

→ 4th ESA ADVANCED TRAINING  
ON OCEAN REMOTE SENSING

# Coastal Water Algorithms

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## 1. Definition and spectral signatures of Case-2 waters

Definition

Spectral signatures = (NAP / Chla / CDOM)

## 2. Atmospheric corrections

Methods / Algorithms

Validation

## 3. Level-2 products and inversion algorithms

Bio-optical and biogeochemical products

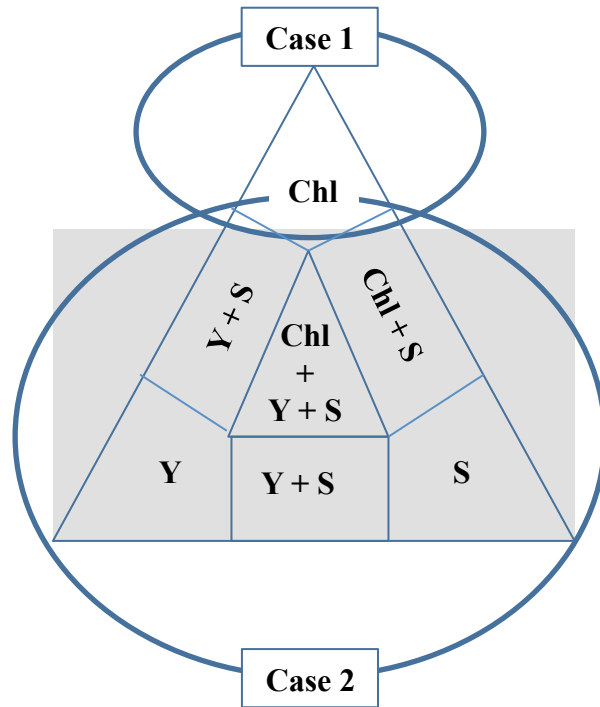
Three main types of inversion algorithms

## 4. Application of the Ocean Colour methods in coastal seas: Experience at Ifremer Brest ([F. Gohin](#))

## 5. Summary / Conclusions

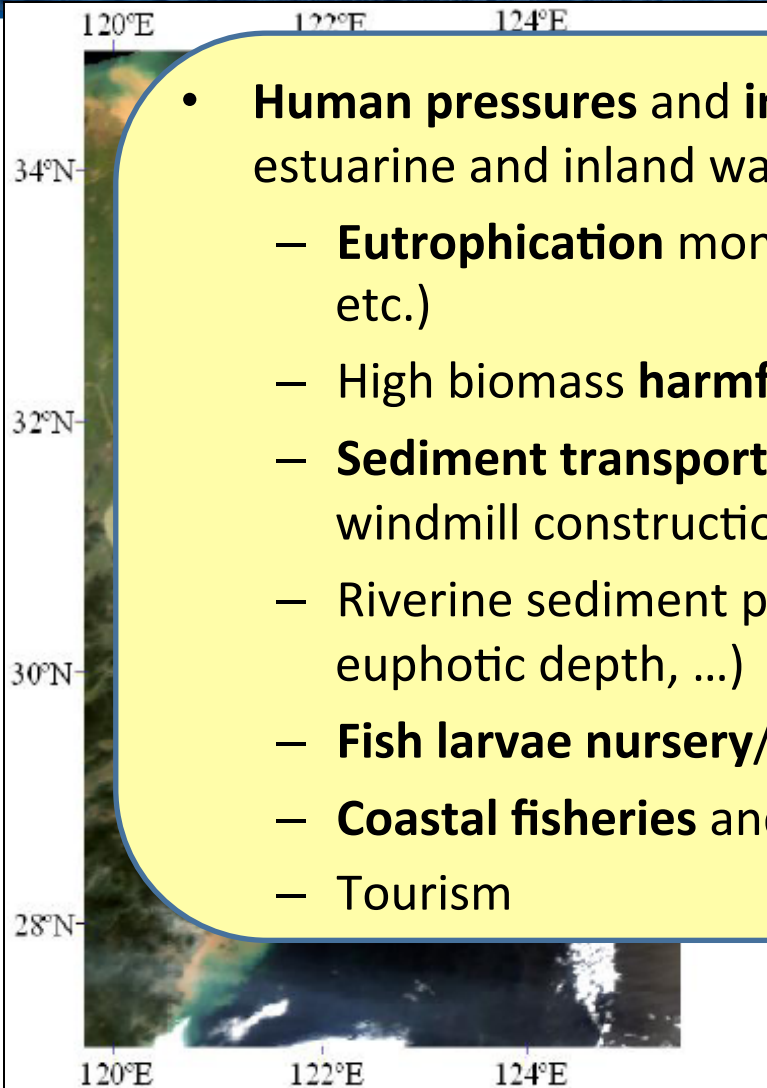


## Classification of natural waters

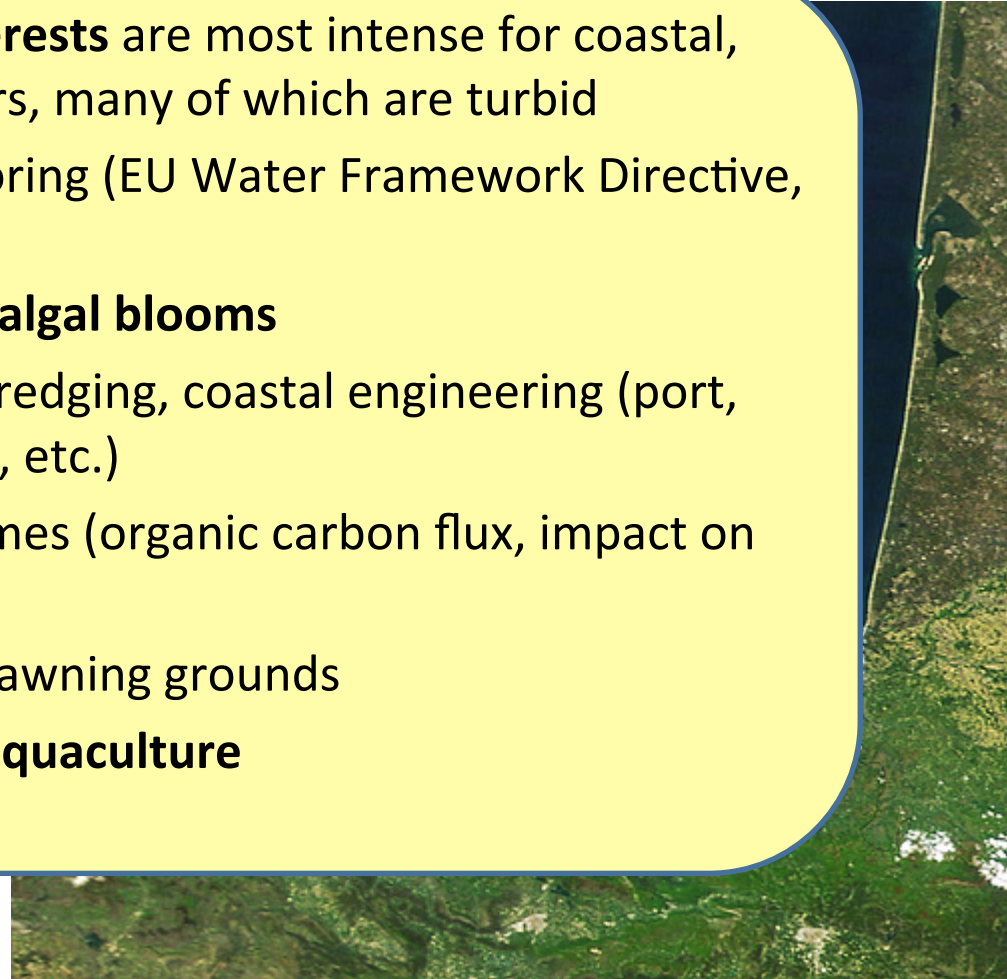


- In Case 2 waters, terrestrial non-algal particles (S) and coloured dissolved organic matter (Y)
- Sources of S and Y: rivers, sediment resuspension, coastal erosion, wind
- Highly productive waters (high nutrient supply)

