



**The Abdus Salam
International Centre for Theoretical Physics**



2018-1

Winter College on Optics in Environmental Science

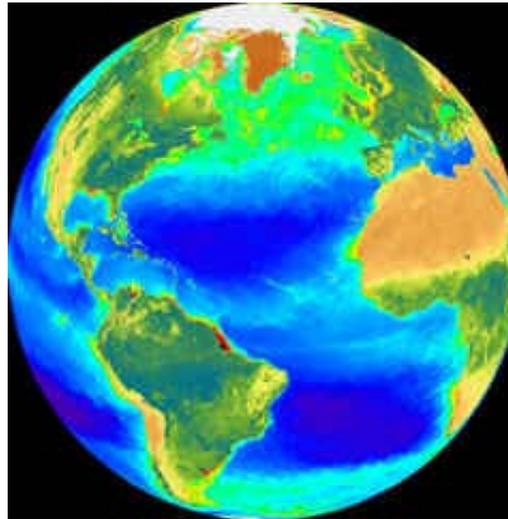
2 - 18 February 2009

Optical Remote Sensing of the Sea: Radiometry from Space

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Optical Remote Sensing of the Sea: Radiometry from Space



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Global Environment Monitoring Unit

Joint Research Centre of EC



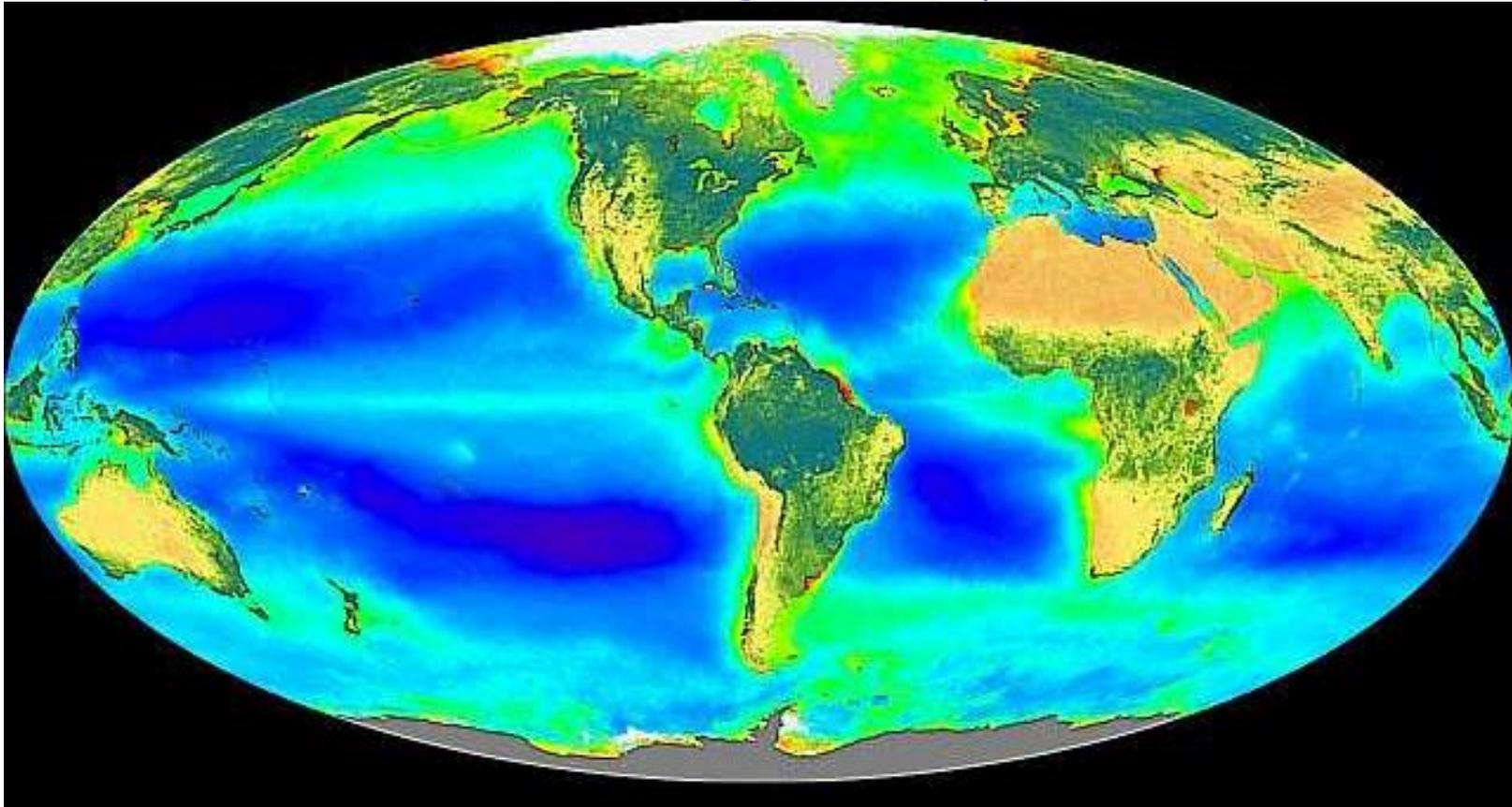
EUROPEAN COMMISSION

DIRECTORATE-GENERAL

Joint Research Centre

Chlorophyll-a (sea) and vegetation index (land) map
inferred from remote sensing data (up to 1 km resolution)

