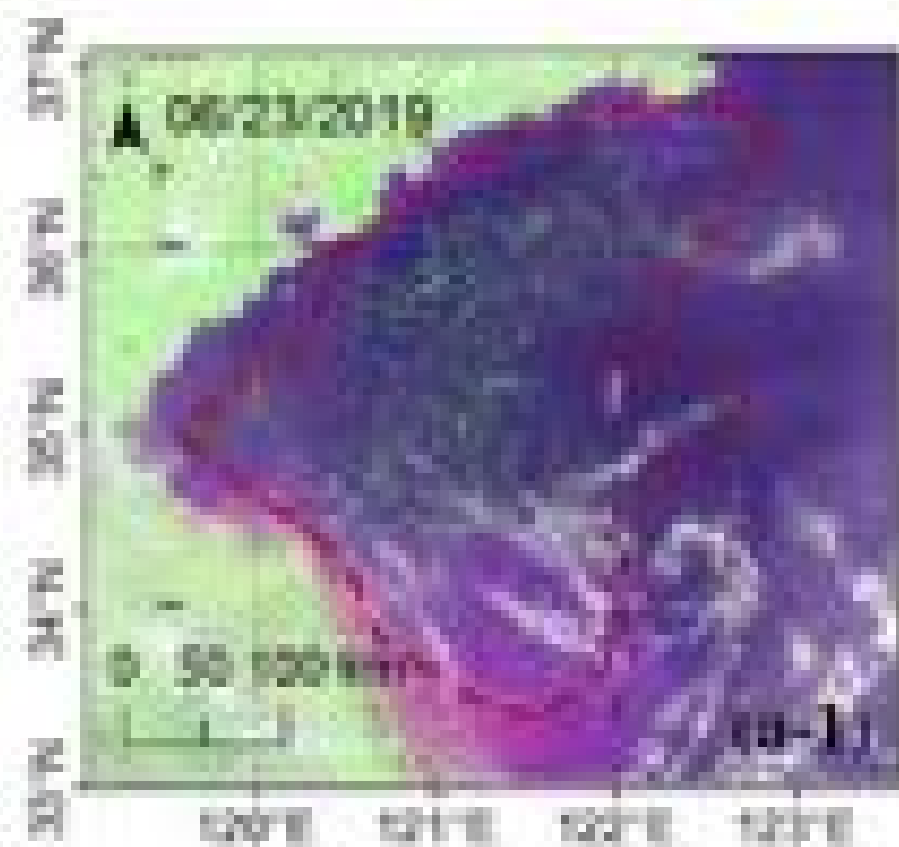


Fig. 1. (a) VIIRS FOCI image shows an Ulva bloom in the western Yellow Sea (Box 3 in Fig. 1 & Table 1). (b) Typical OLCI $R_{rs}(\lambda)$ of several pixels selected from the greenish image features.

High spatial resolution required



If too coarse spatial resolution

→ bloom “diluted” with non-bloom pixels

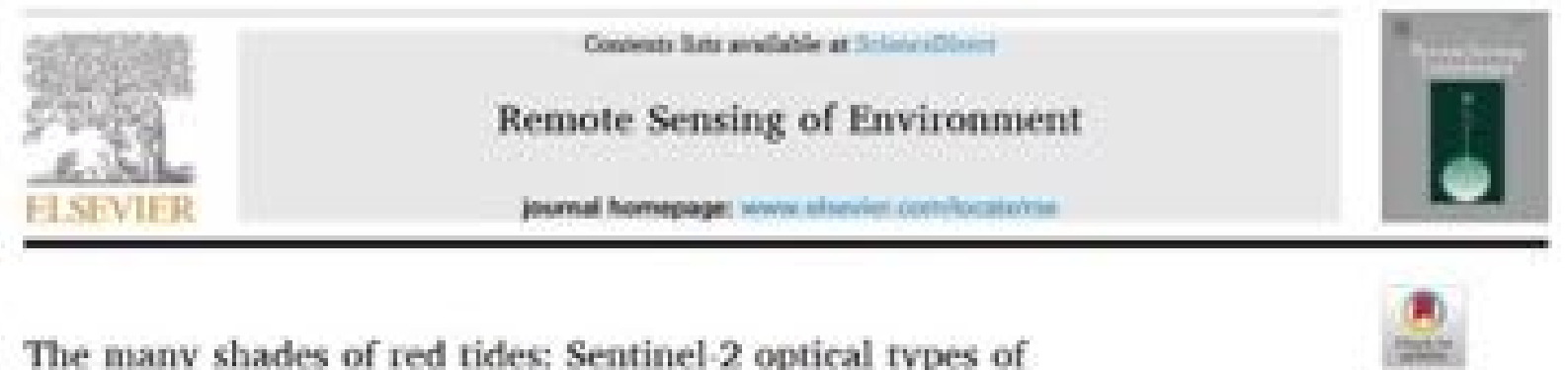
→ optical fingerprint is not “pure”



- ✓ Sentinel-2
- ✓ Landsat-8/9
- ✓ PRISMA
- ✓ EnMAP

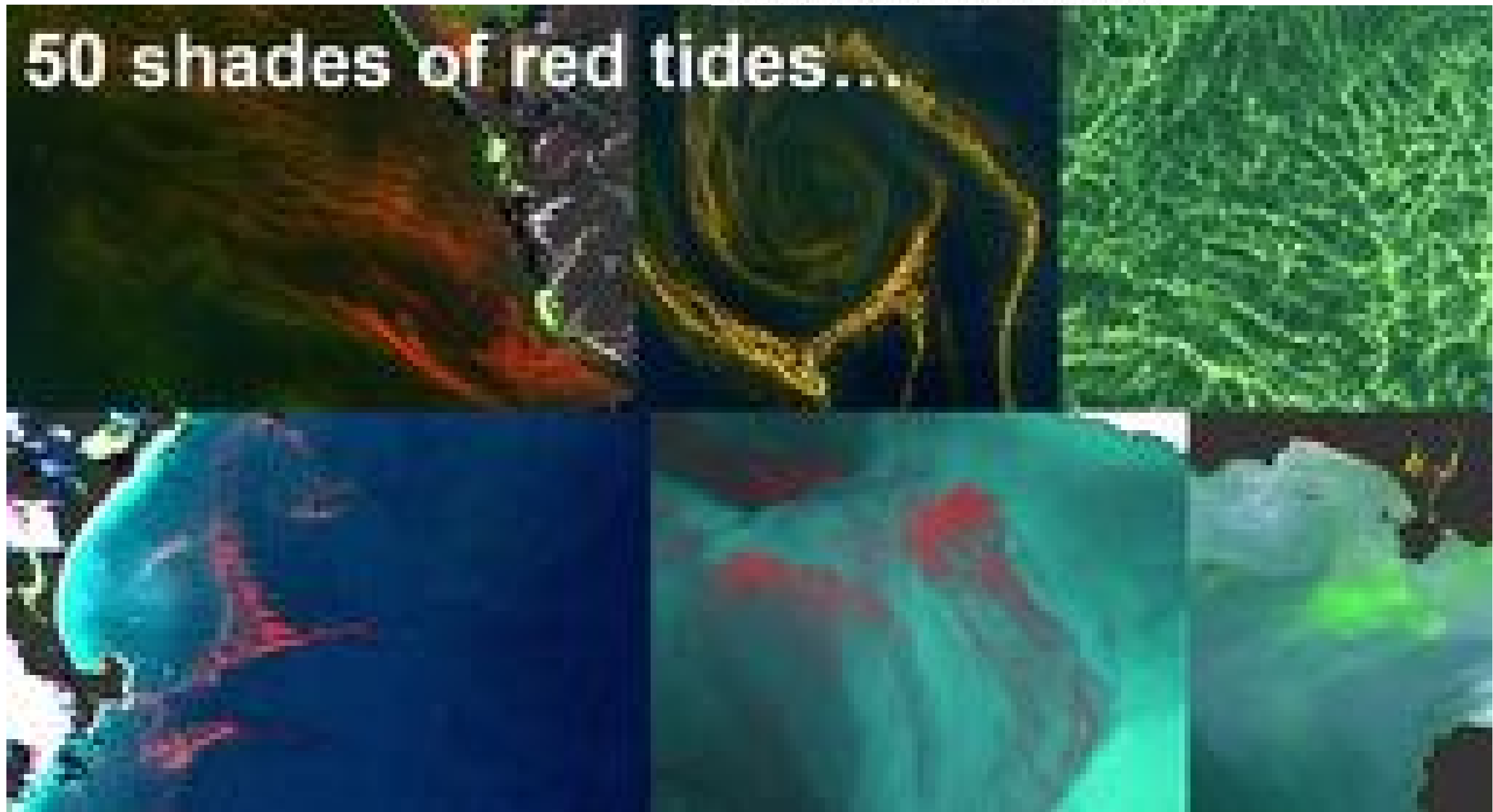
- × Sentinel-3
- × MODIS
- × VIIRS
- × PACE

S2 spectral library



The many shades of red tides: Sentinel-2 optical types of highly-concentrated harmful algal blooms

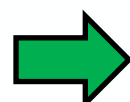
Pierre Gerez^{a,*}, Maria Laura Zoffoli^b, Thomas Lacour^c, Tania Hernández Parillas^d, Gabriel Navarro^e, Isabel Caballero^e, Tristan Harmel^f



Searched bloom events in harmful algae event database (**HAEDAT**) from 2015 – 2021
(304 blooms with known causative species)

Also searched in French phytoplankton and phycotoxin monitoring network (**REPHY**)
Criterion: $> 10^5$ cells / L of a single dominant genus / species ($> 85\%$ diversity)

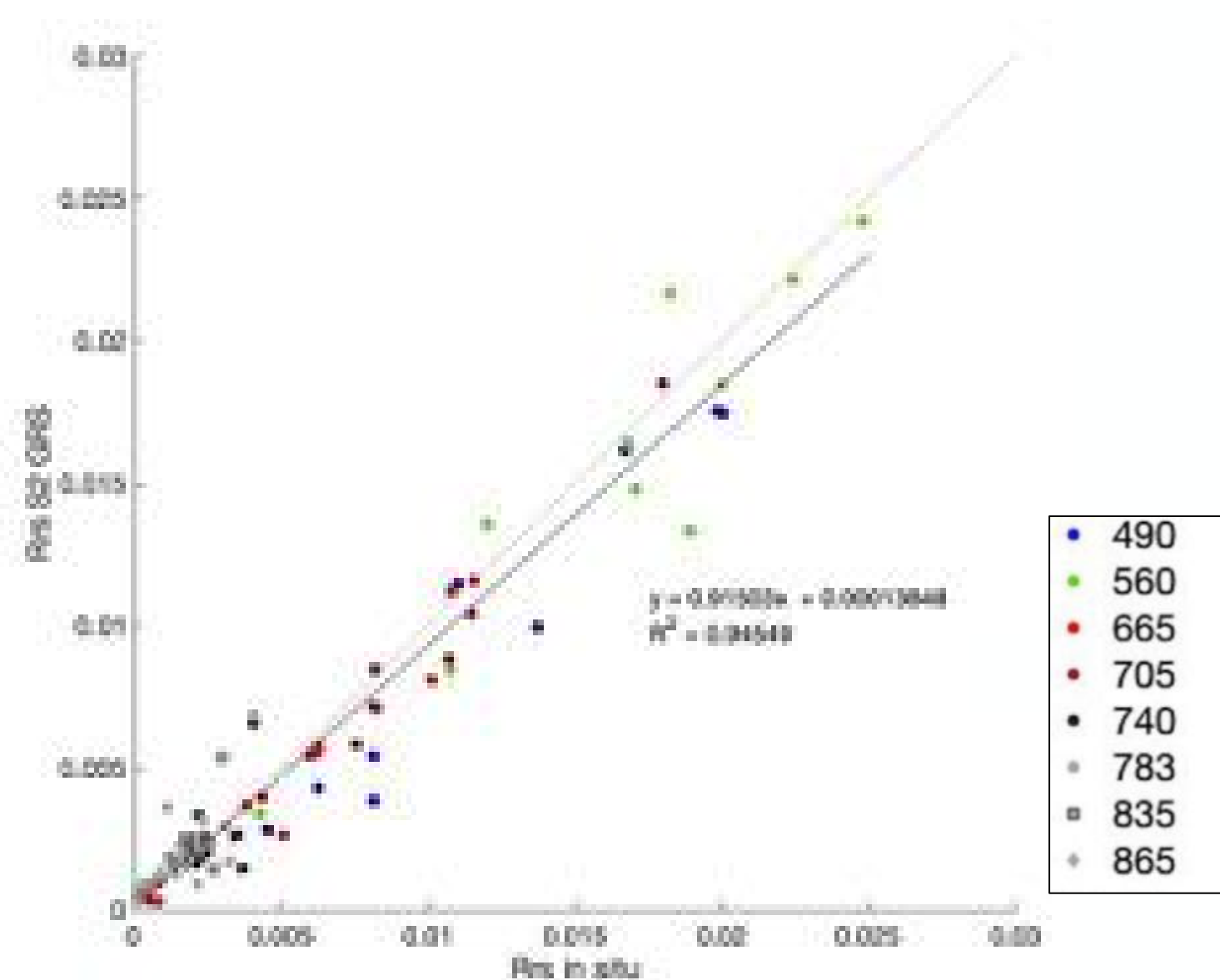
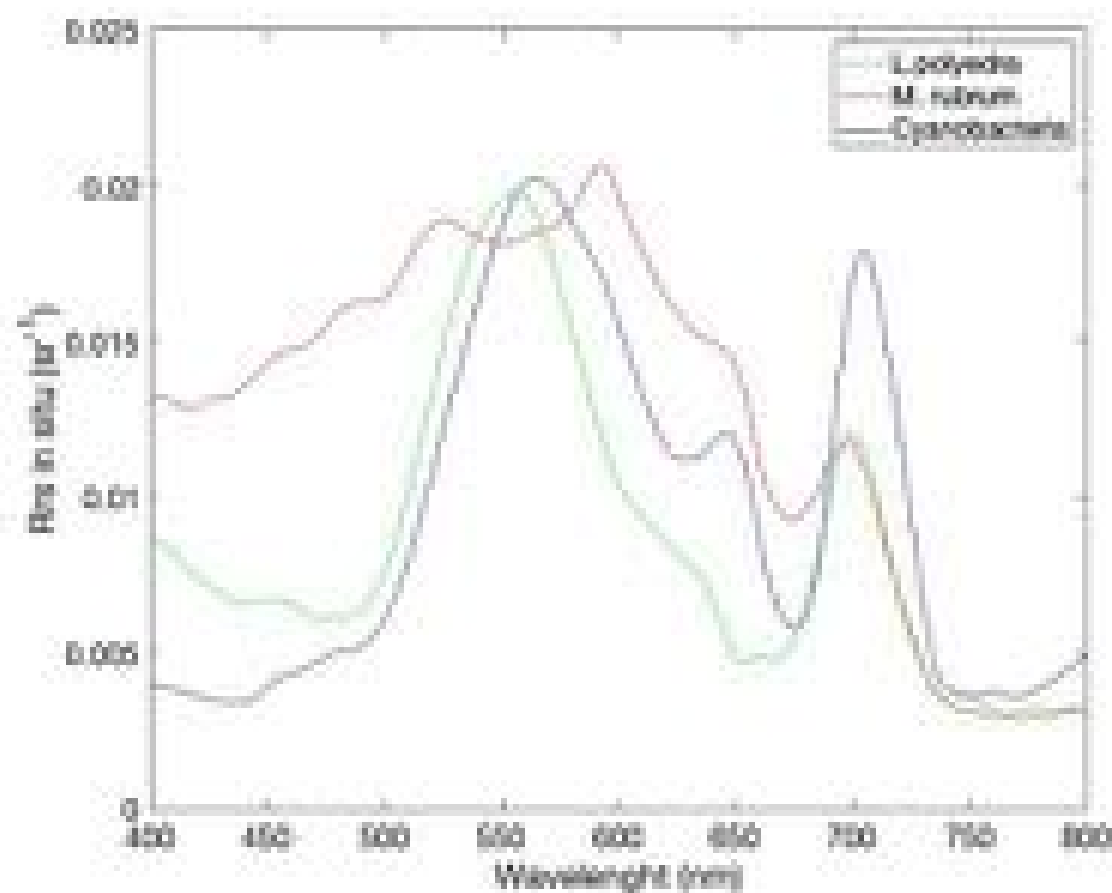
Additional reports: scientific **literature**, regional **monitoring reports**, citizen science

 **Selection of events documented in situ,
synchronous to cloud free S2 acquisition,
with conspicuous seawater discoloration**

Remote-sensing reflectance $R_{rs}(\lambda)$ re-sampled at 20 m

Atmospheric and sunglint correction
(GRS, Harmel et al. 2018)

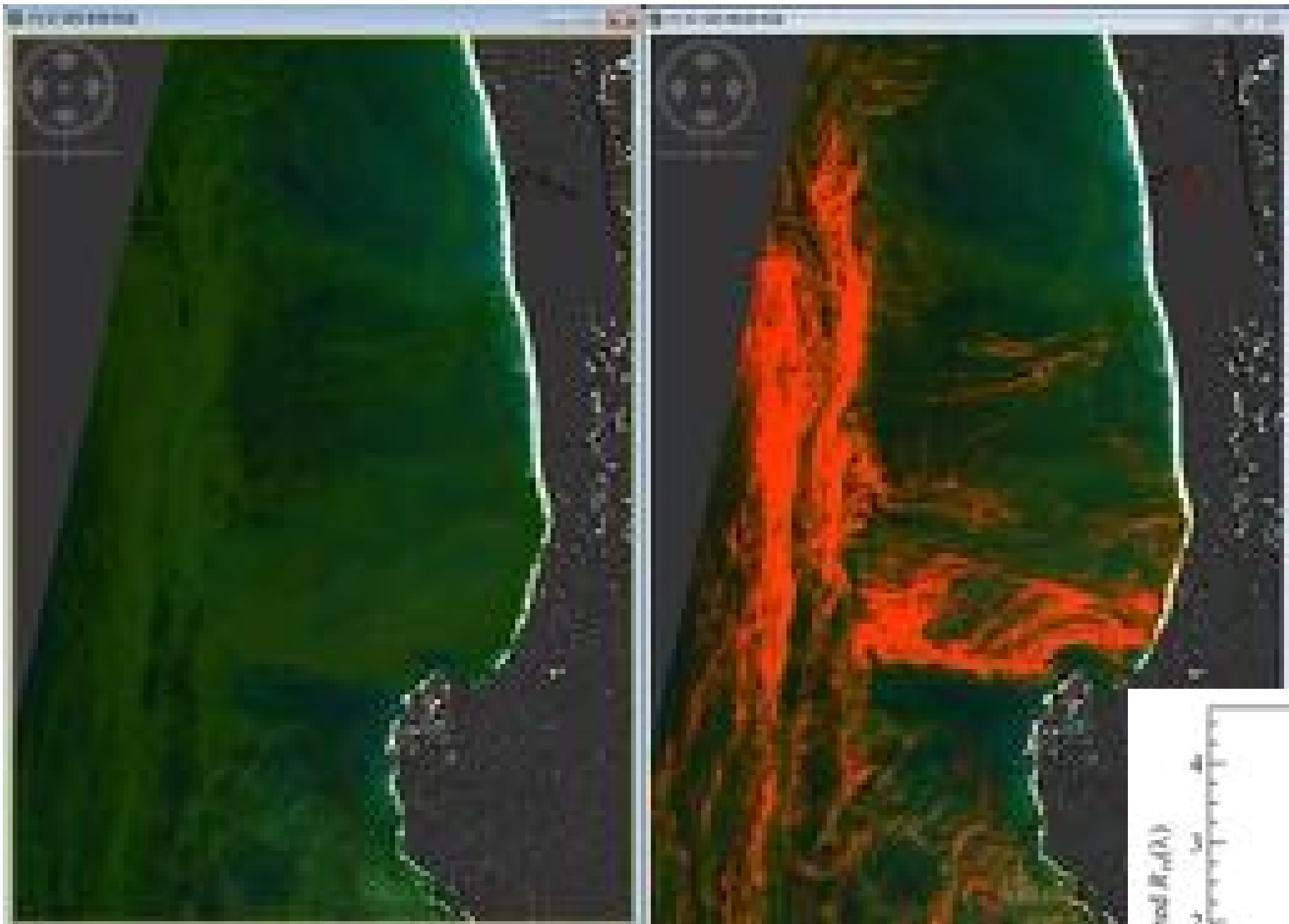
Evaluation of AC & deglint performance
($N = 24$, $r^2 = 0.95$, slope = 0.92, intercept = 0)



