

→ 4th ESA ADVANCED TRAINING  
ON OCEAN REMOTE SENSING

# Coastal Water Algorithms

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# Outline

## 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

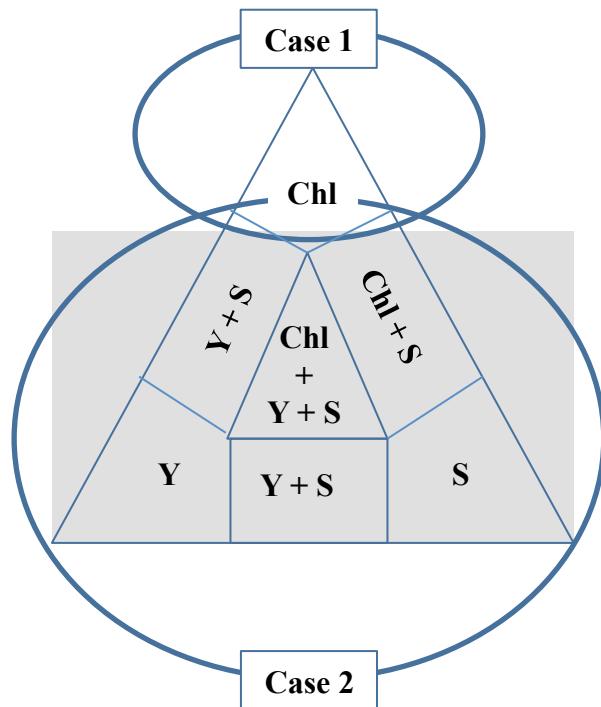
Three main types of inversion algorithms