Morel / Antoine MERIS Case 1 water ATBD



- **Human pressures and interests** are most intense for coastal, estuarine and inland waters, many of which are turbid
  - **Eutrophication** monitoring (EU Water Framework Directive, etc.)
  - High biomass **harmful algal blooms**
  - **Sediment transport**, dredging, coastal engineering (port, windmill constructions, etc.)
  - Riverine sediment plumes (organic carbon flux, impact on euphotic depth, ...)
  - **Fish larvae nursery**/spawning grounds
  - **Coastal fisheries and aquaculture**
  - Tourism





# 1. Spectral signatures

$$R_{rs} = f' \times b_b / (a + b_b)$$

→  $R_{rs}$  increases with increasing light scattering,  
 increases with increasing light absorption  
 decreases

$$a = a_w + a_{Chla} + a_S + a_y$$

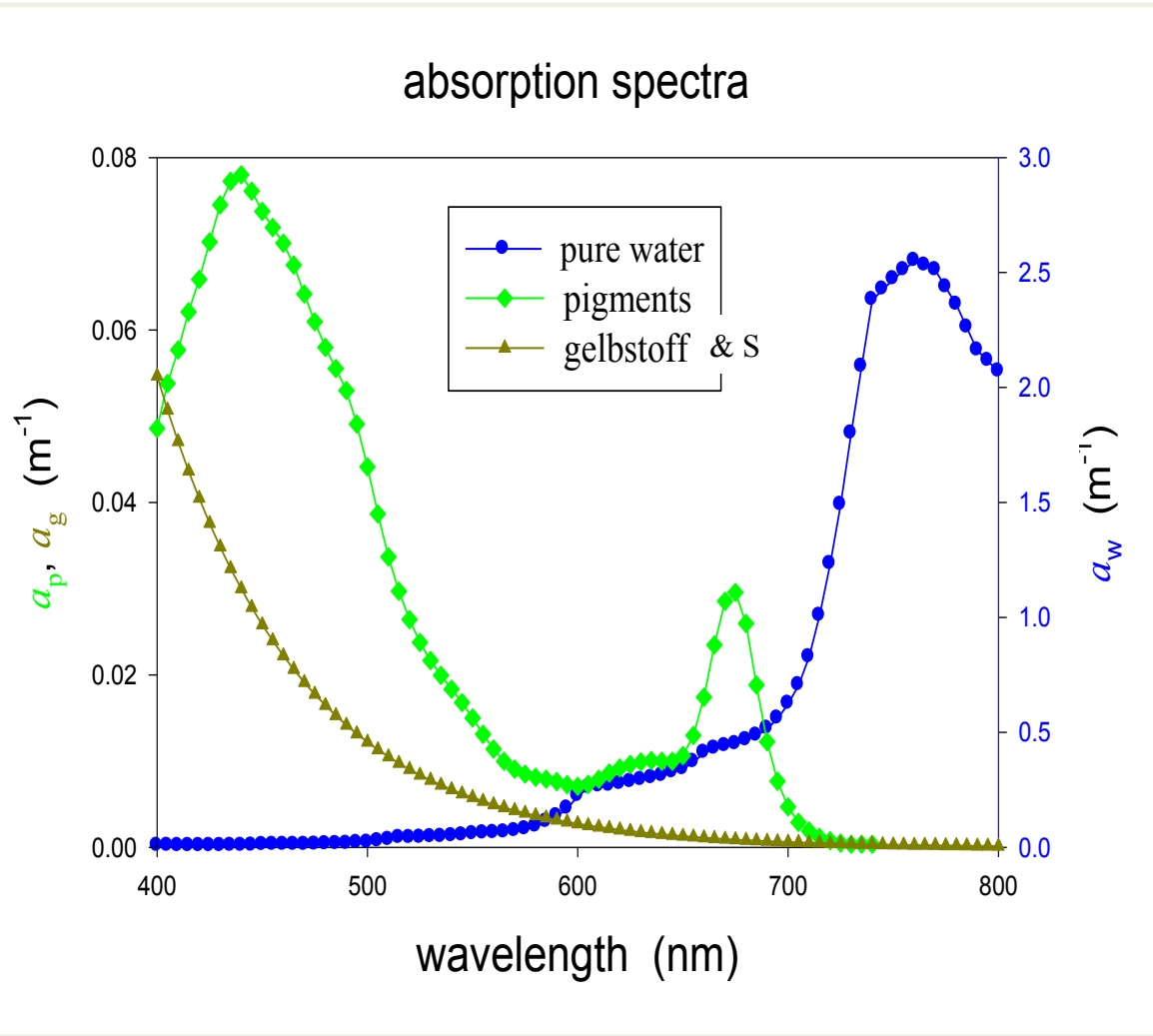