Joint Research Centre



This is not a picture taken from space

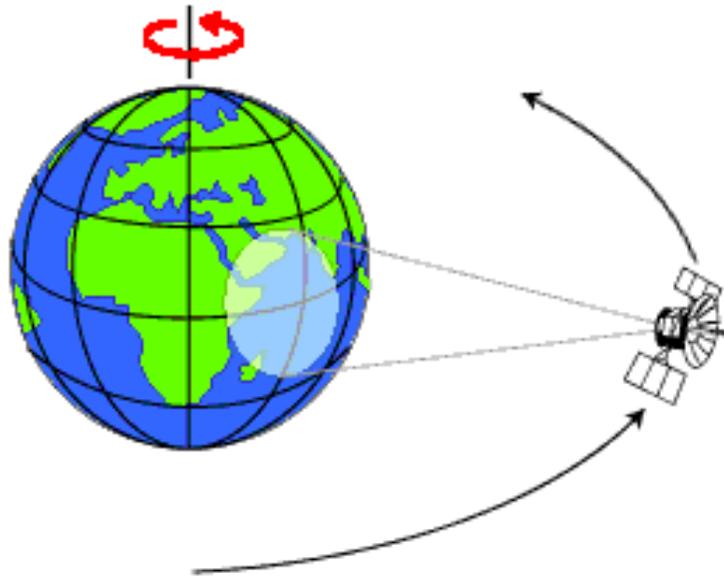
It results from the physical first and bio-optical later,
interpretation of multispectral radiometric measurements
of the Earth surface from space



