

SCIAMACHY on ENVISAT

OMI on AURA

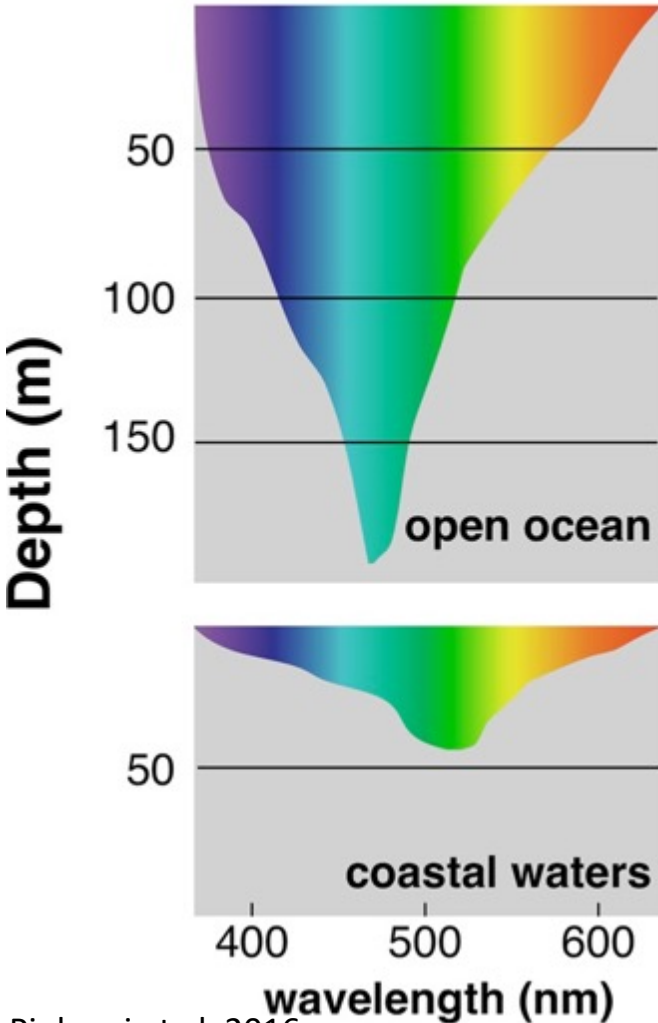
TROPOMI
on S5P

GOME-2
on
Metop

Hijacking other thematic satellite sensors for ocean color application

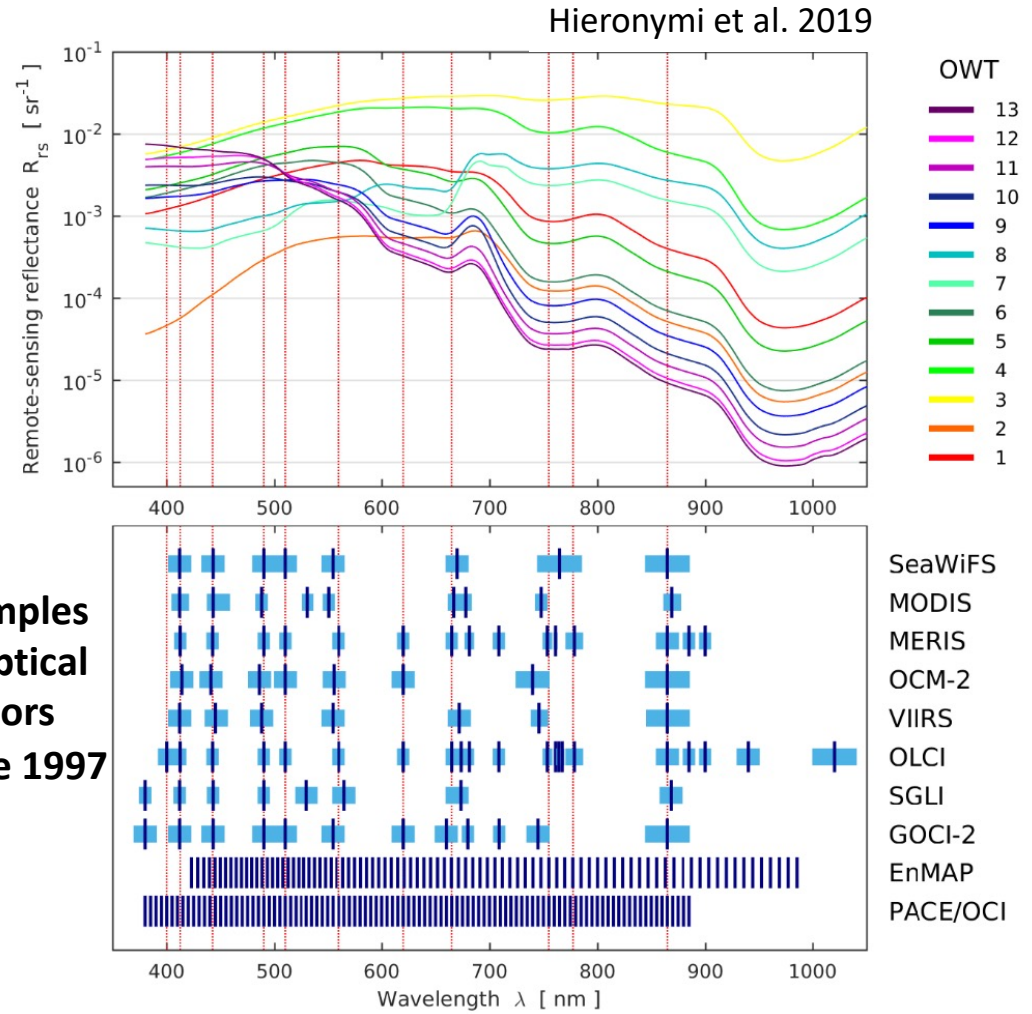
Astrid Bracher

Common Sensors for Ocean Colour & Products



Pinhassi et al. 2016

Examples of optical sensors since 1997



Atmospheric Satellite Sensors for Novel Ocean Colour Products

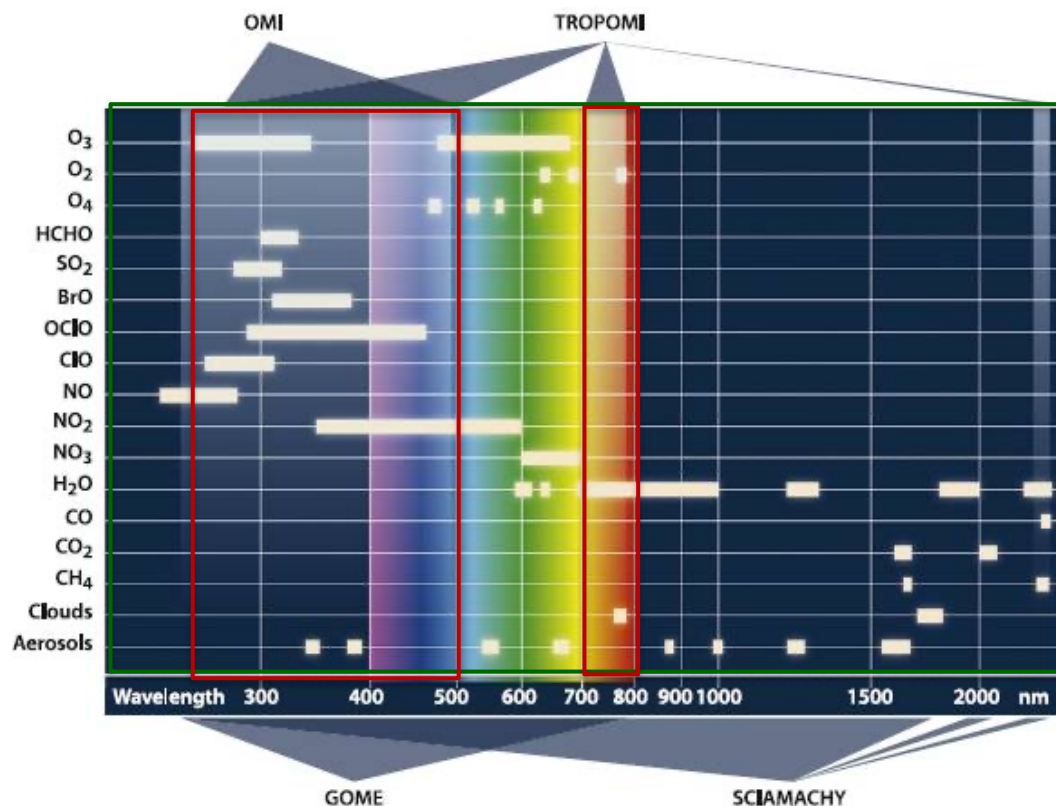
SCIAMACHY

(Scanning Imaging Absorption Spectrometer for Atmospheric Chartography)

ENVISAT

Operation time	2002-2012
Global coverage	6 days
Local overpass time	10:00
Spatial resolution	30 km x 60 km
Spectral resolution*	0.47 nm
Wavelength window	280-2400nm
SNR*	>2000

*for ~450 nm



Atmospheric Satellite Sensors for Novel Ocean Colour Products

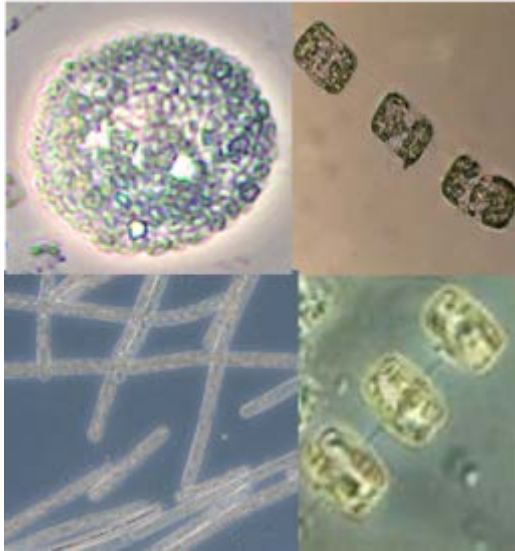
	SCIAMACHY (Scanning Imaging Absorption Spectrometer for Atmospheric Chartography) <i>ENVISAT</i>	OMI (Ozone Monitoring Instrument) <i>AURA</i>	GOME-2 (Global Ozone Monitoring Experiment-2) <i>Metop-A, B, C</i>	TROPOMI (TROPOspheric Monitoring Instrument) <i>Sentinel-5P</i>
Operation time	2002-2012	Since 2004	Since 2006	Launch 2017
Global coverage	6 days	daily	~ 1.5 days	daily
Local overpass time	10:00	13:30	9:30	13:30
Spatial resolution	30 km x 60 km	13 km x 24 km	40 km x 40/80 km	3.5 km x 7 km
Spectral resolution*	0.47 nm	~0.63 nm	~0.51 nm	~0.54 nm
SNR*	>2000	500	1000	1000
Wavelength window	280-2400nm	280-500 nm	280-800 nm	270-500 nm & 675-775 nm

*for ~450 nm

- ✓ Sources of CDOM, UV-absorbing compounds.
- ✓ Shortwave radiation budget in the ocean
- ✓ Better descriptors of primary production & ocean carbon pool