$$b_b = b_{bw} + b_{bChla} + b_{bS}$$

$$a_{Chla} = Chla \times a_{Chla}^*$$

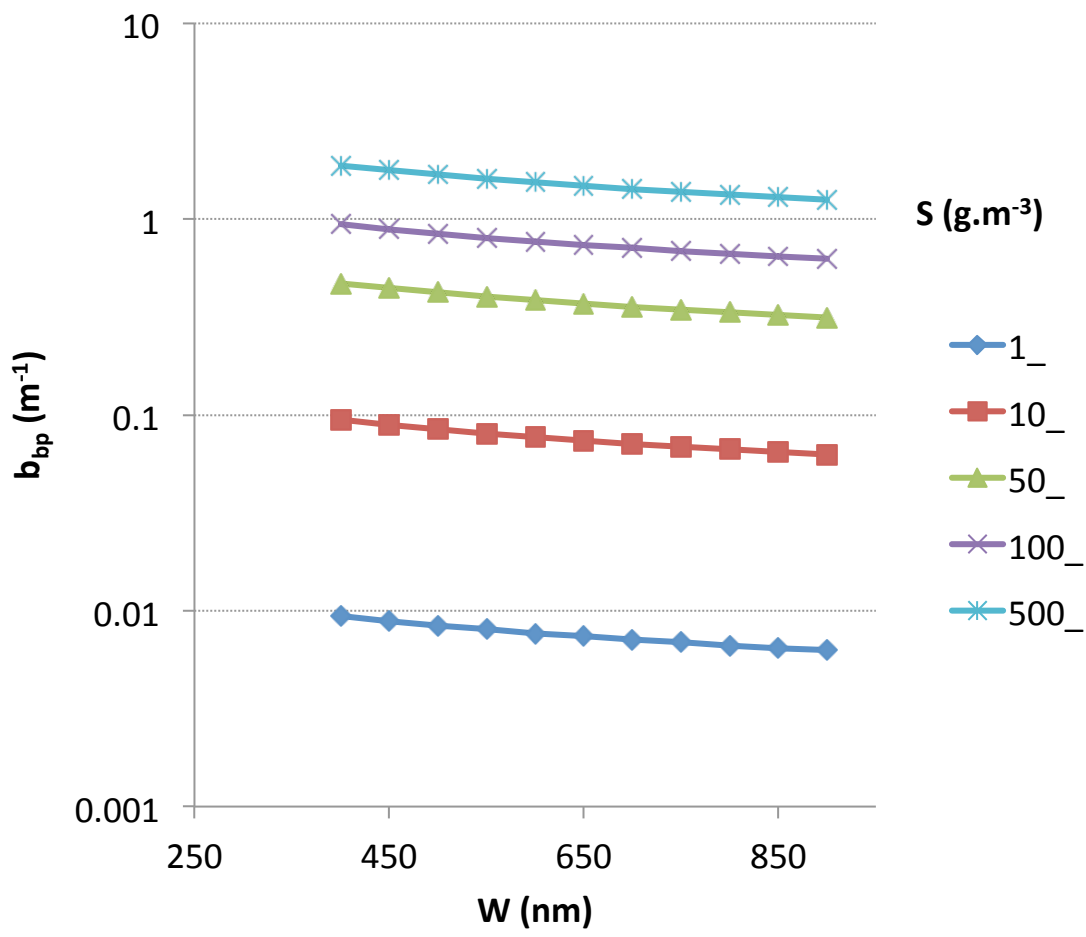
$$b_{bS} = S \times b_{bS}^*$$

→ The contributions of coloured water constituent are additive

→ Concentration × mass-specific absorption/backscattering coefficients



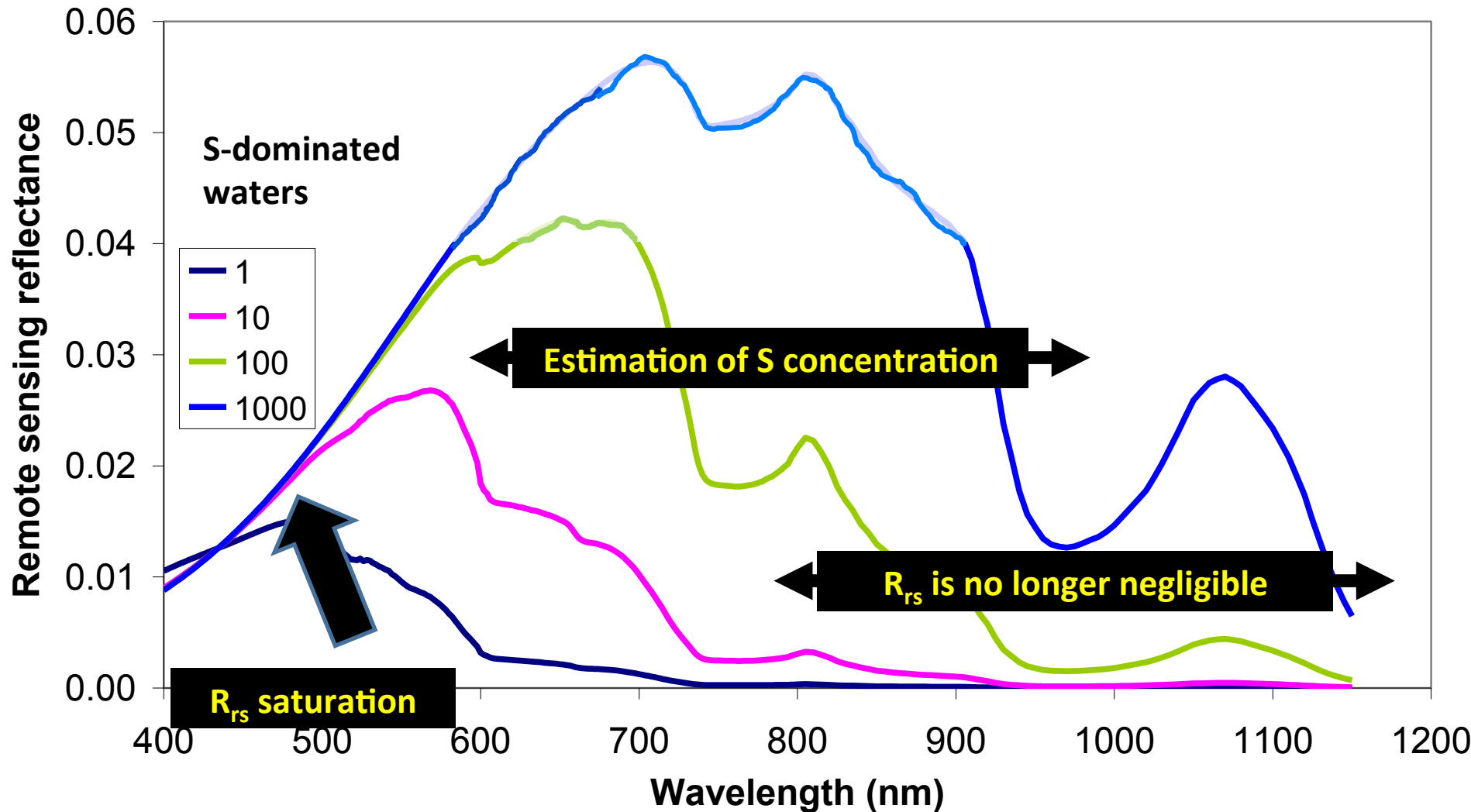
- **Pure water** absorbs light in the near-infrared
- **Chla** has two peaks of absorption (443 and 675 nm)
- **Non-algal particles (S)** and **gelbstoff (Y)** both absorb at short (blue) wavelengths)



- **Particles, mainly non-algal (S), backscatter light**
- $b_{bp}$  is a **proxy of particles concentration**
- **Smooth (power-law) spectral variations** (slope depends on size distribution)

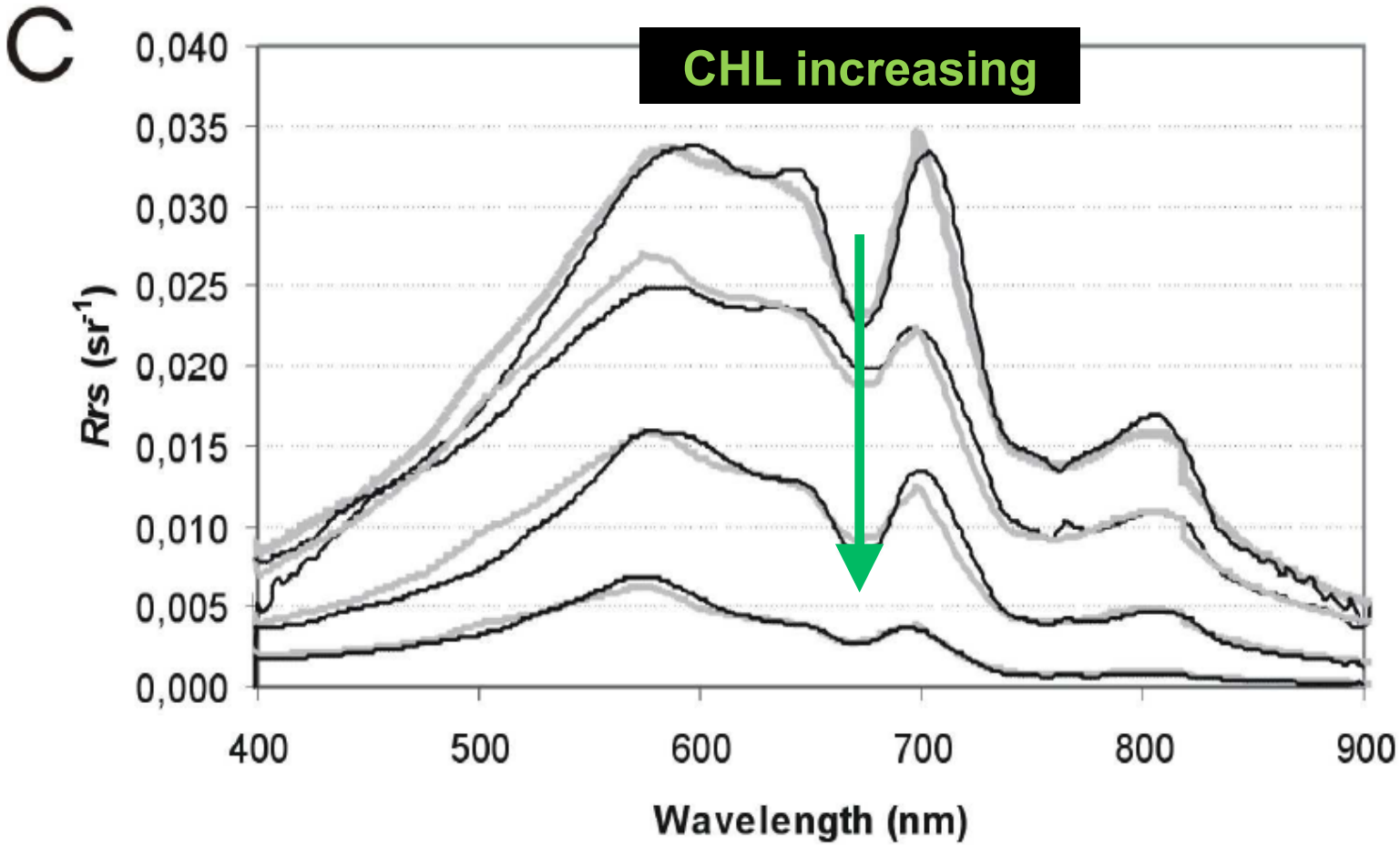


Varying Total Suspended matter concentration (mg/m<sup>3</sup>)