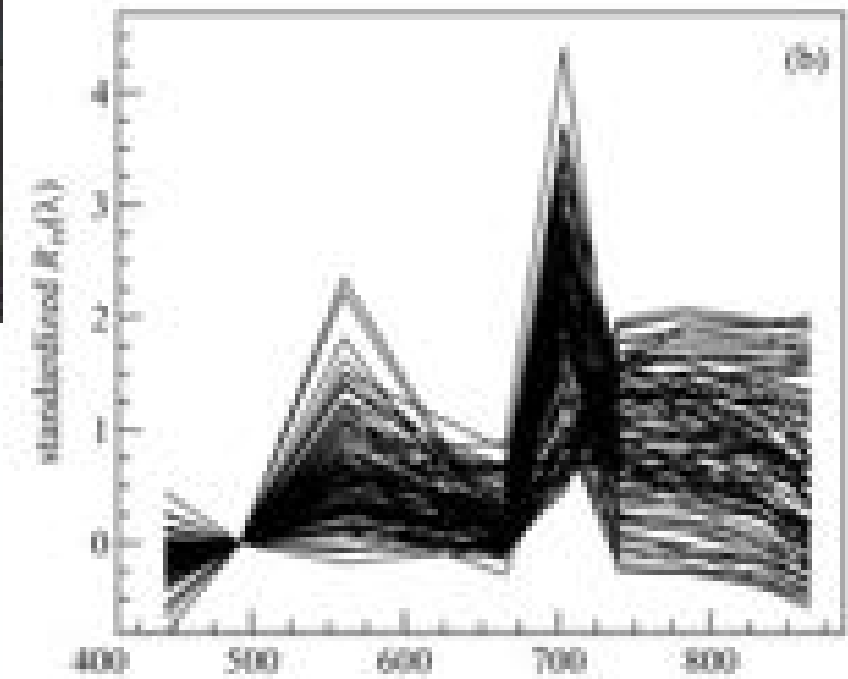
In situ R_{rs} matchups, high phytoplankton

Red-edge Green Blue colour composite (REGB)

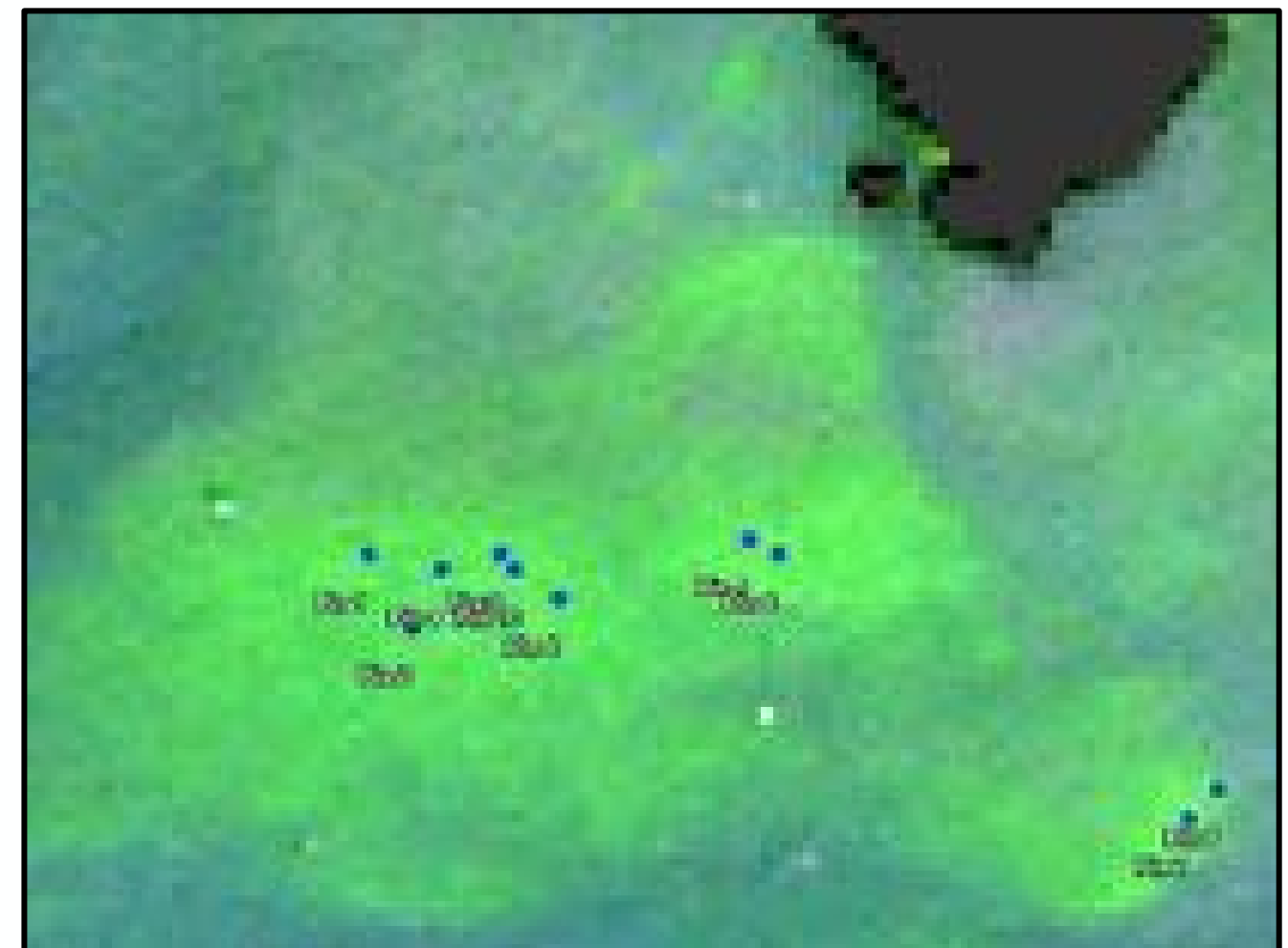
RGB
665 nm
560 nm
490 nm



REGB
705 nm
560 nm
490 nm



S2 high resolution → low spectral mixing
(limit influence of CDOM, NAP, other phyto)



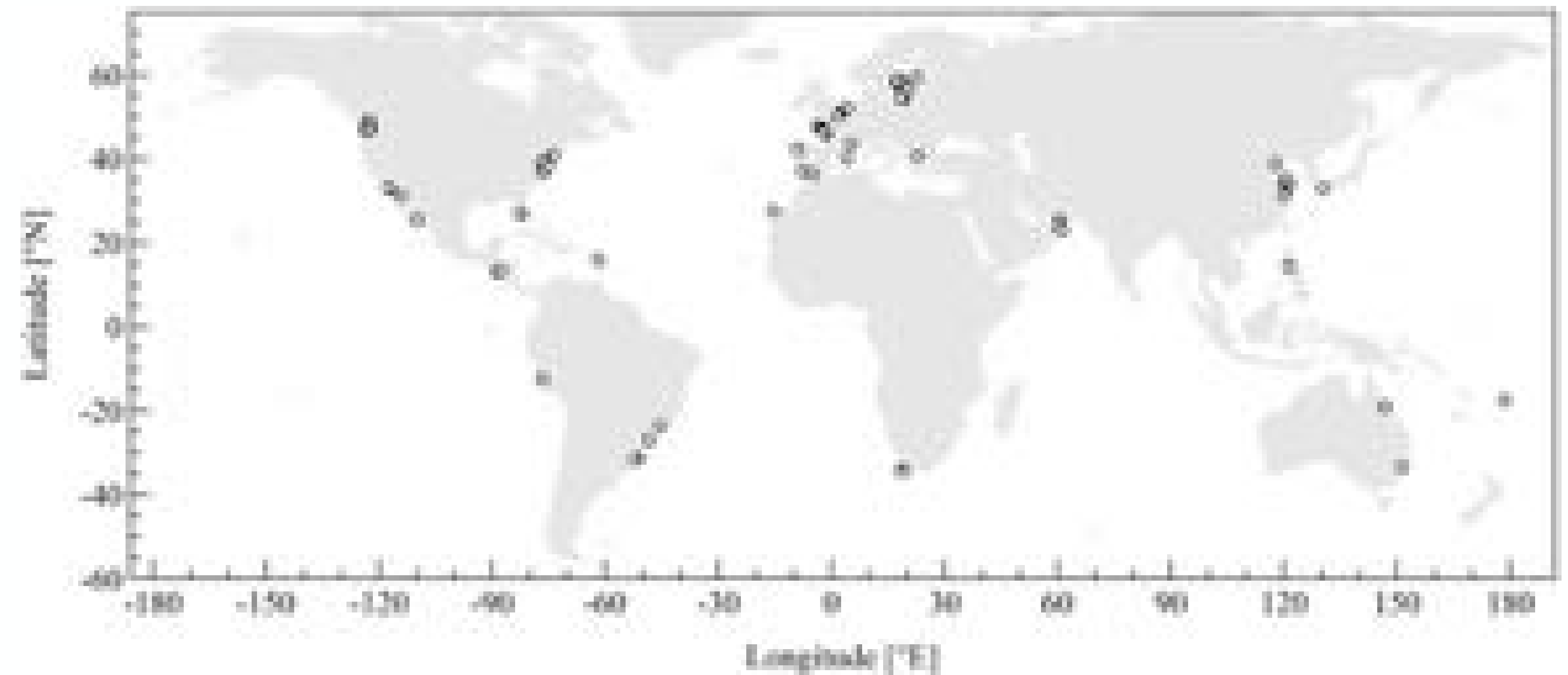
Computed averaged $R_{rs}(\lambda)$ when

- ❖ several blooms of same species in different regions → geographical robustness
- ❖ Several species from same genus → reduce interspecific variability

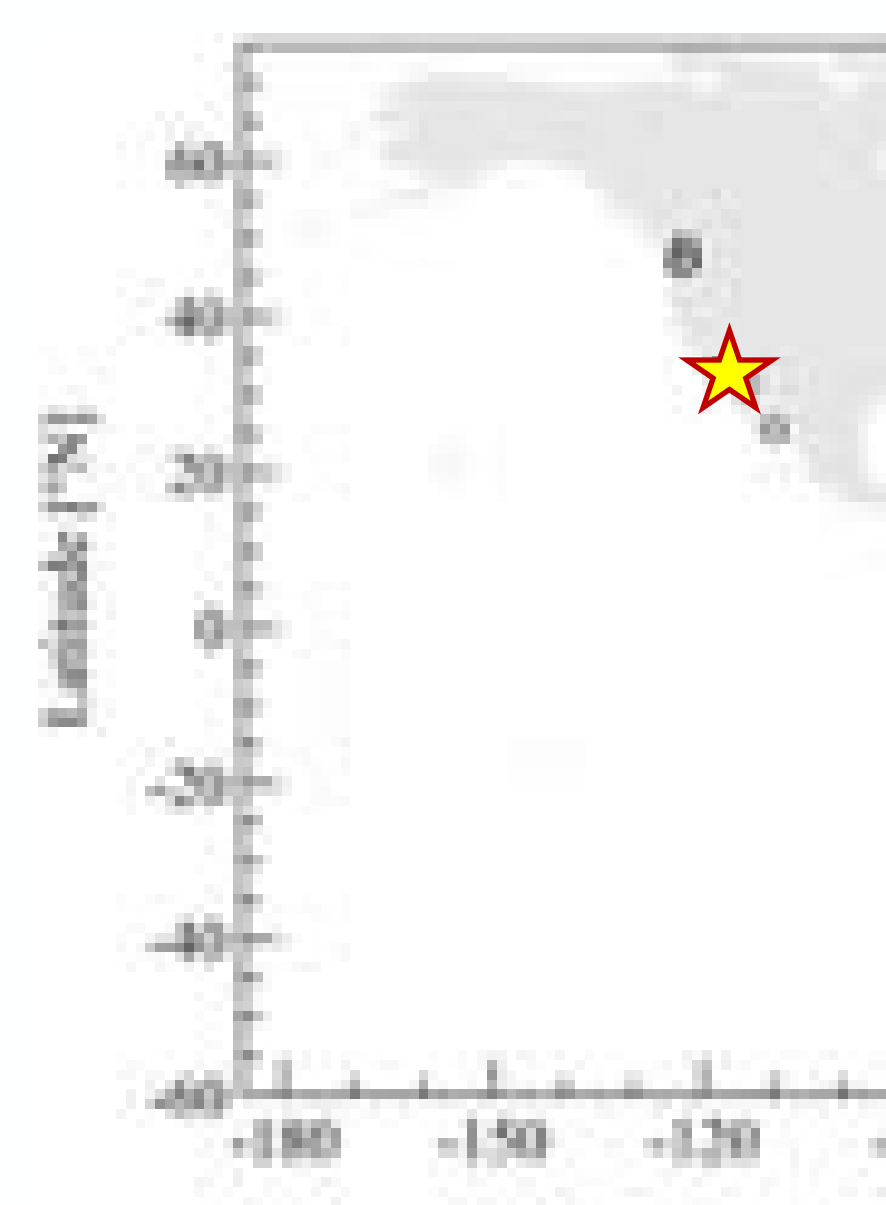
Standardized $R_{rs}(\lambda)$ to focus on spectral shape variability (Spyrakos et al., 2018)

Optical Bloom Types identified using Hierarchical Cluster Analysis (HCA) with cosine distance (Millie et al., 1997, Torrecilla et al., 2011)

S2 RED TIDE DATABASE = 108 bloom events worldwide

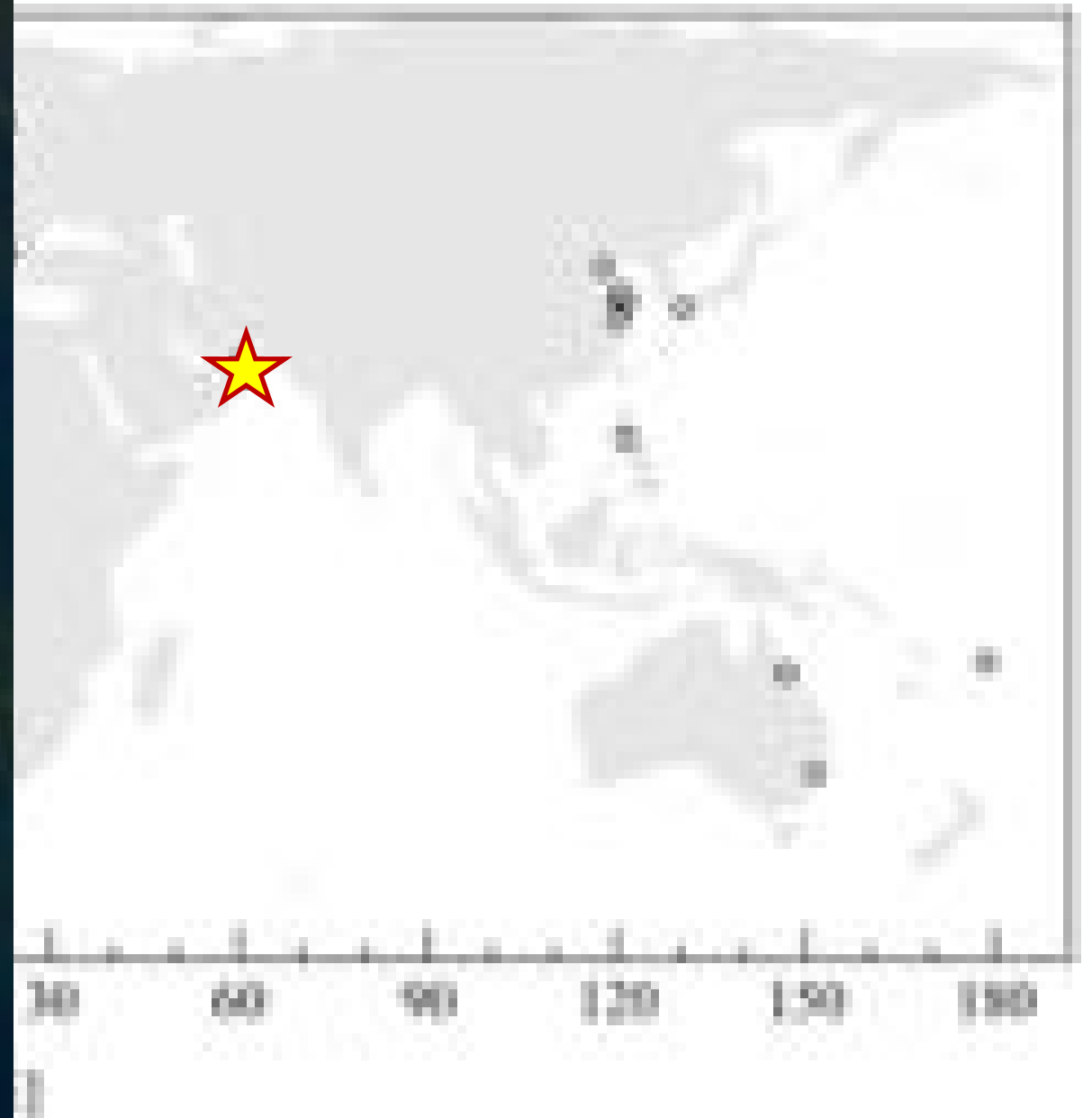


S2 RED TIDE DATABASE = 108 bloom events worldwide



S2 RED TIDE DATABASE = 108 bloom events worldwide

Bloom of *Noctiluca scintillans* (Qi et al., 2019)



20 genera

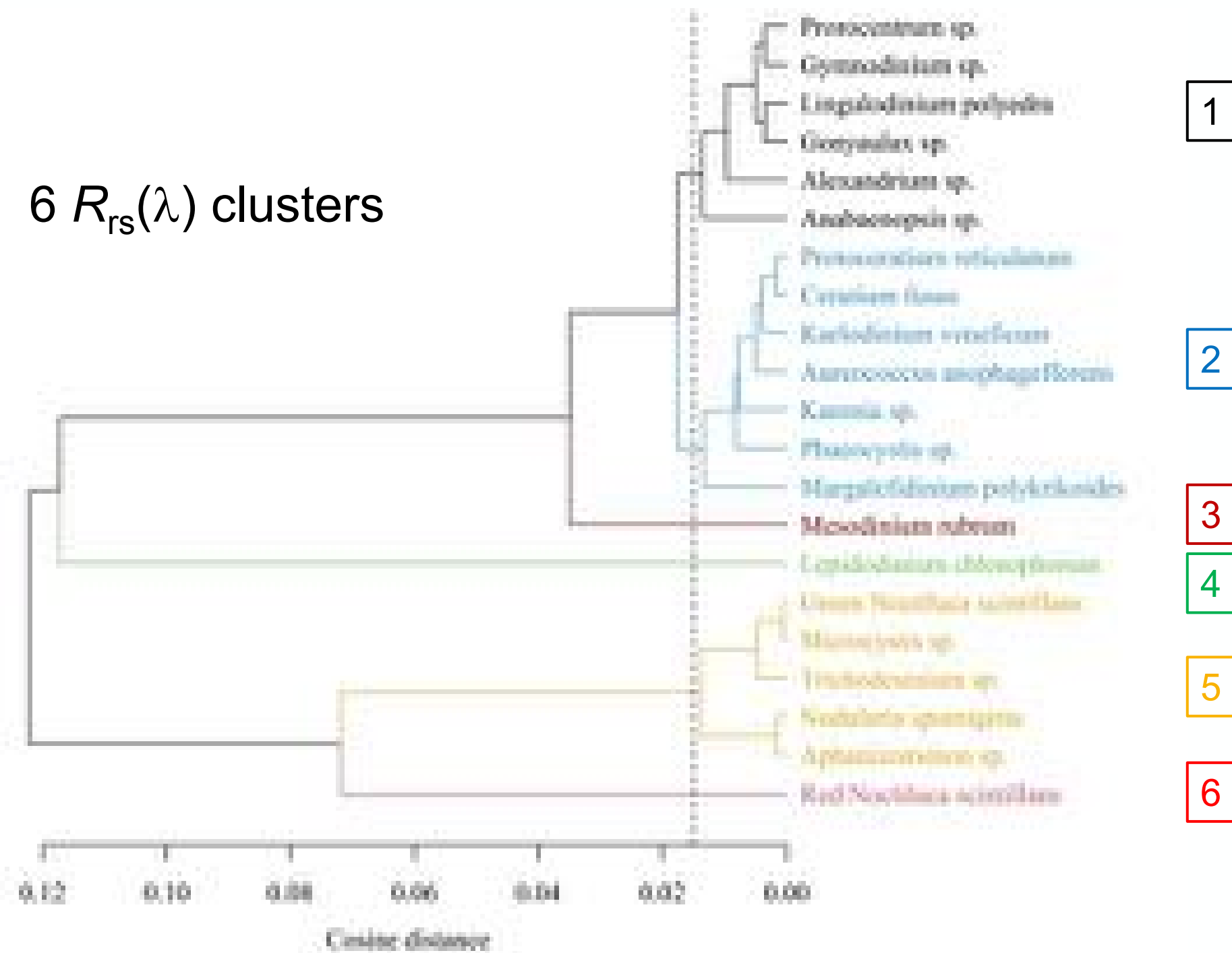
5 classes

- ❖ Many dinoflagellates (55%)
- ❖ Many cyanobacteria (30%)
- ❖ *M. rubrum* (Ciliate, 7%)
- ❖ *Phaeocystis* (Haptophyte, 5%)
- ❖ *Aureococcus* (Pelagophyte, 3%)