Novel Ocean Color Products from Atmospheric Sensors

PFTs: phytoplankton functional types



PFT-CHL from spectral inversion



Only empirical multispectral retrievals

$K_d(\lambda)$: Spectral diffuse attenuation



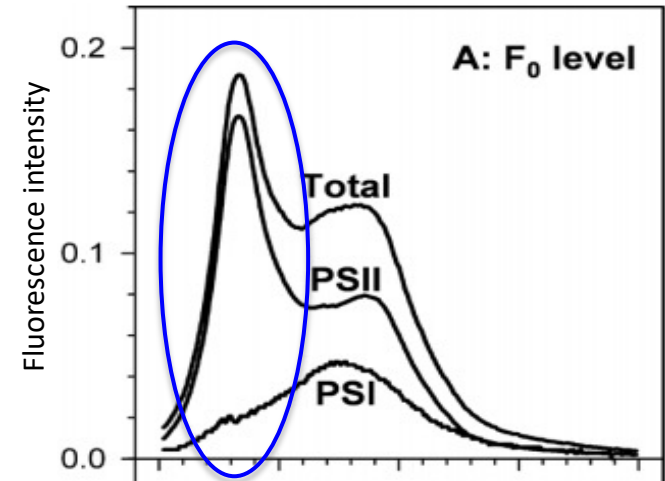
Oelker PhD thesis 2021

Distribution of underwater radiation in UV and short blue



Only KD490 from multispectral

FLH: Chlorophyll fluorescence line height



Spectral signatures of fluorescence in hyperspectral data



FLH multispectral not at peak and not accounting for CHL absorption effect

Atmospheric Sensors for Ocean Colour: Retrieval



PhytoDOAS (Phytoplankton Differential Optical Absorption Spectroscopy)

Bracher et al. 2009, Sadeghi et al. 2012, Dinter et al. 2015, Losa et al. 2017, Oelker et al. 2019 & 2022

Based on Beer-Lambert law



absorption / scattering cross sections of all constituents

concentration of all constituents

light path

PhytoDOAS equation



Polynomial Oceanic contribution

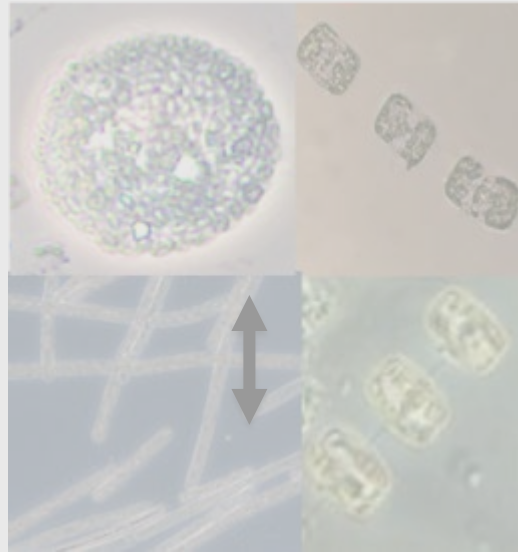
Slant column density (fit factors)

Atmospheric contribution

Differential absorption cross sections

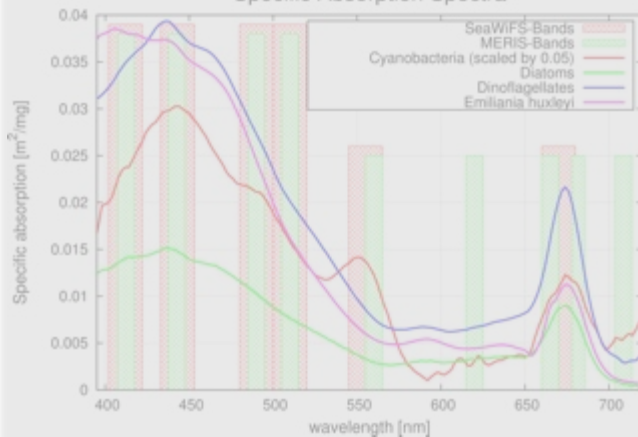
Novel Ocean Color Products from Atmospheric Sensors

PFTs: phytoplankton functional types



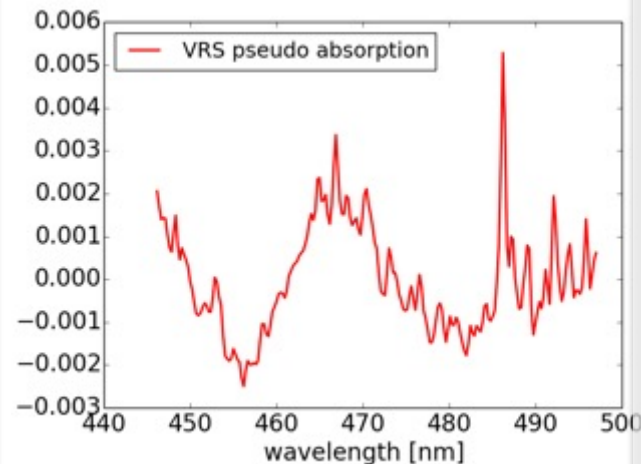
Photos by Sonja Wiegmann - AWI

Specific Absorption Spectra



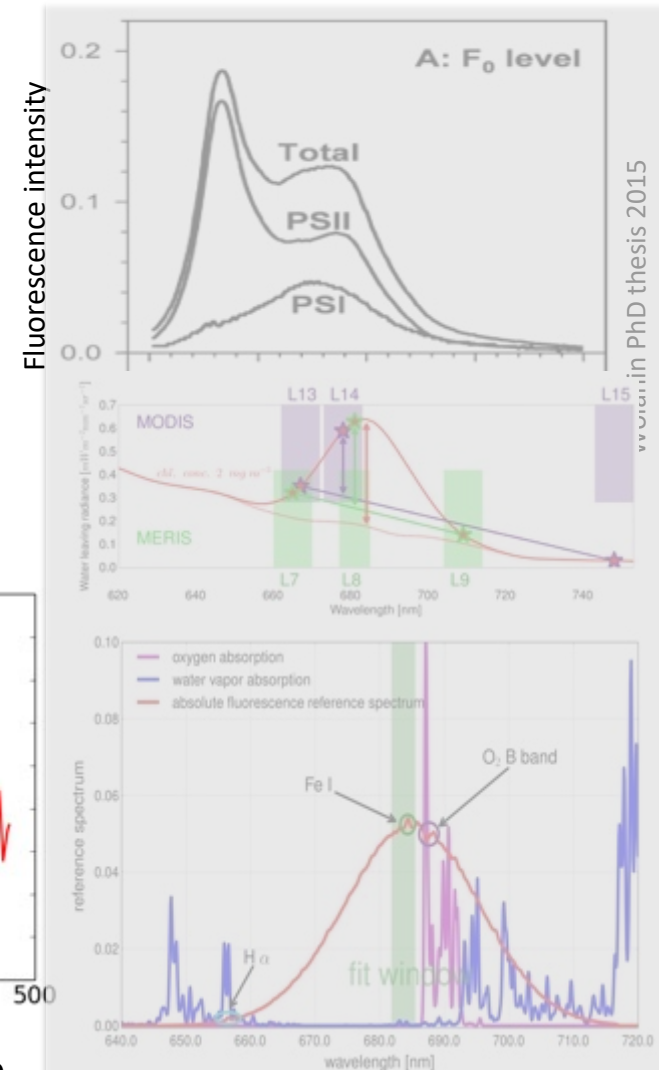
Bracher et al. 2009, Sadeghi et al. 2012, Losa et al. 2017

$K_d(\lambda)$: Spectral diffuse attenuation



Dinter et al. 2015, Oelker et al. 2019, Oelker et al. 2022

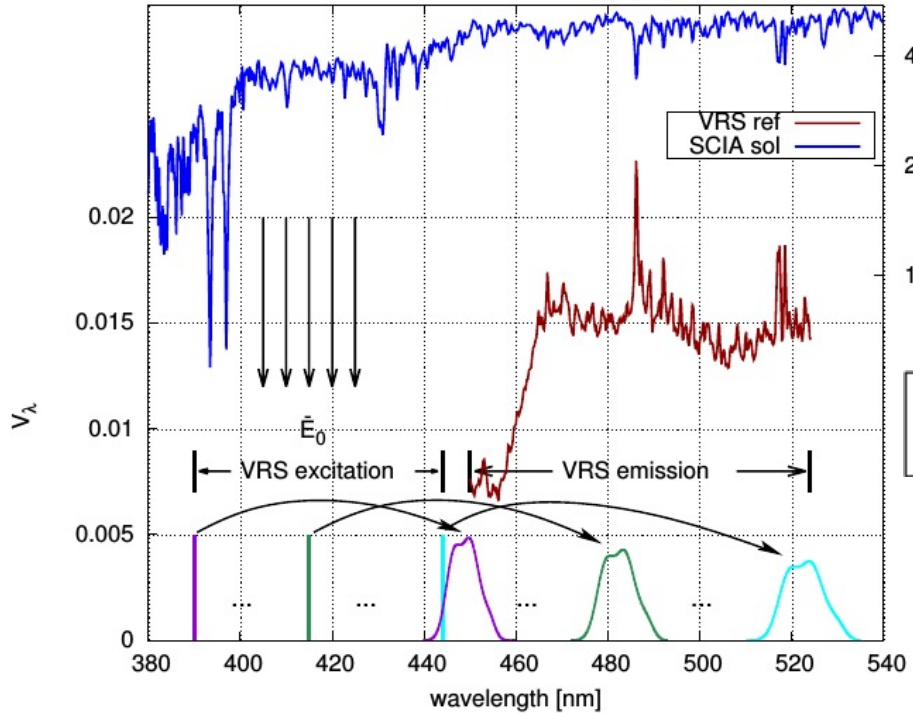
FLH: Chlorophyll fluorescence line height



Wolanin et al. 2015, Joiner et al. 2016, Köhler et al. 2020

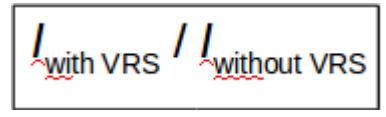
v.vorgerin PhD thesis 2015

Ocean light attenuation (Kd) utilizing Vibrational Raman Scattering (VRS) identified in hyperspectral data



Derived VRS fit-factor used to calculate

- diffuse attenuation (K_d)
- PFT Chla (= PFT-fit factor / VRS-fit factor * X)



Dinter et al. (2015)

- Inelastic scattering process: photons \leftrightarrow water molecules
- The more photons, the more inelastic scattering events
- Causes filling in of Fraunhofer lines in top-of-atmosphere (TOA) radiances
- VRS in atmosphere is negligible



Spectral signature of VRS in TOA radiances \leftrightarrow **amount of photons in the ocean**
 seen in GOME: Vasilkov et al. 2002, Vountas et al 2003;