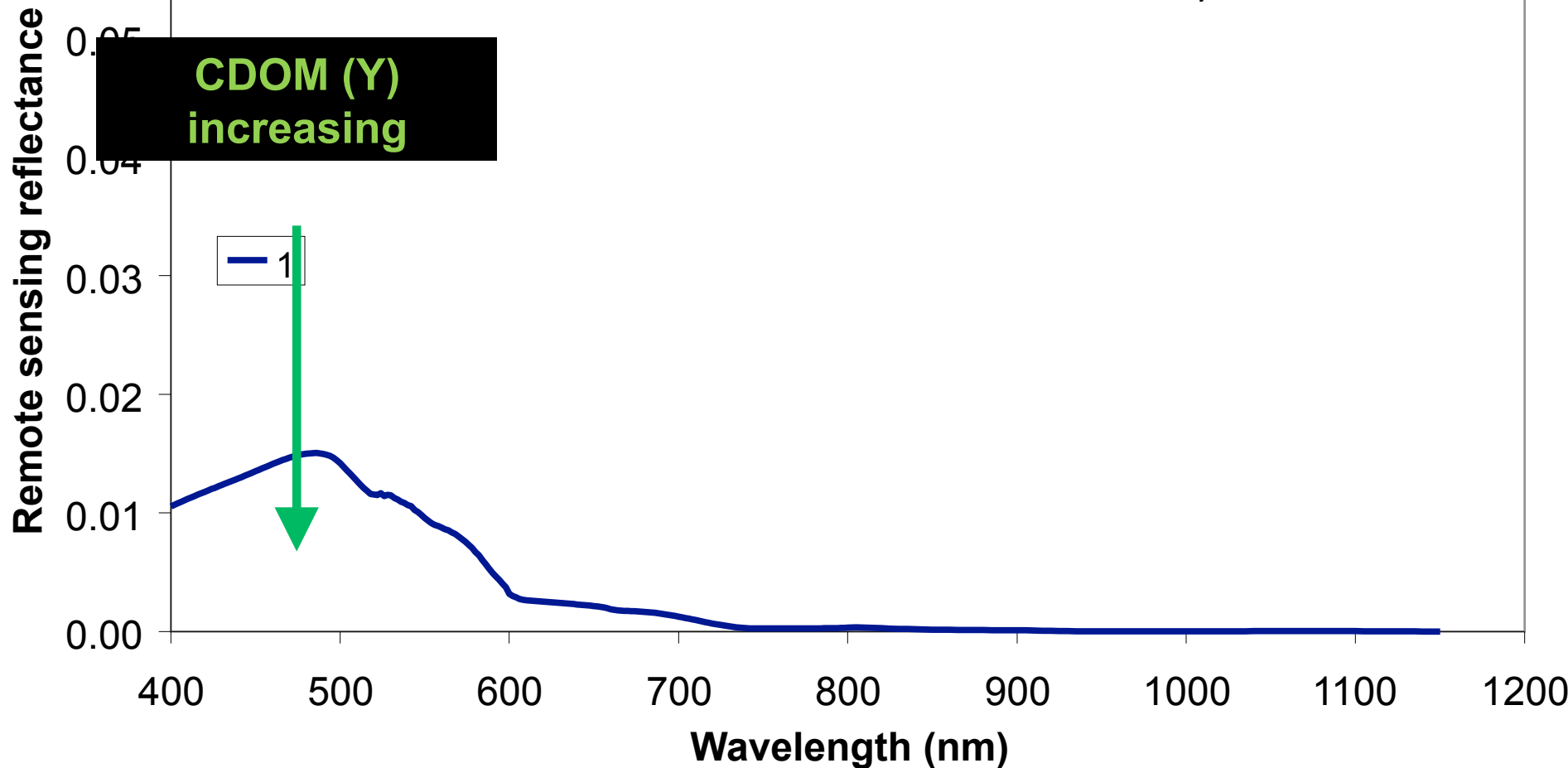


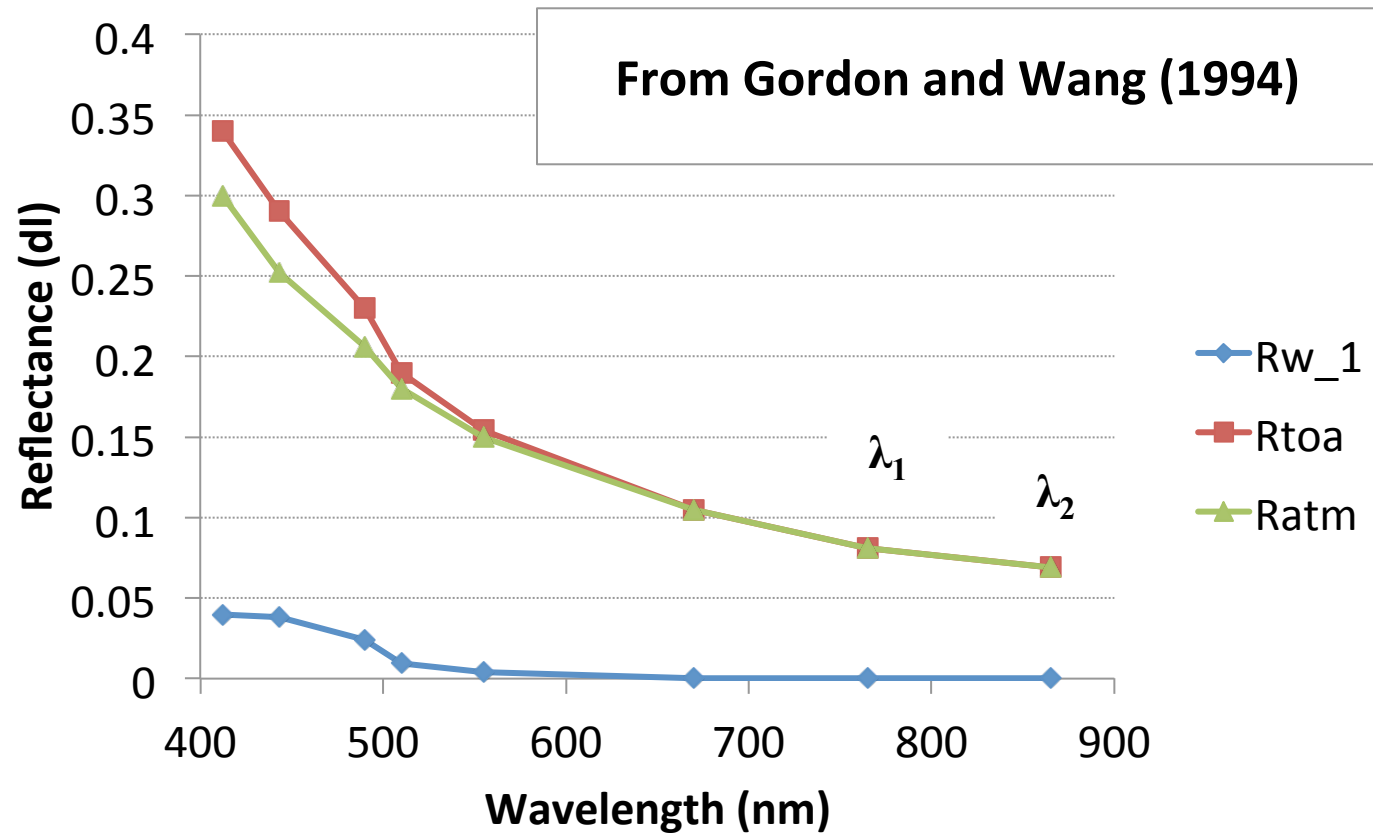


# 1. Spectral signatures



➔ In turbid (absorbing and scattering) waters, light absorption by Chla at 443 nm disappears but light absorption by Chla is clearly detected