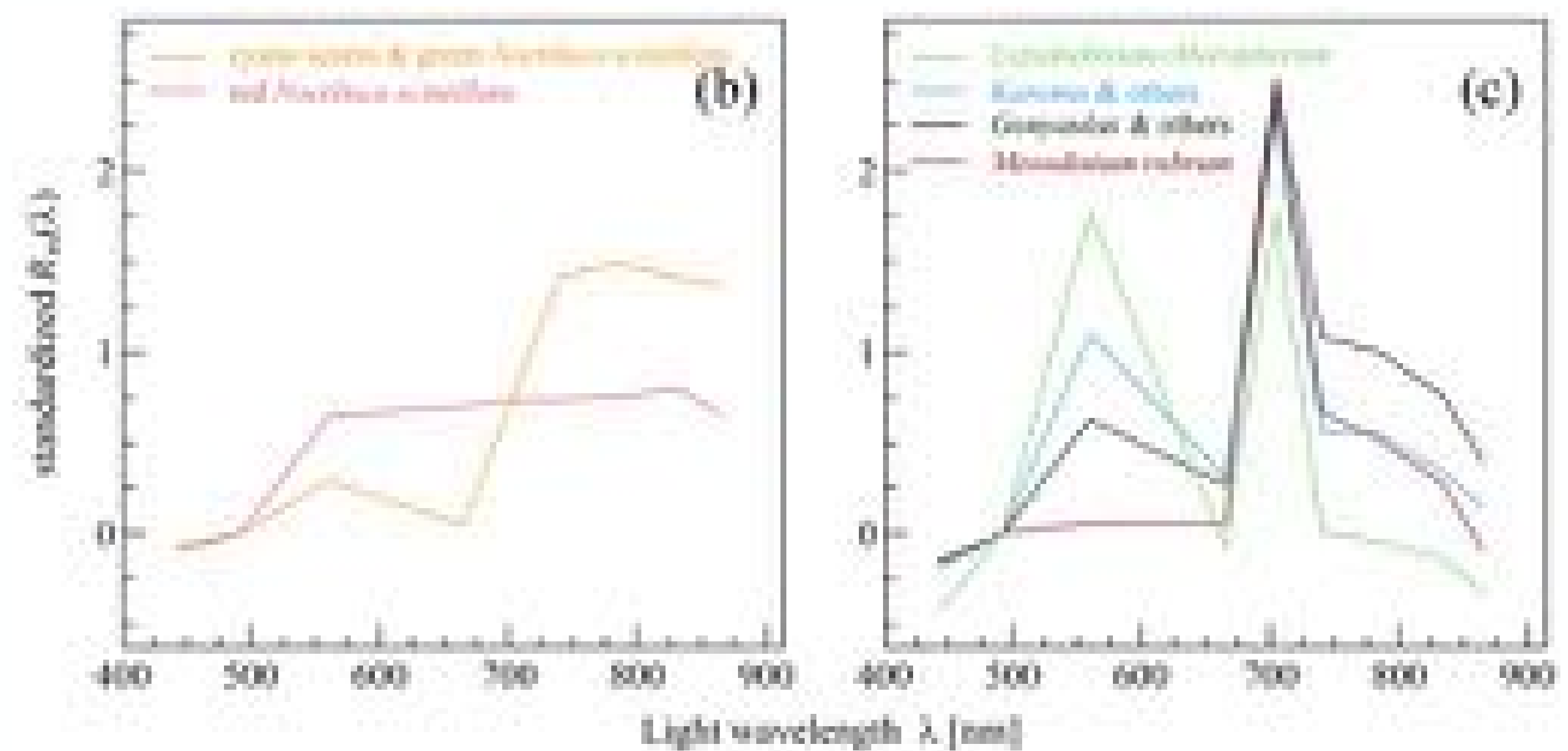
➡ 21 avg $R_{rs}(\lambda)$ spectra

Class	Genus	Species	Blotches number
Dinophyceae	Alexandrium	equale, or oestrupii	3
	Ceratium	form	2
	Gonyaulax	digitata, polygramma, or spinifera	6
	Gymnodinium	aureolum, catenatum, or impudicum	3
	Karreri	brevis or multisetos	3
	Karlodinium	veneficum	2
	Lepidodinium	chlorophorum	3
	Longicollum	polystrum	8
	Margalefidinium	polytrichoides	3
	Nautlella	acutellans (green form)	7
	Nautlella	acutellans (red form)	6
	Prorocentrum	micans, or minimum	3
	Protoperidinium	reticulatum	3
Cyanophyceae	Anabaenopsis	sp.	3
	Lyngbya	flourensii, or spp.	3
	Microcystis	sp.	3
	Nodularia	spumigena	4
	Trichodesmium	sp.	7
Leptostreptus	Micodinium	rubrum	3
Prymnesiophyceae	Phaeocystis	sp.	6
Pelagophyceae	Aureococcus	anophagefferens	2

6 $R_{rs}(\lambda)$ clusters

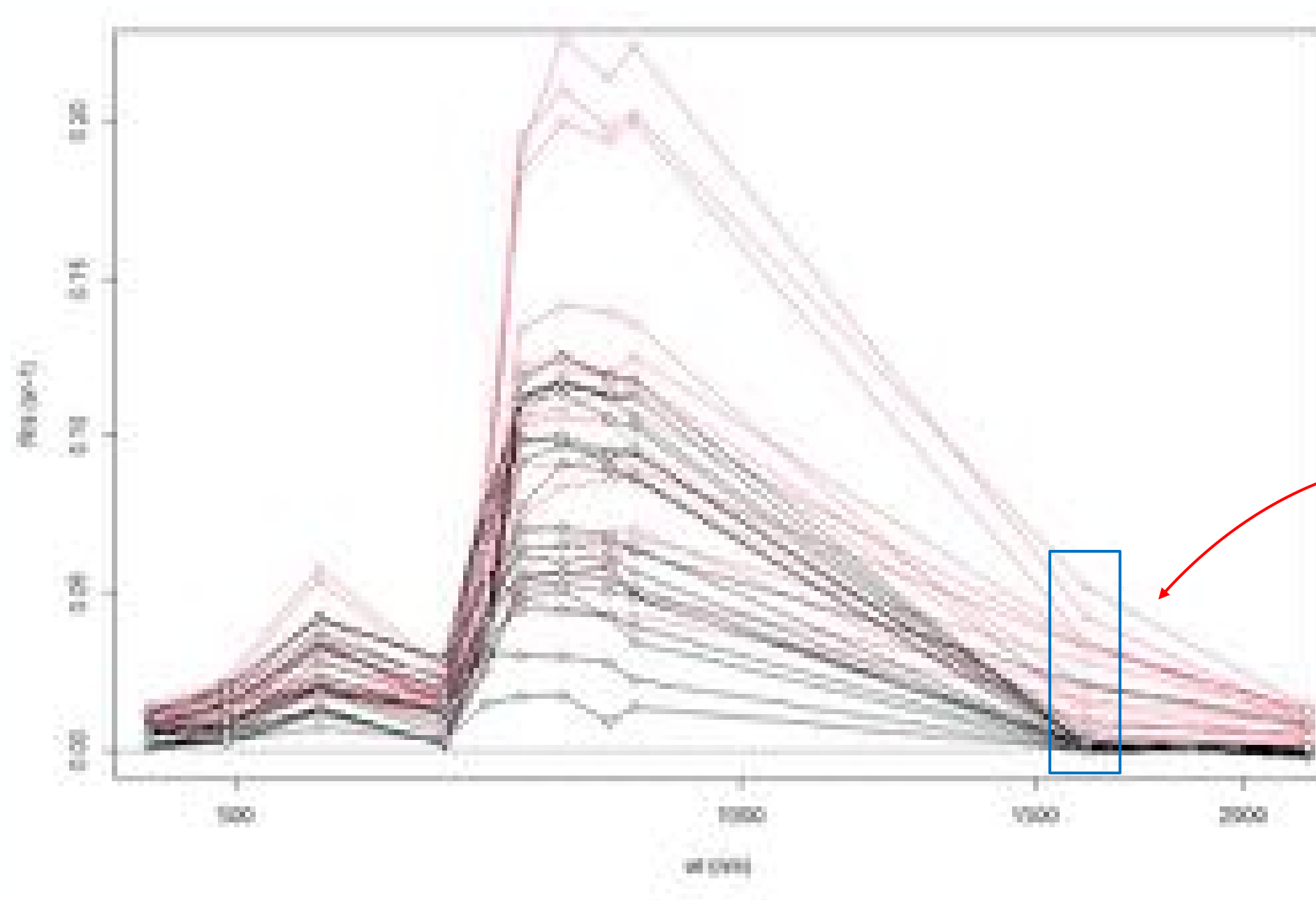


Optical Bloom Types with Sentinel-2



Gernez, P., Zoffoli, M. L., Lacour, T., Fariñas, T. H., Navarro, G., Caballero, I., & Harmel, T. (2023). The many shades of red tides: Sentinel-2 optical types of highly-concentrated harmful algal blooms. *Remote Sensing of Environment*, 287, 113486.

Appendix: cyanoHABs with high SWIR



Microcystis
Trichodesmium
 $R_{rs}(1610) > 0$

Green *Noctiluca*,
Nodularia
Aphanizomenon

Flat $R_{rs}(\lambda)$ from 560 – 865 nm

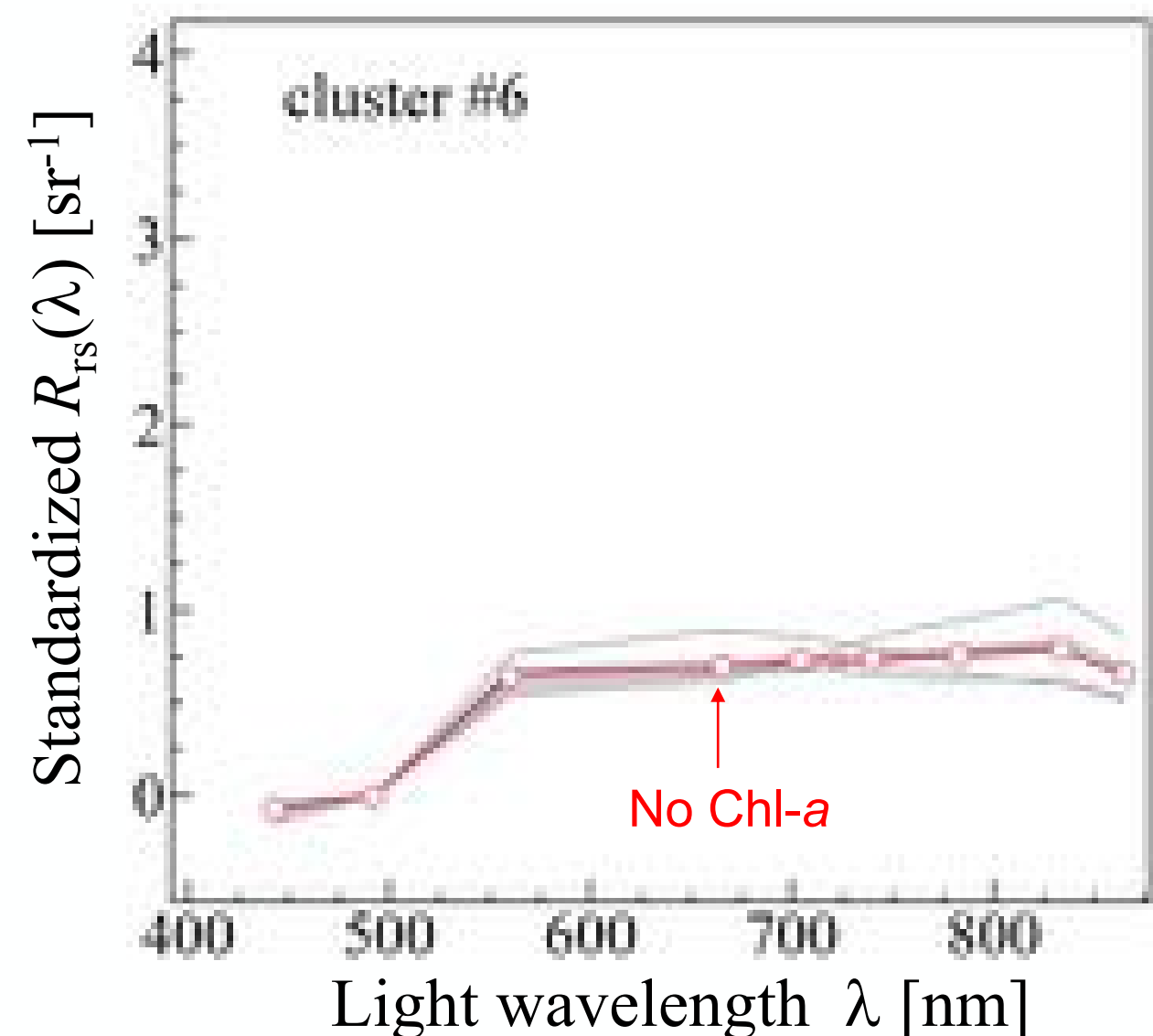
Monospecific cluster

Low $R_{rs}(\lambda)$ variability (6 blooms / 5 countries)

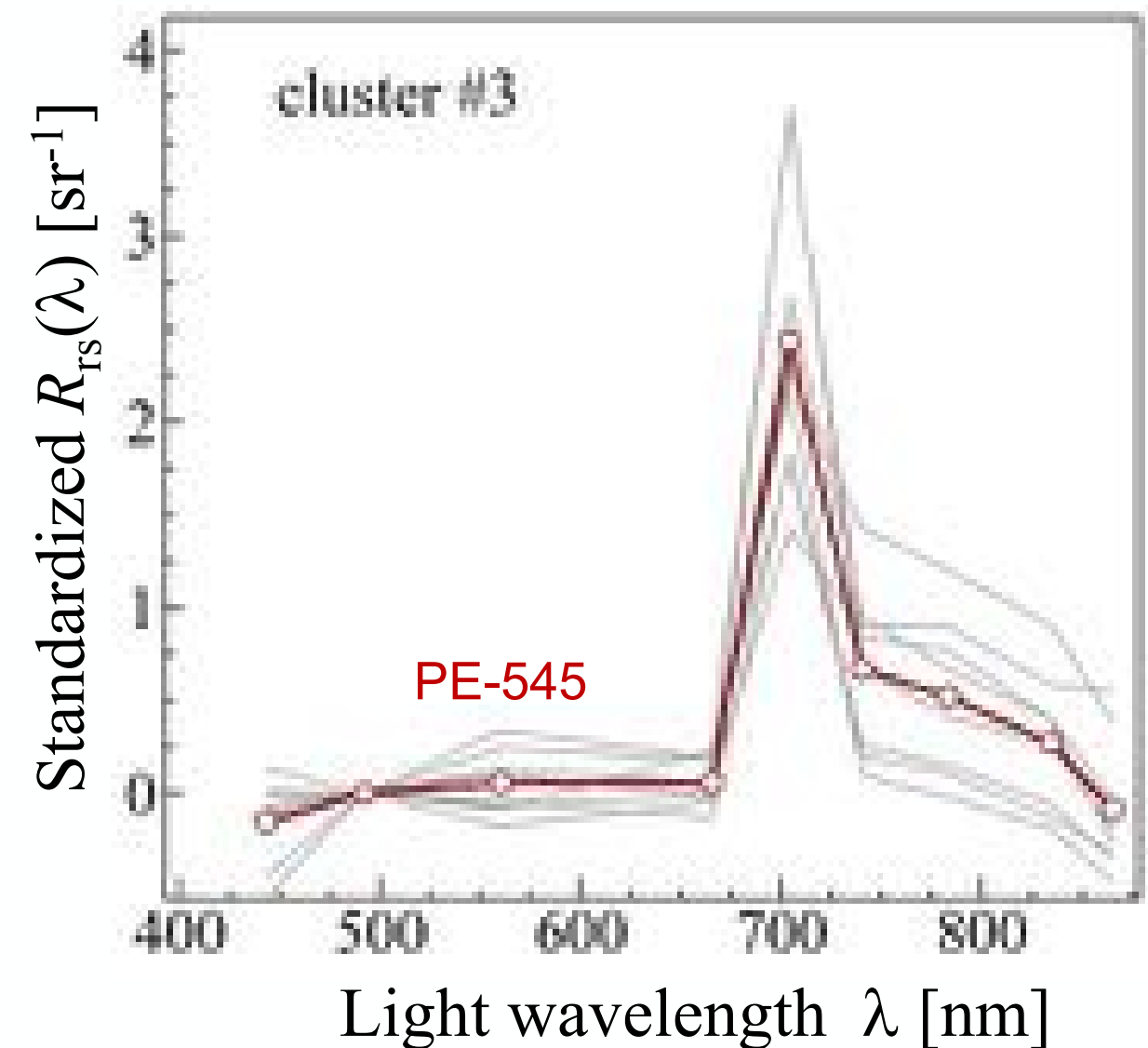
High biomass, **Chl-a absent**

(pure heterotroph with carotenoids)

Distinct from green *Noctiluca* (contains Chl-a due to photosynthetic symbiont)

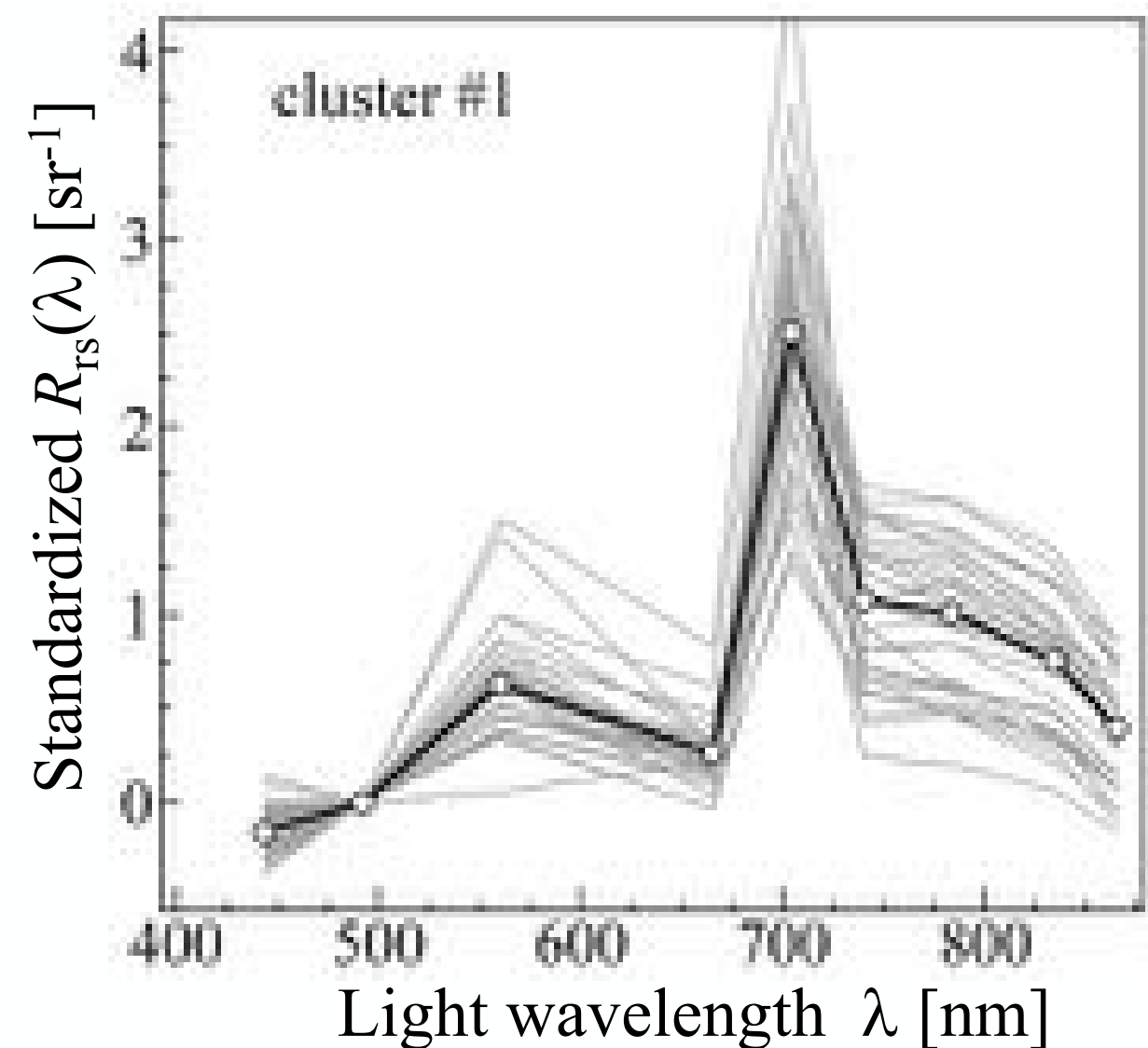


~ *pigment type 5* in Zapata et al., 2012

Manuel Zapata^{1,†}, Santiago Fraga², Francisco Rodriguez^{2,*}, José L. Garrido¹