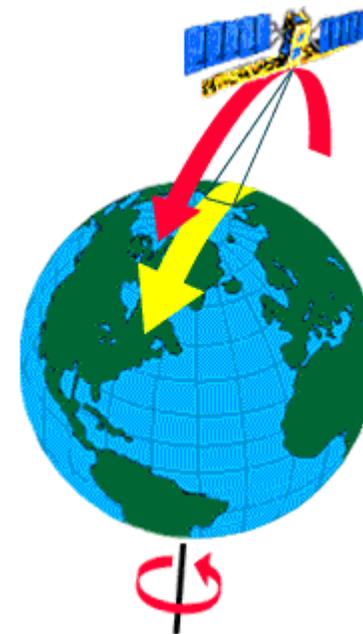
Optical remote sensing



Satellite based observing systems



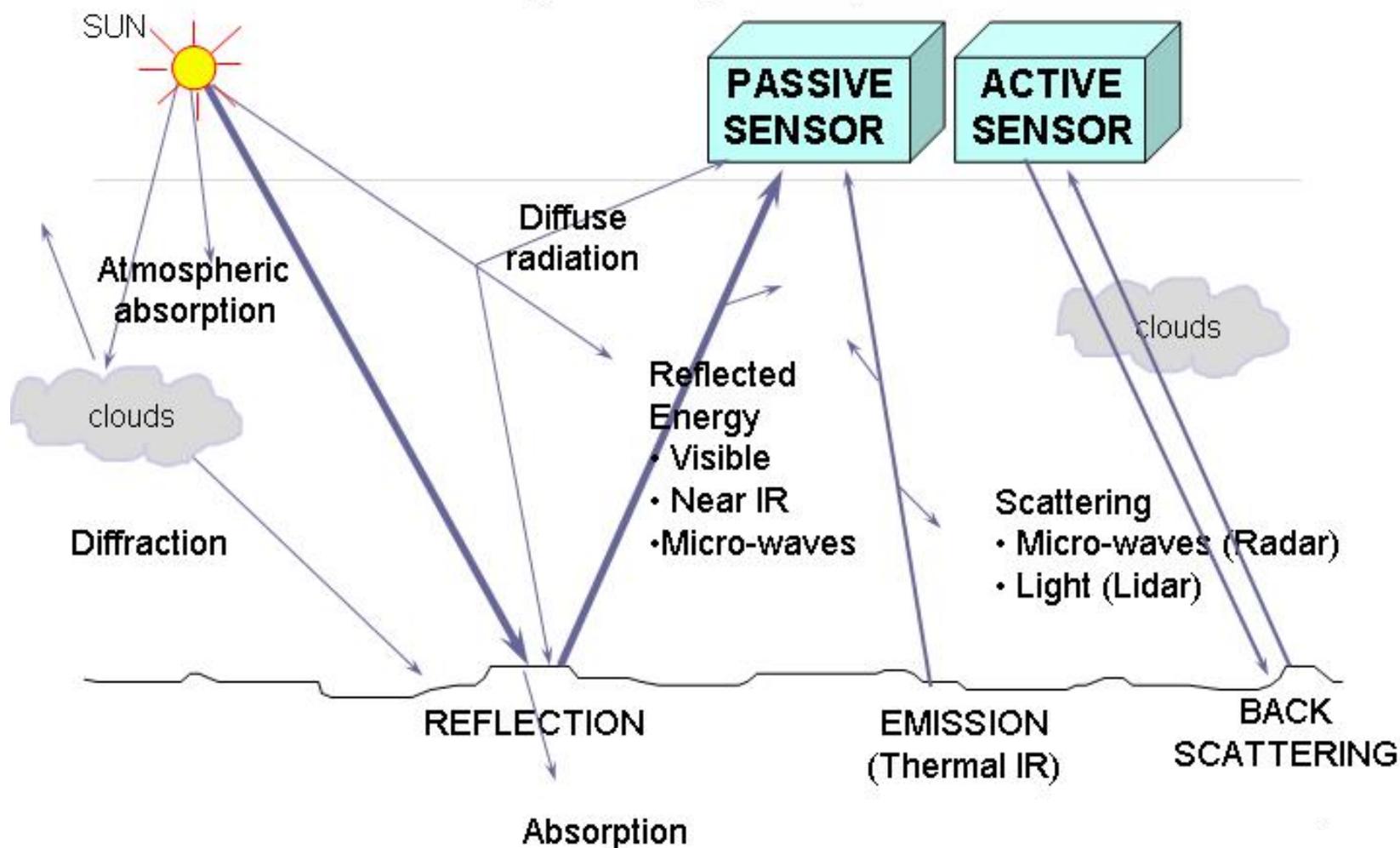
Geostationary



Near-Polar Orbiting

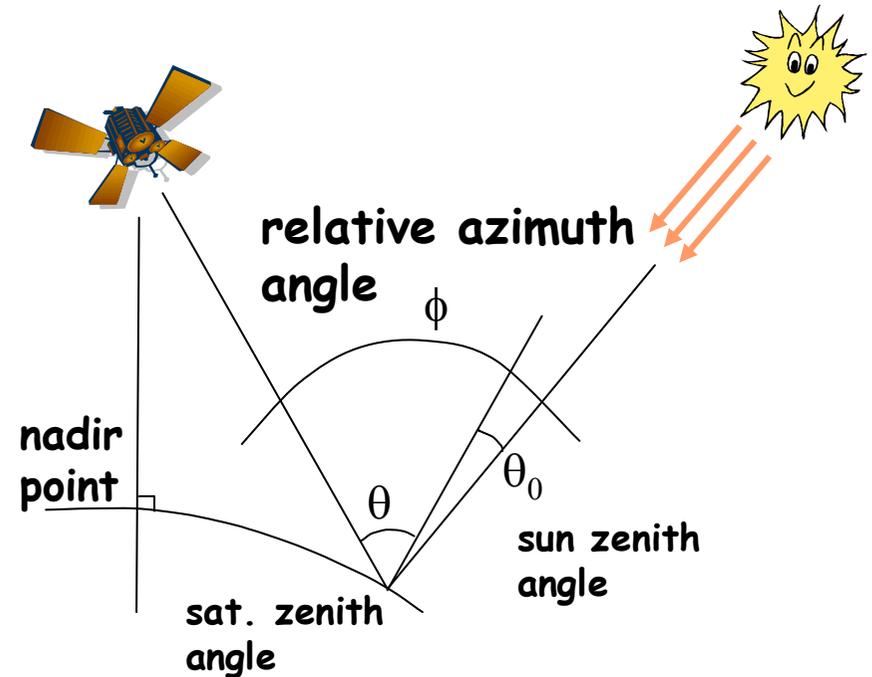
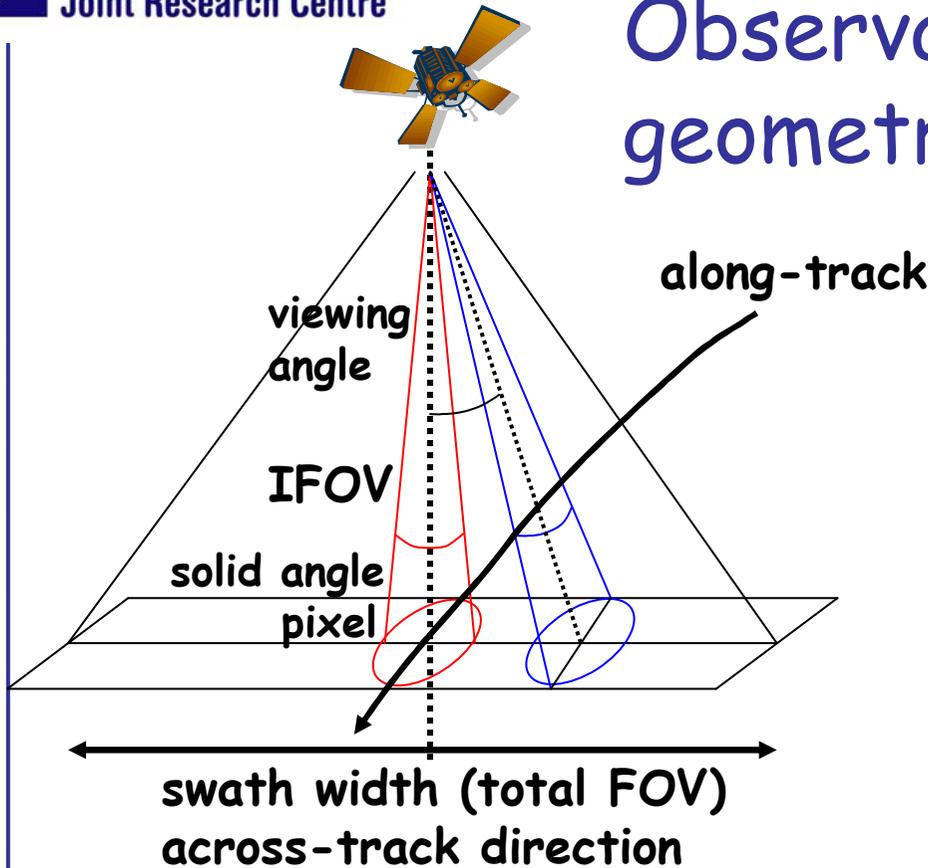