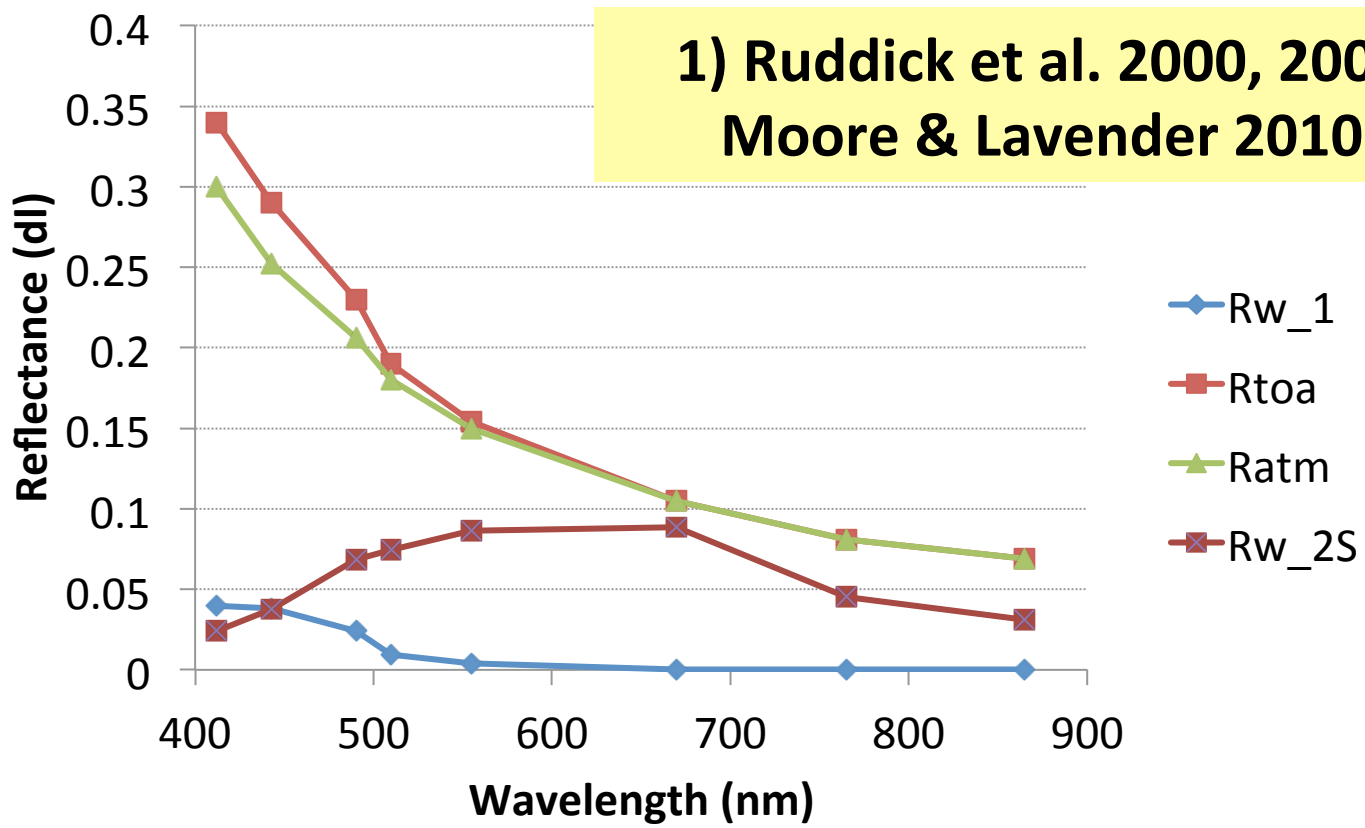
$$R_{toa}(\lambda) = R_r(\lambda) + R_{ra}(\lambda) + t \times R_w(\lambda),$$

Light absorption and scattering by air molecules (Rayleigh) (LUT)

$$\epsilon(\lambda_1, \lambda_2) = R_{ra}(\lambda_1) / R_{ra}(\lambda_2) = (\lambda_1 / \lambda_2)^n$$

Light scattering by aerosols is computed in the NIR and extrapolated to the visible





$$R_{toa}(\lambda) = R_r(\lambda) + R_{ra}(\lambda) + t \times R_w(\lambda),$$

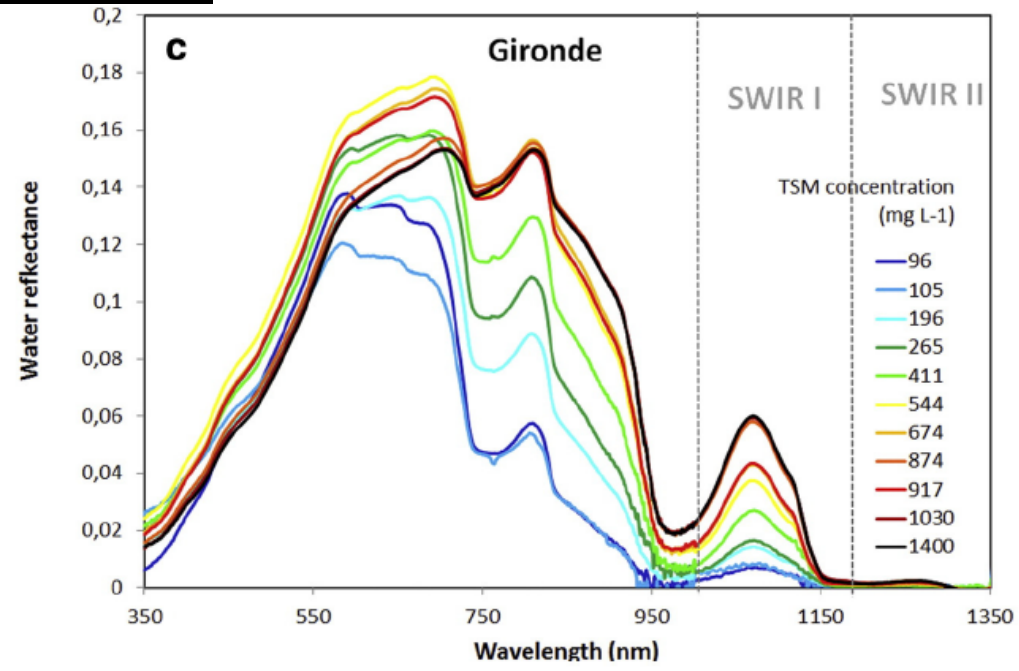
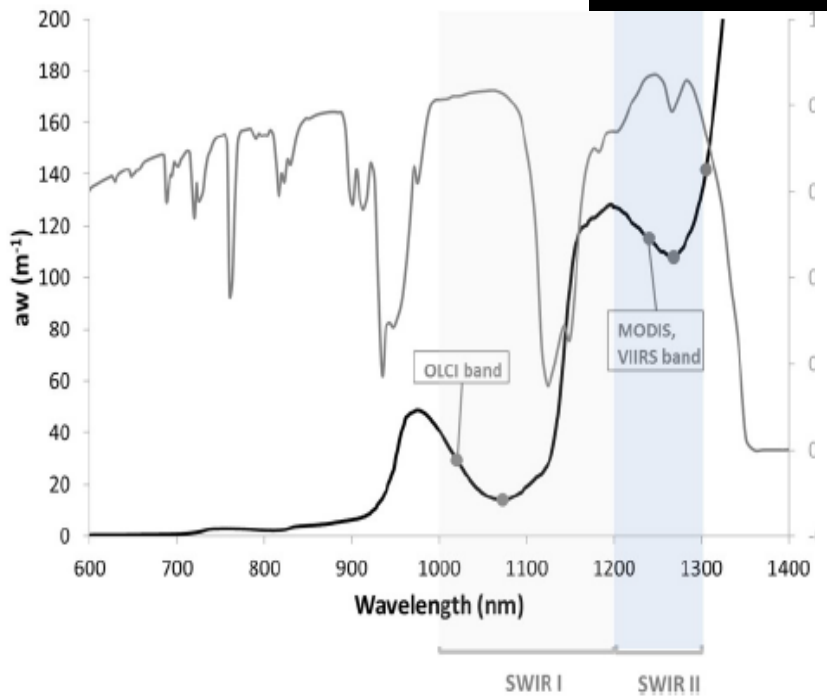
$$\epsilon_w(\lambda_1, \lambda_2) = \text{constant (pure water)}$$