- ❖ *Prorocentrum*
- ❖ *Gymnodinium*
- ❖ *Lingulodinium polyedra*
- ❖ *Gonyaulax*
- ❖ *Alexandrium*

Vol. 463: 33–52, 2012 doi: 10.3354/meps09879	MARINE ECOLOGY PROGRESS SERIES Mar Ecol Prog Ser	Published September 28
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Manuel Zapata^{1,†}, Santiago Fraga², Francisco Rodríguez^{2,*}, José L. Garrido¹

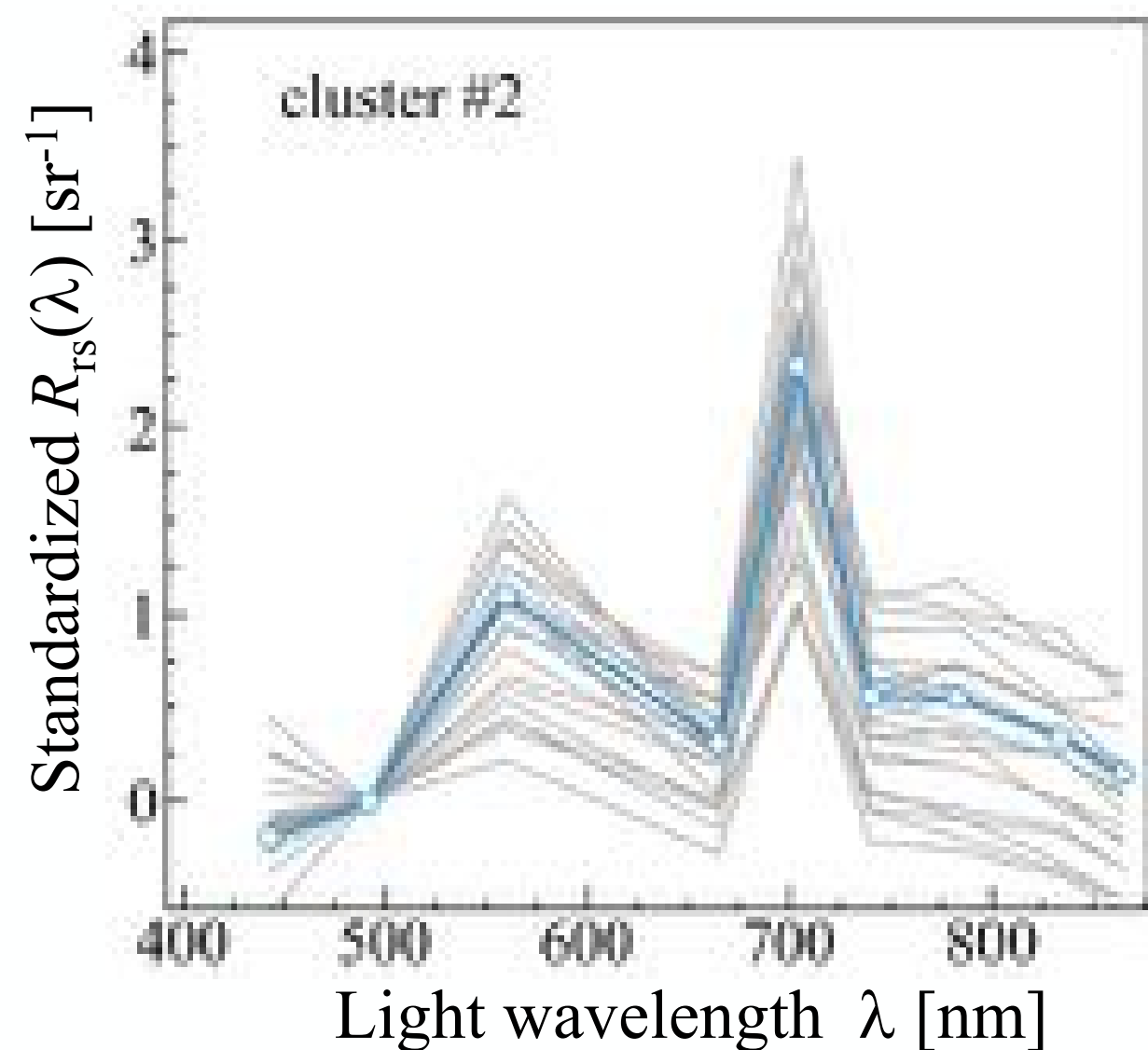
Dominated by dinos (5/7 genera),

- ❖ *Protoceratium reticulatum*
- ❖ *Ceratium fusus*
- ❖ *Karlodinium veneficum* *
- ❖ *Karenia sp.* *
- ❖ *Margalefidinium polykrikoides*

Also contains 2 other classes

- ❖ *Phaeocystis sp.* *
- ❖ *Aureococcus anophagefferens* *

Fucoxanthin-bearing species (4/7)



Three species had unique $R_{rs}(\lambda)$ fingerprint (specific pigment composition)

❖ *Lepidodinium chlorophorum* (Chl-*b*), *Mesodinium rubrum* (PE-545), Red *Noctiluca scintillans* (absence of Chl-*a*)

Floating cyanoHABs (+ green *Noctiluca*) could be discriminated (high NIR & SWIR)

Clusters dominated by Dinoflagellates: very diverse, risk of confusion (e.g. *Phaeocystis*), discrimination between Peridin- vs. Fucoxanthin-bearing species?

Blooms of Diatoms (absent of the database) should be further investigated

Sentinel-2 is relevant for HAB monitoring (in synergy with field observations)

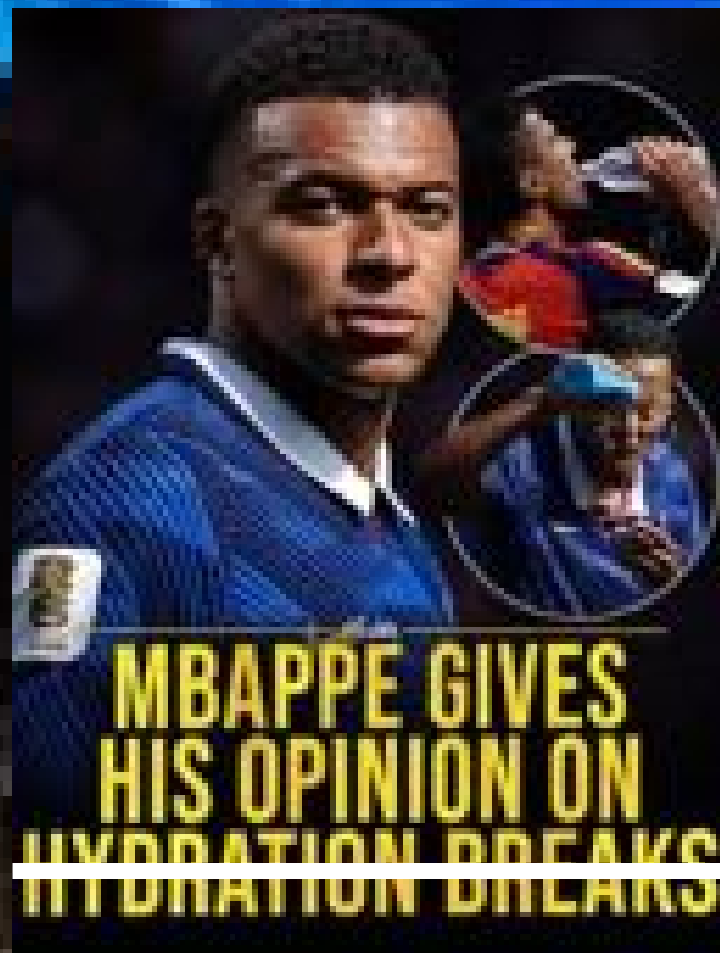
In the absence of in situ measurement, S2 could identify a few optical bloom types

S2-HABs database could be use for algorithms development and validation

“Spectral library approach” could be improved using hyperspectral sensors

5 min break

HYDRATION BREAK



PHYTO ABSORPTION

Hyperspectral approaches

Pigments line height

Spectral derivatives

Similarity index

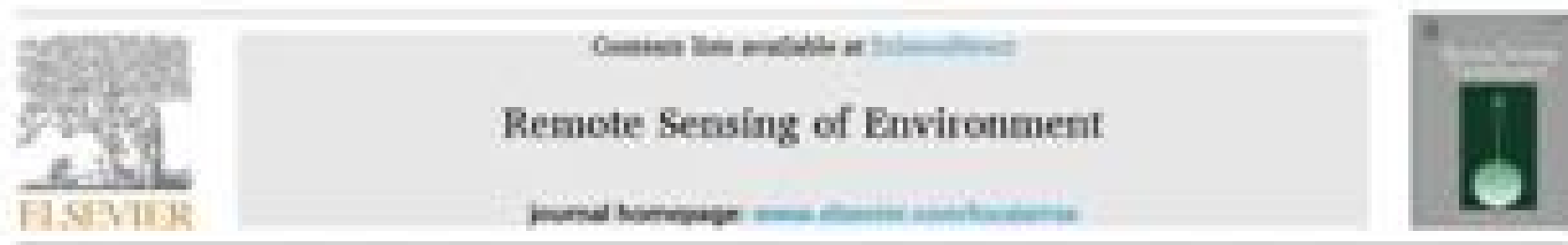
Optical water type (OWT) classification

Many others: see lectures by Jeremy Werdell and Ivona Cetinić

Pigments line height

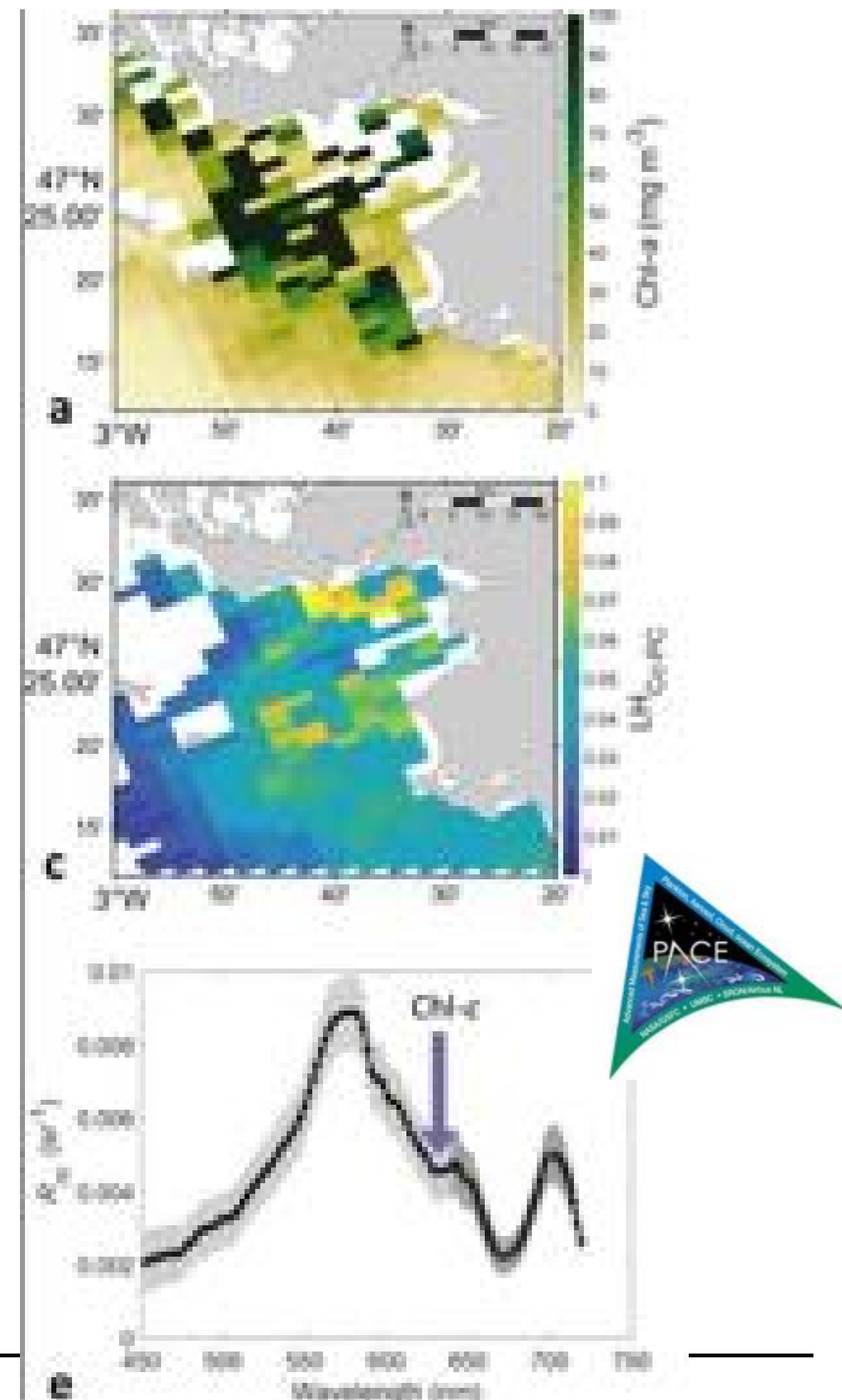
$$LH = 100 * [R_{rs}(\lambda_1) + \tau \bullet (R_{rs}(\lambda_3) - R_{rs}(\lambda_1)) - R_{rs}(\lambda_2)]$$

$$\tau = \frac{\lambda_2 - \lambda_1}{\lambda_3 - \lambda_1}$$



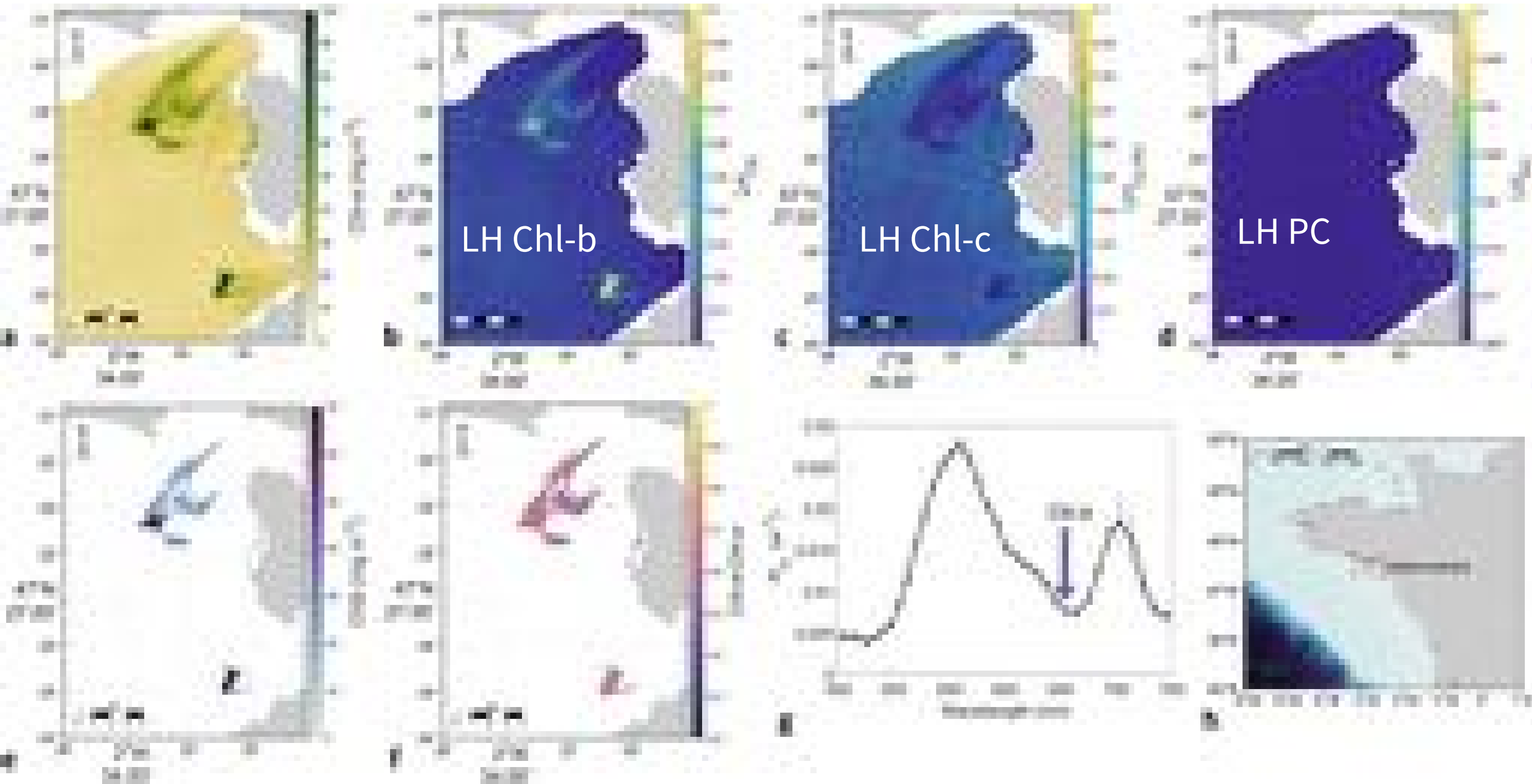
Toward elucidating accessory pigments in intense phytoplankton blooms using hyperspectral satellite remote sensing in support of harmful algal blooms (HAB) monitoring

Maria Laura Zoffoli^{1,*}, Victor Poehle^{2,3}, Mariano Bresciani⁴, Federica Raga⁵, Thomas Lacour⁶, Michael Reitho⁷, Souzig Marach⁸, Rüdiger Röttgers⁹, Amalia Maria Sacilotto Detoni¹⁰, Tristan Harnel¹, Simon Clary¹¹, Pierre Gerner⁷



Chl-*b* line height

(absorption max ~ 650 nm)



Optical water types (OWTs)