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

## 5. Summary / Conclusions

# 1. Definition

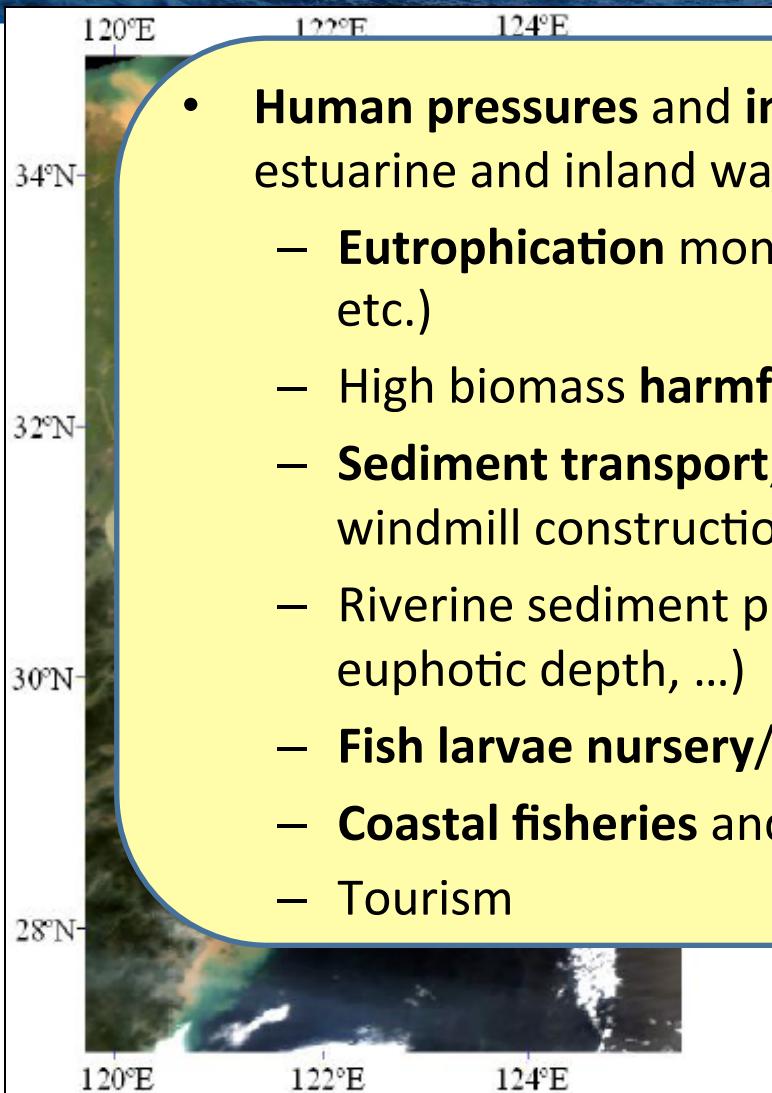
## Classification of natural waters



- 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

# 1. Definition



- **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,

decreases

increasing with increasing light absorption

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

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

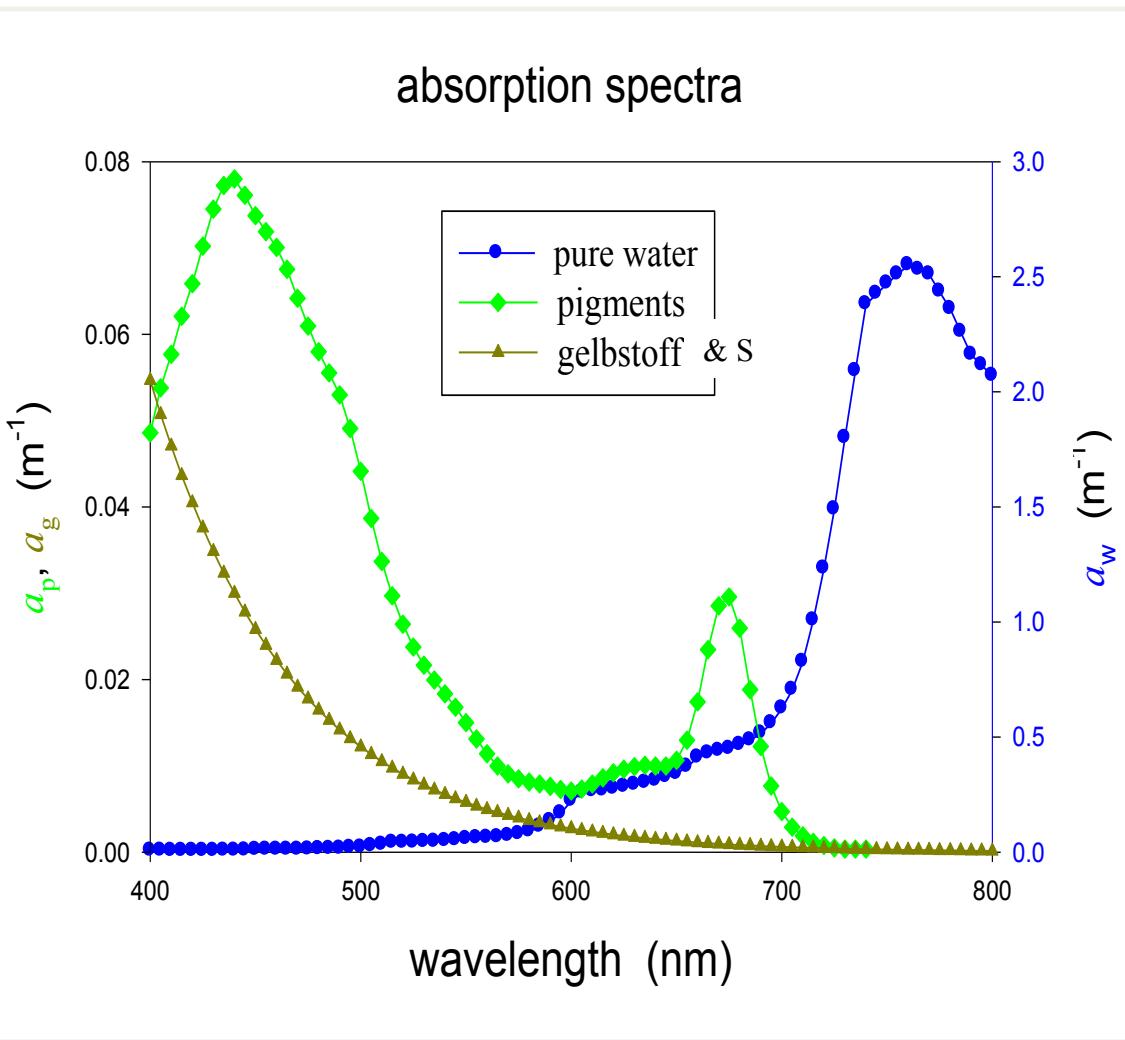
$$a_{Chla} = Chla \times a^*_{Chla}$$

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

→ The contributions of coloured water constituent are additive

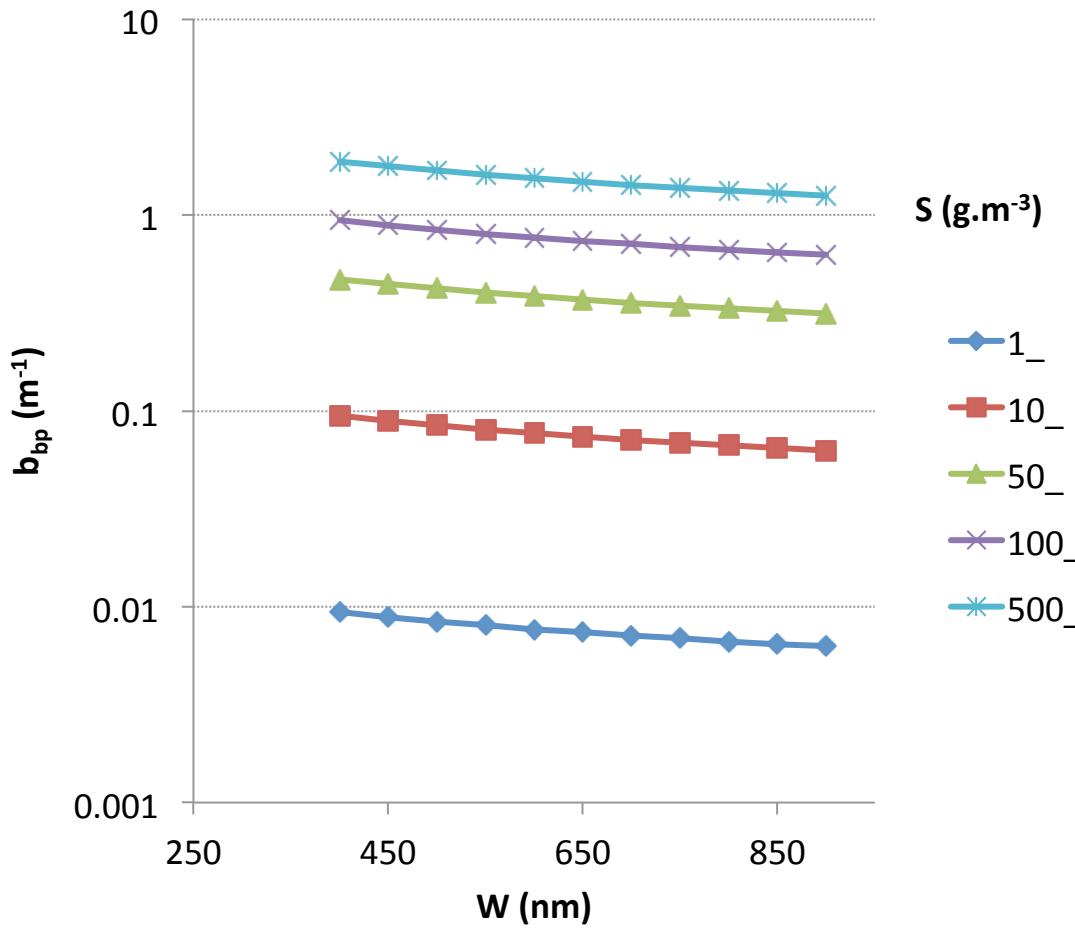
→ Concentration × mass-specific absorption/backscattering coefficients