Light absorption and scattering by air molecules (Rayleigh) (LUT)

Pixel-by-pixel iterations to fit  $R_{atm}(\lambda_1, \lambda_2)$

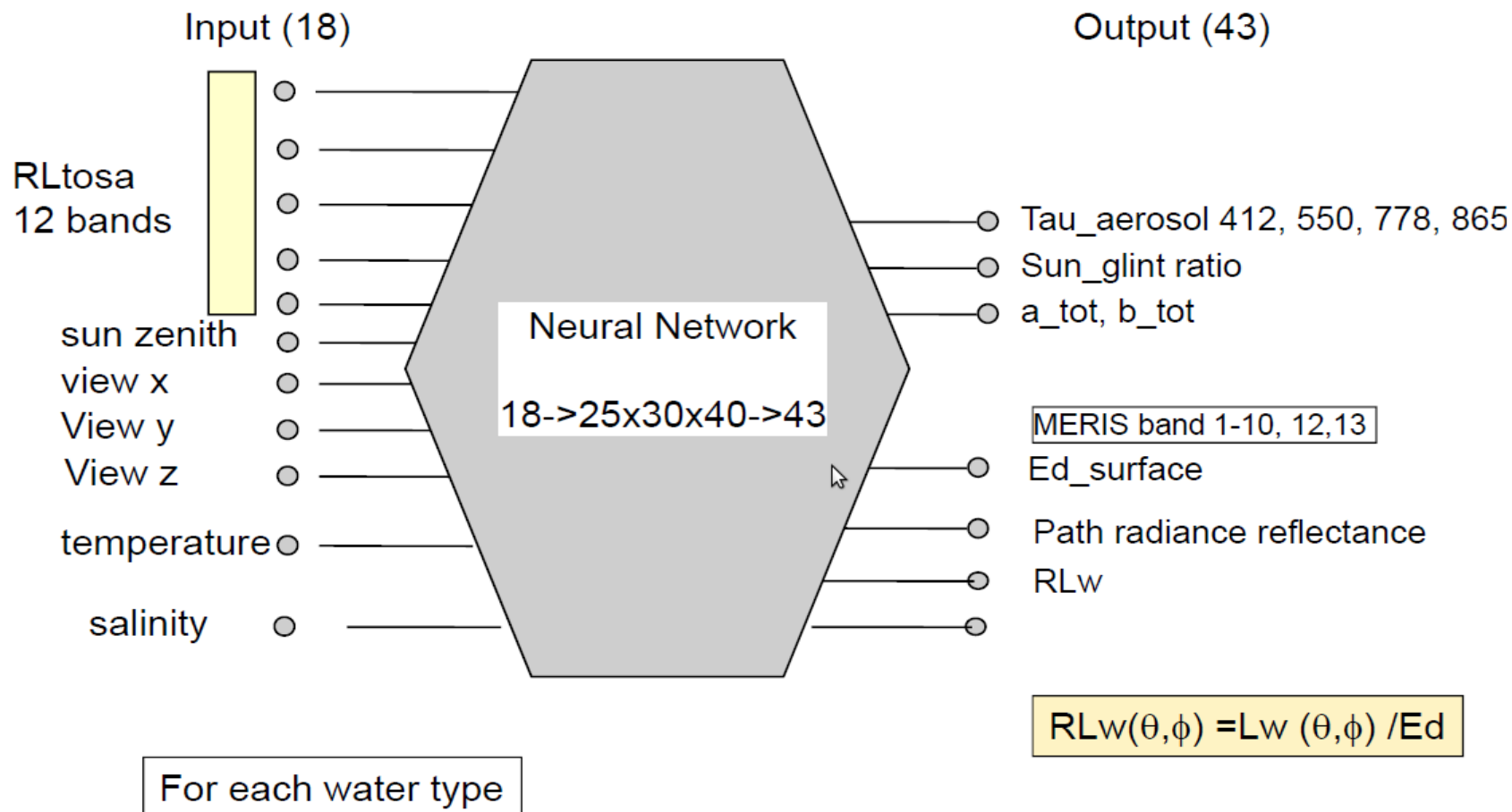
## 2) The NIR-SWIR atmospheric correction (Wang & Shi 2007)\*

Knaeps et al. 2015

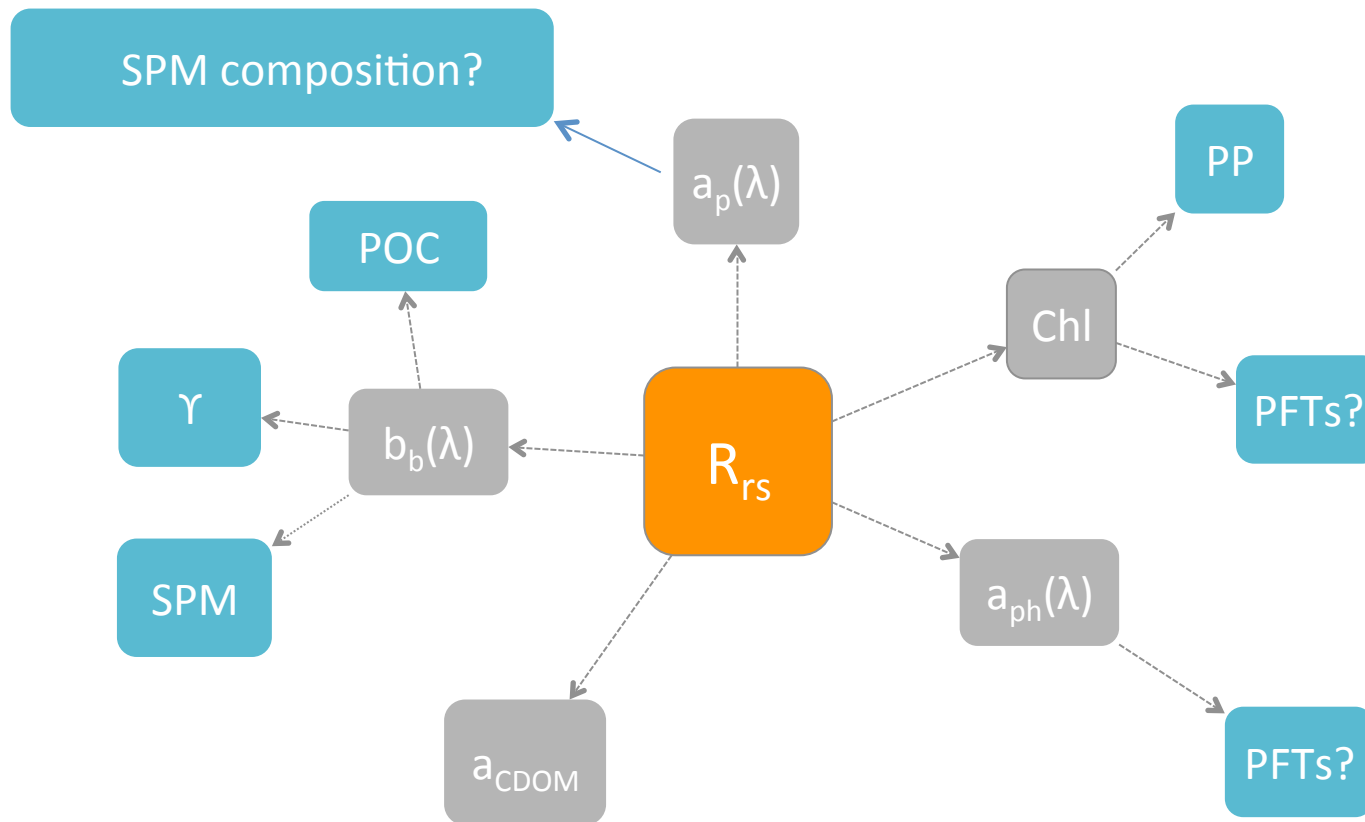


(\* ) → Use of 1240, 1640 and 2130 nm spectral bands (MODIS, VIIRS) to determine aerosol contribution and extrapolate to visible