Physical principles of remote sensing





Observation and illumination geometries



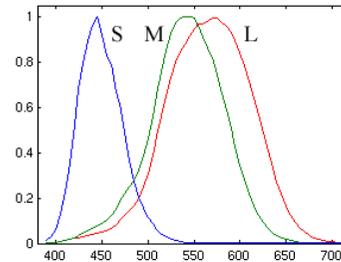
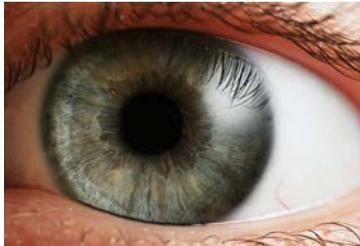
A remote sensing image is generally composed of individual measurements that the satellite acquires as it scans the Earth surface. These individual samples (pixels) have a spatial resolution which depends on the instantaneous field of view (IFOV) of the space sensor, and may vary from a few meters up to several tens kilometers.



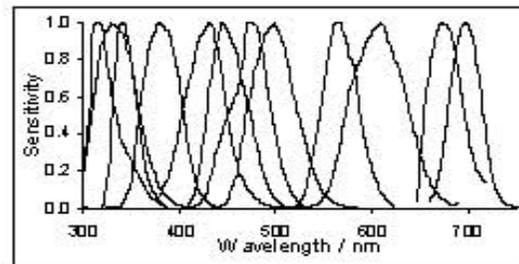
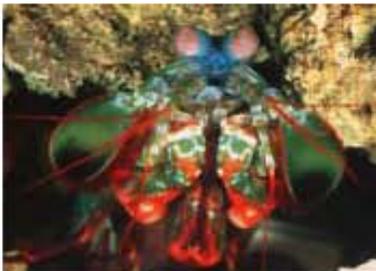
Optical Sensing Systems

Of all the techniques used in remote sensing, the observation of the Earth from optical sensors is perhaps the most easily understood in concept, because it is the most similar to our own personal remote sensing device - the human eye.

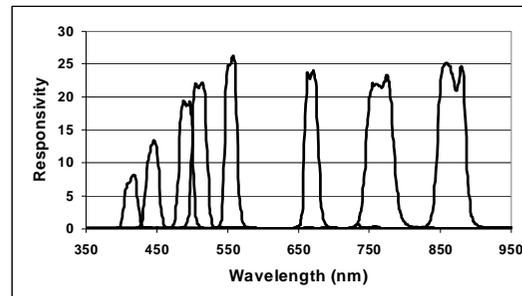
Ian Robinson (2004)



Human eye (tri-chromatic)