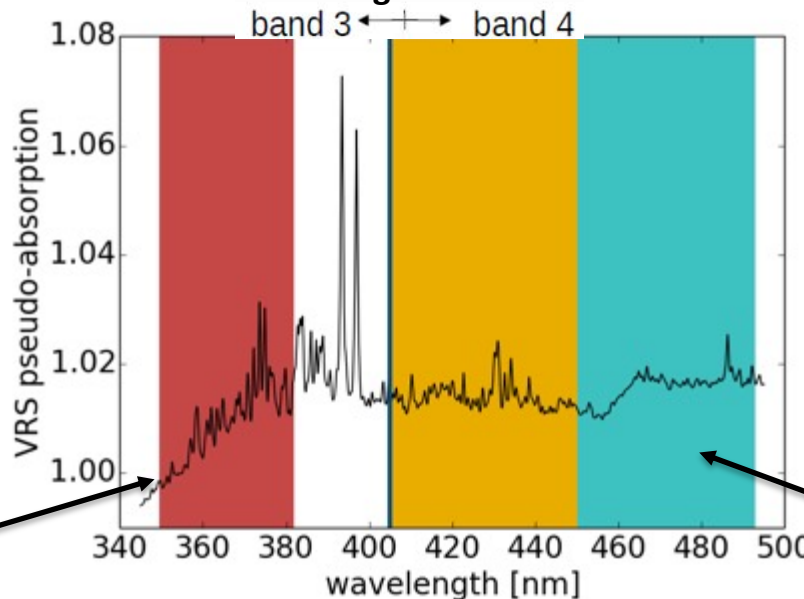
SCIAMACHY: Vountas et al. 2007, Dinter et al. 2015, Oelker et al. 2019
 OMI and TROPOMI: Oelker et al. 2019, Oelker et al. 2022

Diffuse Attenuation (Kd) Products

(derived from VRS via LUT) from hyperspectral atmospheric sensors



Modelled Filling-in of Fraunhofer Lines



Via **PhytoDOAS** to derive inelastic scattering (VRS, RRS) based on their simulated **pseudo-absorption** using coupled Ocean-Atmosphere **RTM SCIATRAN**.

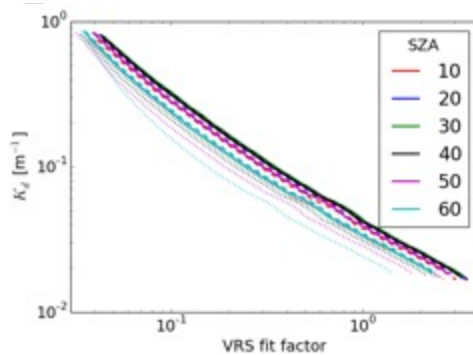
All three Kd in *Oelker et al. FMARS 2022*

Vountas et al. OS 2007

Dinter et al. OS 2015
Oelker et al. OE 2019

VRS fit:
LUT
↓
Kd product:

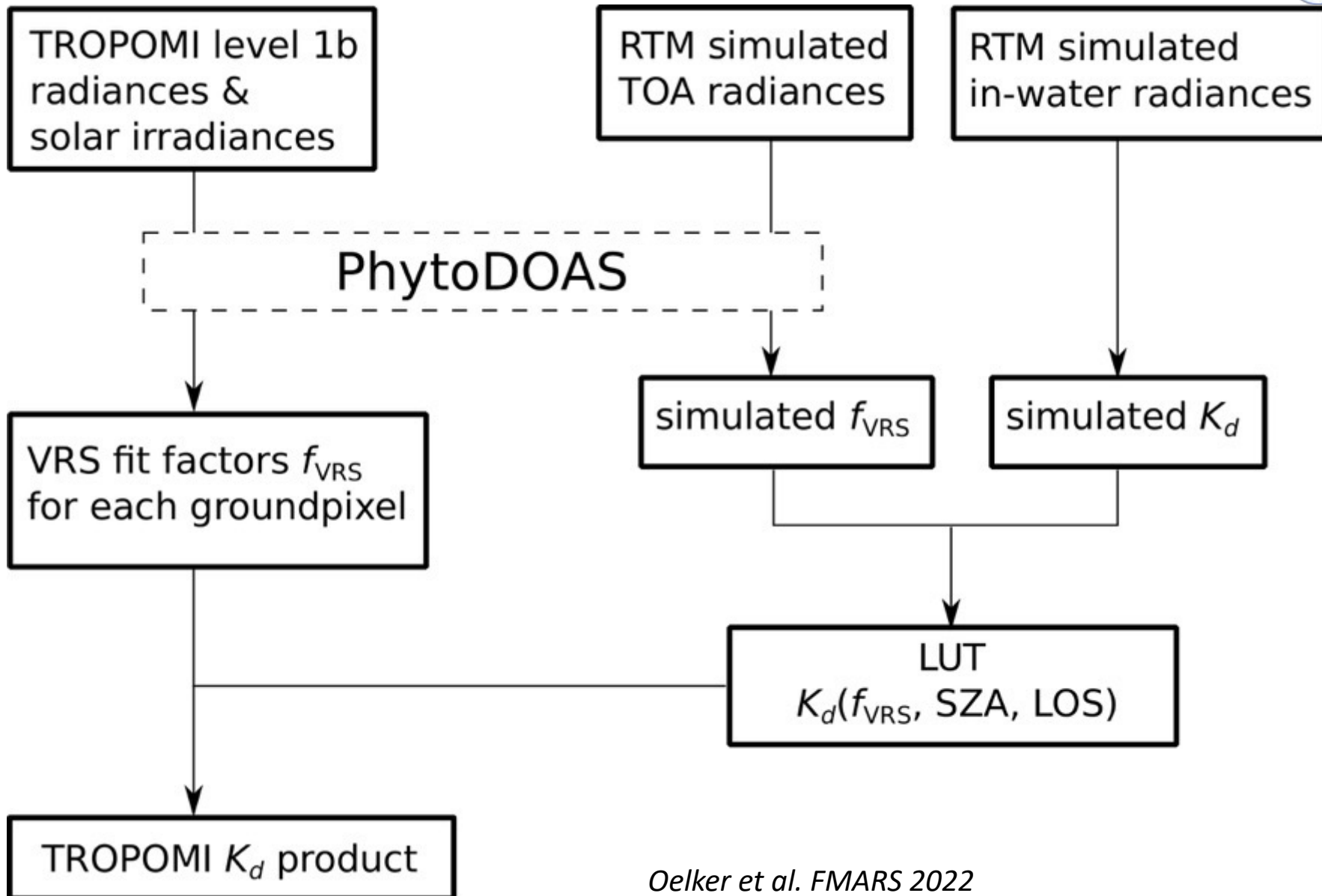
UV	short-blue	blue
349.5 – 382 nm	405 – 450 nm	450 – 493 nm
UVAB	UVA	short-blue
312.5 – 338.5 nm	356.5–390 nm	390 – 423 nm



LUT (Look-Up Table) from coupled O-A RTM links VRS-Fit factor in spectral retrieval range to Kd accounting for **solar zenith and viewing angle geometry**.

Diffuse Attenuation (K_d) Products

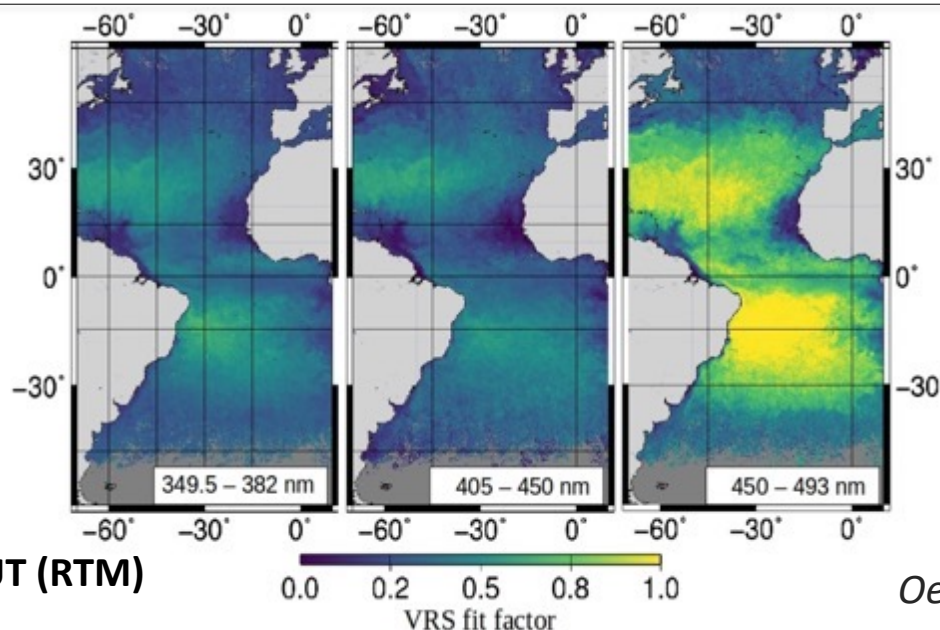
(derived from VRS via LUT) from hyperspectral atmospheric sensors



Diffuse attenuation (K_d) in UVAB, UVA and short-blue from S5P's instrument TROPOMI



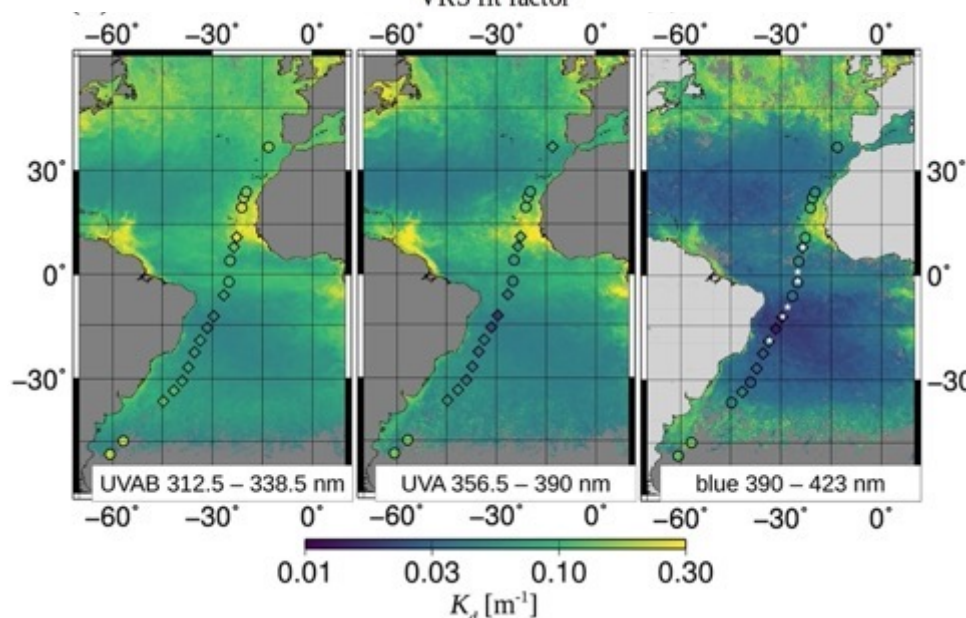
Inelastic Scattering (VRS) in Ocean from TROPOMI



LUT (RTM)