### 3) The Neural Network (NN) approach (Doerffer 1997)







$(SPM = S + Chl)$

# 1) Empirical

based on field (and satellite) data

# 2) Semi-analytical

$$R_{rs} = f' \times b_b / (a + b_b)$$

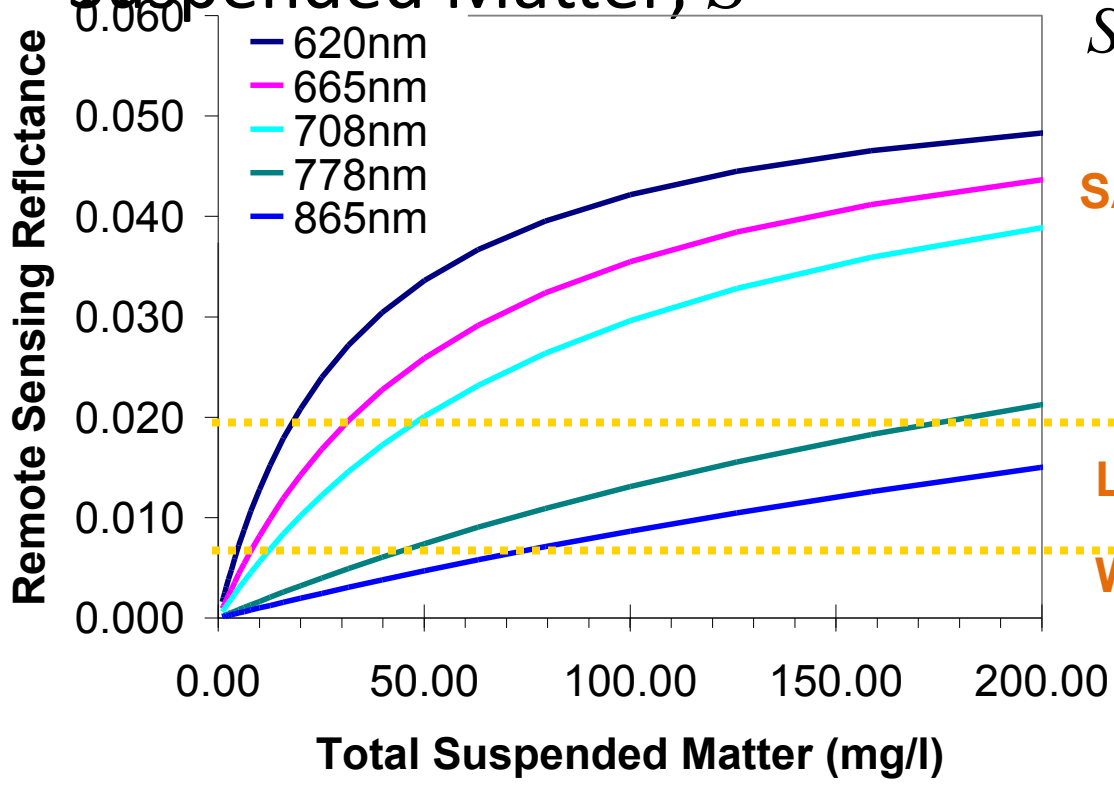
$$a = a_w + a_{Chla} + a_{NAP} + a_y$$

$$b_b = b_{bw} + b_{bChla} + b_{bNAP}$$

# 3) Neural Network

Trained using field data and/or RT simulations

- Remote-sensing reflectance,  $R_{rs}$ , at any single wavelength,  $\lambda$ , is almost linearly related to Total Suspended Matter,  $S$



$$S = \left\{ \frac{A(\lambda)}{1 - R_{rs}(\lambda)/C} \right\} R_{rs}(\lambda)$$

**SATURATION**

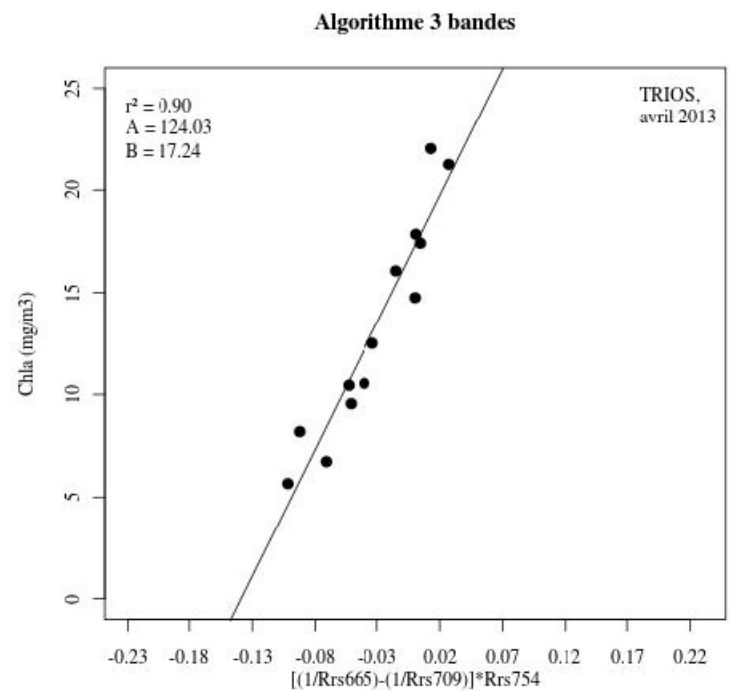
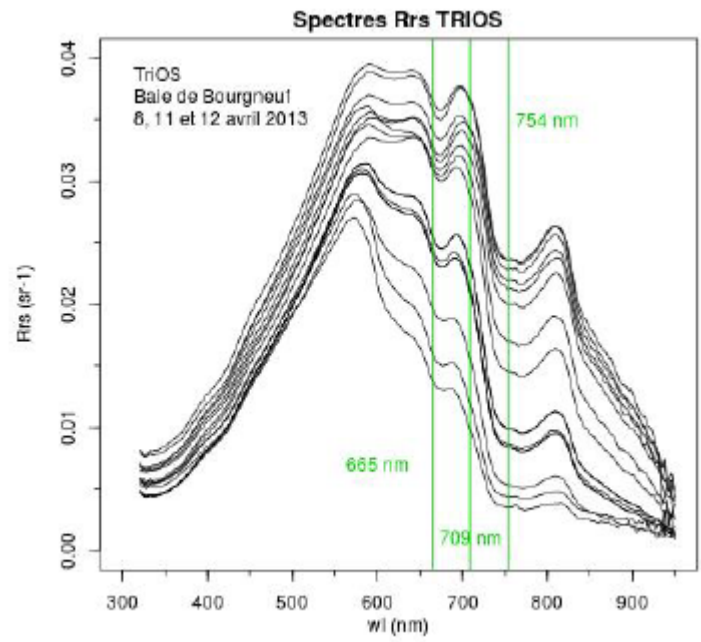
**LINEAR (optimal)**