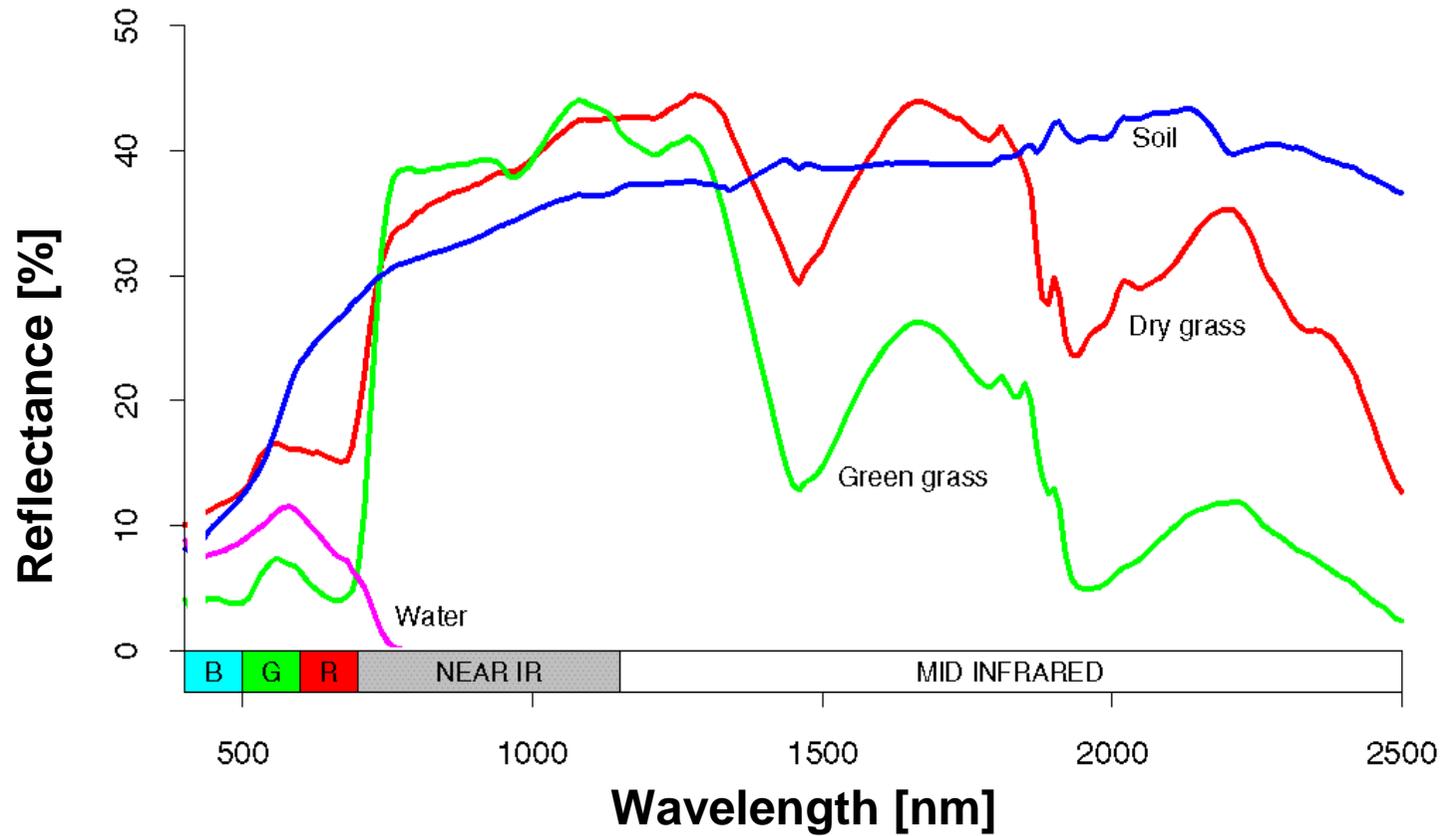
Mantis shrimp eye (multi-spectral with UV and polarization sensitivity)



SeaWiFS remote sensor (multi-spectral with NIR sensitivity)