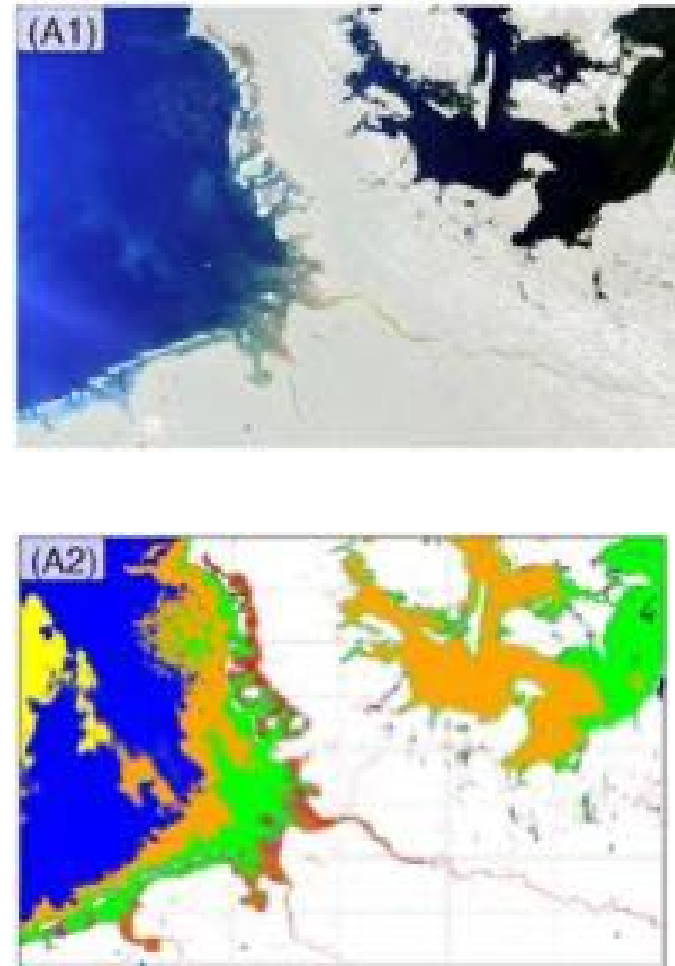
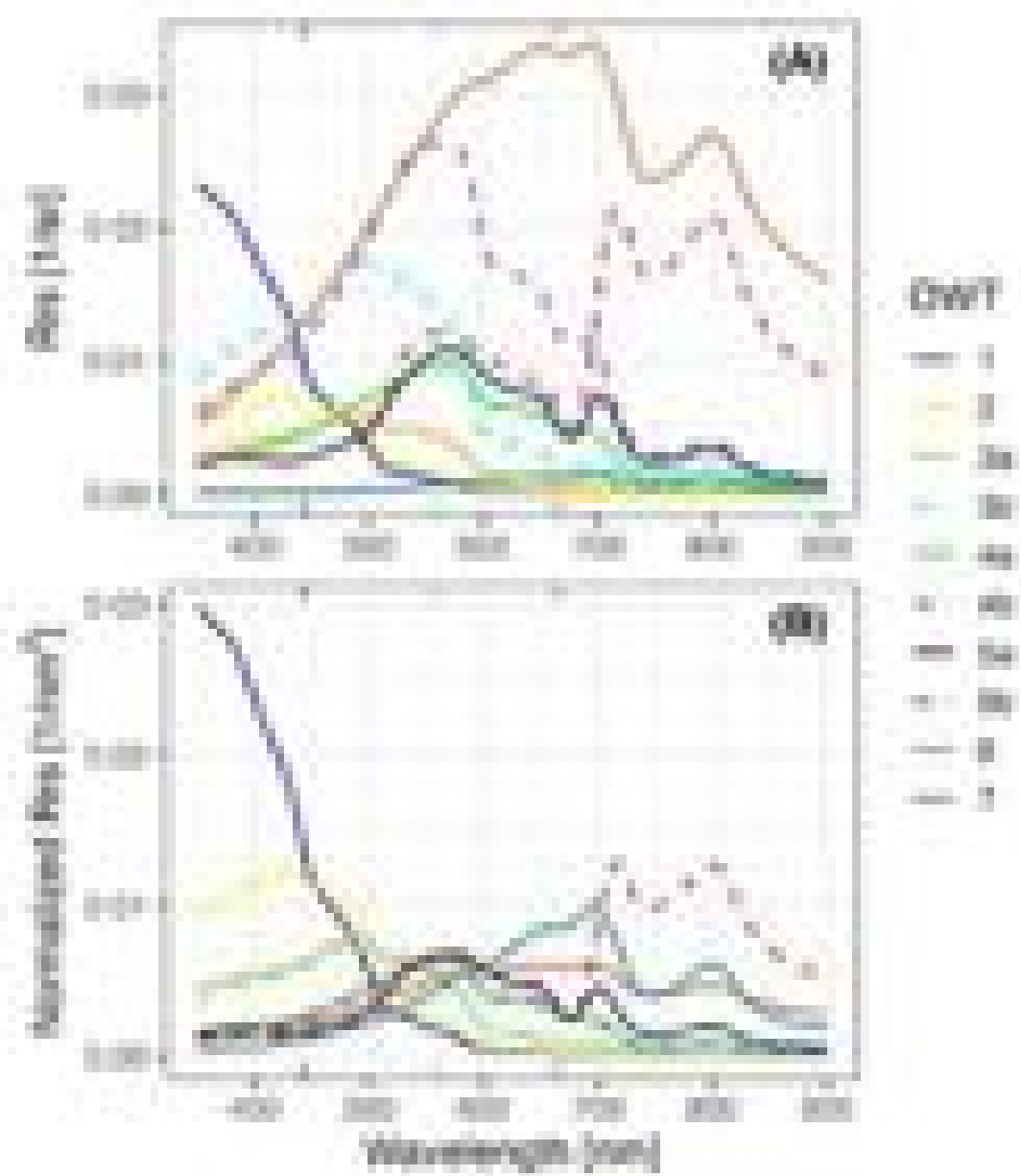
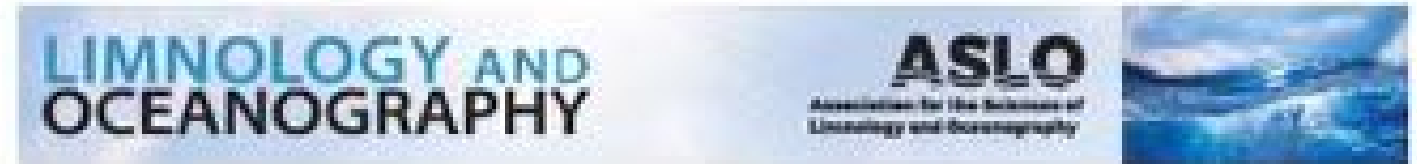
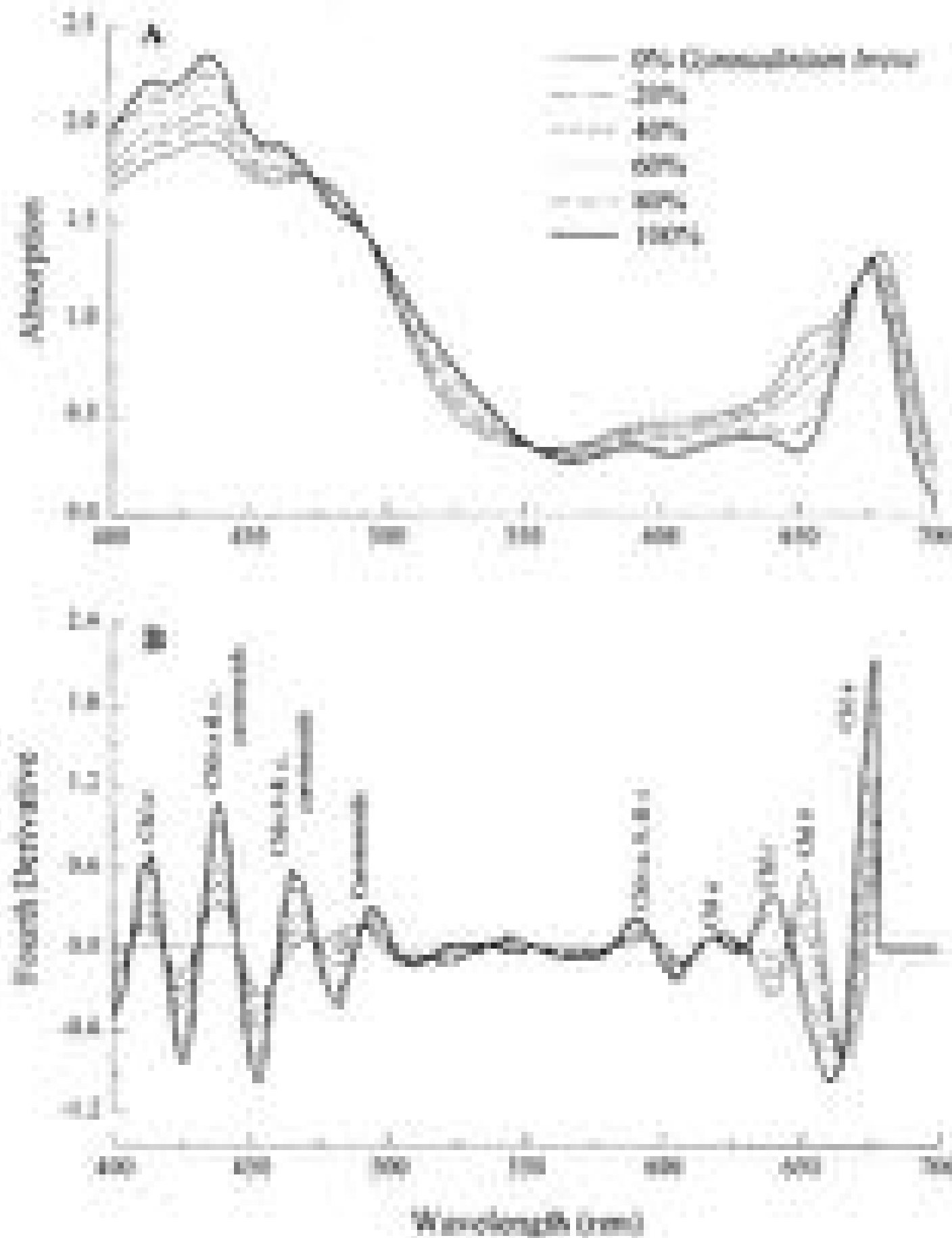


Table 1. Conceptual description of optical water types.

OWT	Description
1	Extremely clear and oligotrophic, deep-blue water with high reflectance in the short visible wavelengths.
2	Blue water with similar biomass level as OWT 1 but with slightly higher detritus and CDOM content.
3a	Turbid water with slightly higher phytoplankton, detritus, and CDOM compared to the first two types.
3b	A special case of OWT 3a with similar detritus and CDOM distribution but with strong scattering and little absorbing particles like in the case of <i>Leptodinium</i> blooms. This type usually appears brighter and exhibits a remarkably ~ 680 nm reflectance peak.
4a	Coastal water found in coastal and inland environments, with higher biomass compared to the previous water types. Reflectance in short wavelengths is usually depressed by the absorption of particles and CDOM.
4b	A special case of OWT 4a, sharing similar detritus and CDOM distribution, exhibiting phytoplankton blooms with higher scattering coefficients, e.g., <i>Cyclotella</i> blooms. The color of this type shows a very bright green.
5a	Green eutrophic water, with significantly higher phytoplankton biomass, exhibiting a bimodal reflectance shape with typical peaks at ~ 550 and ~ 700 nm.
5b	Green hyper-eutrophic water, with even higher biomass than that of OWT 5a (over several orders of magnitude), displaying a reflectance pattern in the near infrared region, 700–900 (vegetation-like spectrum).
6	Bright brown water with high detritus concentrations, which has a high reflectance determined by scattering.
7	Dark brown to black water with very high CDOM concentration, which has low reflectance in the entire visible range and is dominated by absorption.

Spectral derivative (using absorption)



Article [Free to Read](#)

Detection of harmful algal blooms using photopigments and absorption signatures: A case study of the Florida red tide dinoflagellate, *Gymnodinium breve*

David F. Millie, Oscar M. Schofield, Gary J. Kirkpatrick, Geir Johnsen, Patricia A. Tester, Bryan T. Vinyard

Millie et al., (1997)

Spectral derivative (using reflectance)

Applied Optics Vol. 45, [Issue 21](#), pp. 5414-5425 (2006) • <https://doi.org/10.1364/AO.45.005414>



Use of hyperspectral remote sensing reflectance for detection and assessment of the harmful alga, *Karenia brevis*

Susanne E. Craig, Steven E. Lohrenz, Zhongping Lee, Kevin L. Mahoney, Gary J. Kirkpatrick, Oscar M. Schofield, and Robert G. Steward

Abstract

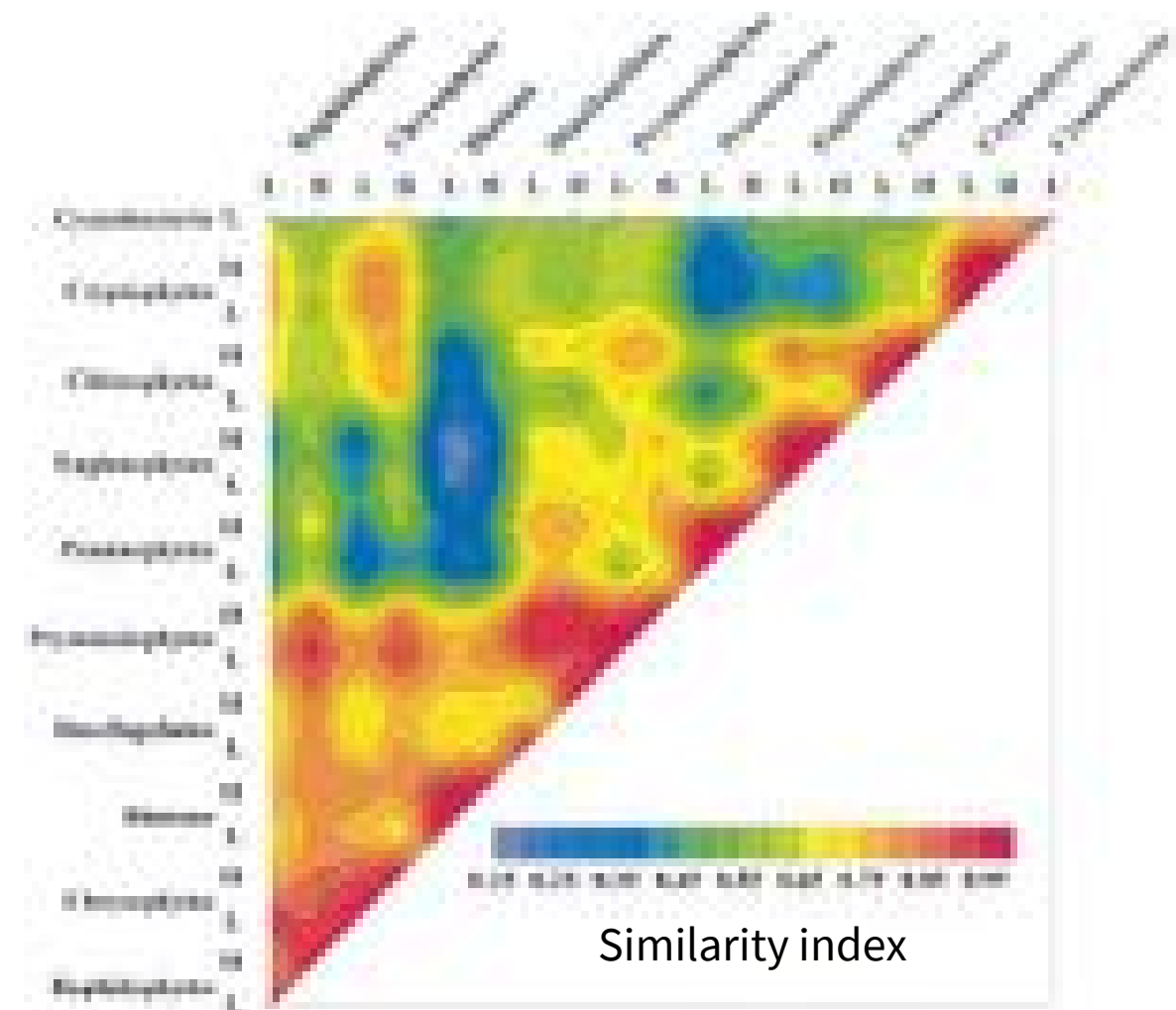
We applied two numerical methods to *in situ* hyperspectral measurements of remote sensing reflectance R_{rs} to assess the feasibility of remote detection and monitoring of the toxic dinoflagellate, *Karenia brevis*, which has been shown to exhibit unique absorption properties. First, an existing quasi-analytical algorithm was used to invert remote sensing reflectance spectra, $R_{rs}(\lambda)$, to derive phytoplankton absorption spectra, $a_p^{Rrs}(\lambda)$. Second, the fourth derivatives of the $a_p^{Rrs}(\lambda)$ spectra were compared to the fourth derivative of a reference *K. brevis* absorption spectrum by means of a similarity index (SI) analysis. Comparison of reflectance-derived a_p with

Similarity index (2 spectra are similar if SI = 1) aka spectral angle mapper (SAM)

Within each comparison, the degree of similarity between spectra then was computed using an index (SI) algorithm:

$$SI = \frac{2}{\pi} \arccosine \frac{(S_1 \bullet S_2)}{|S_1| \times |S_2|}$$

where $\frac{(S_1 \bullet S_2)}{|S_1| \times |S_2|}$ is the cosine of the angle between two vectors, S_1 is the vector of fourth-derivative spectrum values $S_1(\lambda_1), S_1(\lambda_2), S_1(\lambda_3), \dots, S_1(\lambda_n)$ for the first spectrum, and, similarly, S_2 is the vector of the fourth-derivative spectrum values for the second absorbance or



Using absorbance and fluorescence spectra to discriminate microalgae

Published online by Cambridge University Press: 08 October 2002

DAVID F. MILLIE, OSCAR M. E. SCHOFIELD, GARY J. KIRKPATRICK, GEIR JOHNSEN and
TERENCE J. EVENS

Millie et al., (2002)

Johnsen et al., (1994)

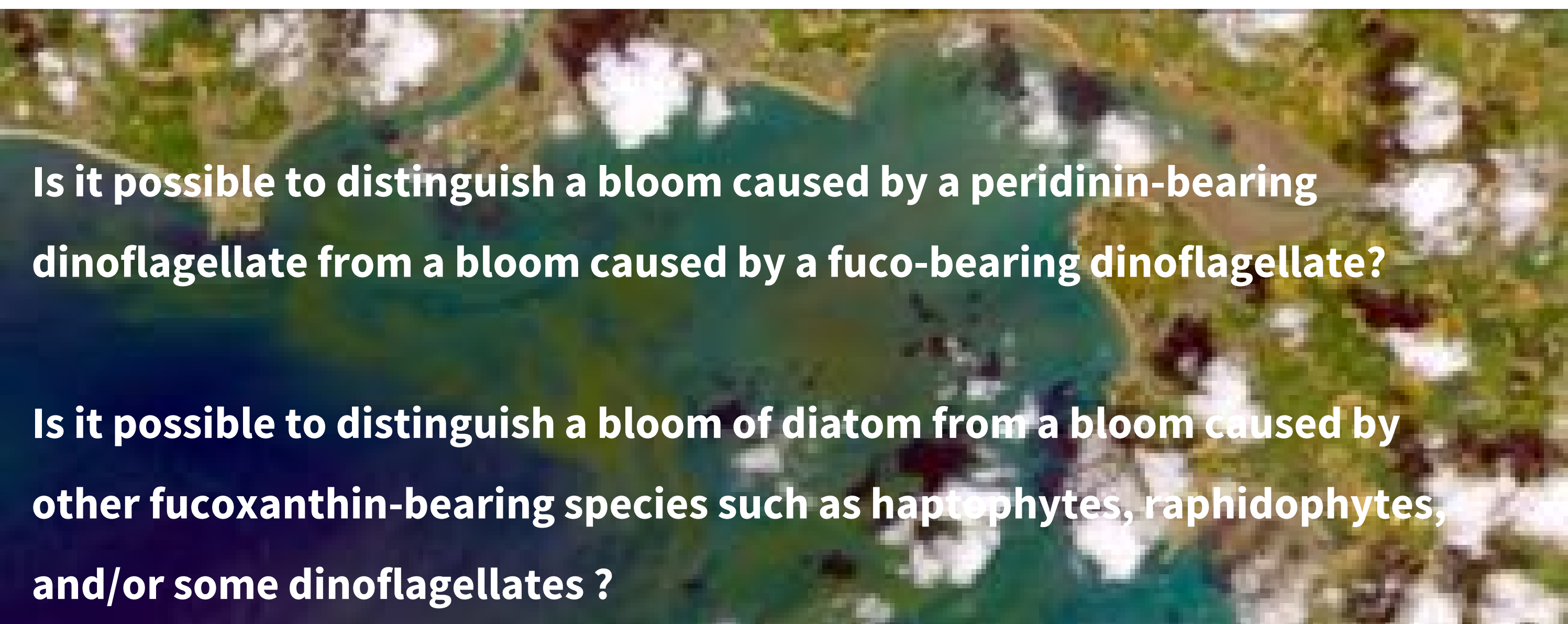
Vol. 100, 100–110, 1994	MAJOR BIOLOGICAL PROCESSES SERIES Mar. Ecol. Prog. Ser.	Published February 02
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***In vivo* absorption characteristics in 10 classes of bloom-forming phytoplankton: taxonomic characteristics and responses to photoadaptation by means of discriminant and HPLC analysis**

Geir Johnsen¹, Oddgeir Samset², Laila Gramslog¹, Egil Sakshang¹

¹Trondheim Biological Station, The Museum, University of Trondheim, Postboks 40, N-7001 Trondheim, Norway
²NTNF Industrial Mathematics, NTH, N-7001 Trondheim, Norway

Assessing potential of hyperspectral RS



Is it possible to distinguish a bloom caused by a peridinin-bearing dinoflagellate from a bloom caused by a fuco-bearing dinoflagellate?

Is it possible to distinguish a bloom of diatom from a bloom caused by other fucoxanthin-bearing species such as haptophytes, raphidophytes, and/or some dinoflagellates ?

Material & Methods



Particulate absorption coefficient $a_p(\lambda)$ measured on monospecific algal cultures using filter pads (inside sphere) or cells suspension (inside PSICAM).

Spectral library of 60 bloom-forming phytoplankton species

164 hyperspectral spectra covering 8 taxonomic classes compiled from Xi et al. (2015), Lomas et al. (2024), and a new IOP library of bloom-forming species.

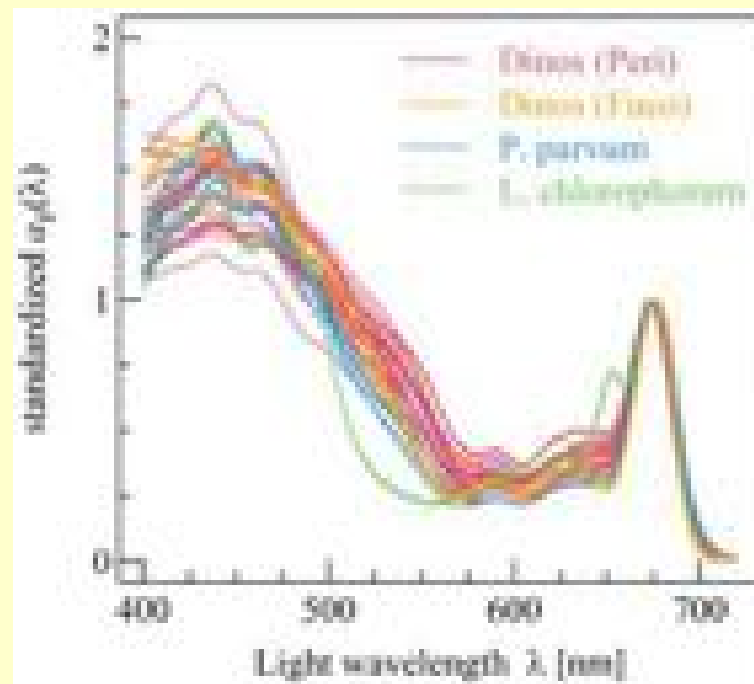
Simple classification method of $a_p(\lambda)$ spectra

Spectra were smoothed (9nm moving average), and 2nd derivative computed. Hierarchical cluster analysis was performed using a similarity index distance.