Diffuse Attenuation in Ocean from TROPOMI and matching in situ data

Data at:
<https://doi.org/10.1594/PANGAEA.940352>

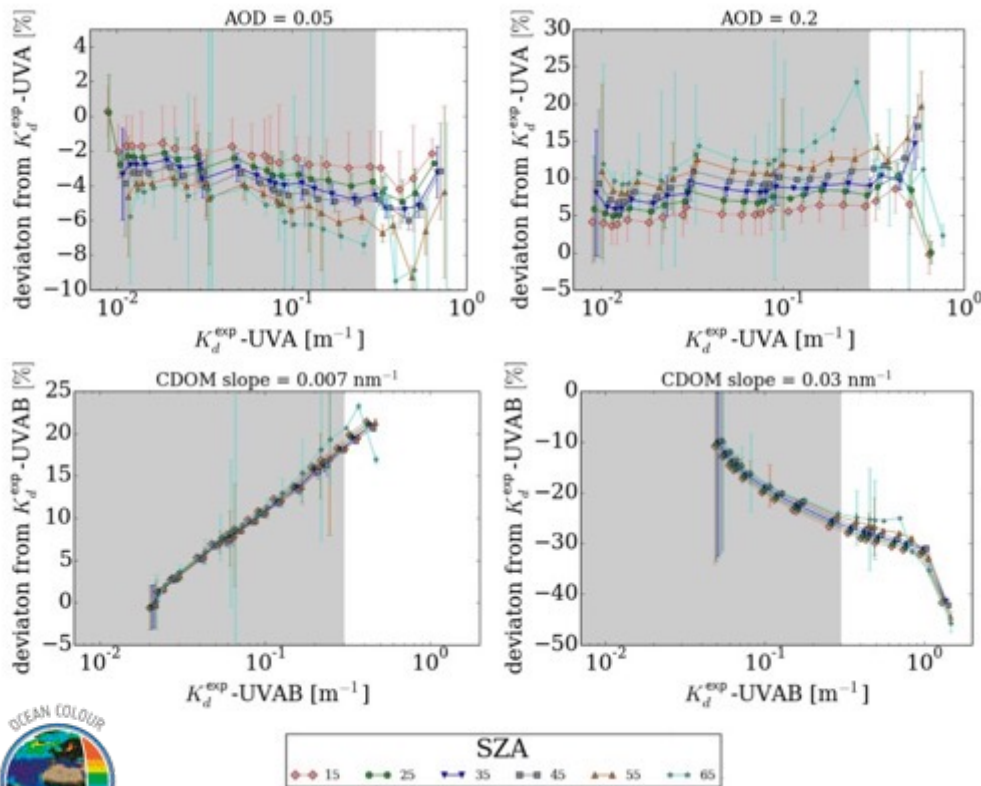


First time K_d retrievals in UV-AB & UV-A from satellite UV data (globally, via inversion).

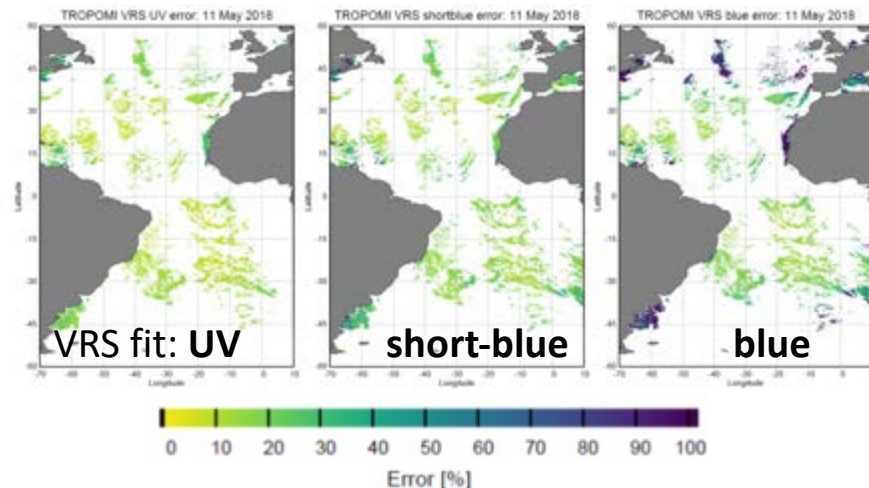


TROPOMI Diffuse Attenuation (Kd) in UVAB, UVA and short blue: uncertainties

Retrieval sensitivities to parametrizations of Look-Up-Table



Specific fit factor error [%]



Error contributions as input for pixel-by-pixel uncertainty.

Overall uncertainty (max. 35%-45%) ~ multispectral Kd490 & CHL retrieval errors.

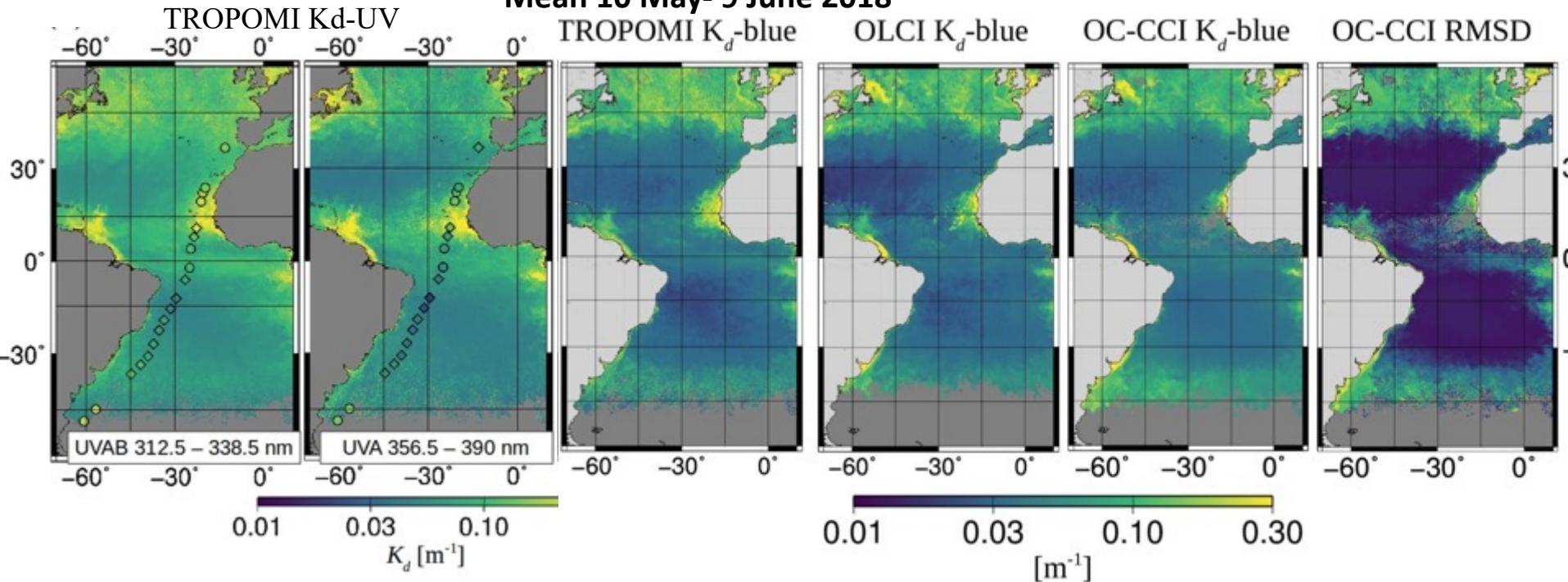
Max. fit and model error [%] of TROPOMI Kd (<0.3

Kd	Fit	α_{CDOM}	CDOM (nm ⁻¹)	α_{ph}^*	AOD	WS	O ₃
UVAB	10	10	30	20	5	3	3
UVA	15	10	25	20	15	5	0
blue	20	20	25	10	20	10	0



TROPOMI Kd comparison to in-situ and OLCI & OC-CCI KD490

Mean 10 May- 9 June 2018



TROPOMI Kd against in situ [n=42]

TROPOMI (TRP) vs OLCI (OL) vs OC-CCI (OC)

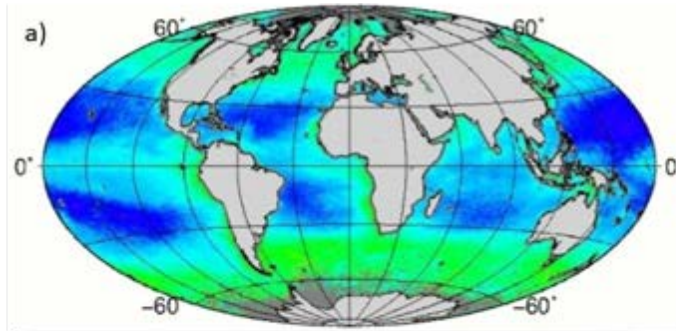
Criteria	Kd-UVAB	Kd-UVA	Kd-blues	Criteria	TRP vs OC	TRP vs OC	OL vs OC
r	0.72	0.54	0.54	r	0.65	0.58	0.80
Bias (m^{-1})	-0.026	0.015	-0.008	Bias (m^{-1})	-0.00	-0.01	-0.01
MAE (m^{-1})	0.026	0.026	0.012	MAE (m^{-1})	0.03	0.03	0.02
RMSD (m^{-1})	0.031	0.029	0.015	RMSD (m^{-1})	0.07	0.08	0.06

Short-blue light attenuation derived from different atmospheric hyper-spectral sensors

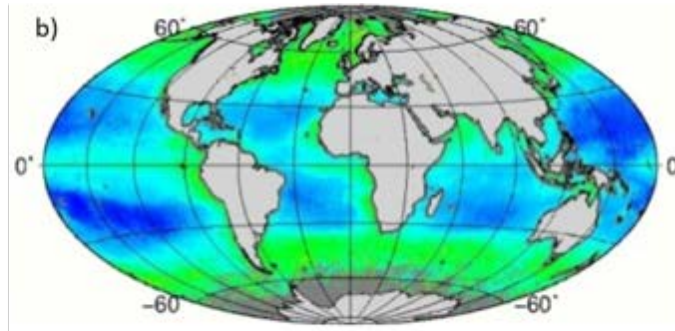


Important for understanding **biogeochemical processes** and **heat budget** of global ocean