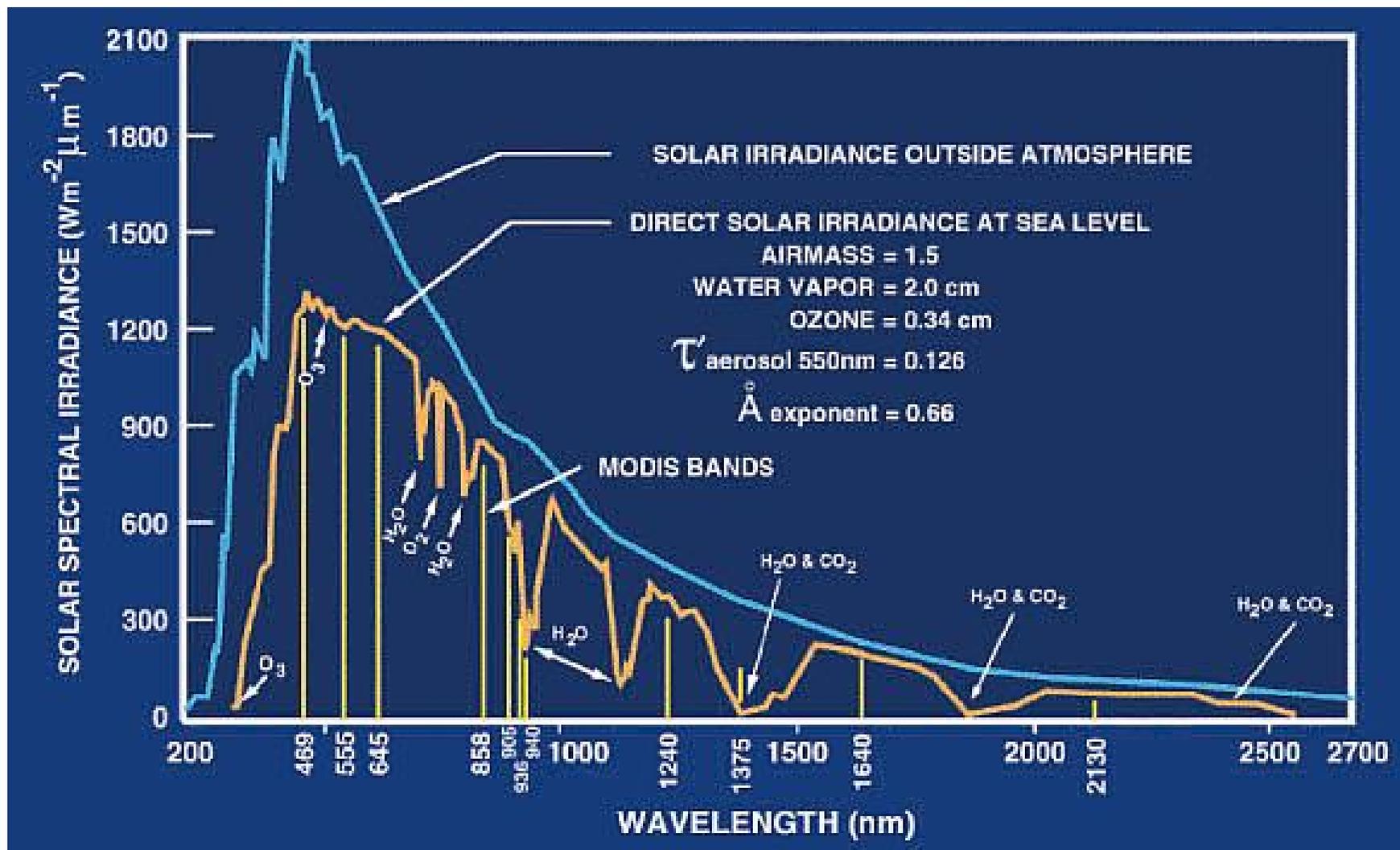


Reflectance spectra of natural surfaces





The Visible and Near-Infrared

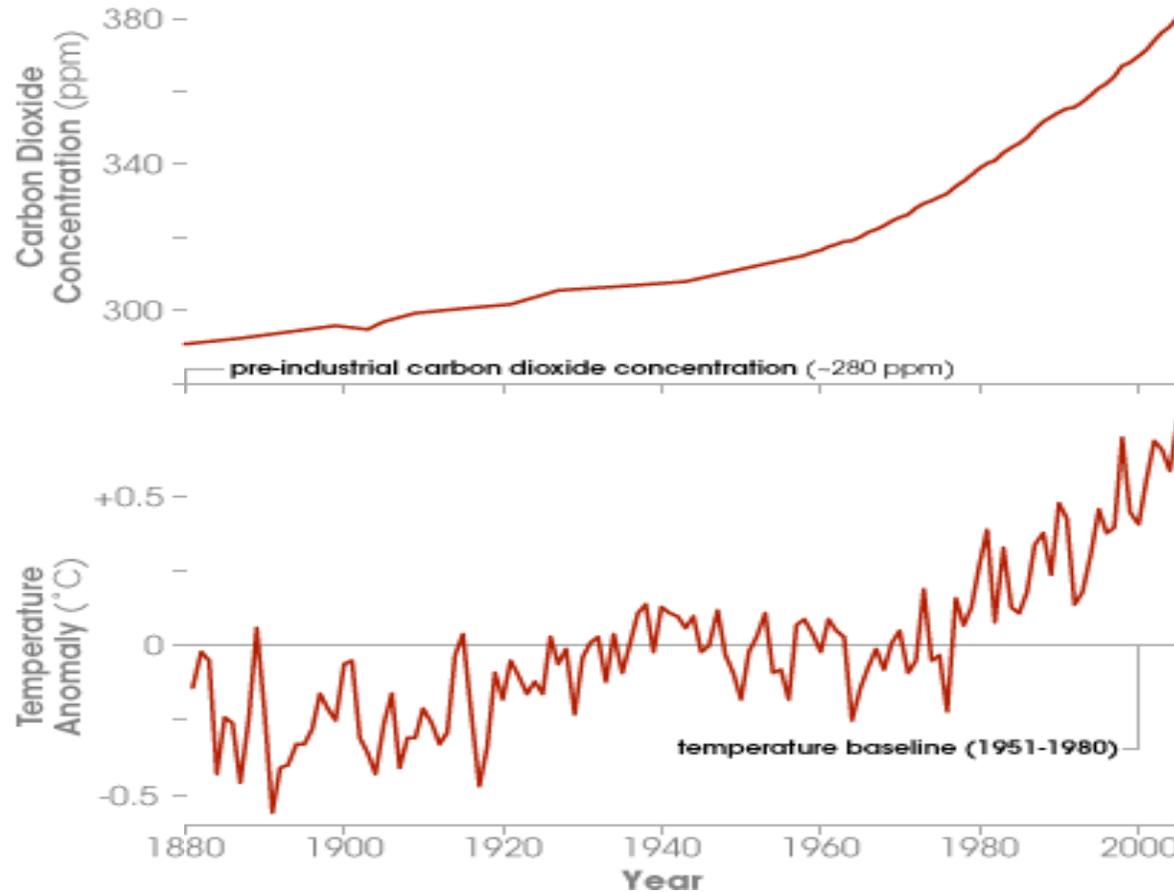




Satellites and climate change



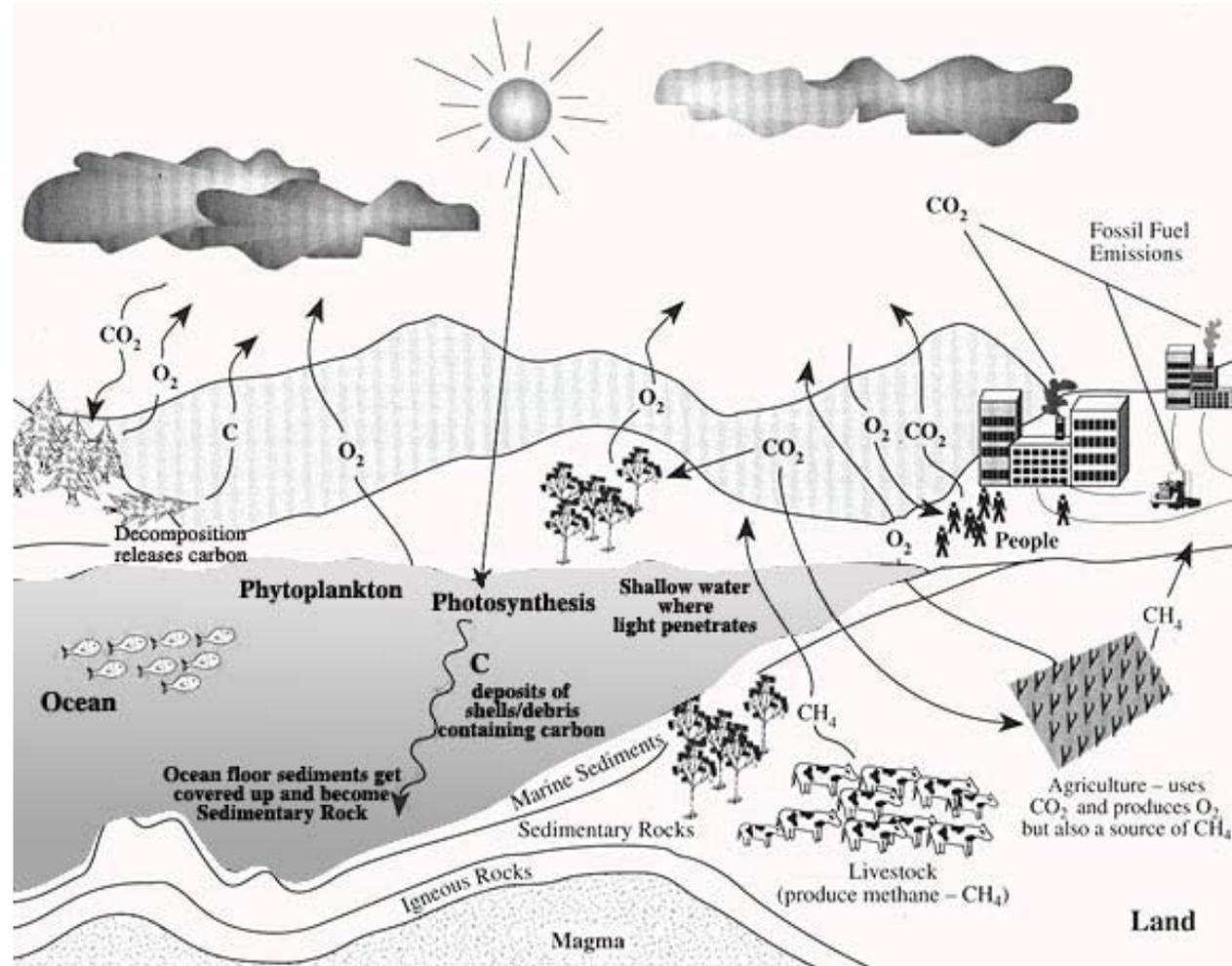
Climate change: a planetary emergency



The Earth's surface temperature has risen by about 0.5°C over the last 100 years. Computer simulations indicate that given a steady rise in levels of greenhouse gases, such as carbon dioxide, there will be a corresponding increase in surface temperatures.



Carbon Budget



Any forecast of climate change requires an accurate understanding and quantification of variables such as clouds, aerosols, and oceans.

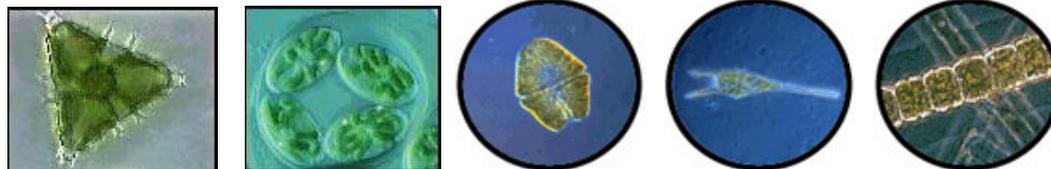


Phytoplankton: A Climate Variable

Phytoplankton is composed of microscopic unicellular marine plants which contain chlorophyll. This is vital for photosynthesis: sunlight is absorbed and utilized to fuse water molecules and carbon dioxide into carbohydrates (the plant food).