# 1. Spectral signatures



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

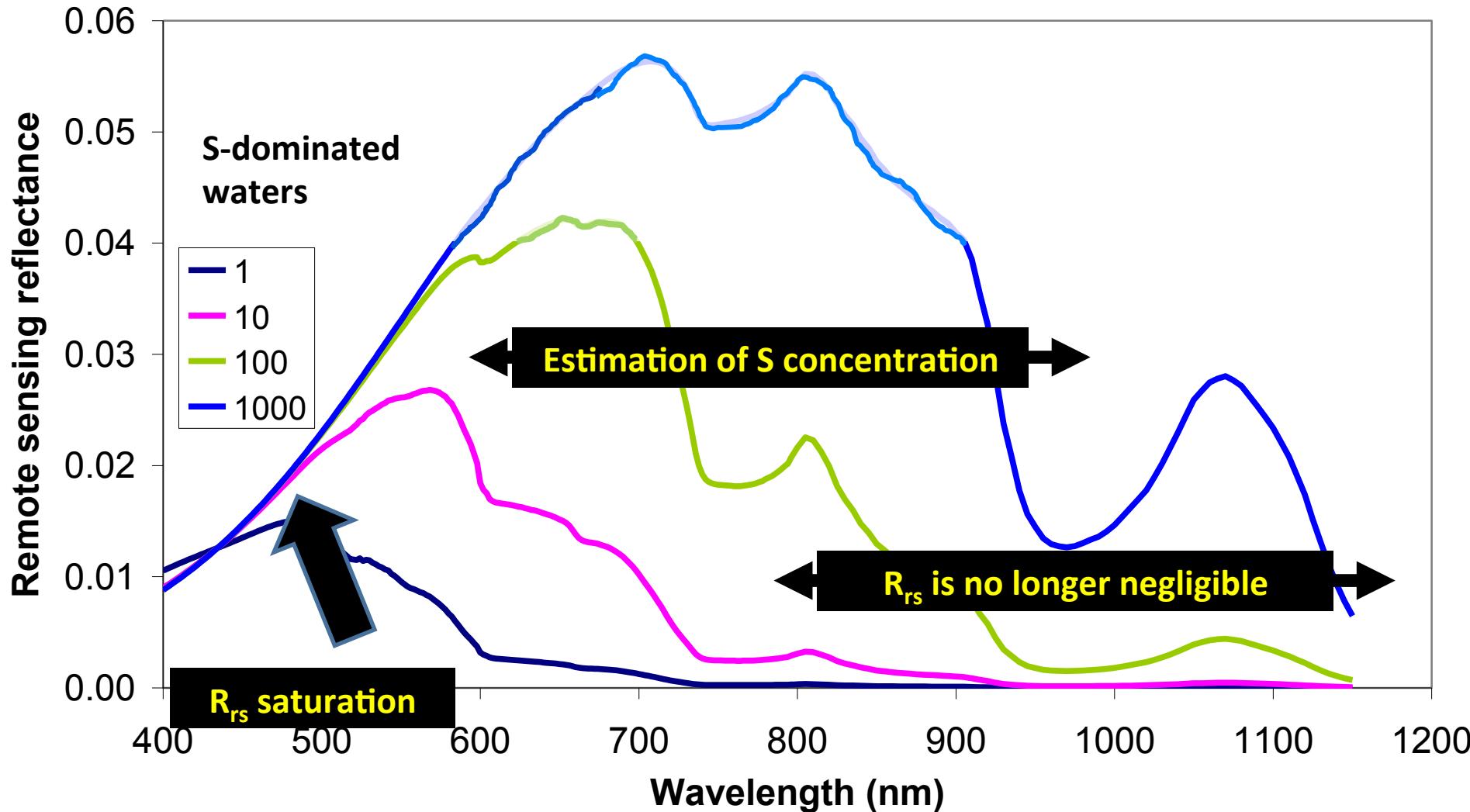
# 1. Spectral signatures



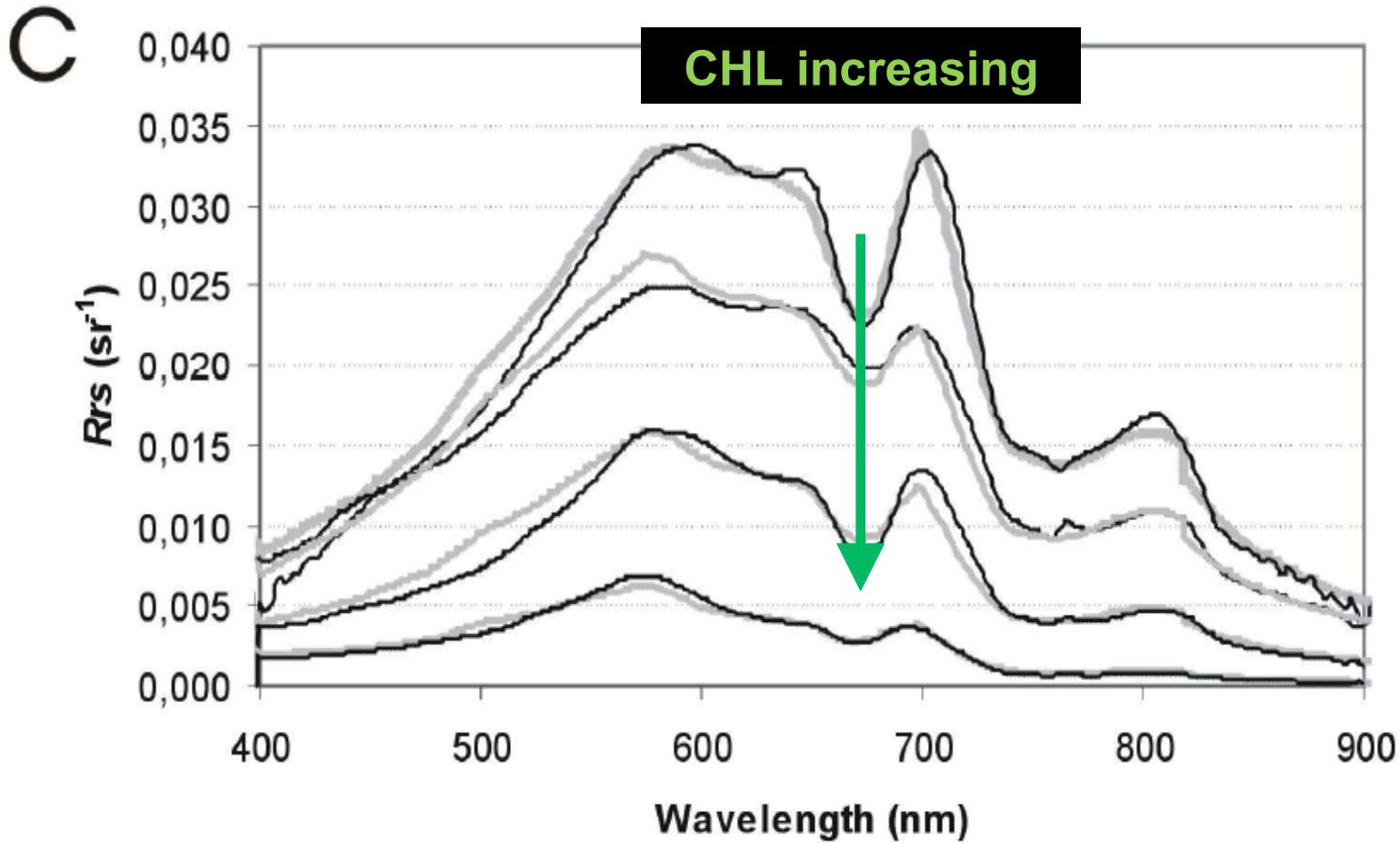
- Particles, mainly non-algal (S), backscatter light
- $b_{bp}$  is a proxy of particles concentration
- Smooth (power-law) spectral variations (slope depends on size distribution)

# 1. Spectral signatures

Varying Total Suspended matter concentration ( $\text{mg/m}^3$ )