Diatoms and/or raphidophytes

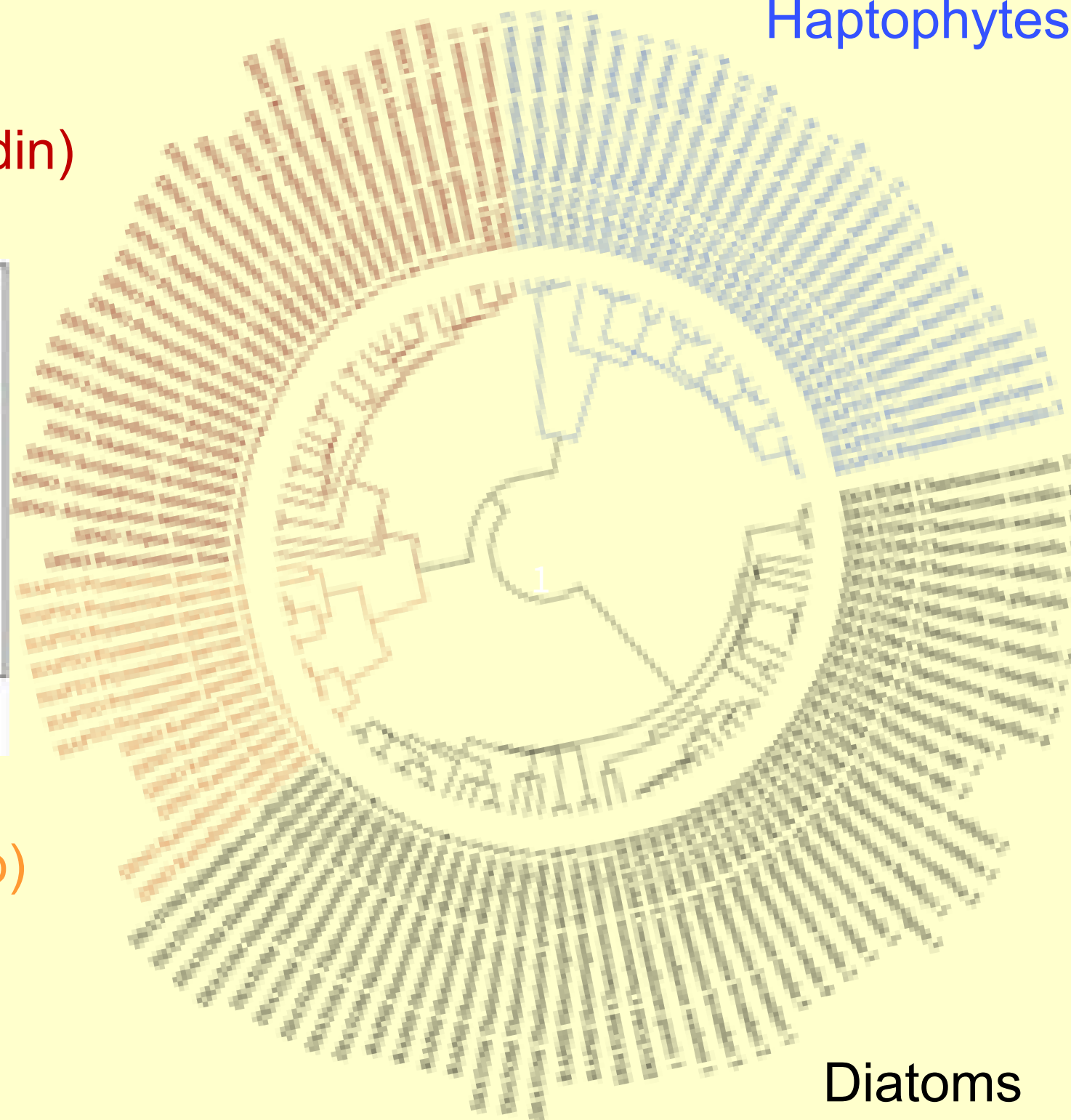
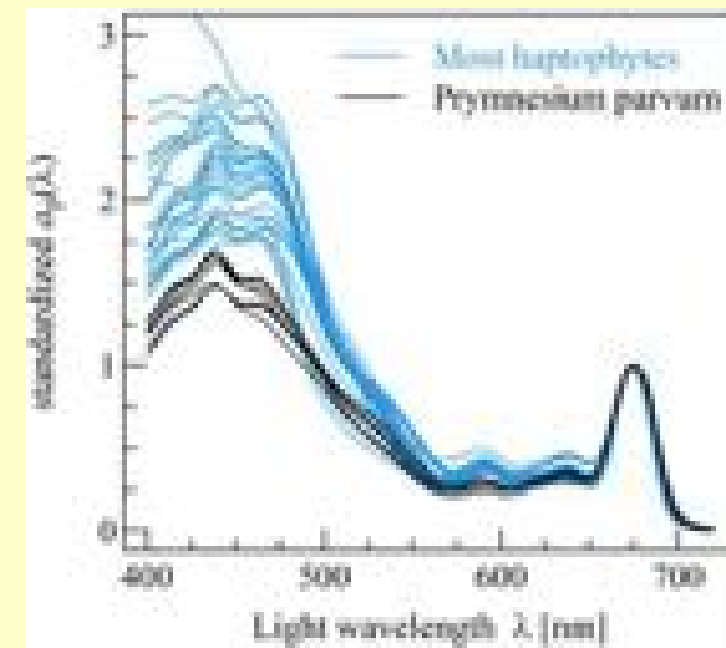
more
challenging
(next slide)

Dinoflagellates (Peridin)

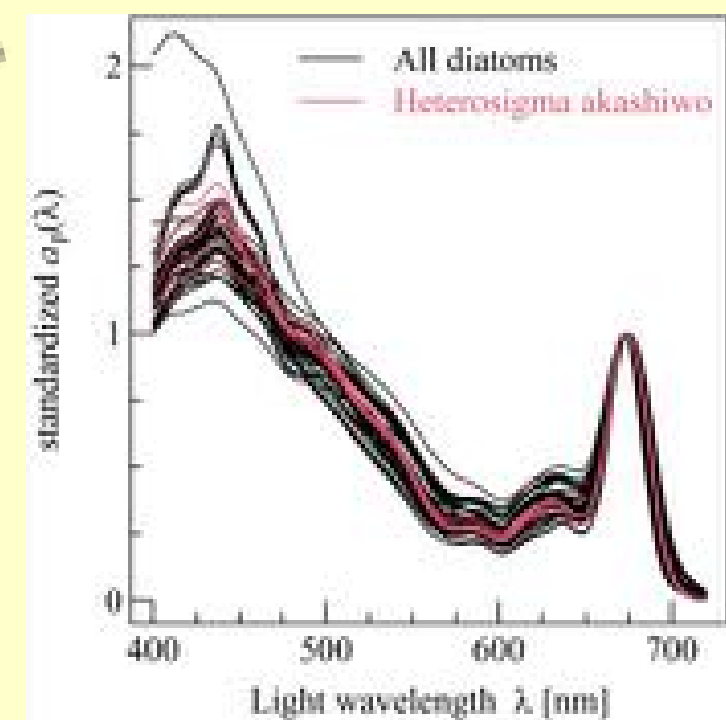


Dinoflagellates (Fucox)

Haptophytes



Diatoms



Scattering coef. of phytoplankton cultures

HABLAB: full characterization of phytoplankton scattering using advanced instruments



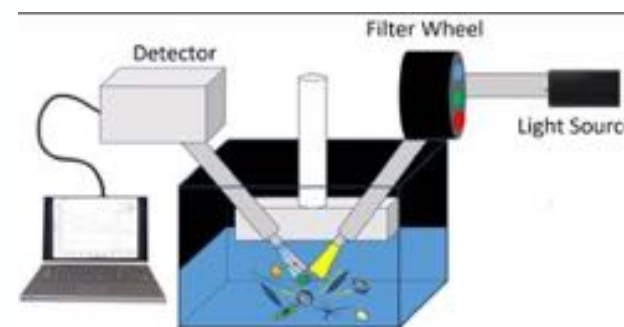
Attenuation $c_p(\lambda)$ and scattering $b_p(\lambda)$ coefficients (ac-s)

Angular scattering at 515 nm $\beta_p(\theta)$

Hyperspectral backscattering coefficient $b_{bp}(\lambda)$

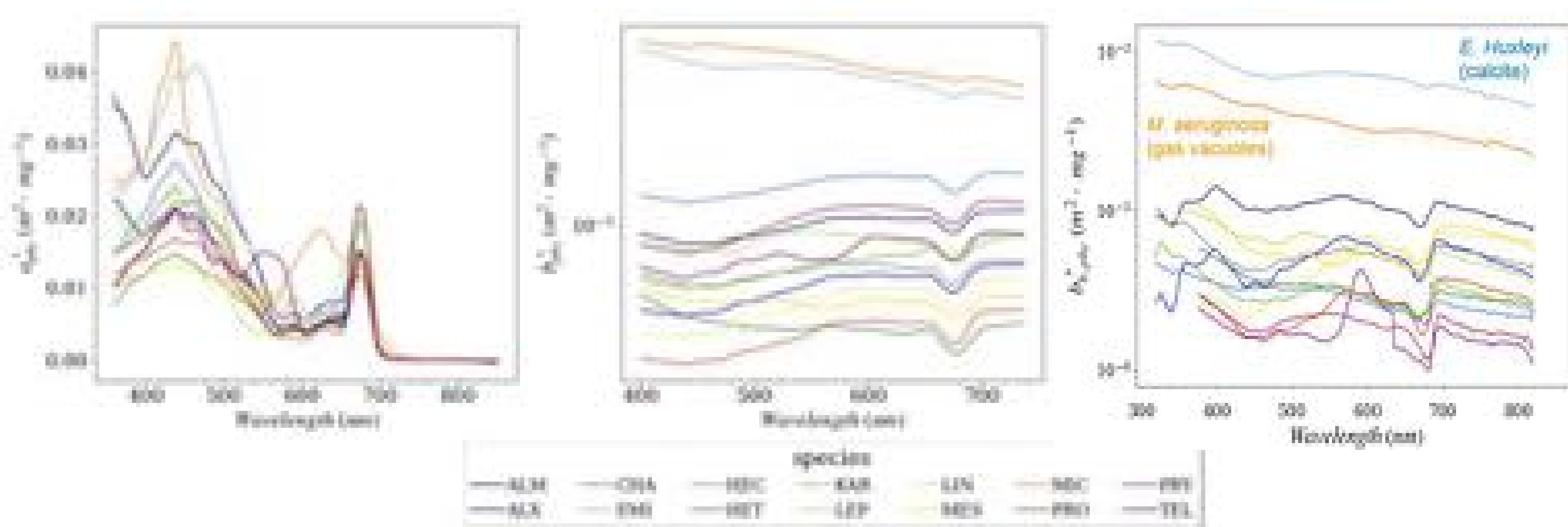


LISST-VSF: directional scattering at 515 nm (phase function + polarization terms)



Hyfi-bb: hyperspectral backscattering through monodirectional VSF measurement (Novak et. al, 2024)

<https://doi.org/10.1364/OE.529061>

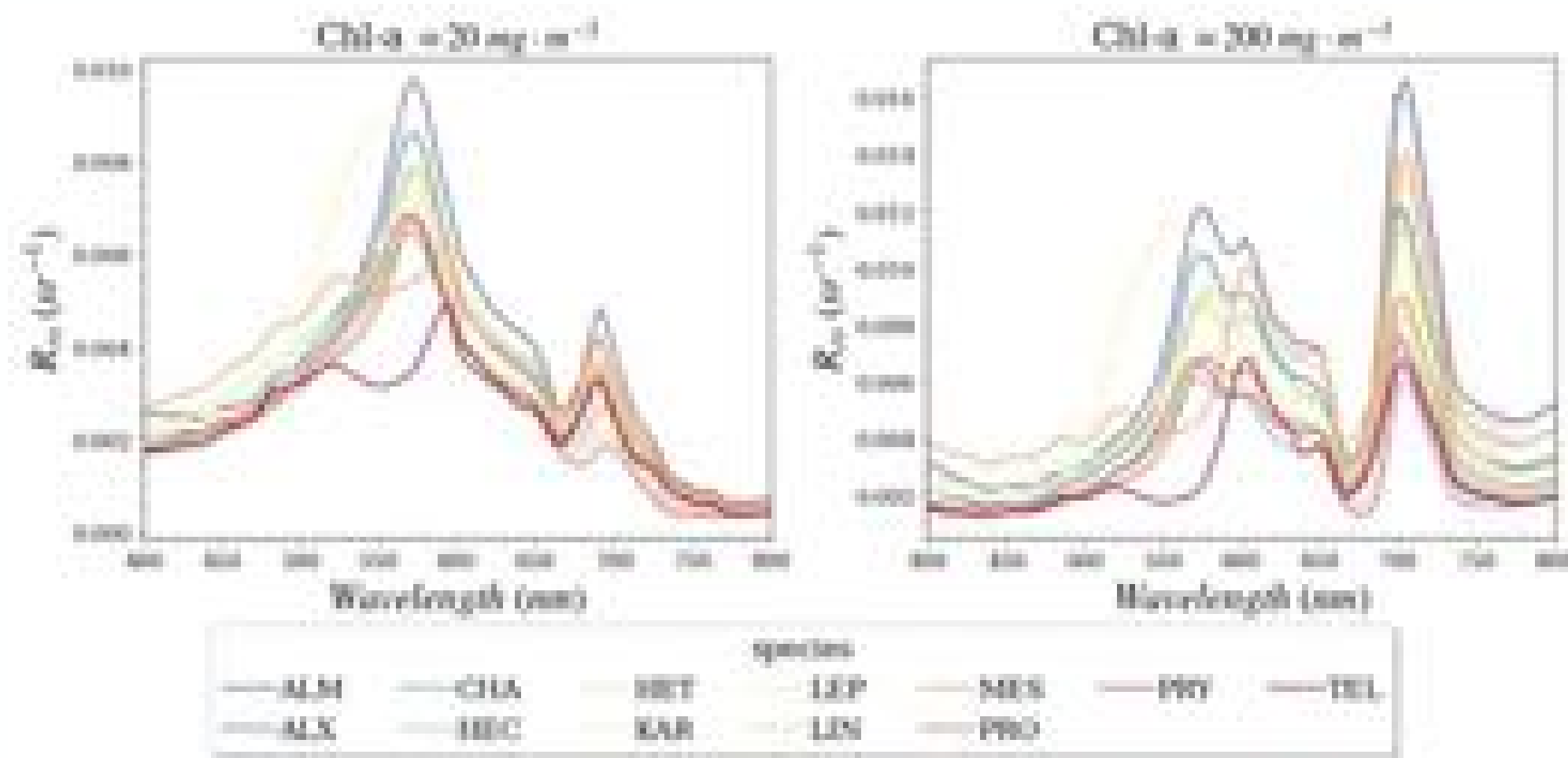


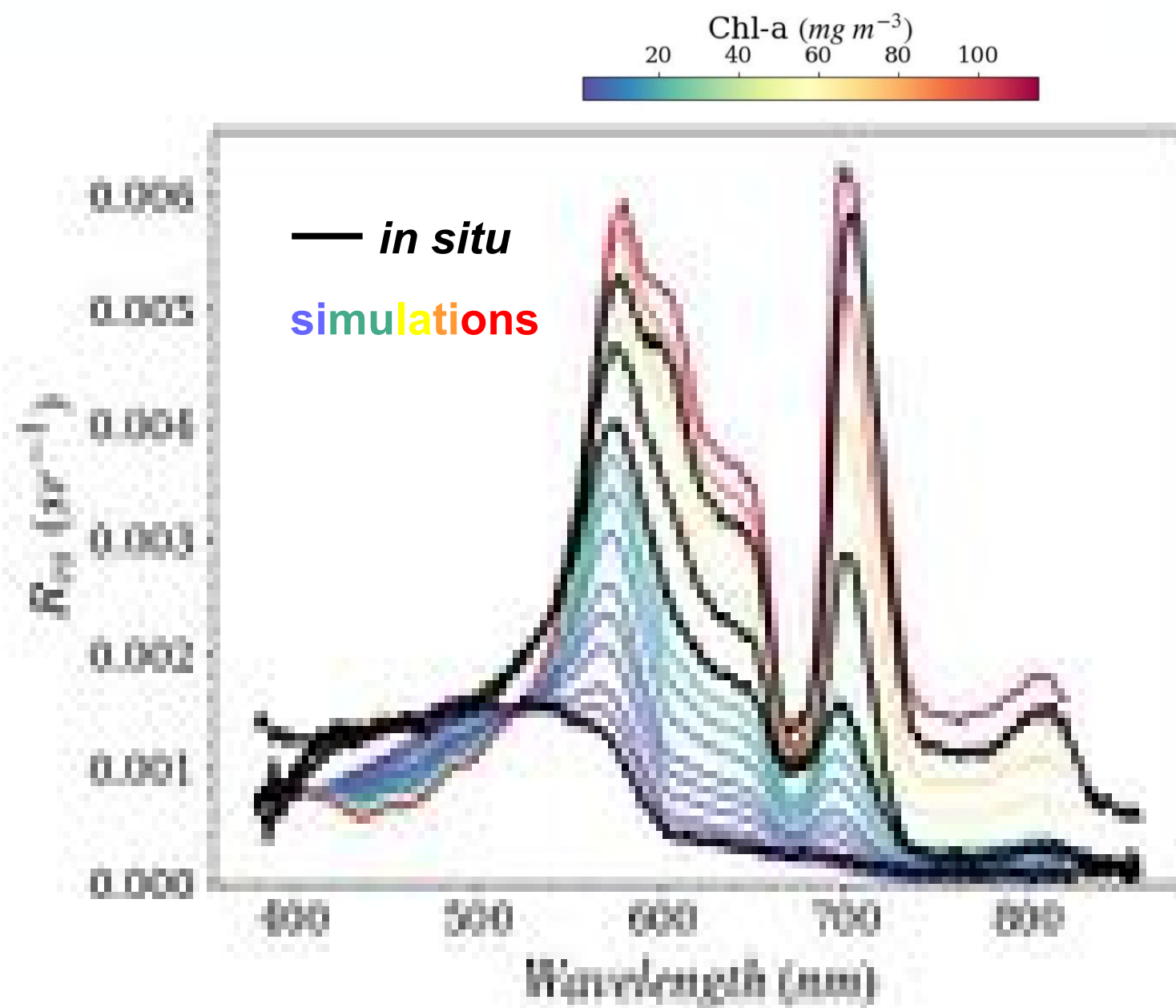
ALM & ALX = *Alexandrium minutum*, CHA = *Chaetoceros muelleri*, HEC = *Heterocapsa triquetra*, KAR = *Karlodinium veneficum*, LIN = *Lingulaulax polyedra*, MIC = *Microcystis aeruginosa*, PRY = *Prymnesium parvum*, EMI = *Emiliania huxleyi*, HET = *Heterosigma akashiwo*, LEP = *Lepidodinium chlorophorum*, MES = *Mesodinium rubrum*, PRO = *Prorocentrum micans*, TEL = *Teleaulax amphioxeia*

RT model from Harmel et al., (in prep.)

$$R_{rs}(\lambda, \theta_s, \theta_v, \Delta\phi) = f\left(\frac{b_b(\lambda)}{a(\lambda)}, \theta_s, \theta_v, \Delta\phi\right)$$

IOP model parameterized using **lab data** $\left\{ \begin{array}{l} a(\lambda) = a_w(\lambda) + a_{\text{cdom}}(\lambda) + [\text{Chl-a}] \mathbf{a}^*_{\text{phy}}(\lambda) + a_{\text{others}}(\lambda) \\ b_b(\lambda) = b_{b,w}(\lambda) + [\text{Chl-a}] \mathbf{b}^*_{b,\text{phy}}(\lambda) + b_{b,\text{others}}(\lambda) \end{array} \right.$





Above-water $R_{rs}(\lambda)$ measurements during red tide

Best fit for *Lingulodinium polyedra*

Confirmed by microscope obs.