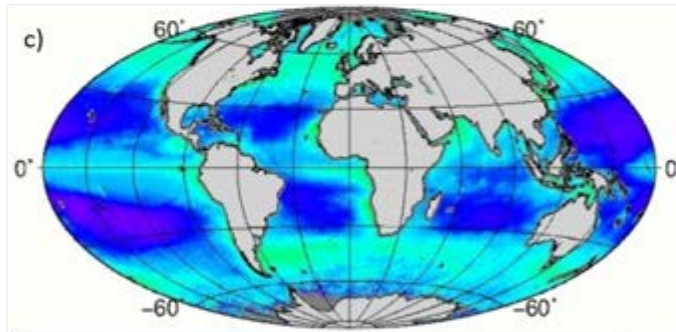
SCIAMACHY



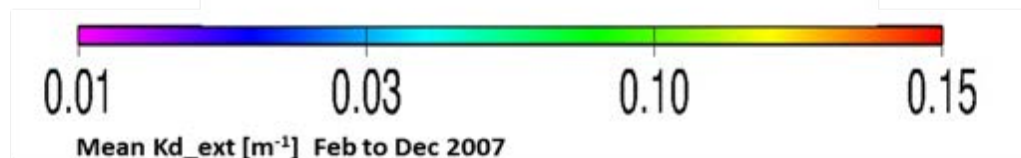
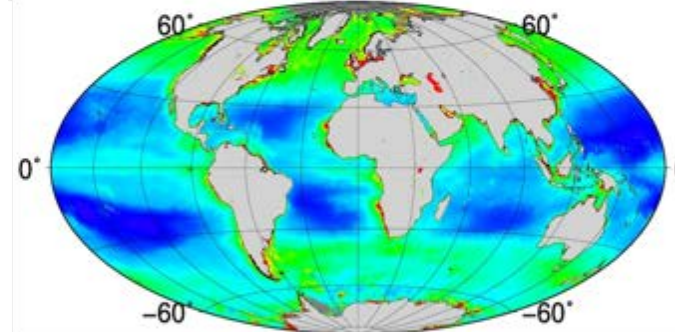
GOME-2



OMI



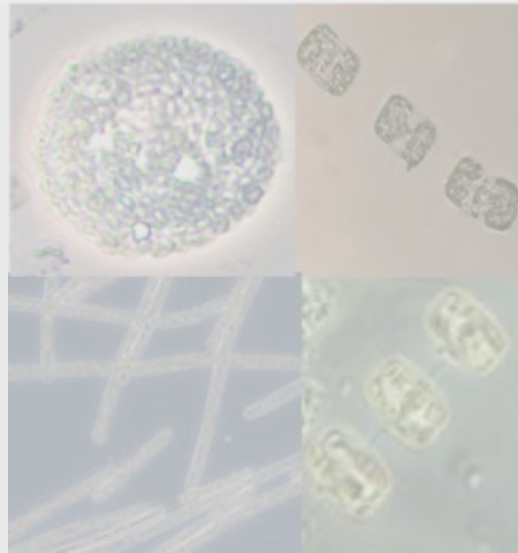
OC-CCI product
(merged
SeaWiFS-
MODIS-MERIS)



Oelker et al. 2019. Optics Express

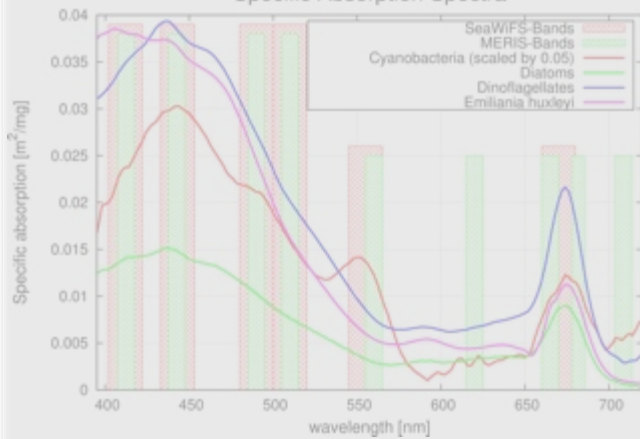
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PFTs: phytoplankton functional types



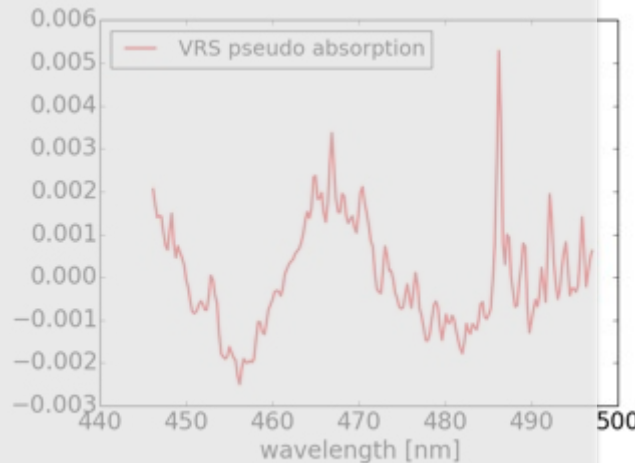
Photos by Sonja Wiegmann - AWI

Specific Absorption Spectra



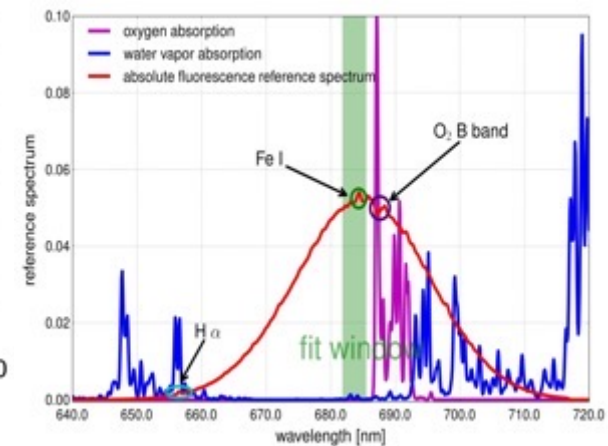
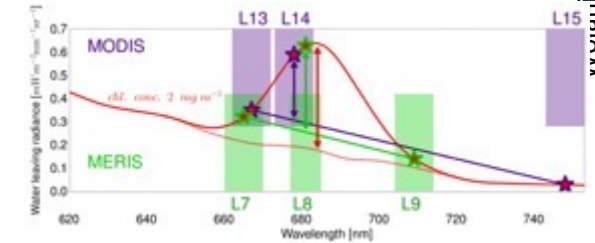
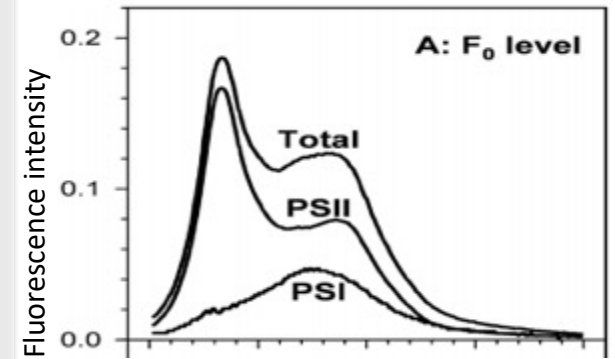
Bracher et al. 2009, Sadeghi et al. 2012, Losa et al. 2017

$K_d(\lambda)$: Spectral diffuse attenuation (K_d)



Dinter et al. 2015, Oelker et al. 2019, Oelker et al. 2022

FLH: Chlorophyll fluorescence line height



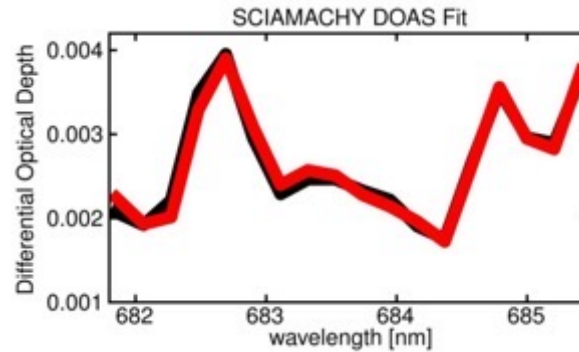
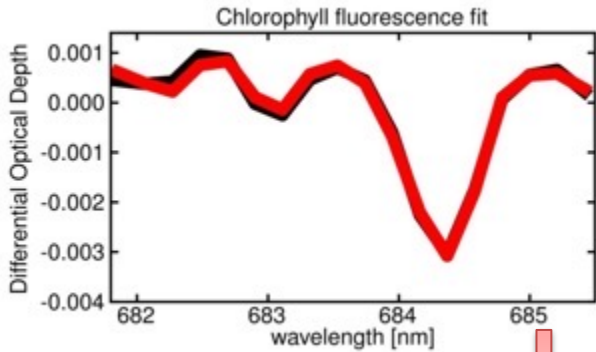
Wolanin et al. 2015, Joiner et al. 2016, Köhler et al. 2020

v.v.v.v.v.in PhD thesis 2015

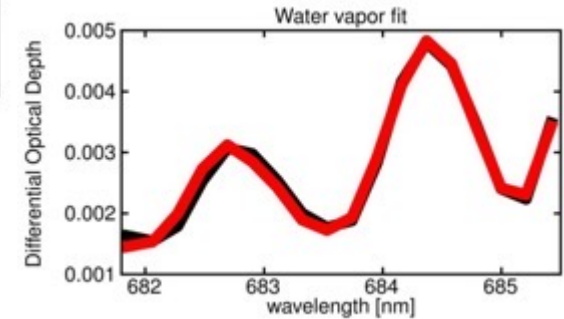
Chl fluorescence from SCIAMACHY



Wolanin et al. Remote Sensing of Environment 2015



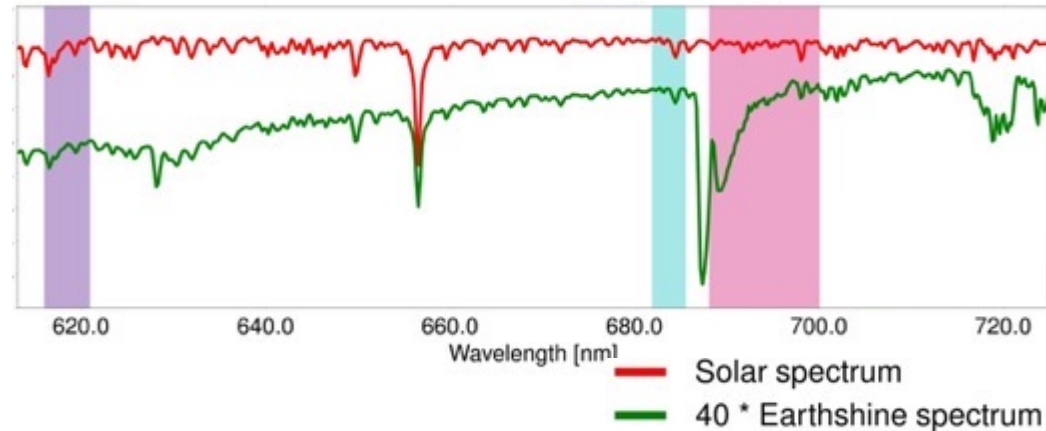
Chl fluorescence
DOAS Fit window
681.8 nm – 685.5 nm



Raman scattering

Fluorescence

Water vapor

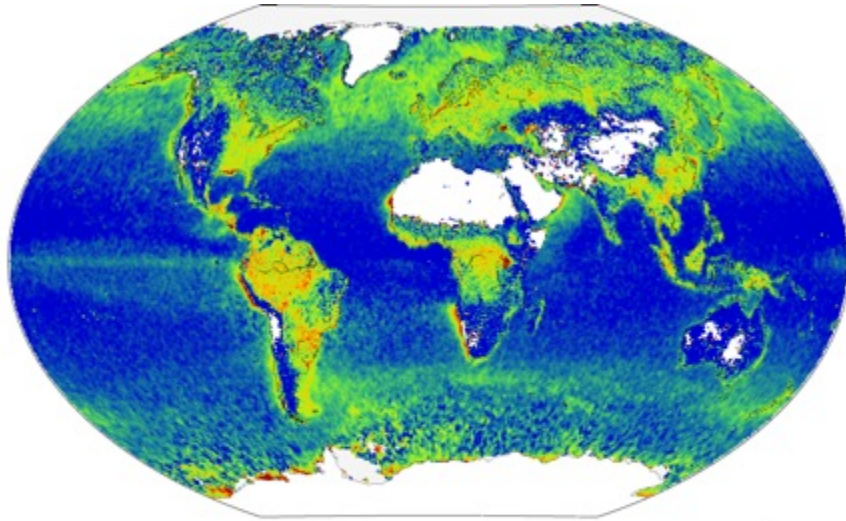


Account for:

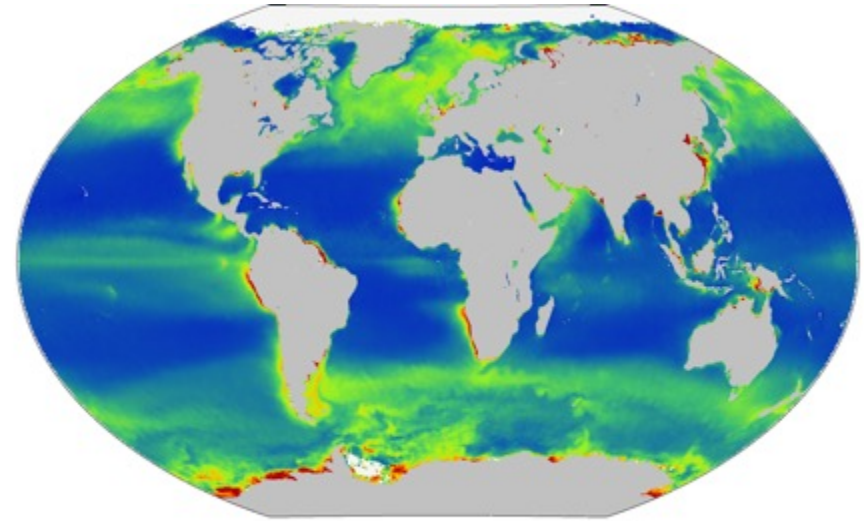
- water vapor,
- inelastic scattering (rotational and vibrational Raman scattering),
- background radiance (atmospheric transmittance & surface reflectance).

Fluorescence 2003-2011 SCIAMACHY & MODIS

Wolanin et al. RSE 2015



SCIAMACHY fluorescence 684 nm ($\text{mW}/\text{m}^2/\text{nm}/\text{sr}$)



MODIS-Aqua nFLH ($\text{mW}/\text{m}^2/\text{nm}/\text{sr}$)

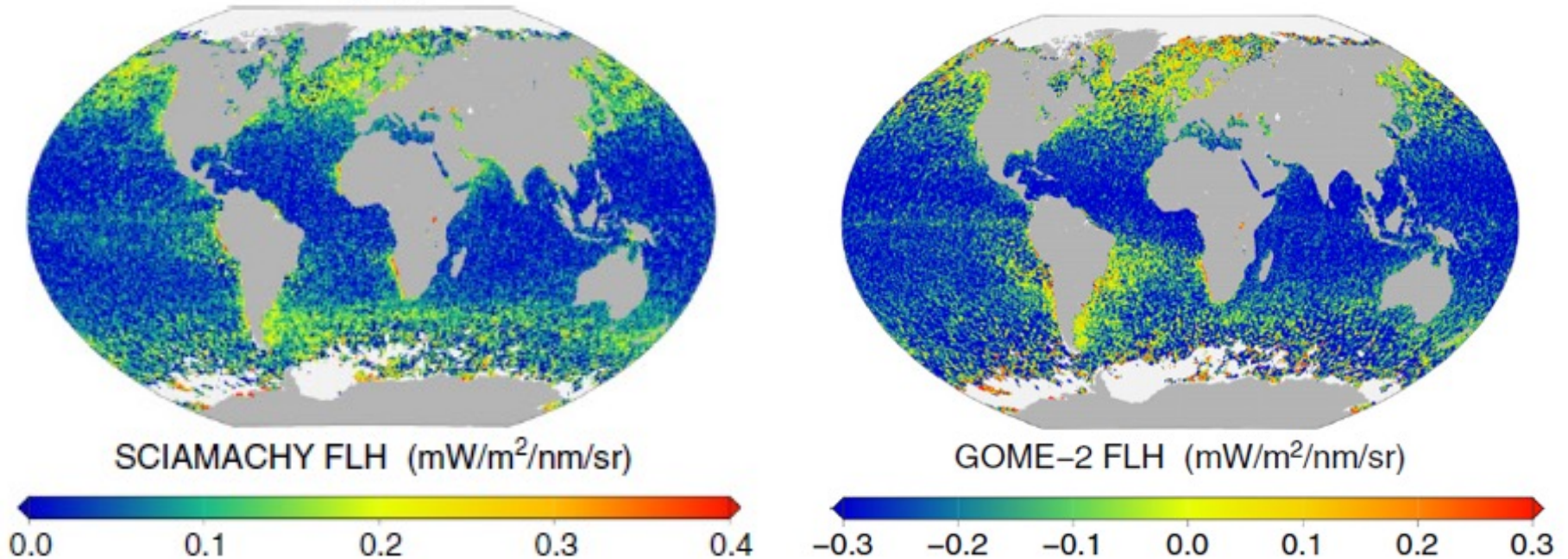


Wolanin et al. 2019: SCIAMACHY 2003-2011 data
<https://doi.org/10.1594/PANGAEA.897169>.

FCHL: Generally representative of the **photophysiological state of phytoplankton** and linked to primary production.

Fluorescence 2009 SCIAMACHY & GOME-2

Wolanin et al. RSE 2015

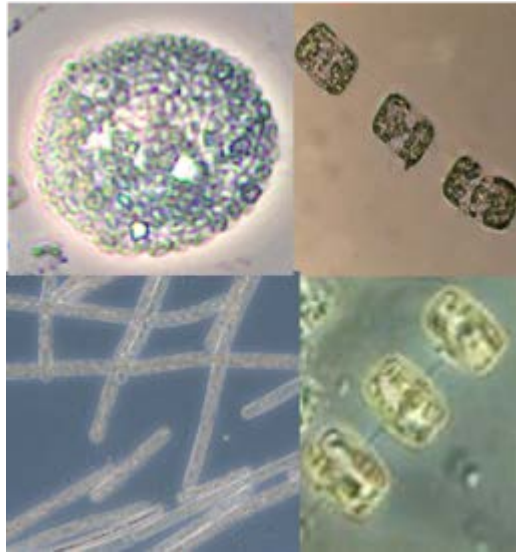


Similar methods to retrieve red fluorescence in ocean waters:

- 1) Joiner et al. **AMT 2016** with GOME-2 data (data set at avdc.gsfc.nasa.gov)
- 2) Köhler et al. **GRL 2020** <https://doi.org/10.1029/2020GL087541> with TROPOMI data (<ftp://fluo.gps.caltech.edu/data/tropomi/>)

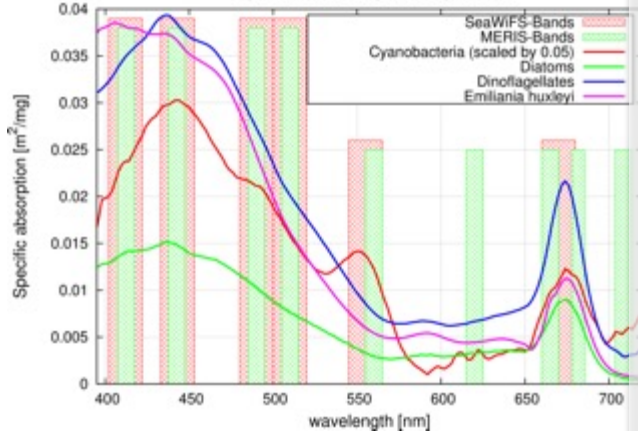
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Photos by Sonja Wiegmann - AWI

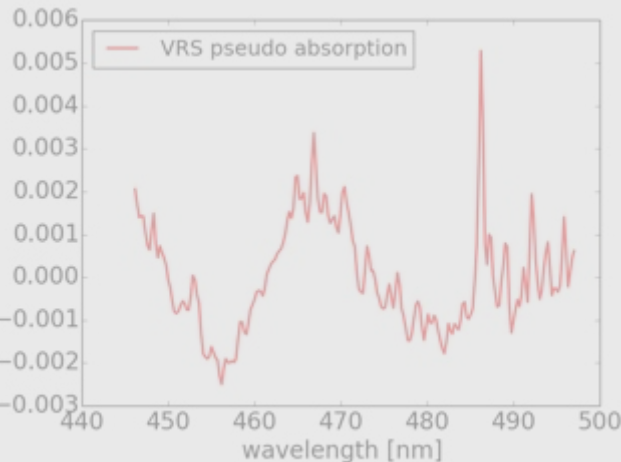
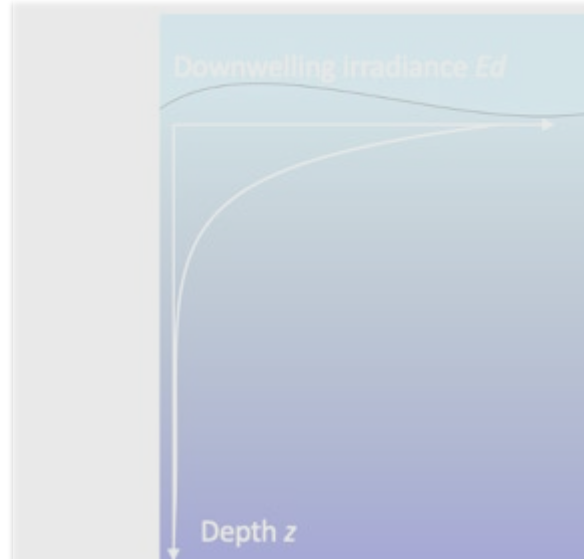
Specific Absorption Spectra