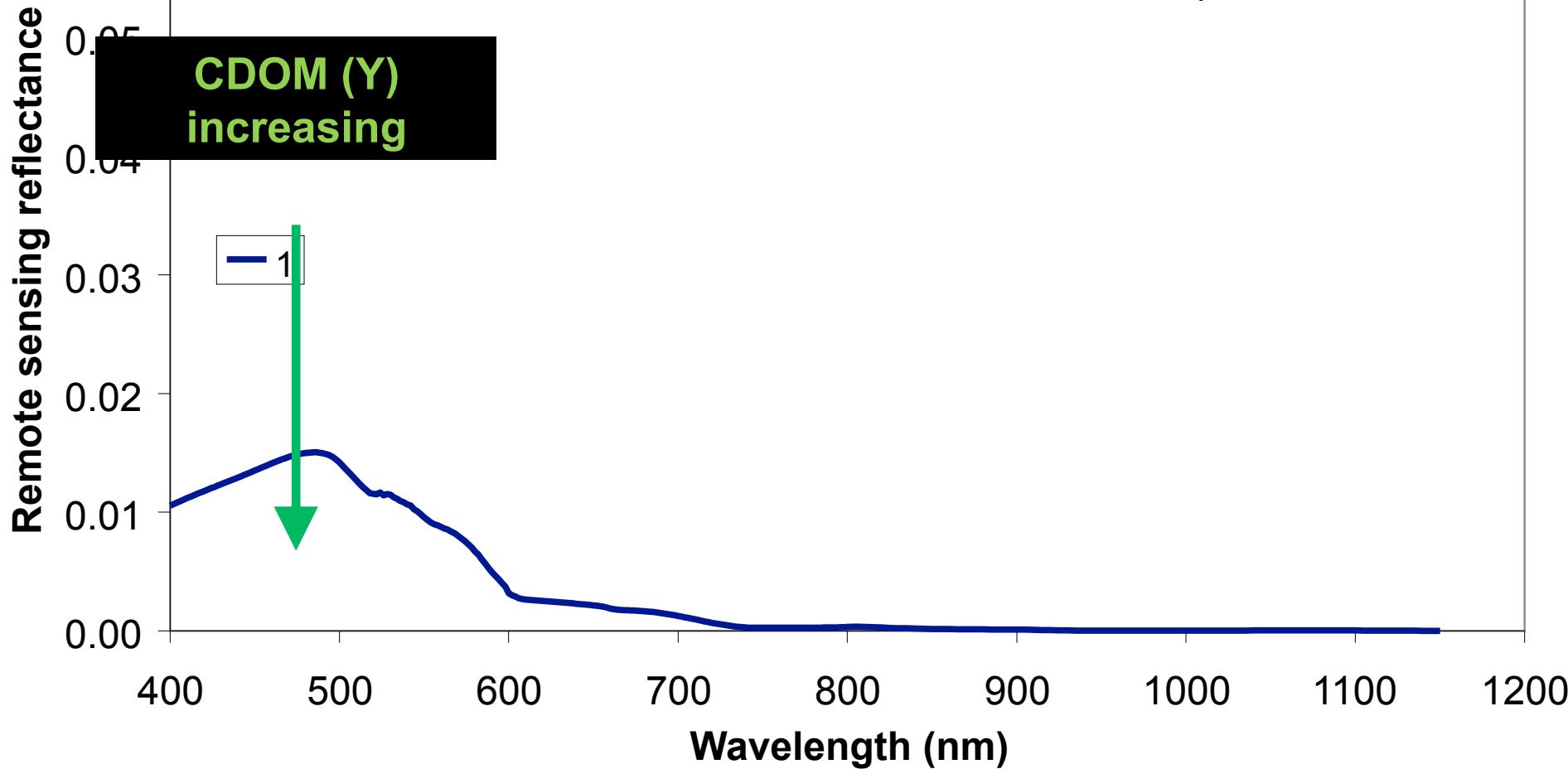
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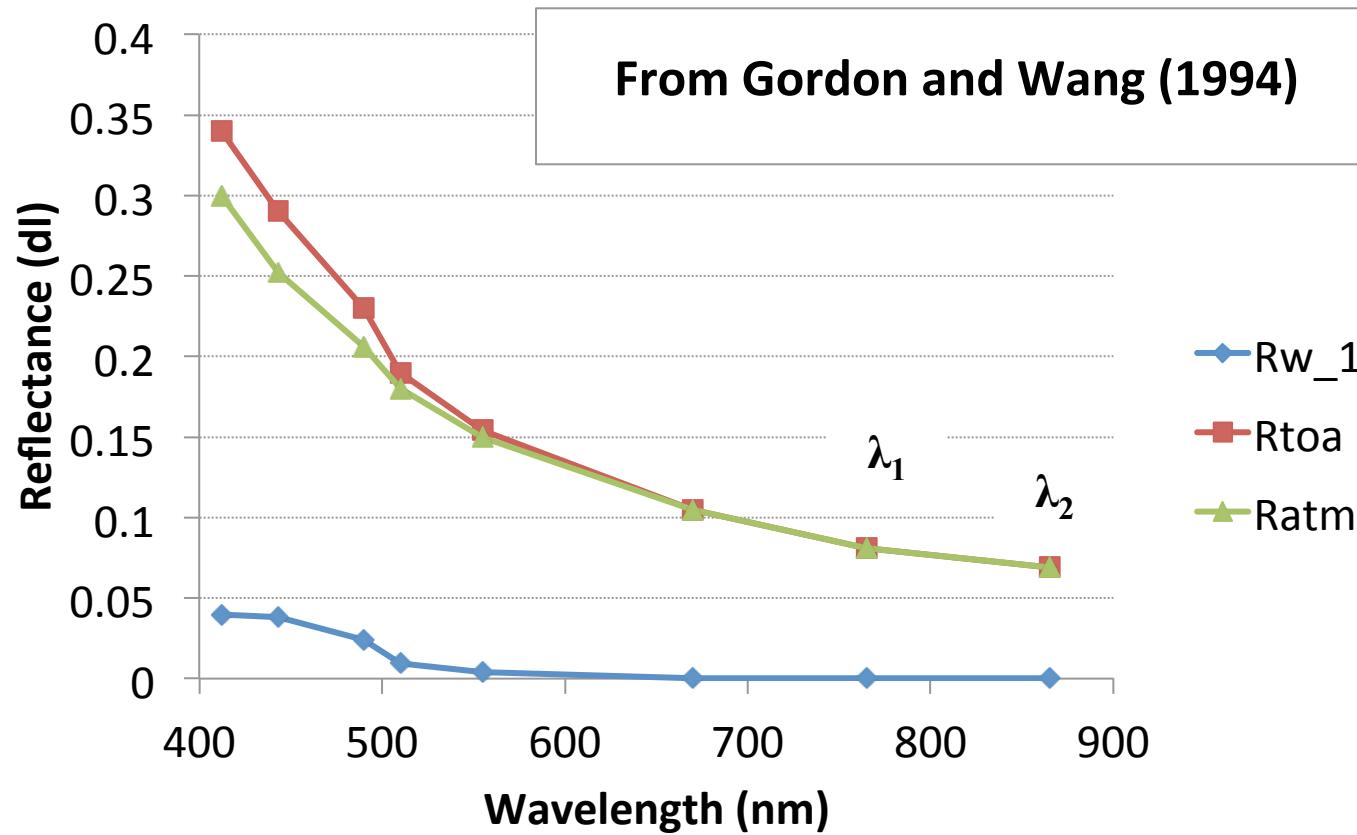
→ In turbid (absorbing and scattering) waters, light absorption by Chla at 443 nm disappears but light absorption by Chla is clearly detected