**WEAK SIGNAL**

[Nechad et al, 2010]  
 Shen et al. 2010  
 Doxaran et al. 2002-2009



➔ Blue/green  $R_{rs}$  ratio usually fail in coastal waters



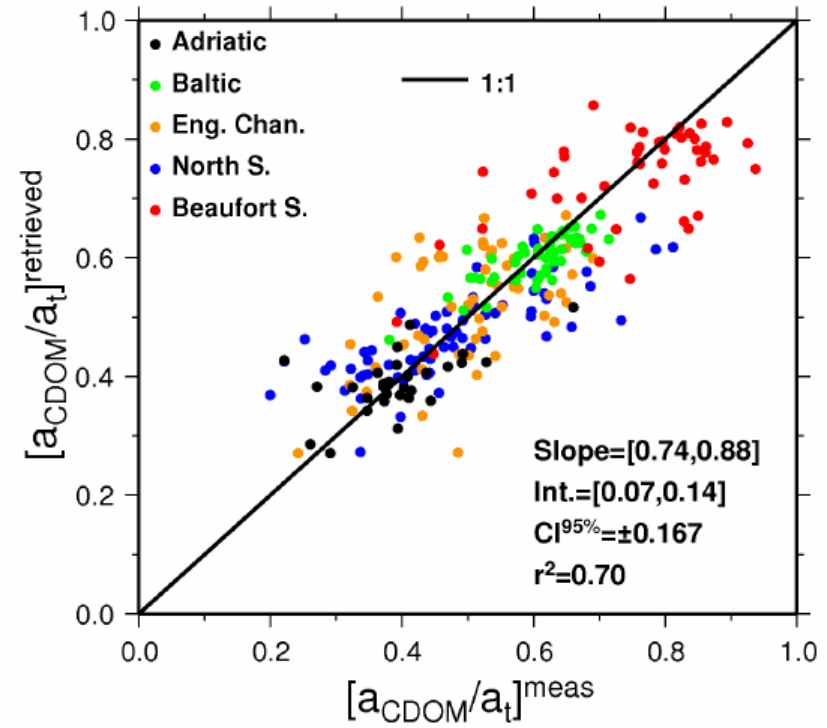
➔ Use of the second absorption peak of Chla (675 nm)

D'après Gitelson et al, 2008

Taking into account the contributions of S and Chla:

$$\left[ \frac{a_{CDOM}}{a_t} \right] (412) = \alpha + \beta \cdot \log_{10} \left( \frac{R_{rs}(412)}{R_{rs}(555)} \right) + \chi \cdot \log_{10} \left( \frac{R_{rs}(490)}{R_{rs}(555)} \right) + \delta \cdot \log_{10} (R_{rs}(555))$$

General retrieval of Light absorption (or concentration) of CDOM or Y (Bélanger 2006), with regional adaptations (Bélanger et al. 2008; Matsuoka et al. 2012)



Semi-analytical (SA) ocean color models allow retrieving multiple ocean properties from a single water-leaving radiance spectrum.

$$\hat{L}_{wN}(\lambda) = \frac{tF_0(\lambda)}{n_w^2} \sum_{i=1}^2 g_i \left\{ \frac{b_{bw}(\lambda) + b_{bp}(\lambda_0)(\lambda/\lambda_0)^{-\eta}}{b_{bw}(\lambda) + b_{bp}(\lambda_0)(\lambda/\lambda_0)^{-\eta} + a_w(\lambda) + \text{Chl} a_{ph}^*(\lambda) + a_{cdm}(\lambda_0) \exp[-S(\lambda - \lambda_0)]} \right\}^i$$

Rrs → IOPs → coloured constituents



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Rrs → IOPs → coloured constituents

The **GSM** (Maritorena et al. 2002), the **QAA** (Lee et al. 2002-2013) and **C2R-NN** (Doerffer and Schiller 2007) algorithms are well-known SA models implemented into SeaDAS and Beam-Visat softwares.

These algorithms may fail in specific coastal environments where regional algorithms will perform better.

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# **Application of the Ocean Colour methods in coastal seas: Experience at Ifremer Brest**

Francis Gohin  
Laboratoire d'Ecologie Pélagique

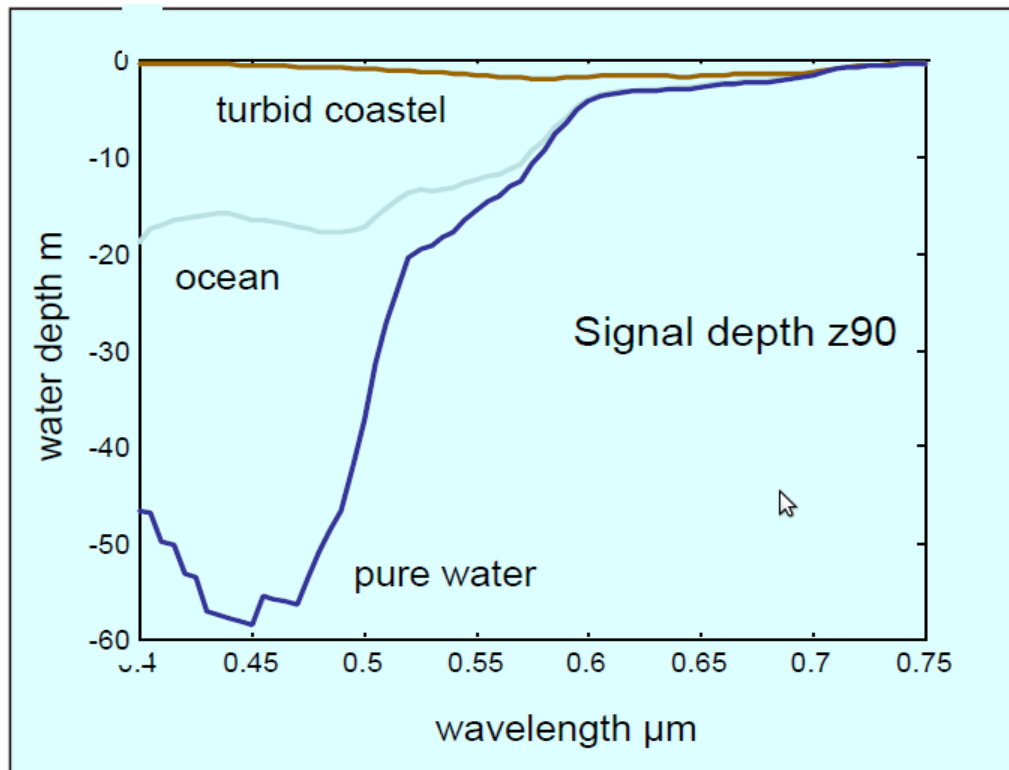
[Francis.Gohin@ifremer.fr](mailto:Francis.Gohin@ifremer.fr)

7–11 September 2015 | IFREMER | Brest, France

# But optical depth is a limit of ocean colour satellite observations

## Signal depth at different spectral bands

Multiband algorithms: the information for each band may come from a different water layer



$$z_{90} = 1/k$$

coastal:  
TSM=5 mg/l  
Chlor.=5 $\mu$ g/l  
Gelb= $a_{380}=1\text{m}^{-1}$

open ocean:  
Chlor.=1 $\mu$ g/l



- **Complex Coastal Water Algorithms (SA, NN) and often regional**
  
- **Two steps:**
  - Atmospheric corrections (TOA  $\gg$  Rrs)
  - Inversion (Rrs  $\gg$  IOPs, coloured water constituents)
  
- **Ocean colour satellite products are (potentially) very useful in coastal waters:**
  - Water quality monitoring
  - Biogeochemistry, Sediment transport
  - Fluxes / exchanges at the land-ocean interface

## **International Ocean Colour Coordinating Group**

<http://www.ioccg.org/>

## **European Space Agency MERIS Handbook**

<https://earth.esa.int/handbooks/meris/>

## **NASA Ocean Color homepage**

<http://oceancolor.gsfc.nasa.gov/cms/>

## **LOV Remote Sensing Group publications as pdf files**

<http://omtab.obs-vlfr.fr/>

## **RBINS Remote Sensing and Ecosystem Modelling team website**

<http://odnature.naturalsciences.be/remsem/>