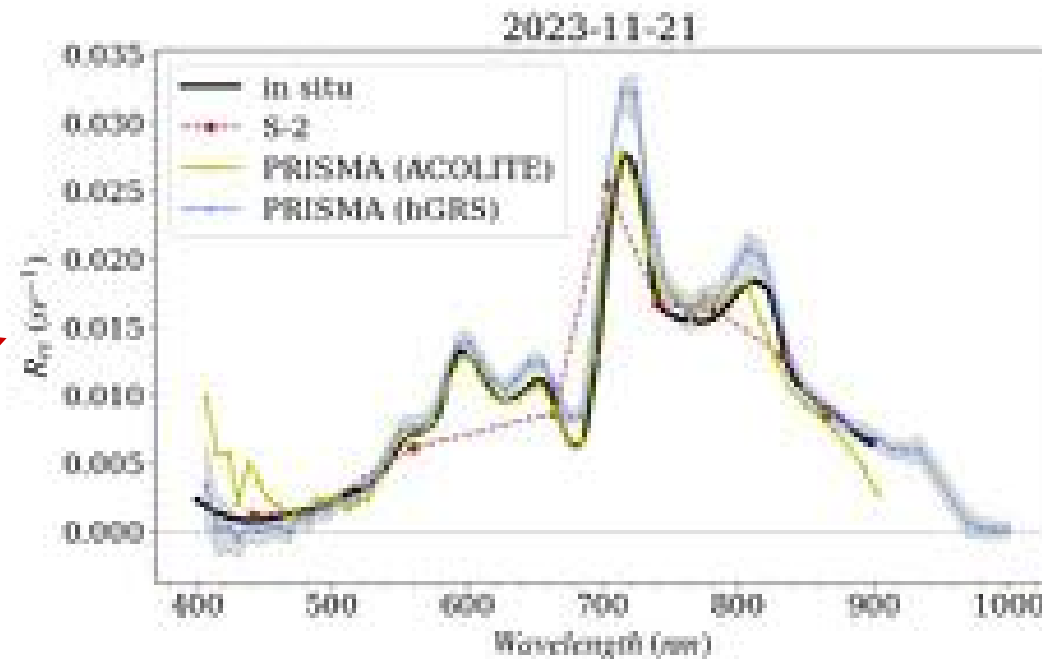


Application to PRISMA

Albufera lagoon (Spain)



Bloom dominated by tiny reddish bacteria
(likely *Raphidiopsis* or *Leptolyngbya*)

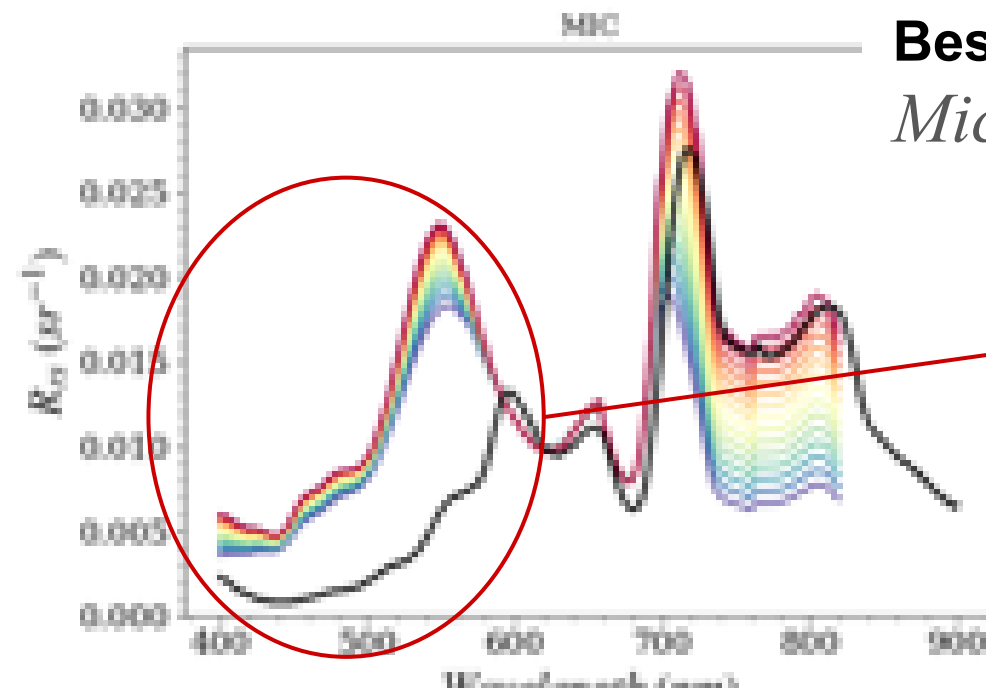


In situ data from
Antonio Ruiz Verdú et al.

Chl-*a*: 230.0 µg/l

CDOM (a_{400}): 2.78 m^{-1}

SPM : 28.57 mg/L



Best fit with cyanobacteria
Microcystis aeruginosa

discrepancies in blue-green

Need to expand IOPs dataset with more cyanobacteria species



Detailed characterization of phytoplankton IOPs for bloom-forming species

Laboratory measurements of volume scattering function, absorption and backscattering coefficients

Lab data used to feed forward & inverse RT model

Preliminary application to HAB monitoring using in situ and PRISMA

Next steps

Expand IOP database to include more cyanobacteria

How many Optical Bloom Types could be identified using hyperspectral satellites?

Validate approach for HAB monitoring and phytoplankton bloom studies

General framework for red tides RS

1. Identify & classify main bloom-forming species

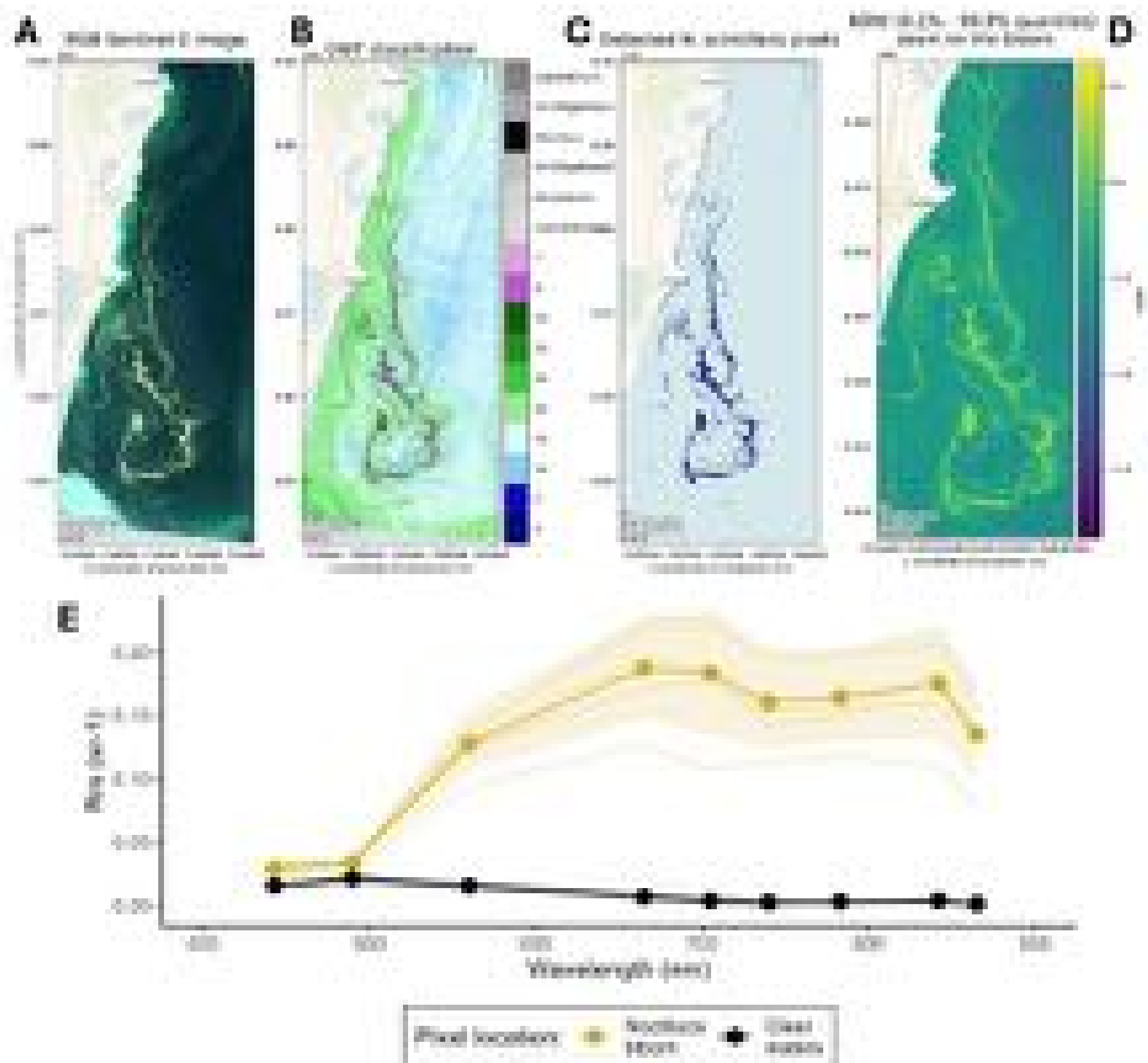
(using i.e. OWT/OBT, spectral derivative, SI, pigments LH, pigments fluorescence, ...)

2. Select most relevant IOP model

3. Inverse IOP, pigments concentration and/or cell abundance

Example using Sentinel-2

results from Victor Pochic and Zoé Besnard



1. OWT & OBT classification

→ *Noctiluca*

2. Specific algorithm

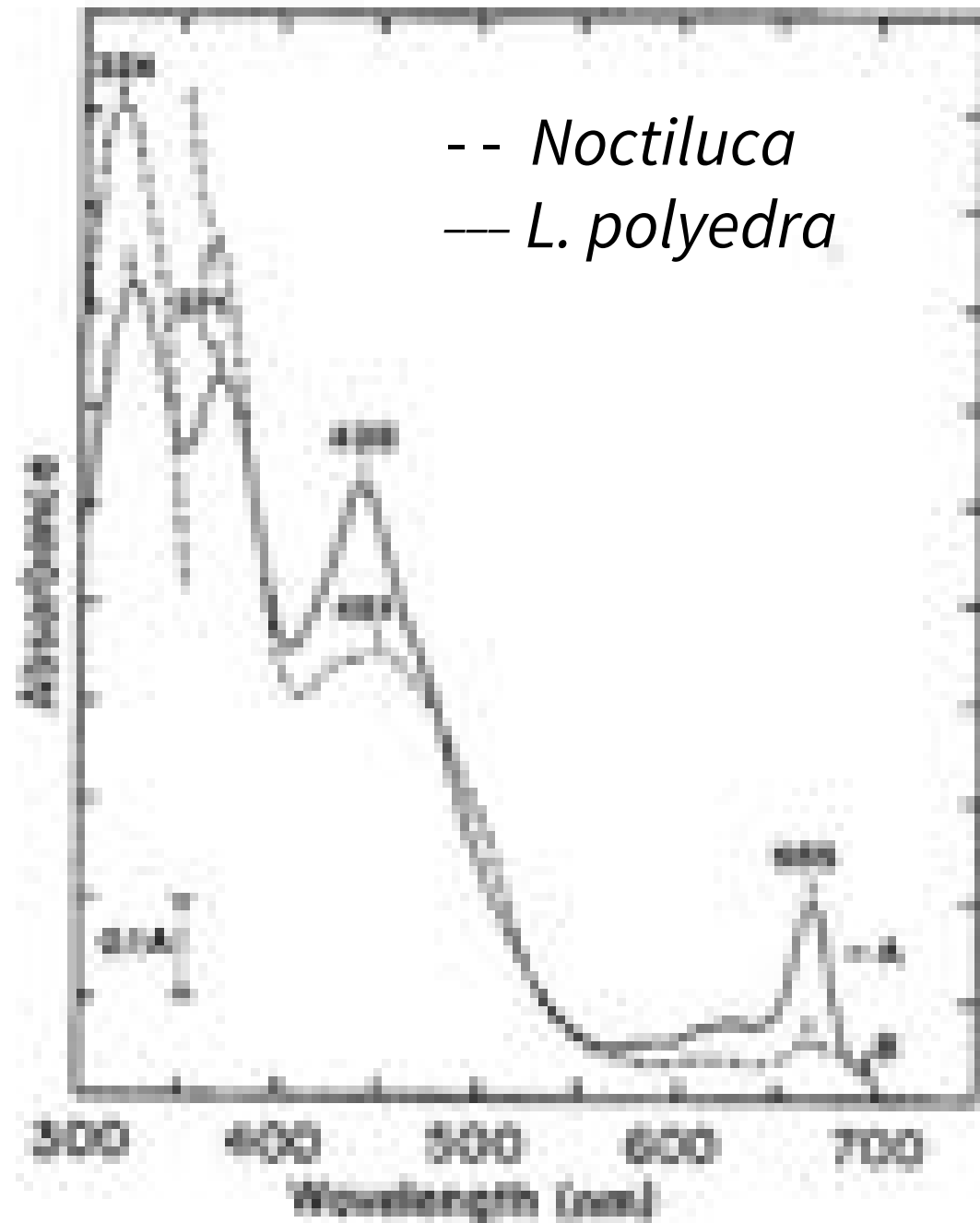
NDNI (Detoni et al., 2023)



Mafra et al., submitted

New research avenues

UV-absorption as signature of dinoflagellates?



Balch & Haxo (1984)



M. Vernet & E. Whitehead

These results indicate that UV-absorbing compounds synthesized by phytoplankton, such as certain dinoflagellates, may be a component of the DOM pool in surface waters of the ocean and contribute to the attenuation of UV radiation in the water column.

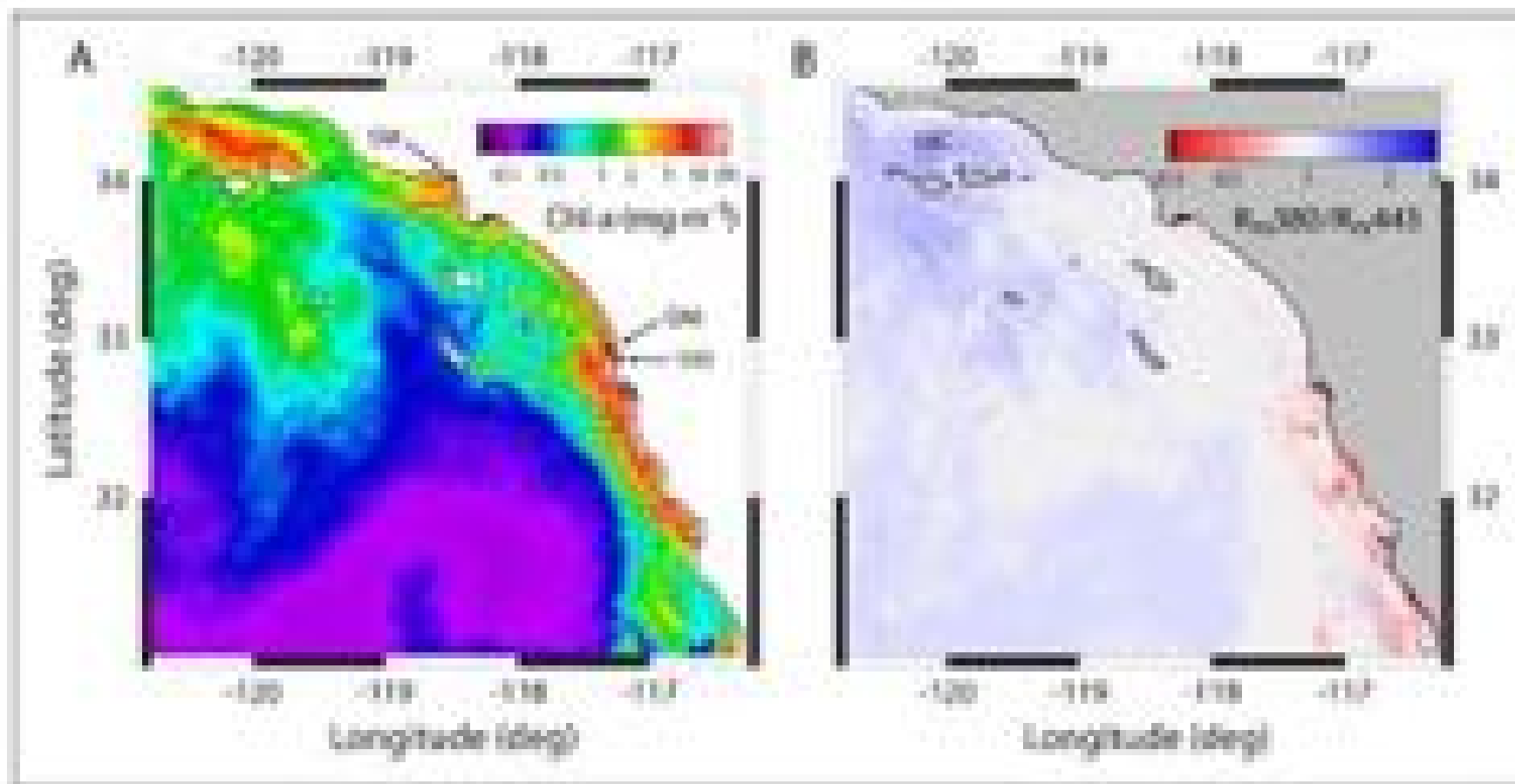
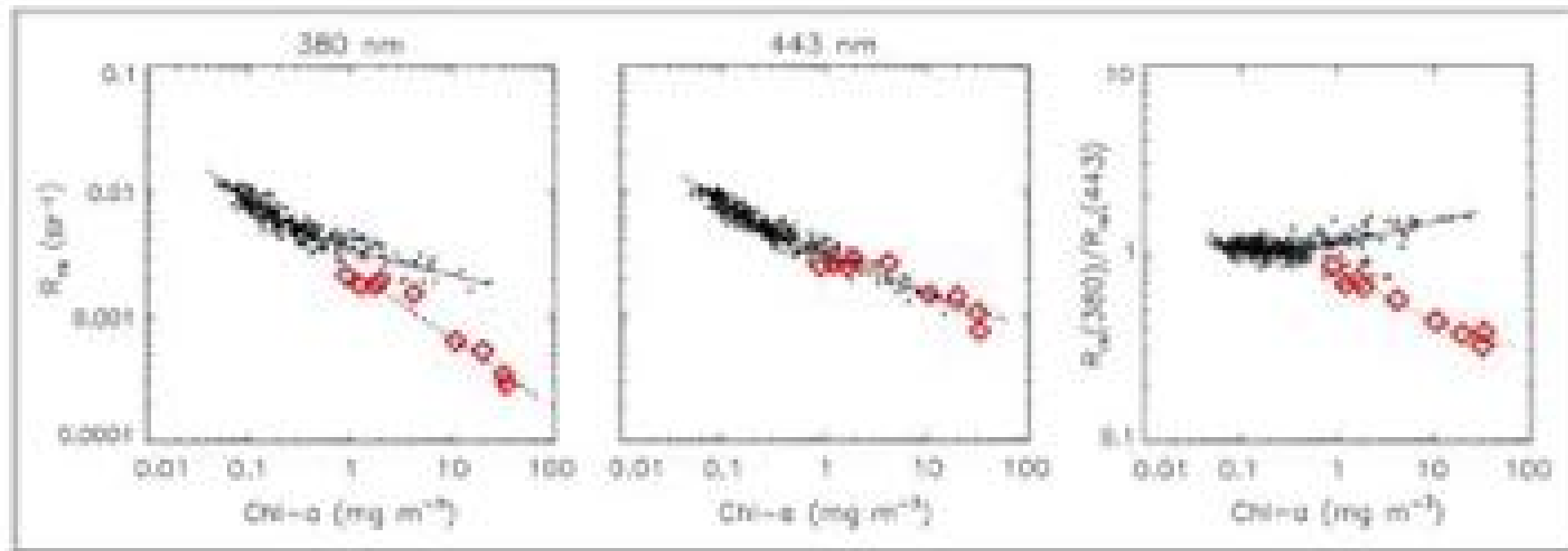
ultraviolet-absorbing compounds:
known as **mycosporine-like amino acids (MAAs)**

Application to HAB remote sensing

RESEARCH ARTICLE

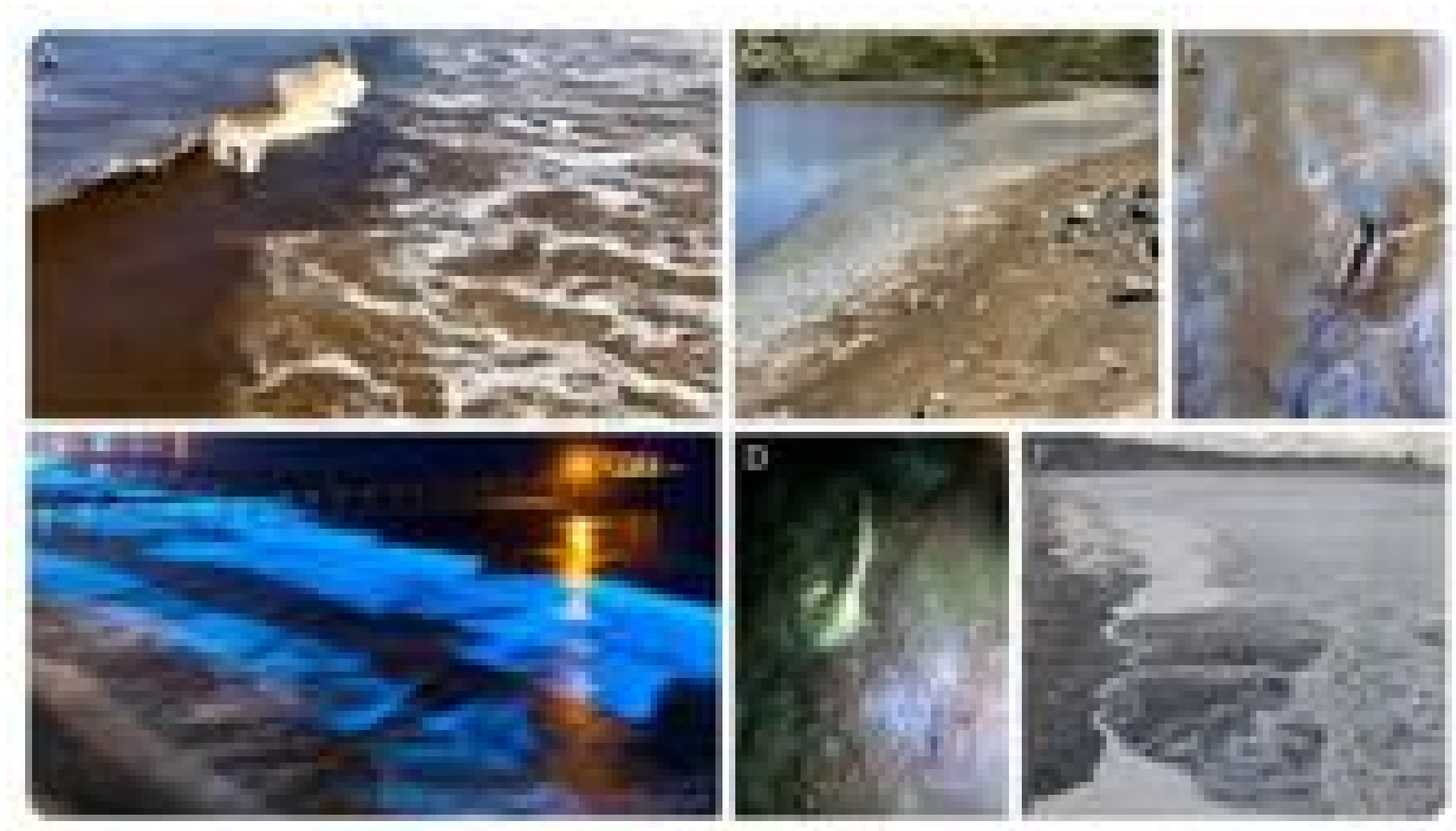
Satellite detection of dinoflagellate blooms off California by UV reflectance ratios

Mati Kahru^{1,*}, Clarissa Anderson², Andrew D. Barton^{1,3}, Melissa L. Carter¹, Dylan Catlett⁴, Uwe Send¹, Heidi M. Sosik⁵, Elliot L. Weiss¹, and B. Greg Mitchell¹



Karhu et al., (2021)

Remote sensing of bioluminescence?