# 1. Spectral signatures

Villefranche, 11-12<sup>th</sup>  
July 2012



# 2. Atmospheric corrections

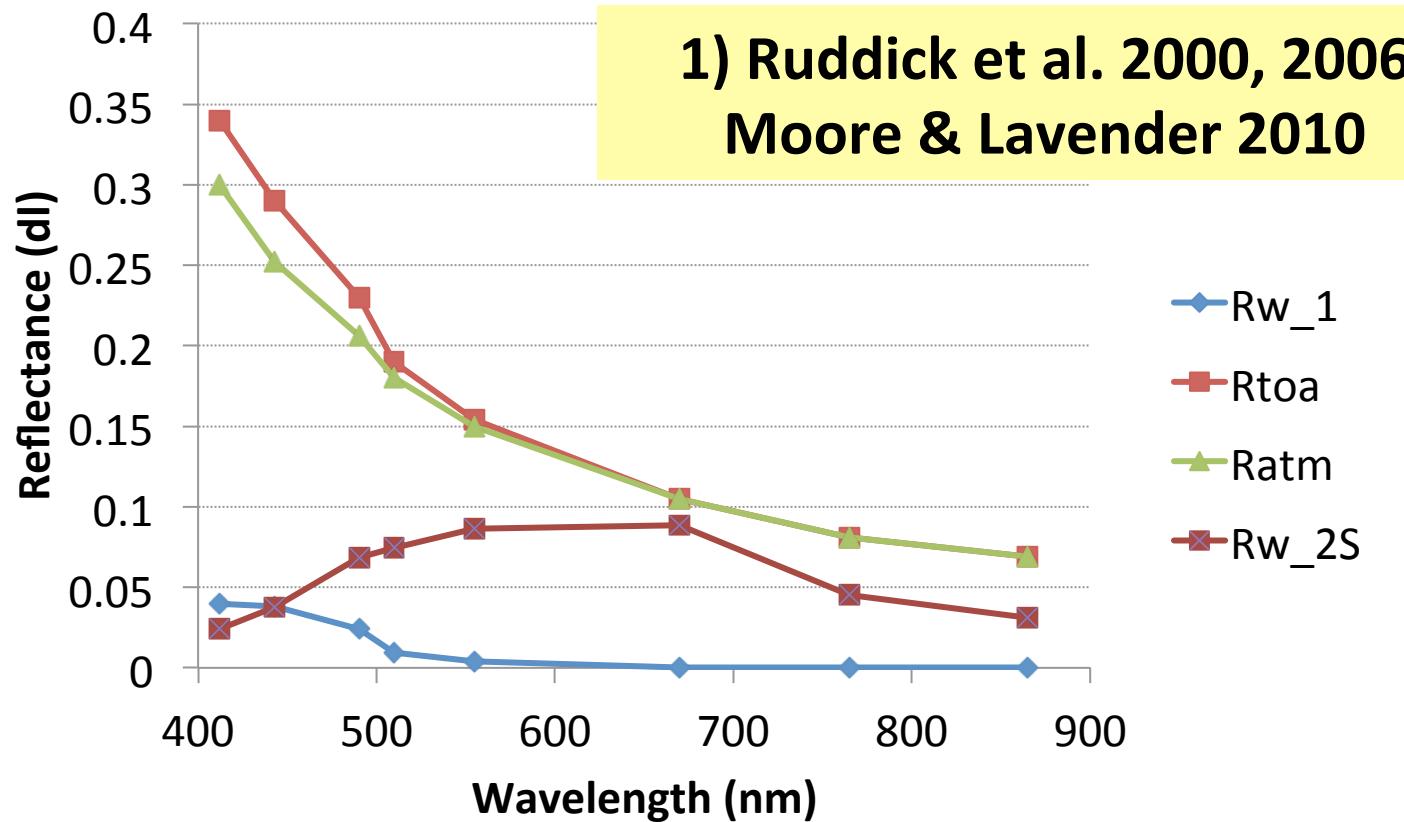


$$Rtoa(\lambda) = Rr(\lambda) + Rra(\lambda) + t \times Rw(\lambda), \quad \varepsilon(\lambda_1, \lambda_2) = Rra(\lambda_1) / Rra(\lambda_2) = (\lambda_1 / \lambda_2)^n$$

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

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

# 2. Atmospheric corrections



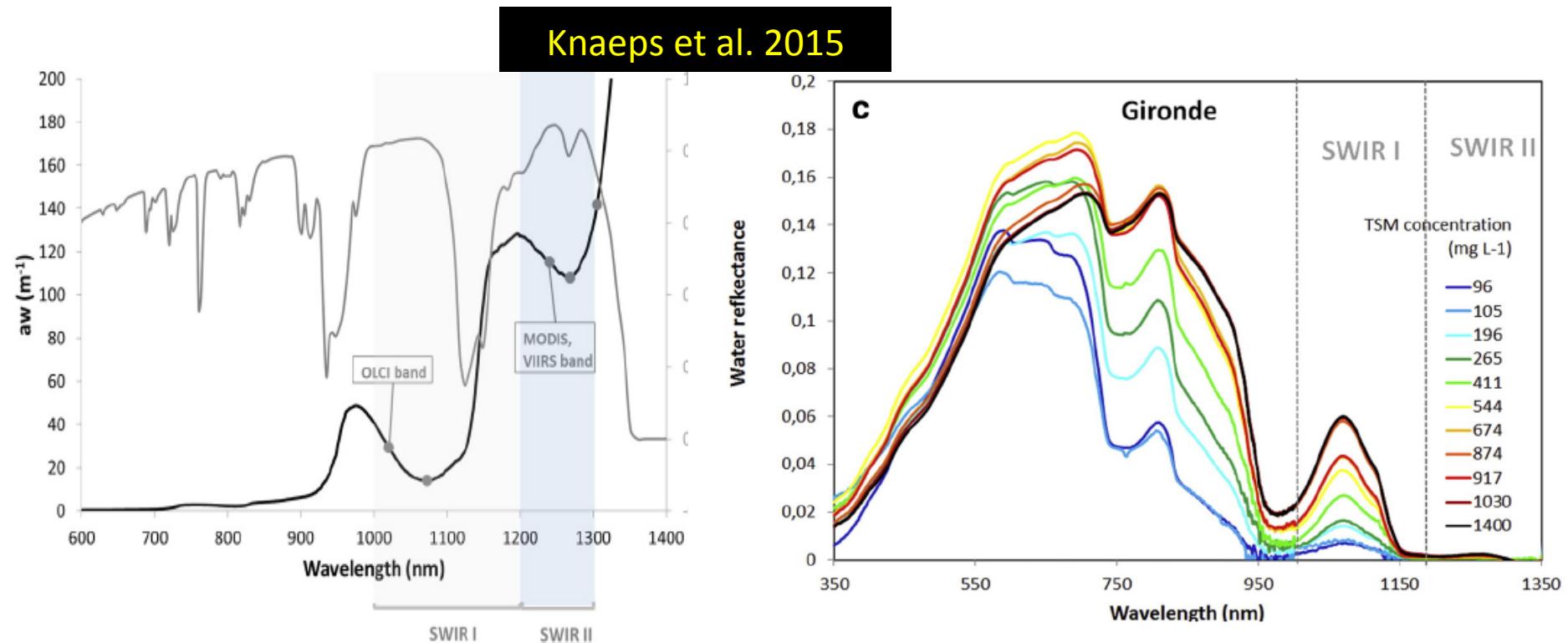
$$Rto(\lambda) = Rr(\lambda) + Rra(\lambda) + t \times Rw(\lambda), \quad \varepsilon_w(\lambda_1, \lambda_2) = \text{constant (pure water)}$$