Bracher et al. 2009, Sadeghi et al. 2012, Losa et al. 2017

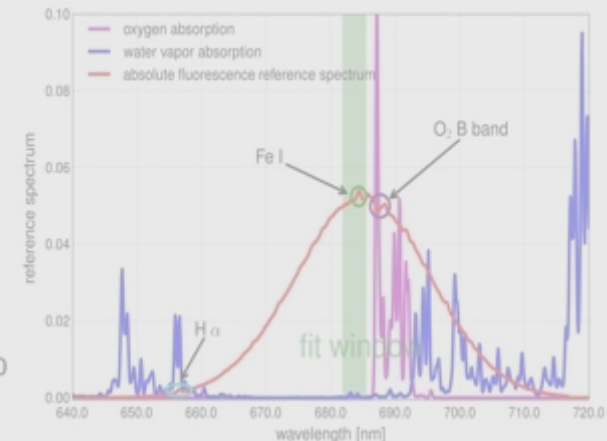
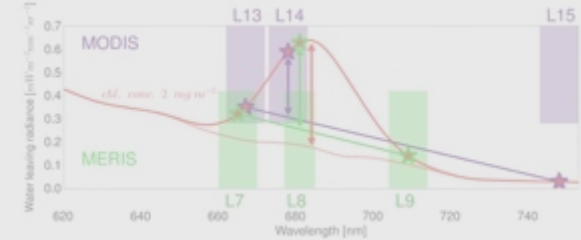
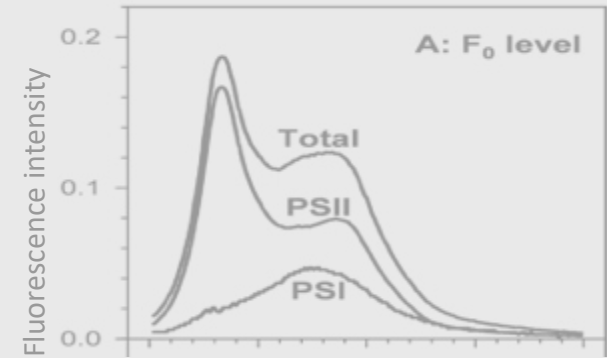
Astrid.Bracher@awi.de

$K_d(\lambda)$: Spectral diffuse attenuation (Kd)



Dinter et al. 2015, Oelker et al. 2019, Oelker et al. 2022

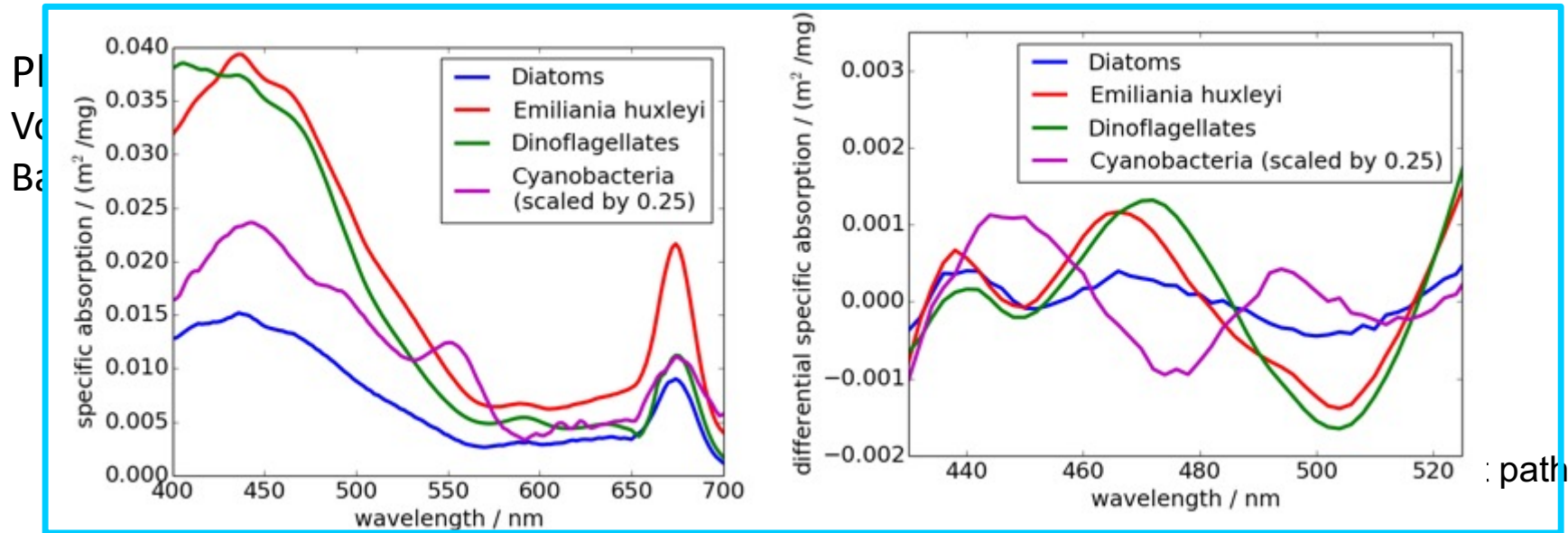
FLH: Chlorophyll fluorescence line height



Wolanin et al. 2015, Joiner et al. 2016, Köhler et al. 2020

v.v.v.v.v.in PhD thesis 2015

PhytoDOAS Retrieval for deriving Chl-a conc. of Phytoplankton Groups (PFT-CHL)



PhytoDOAS equation



Polynomial Oceanic contribution
Slant column density (fit factors)



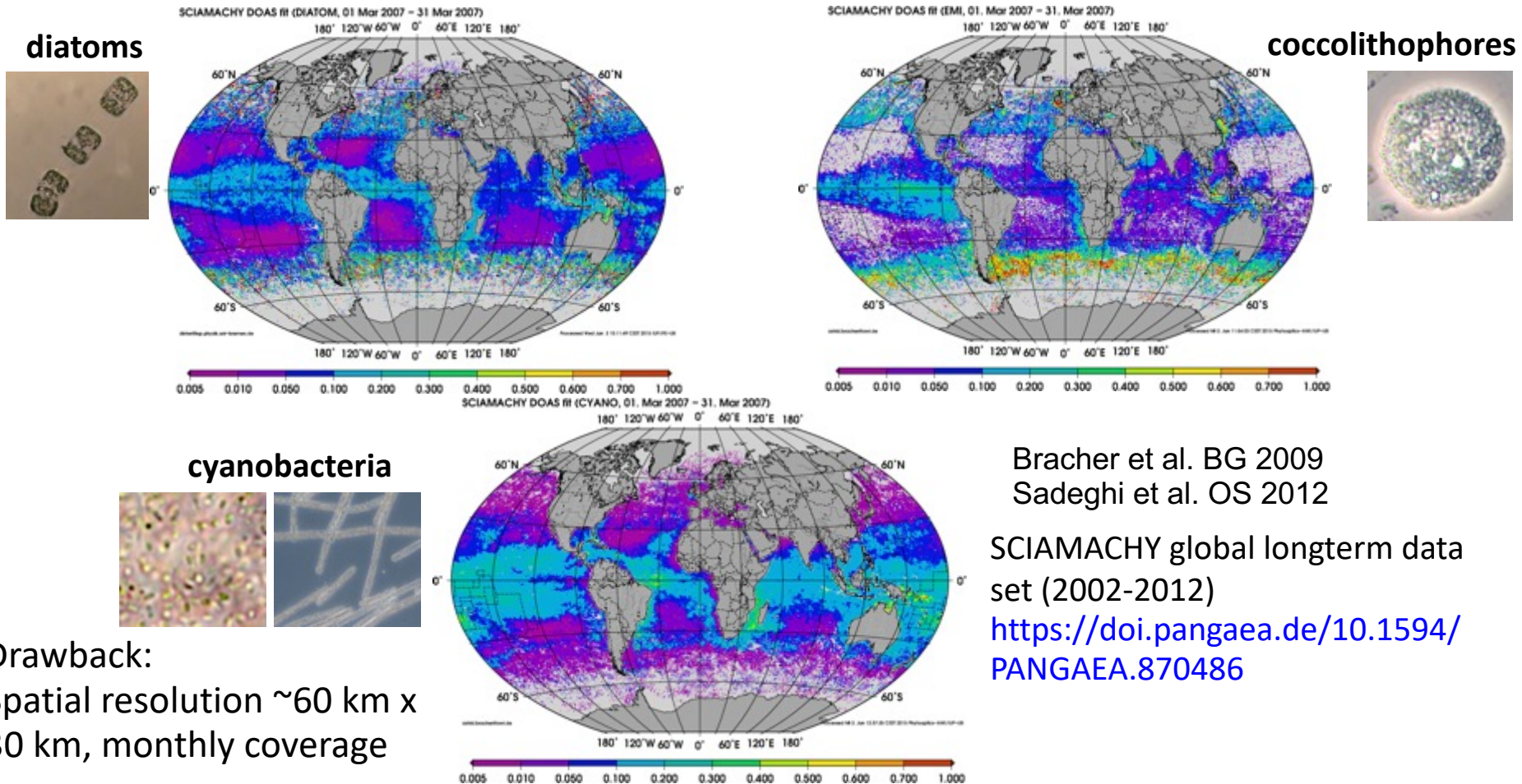
Atmospheric contribution
Differential absorption cross sections

Phytoplankton Groups Chla conc. (PFT-CHL) from SCIAMACHY



Different PFTs play different roles in the biogeochemical cycles of the ocean and food web.

Monthly Concentration (mg m^{-3}) – March 2007



Bracher et al. BG 2009
Sadeghi et al. OS 2012

SCIAMACHY global longterm data
set (2002-2012)

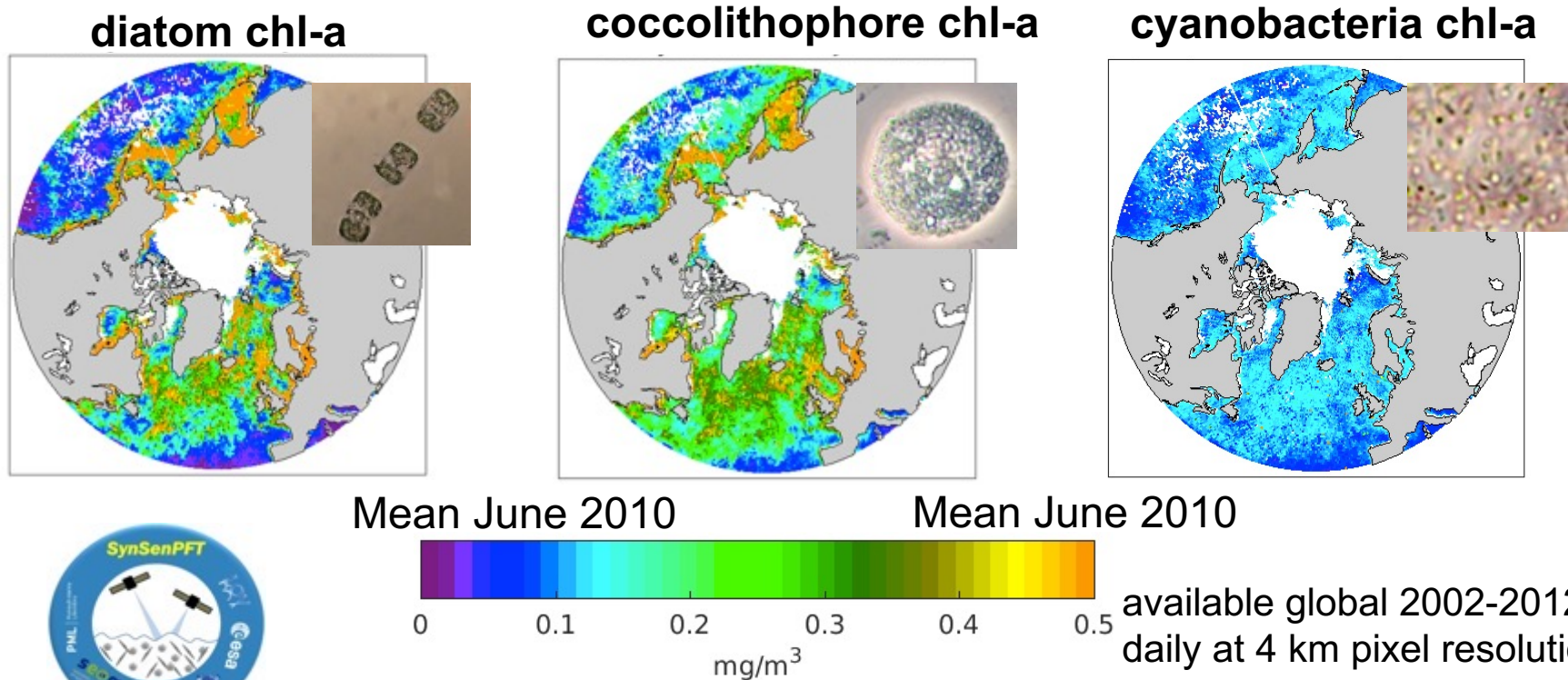
<https://doi.pangaea.de/10.1594/PANGAEA.870486>