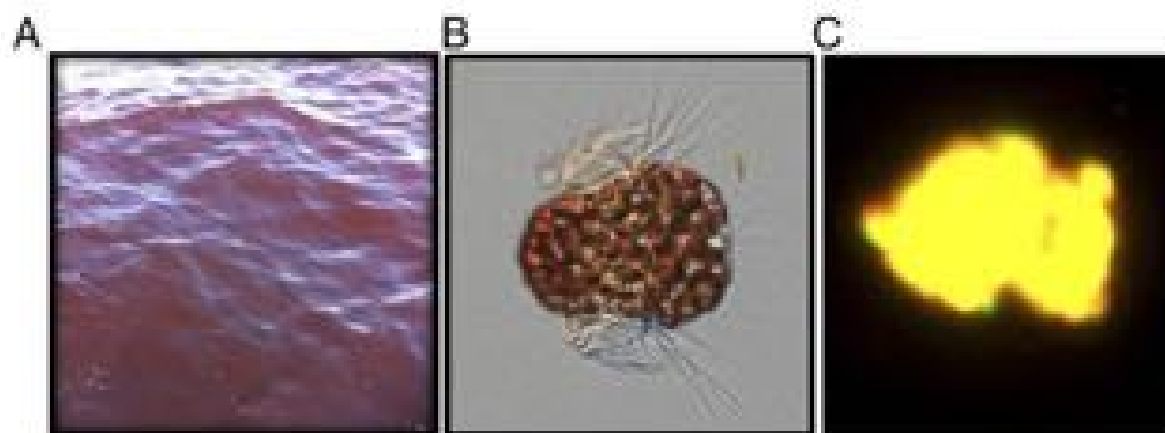
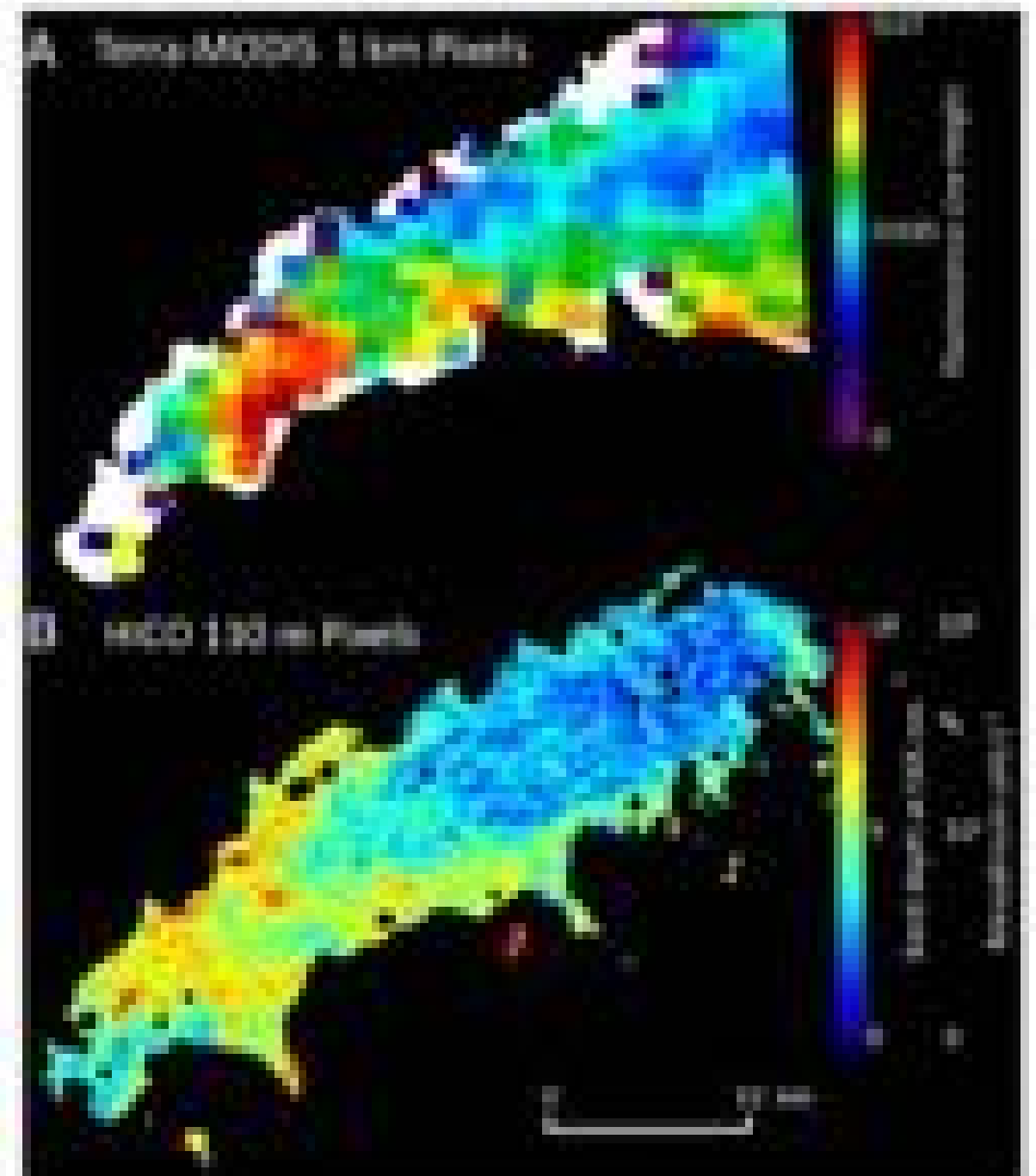
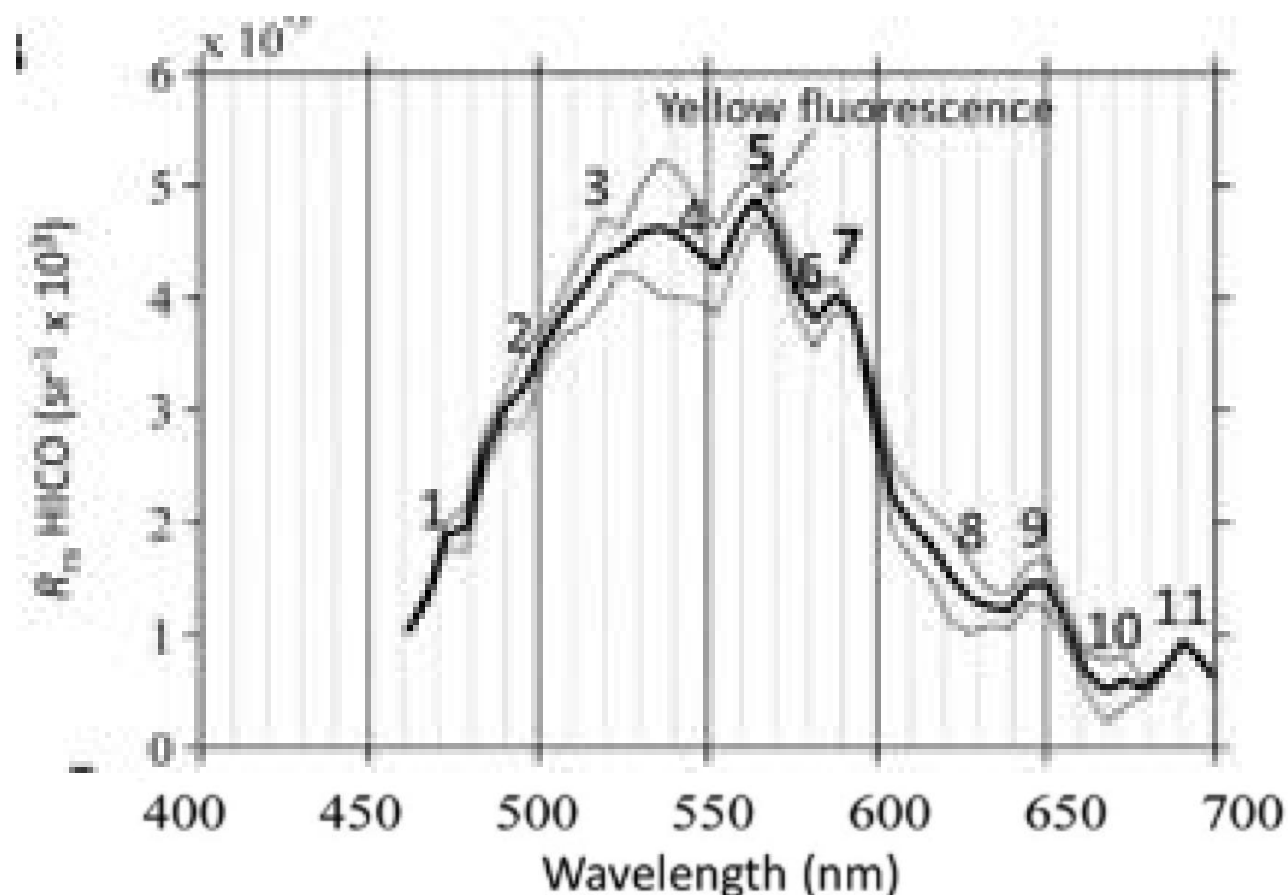


A photo compilation showing aspects of the 2020 *L. polyedra* red tide.

Fluorescence of bio-marker pigments?

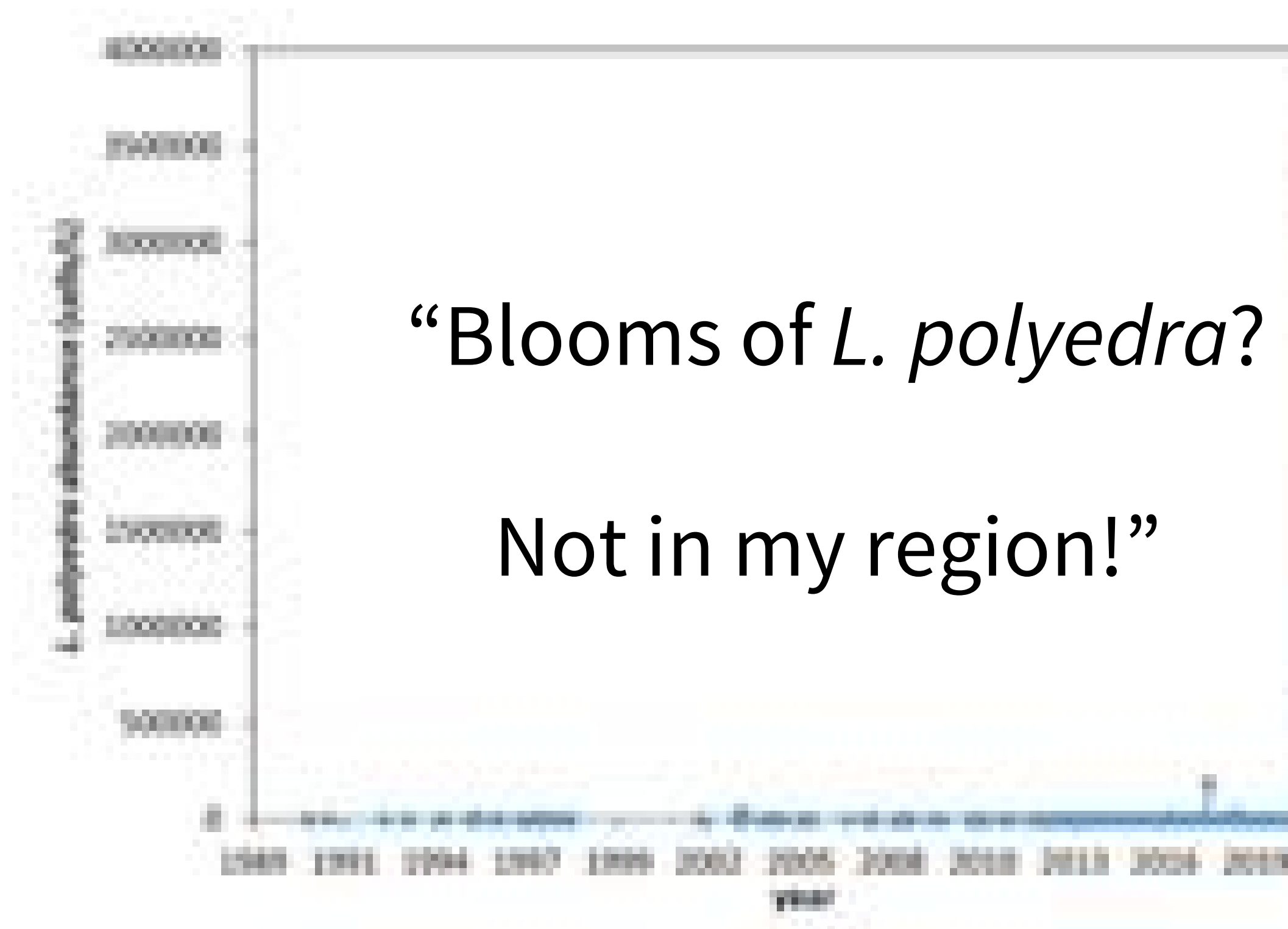
Space station image captures a red tide ciliate bloom at high spectral and spatial resolution

Heidi Dierssen^{a,1}, George B. McManus^a, Adam Chlus^a, Dajun Qiu^{a,b}, Bo-Cai Gao^c, and Senjie Lin^a

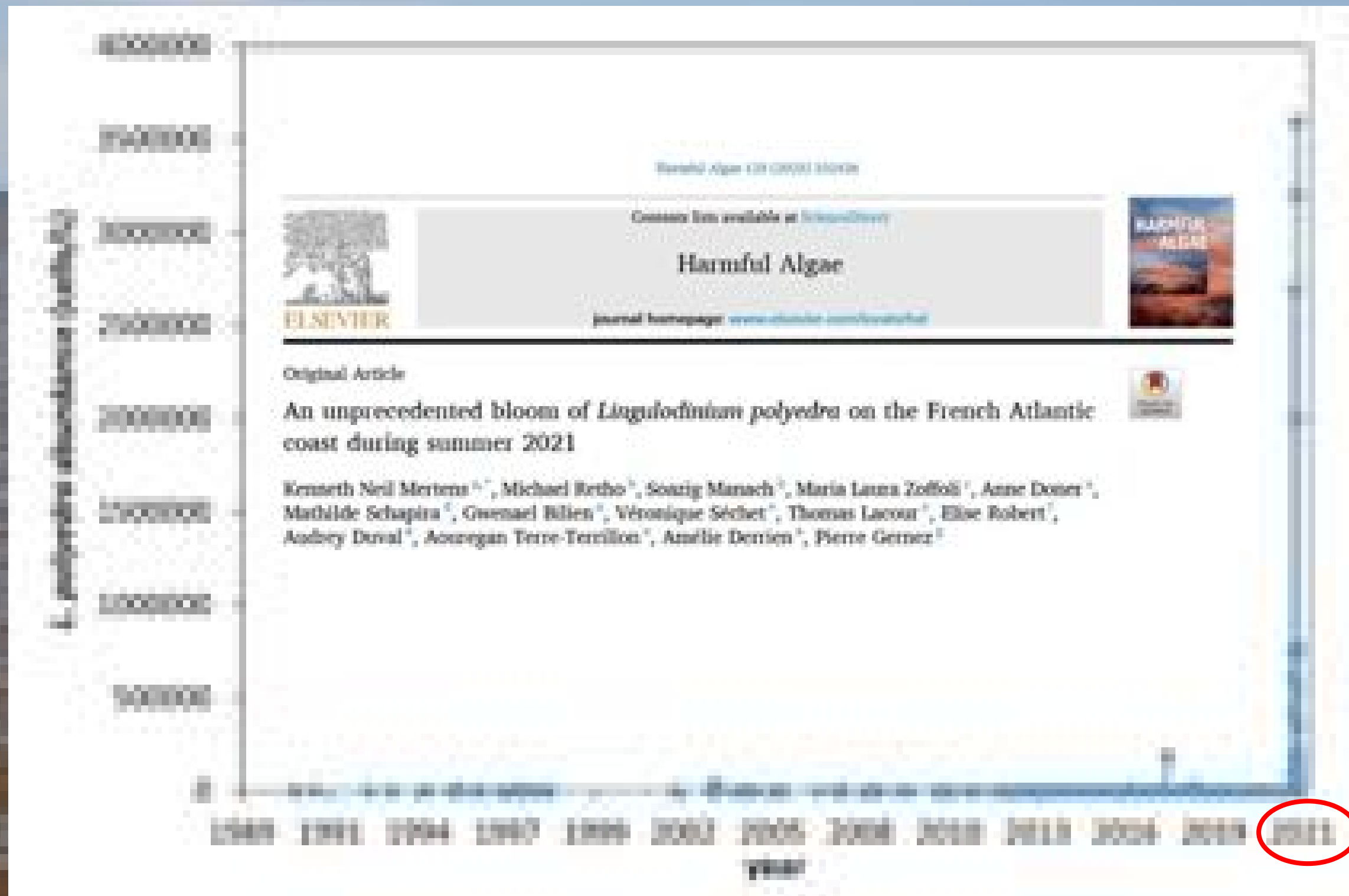


Dierssen et al., (2015)

Do not trust empirical knowledge too much



Surprise!



Mertens et al., (2023)

Emerging HABs

“These findings reveal a BTX-producing *Karenia* that causes substantial detrimental marine ecosystem impacts, representing an emerging international threat with unknown consequences given changing ocean conditions”

nature ecology & evolution

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Received 12 November 2003; accepted 12 November 2003

A catastrophic marine mortality event caused by a complex algal bloom including the brevetoxin producer *Karenia cristata*

Elizabeth A. Minton,¹²³ Christopher A. J. Kelly, Rose North, Chelsea Ko Olson, Mark Douillard, David J. Kelly, Linda G. Lewis, Patricia Goodenough, Gabriel Melgarejo, Spencer Smith, Dorothy E. Lee-Hammond, Robert E. Smith, David M. Donald, Ian M. Macky, Hyeon-Ryong Kim, Shuang Guo, Krista M. Smith, Anne Nelson, Andrew J. Johnson, Scott A. Johnson, Matthew Miller, Anastasia Simionova, Shana Rindler, Cassi Olson, Clifton Williams, & David Smith

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Thank you for your attention!