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

Pixel-by-pixel iterations to fit Ratm( $\lambda_1, \lambda_2$ )

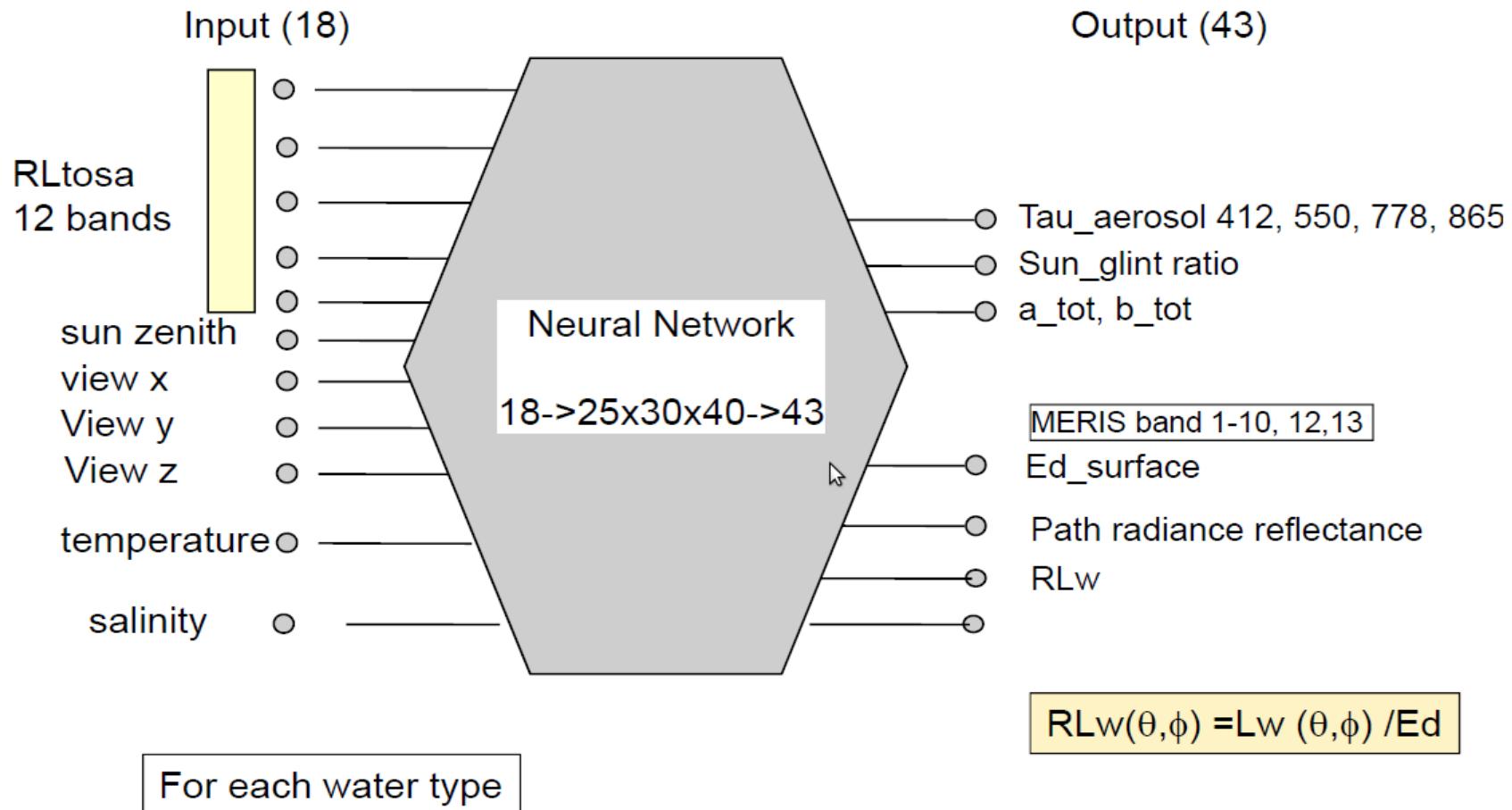
# 2. Atmospheric corrections

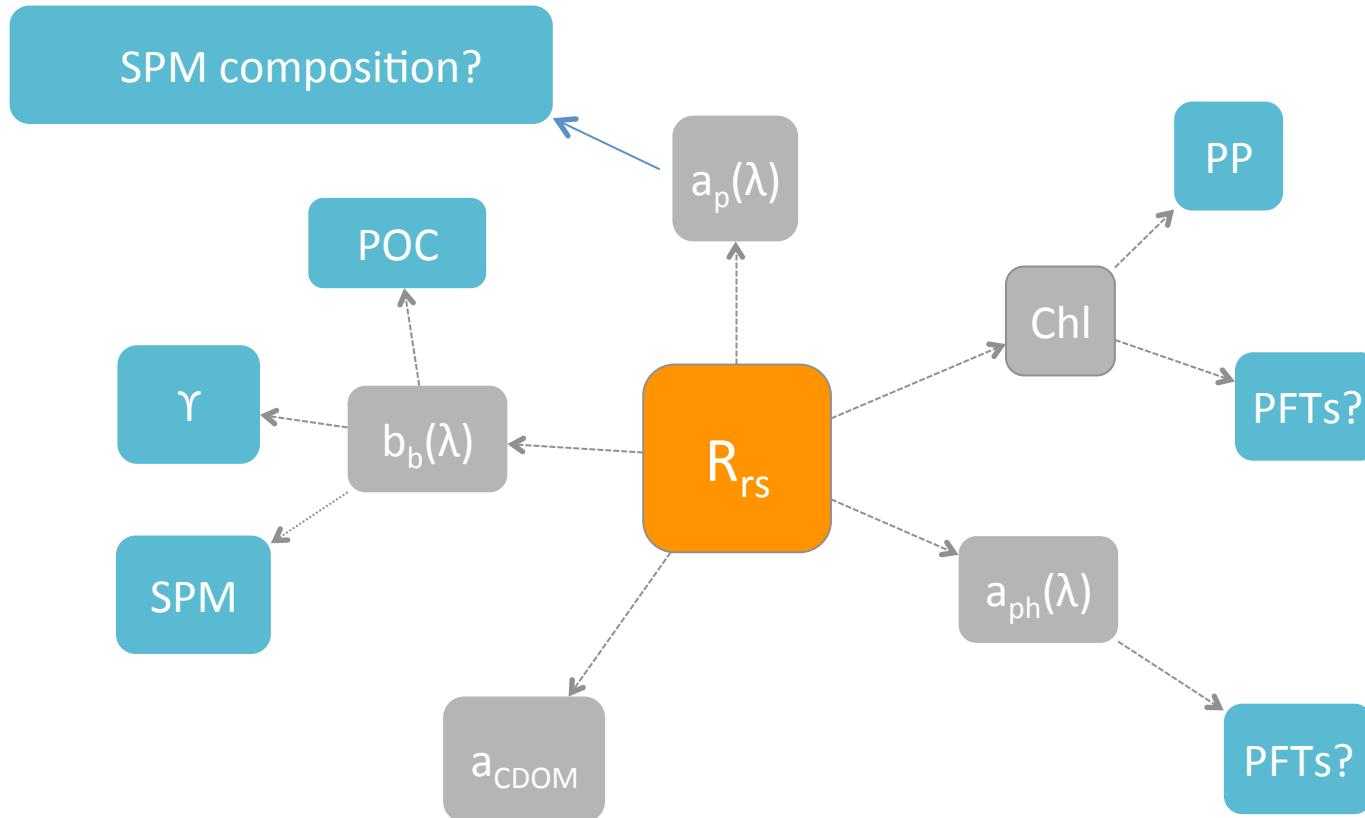
2) The NIR-SWIR atmospheric correction ([Wang & Shi 2007](#))\*