Drawback:
Spatial resolution $\sim 60 \text{ km} \times 30 \text{ km}$, monthly coverage

High temporal & spatial resolution data set on three major phytoplankton groups



Losa S., Soppa M. A., Dinter T., et al. Bracher A. (2017) Synergistic exploitation of hyper- and multispectral precursor Sentinel measurements to determine Phytoplankton Functional Types (SynSenPFT). *Frontiers in Marine Science* 4: 203, doi: 10.3389/fmars.2017.00203



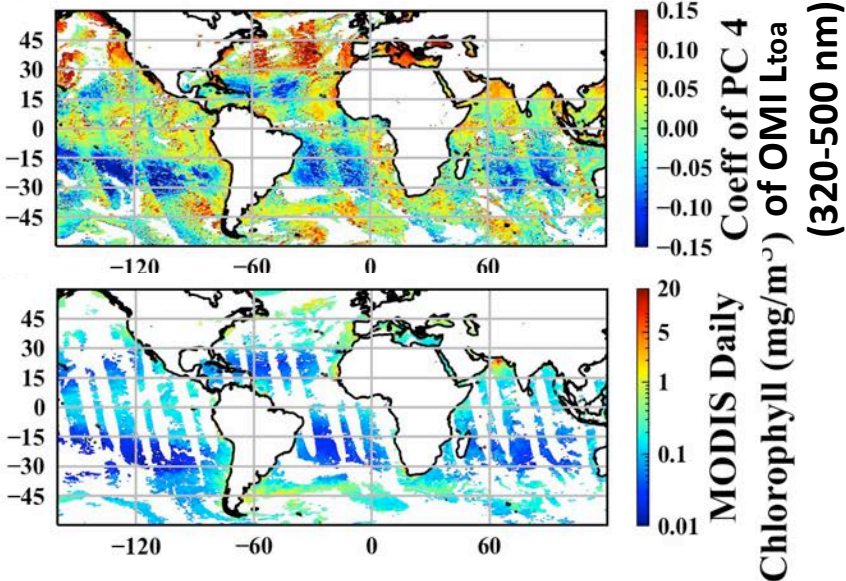
available global 2002-2012
daily at 4 km pixel resolution

<https://doi.org/10.1594/PANGAEA.873210>

Obtained by synergistic use of hyper- (SCIAMACHY/ PhytoDOAS) and multi-spectral (OC-CCI (merged MERIS-MODIS-SeaWiFS)/ OC-PFT) data.

OMI/TROPOMI TOA hyperspectral radiances (L_{TOA}) for predictions of Chla under thin clouds

Via PCA decomposition, matchups with MODIS Chla (+ blue bands) and ML-models

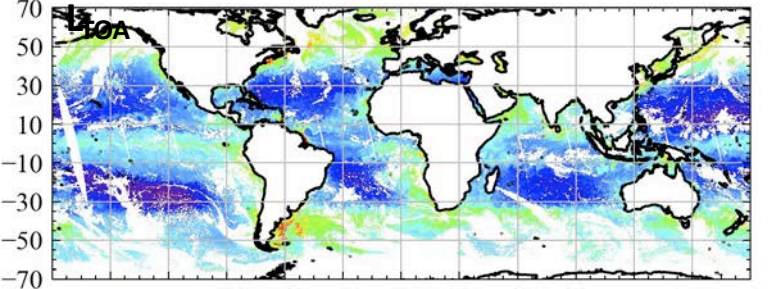


Fasnacht et al. (2022) Using Machine Learning for Timely Estimates of Ocean Color Information From Hyperspectral Satellite Measurements in the Presence of Clouds, Aerosols, and Sunlight. *Frontiers in Remote Sensing*.

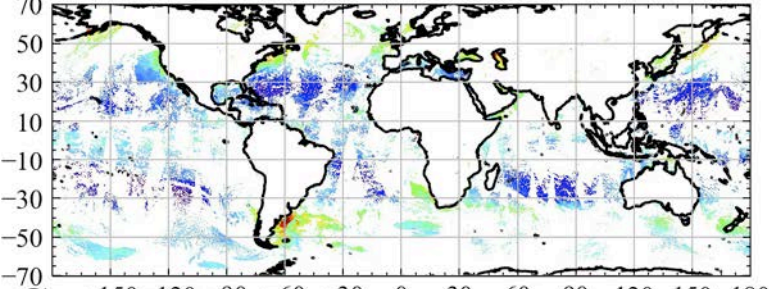
<https://doi.org/10.3389/frsen.2022.846174>

Astrid.Bracher@awi.de

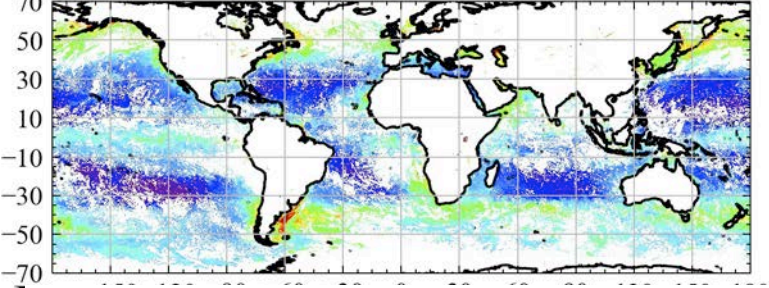
A TROPOMI daily CHL predicted from



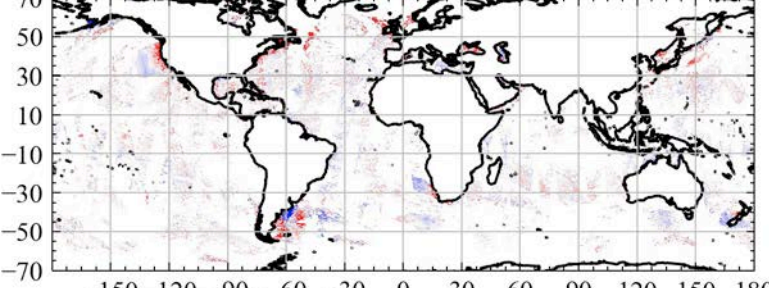
E Merged MODIS Aqua & Terra daily CHL



C Merged MODIS A & T 8 days CHL



I TROPOMI vs. merged MODIS A & T daily CHL



Chlorophyll (mg/m^3)

Percent Difference

Conclusions for atmospheric sensors' OC products

High spectral resolution of the optical atmospheric sensors SCIAMACHY, GOME-2, OMI and TROPOMI enables analytical retrievals of innovative products: PFT-CHL, FLH, Kd-short blue and recently Kd in UVAB & in UVA.

TROPOMI's products with much higher spatial resolution and daily global coverage than SCIA, OMI, ...

- show less retrieval uncertainty and enable a better assessment of their quality with much more matchups to in-situ;
- enhance their application for understanding global processes (e.g. carbon cycling, primary production, radiative SW budget).

Using ML-based prediction methods with high spectral resolution LTOA data from TROPOMI enables gap filling of operational (multispectral) OC-products leading to quasi global daily coverage.

The similar OC products from the atmospheric sensors serve as alternative evaluation of multispectral (mostly empirical / statistical retrievals) sensor products, and serve as test-beds and demonstration for algorithm development and comparisons to recently launched (PRISMA; EnMAP, ...) and soon upcoming hyperspectral (PACE, SBG, CHIME, ...) sensors.

Outlook for atmospheric sensors' OC products

Next 5 years:

Combining TROPOMI retrievals with OLCI (and similar data) to obtain

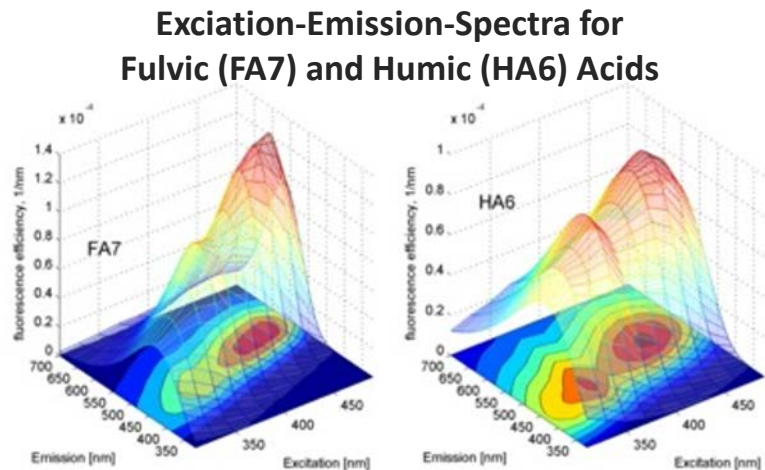
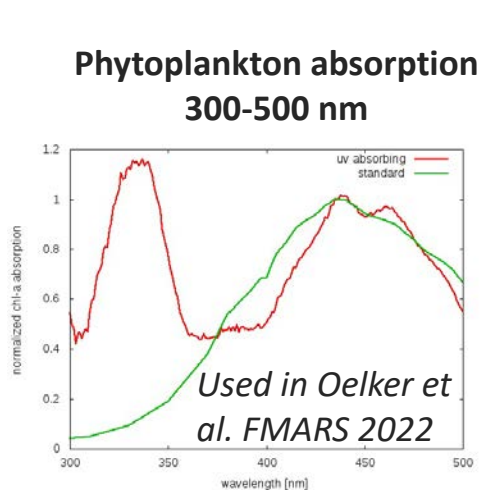
- $K_d(\lambda)$ at ≥ 9 -bands from 325-700 nm (325, 373, 405, 412, 443, 490, 510, 560, 665)
- Synergy of PFT-products
- Alternative evaluation of common sensors' products (FCHL, PFT)

Developing new TROPOMI products: MAAs, CDOM sources, photobleaching

Transferring Know-How: TROPOMI-OC to UVN on S4 (GEO) & S5 – PACE, EnMAP, DESIS, PRISMA

10 years:

Long term data sets at best spatial and temporal coverage including pixel uncertainty from synergy of historic, current and upcoming sensors for $K_d(\lambda)$, PFTs, photobleaching, type of CDOM,... for many applications (climate/ecosystem/bgc ocean modelling, etc.)



Thank you!

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