(\*) → 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

$$Rrs = f' \times b_b / (a + b_b)$$

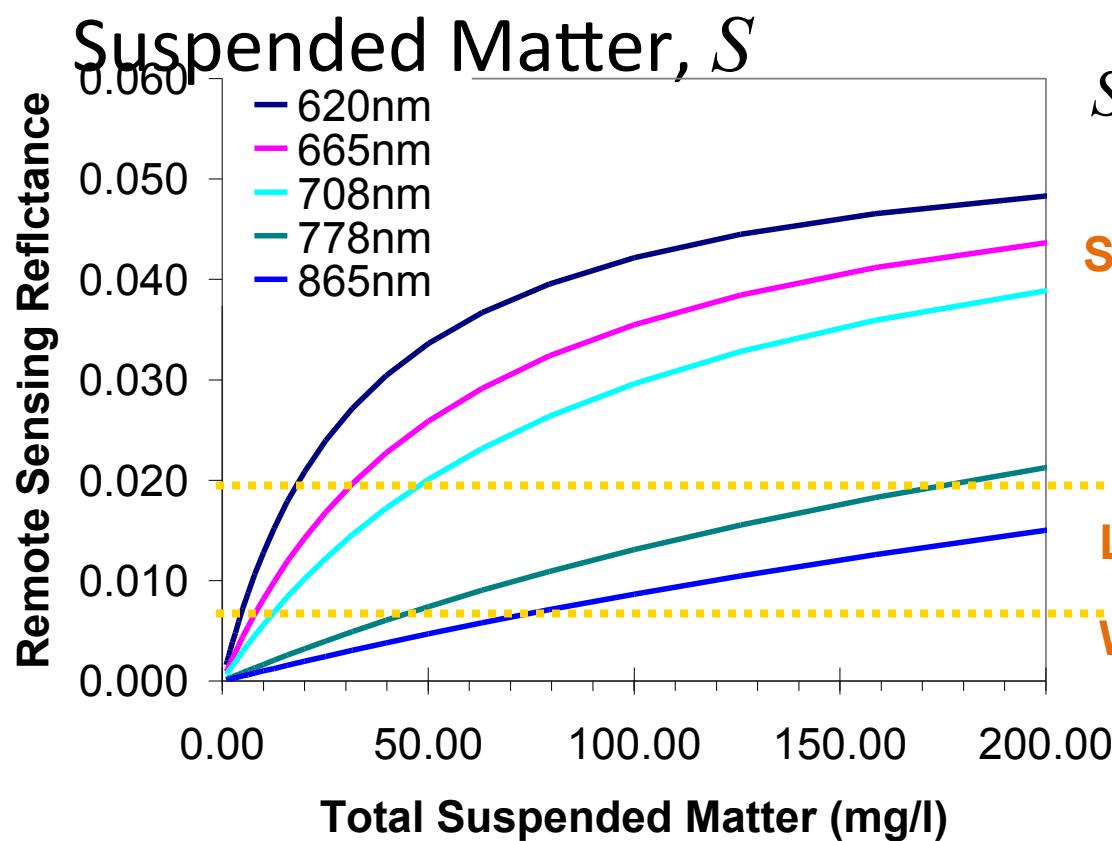
$$a = a_w + a_{Chla} + a_{NAP} + a_y$$

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

## 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

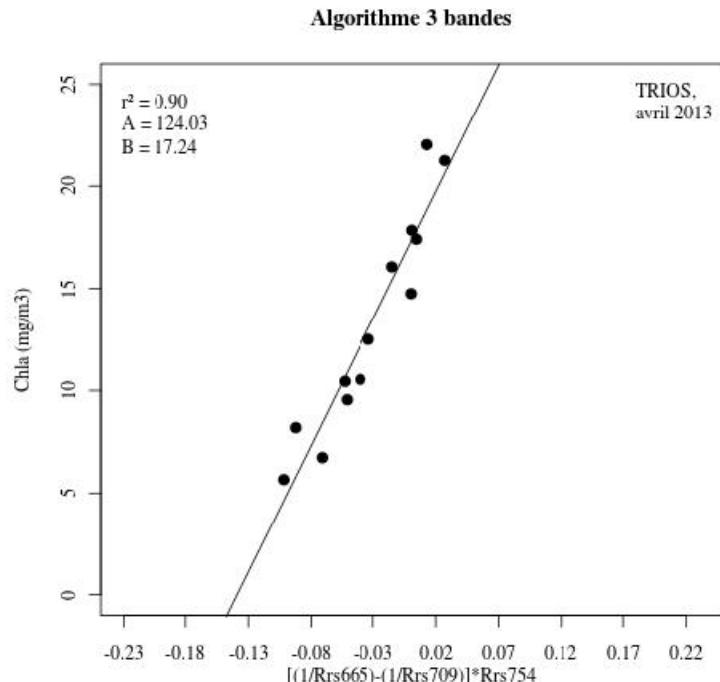
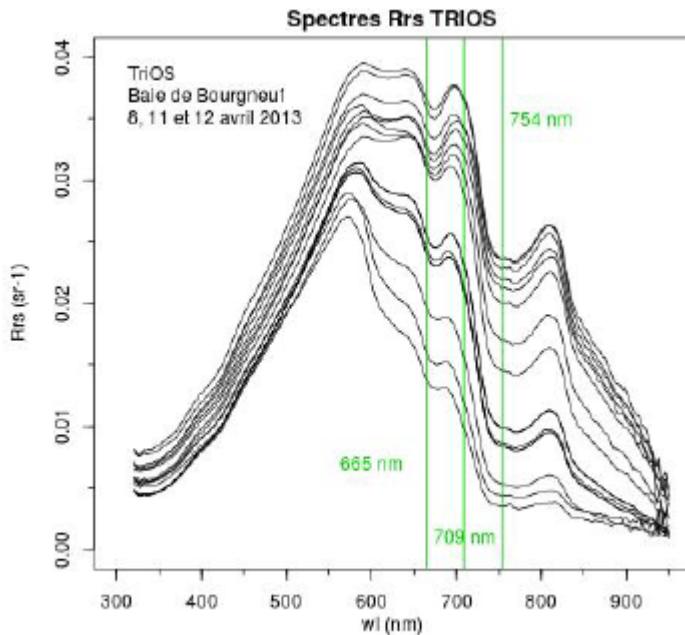
[Nechad et al, 2010]

Shen et al. 2010

Doxaran et al. 2002-2009

# 3. Chl inversion algorithms

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



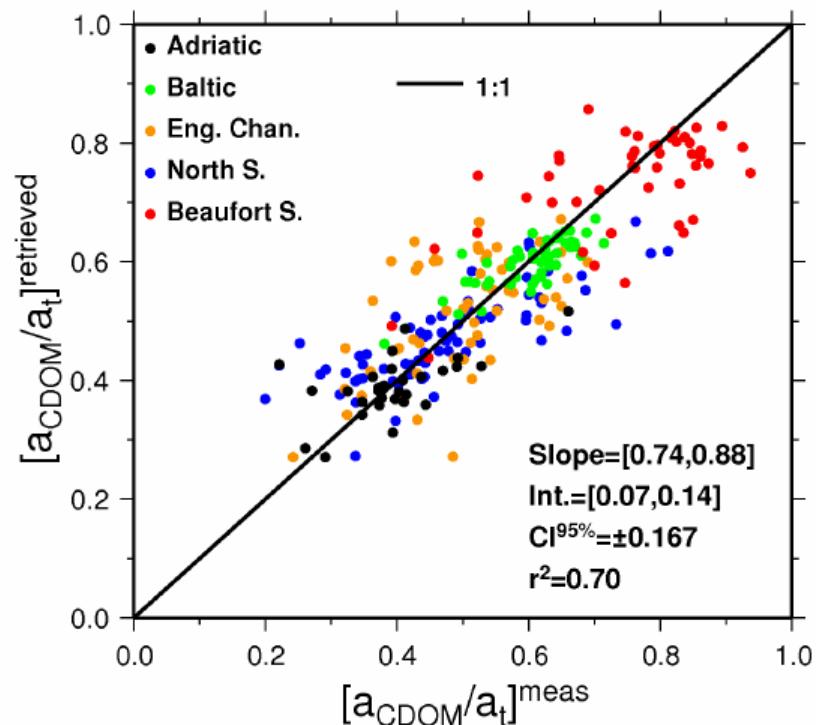
→ Use of the second absorption peak of Chla (675 nm)

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)



# 3. Semi-analytical algorithms

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

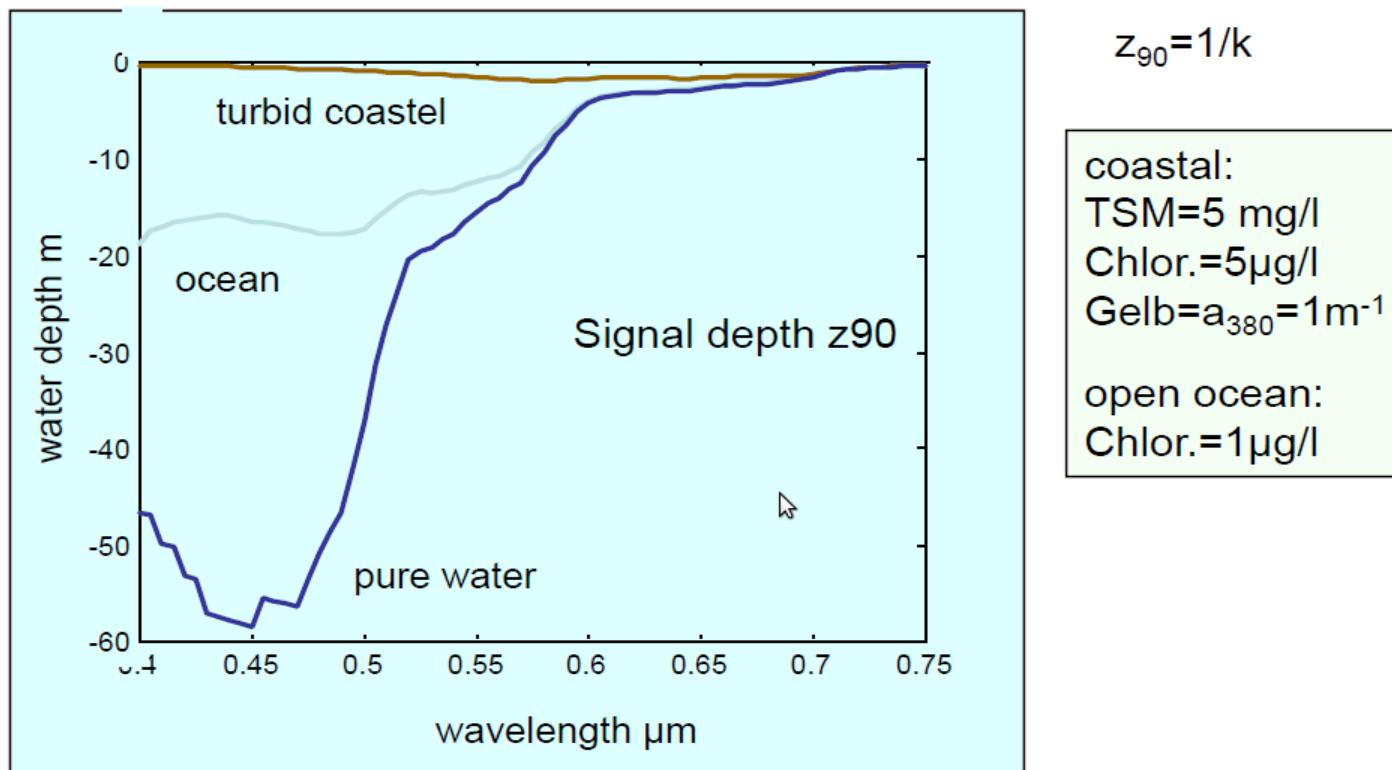
[Francis.Gohin@ifremer.fr](mailto:Francis.Gohin@ifremer.fr)

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



- **Complex Coastal Water Algorithms (SA, NN) and often regional**
- **Two steps:**
  - Atmospheric corrections (TOA >> Rrs)
  - Inversion (Rrs >> 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/>