When phytoplankton dies, its detritus component sinks to the sea bottom. Although phytoplankton account for approximately 50% of the photosynthesis on the Earth, over 99% of all the carbon dioxide that has been incorporated into living things and permanently pulled from atmosphere over geologic time is buried in marine sediments.

Thus, the larger the world's phytoplankton population, the more carbon dioxide gets pulled from the atmosphere.





Rationale for Ocean Color from Space

Satellite observations ensure global coverage of the Earth surface and a consistent mapping of Essential Climate Variables required by the Global Climate Observing System (GCOS). Ocean Color is listed among the Essential Climate Variable for ocean biology.

GCOS, established in 1992 to ensure that the observations and information needed to tackle climate-related issues are obtained and made available to all potential users, is co-sponsored by the: World Meteorological Organization ([WMO](#)), Intergovernmental Oceanographic Commission ([IOC](#)) of UNESCO, United Nations Environment Programme ([UNEP](#)), and International Council for Science ([ICSU](#)).



Satellite Ocean Color



Basics of Ocean Color

Ocean color observations depend on the electromagnetic energy in the 400-700 nm spectral region. This energy is emitted by the sun, transmitted through the atmosphere and reflected by seawater after being absorbed and scattered by the optically significant seawater components.

The **optically significant components** responsible of changes in the color of seawater, are:

1. **Phytoplankton** pigments (chlorophyll-a, carotenoids, ...);
2. **Colored dissolved organic material** (CDOM or yellow substance) from phytoplankton decay or from anthropogenic origin;
3. **Suspended particulate matter** composed of *i.* organic particulates (detritus) consisting of decomposed phytoplankton and zooplankton cell fragments; and *ii.* inorganic particulates (e.g., sand and dust) from soil erosion, sea bottom re-suspension and rivers transport.

Thus the **color of seawater** depends on the relative concentrations of optically significant components.



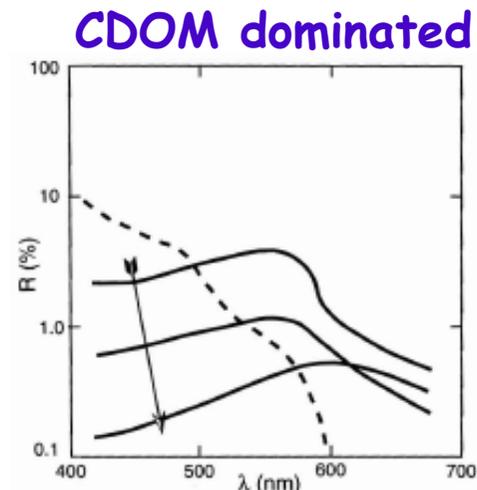
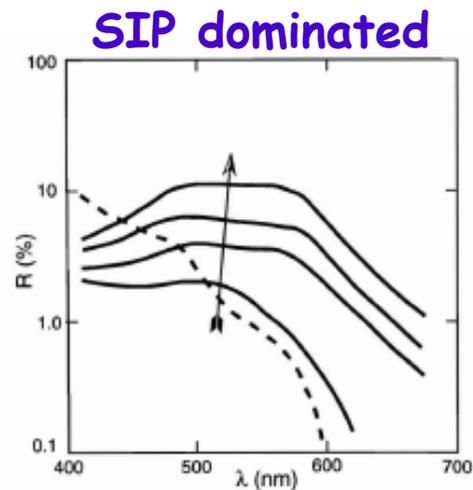
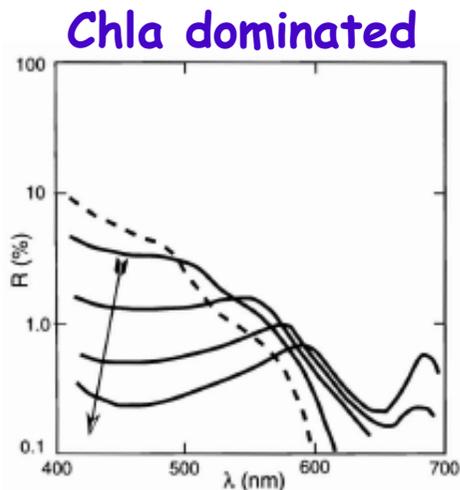
The Color of Seawater

Clear water absorbs in the red spectral region and scatters the light at shorter wavelength: this explains its blue color.

Phytoplankton cells contain chlorophyll-a (**Chla**) that mostly absorbs in the blue spectral region: this explains the green color of phytoplankton dominated seawater.

Suspended inorganic particles (**SIP**) backscatters sunlight at almost all the visible wavelengths: this explains the brightness of sediment dominated seawater.

Colored dissolved organic matter (**CDOM**) highly absorbs in the blue-green spectral region: this explains the brownish color of yellow-substance dominated seawater





Seawater types

The color of seawater is determined from its reflectance spectrum. The concentration of optically significant constituents can be derived the reflectance spectrum.

Such a process is straight forward in oceanic waters (classified as **Case-1** waters) which exhibit correlation between phytoplankton and the other optically significant constituents (i.e., particulate and dissolved matter).

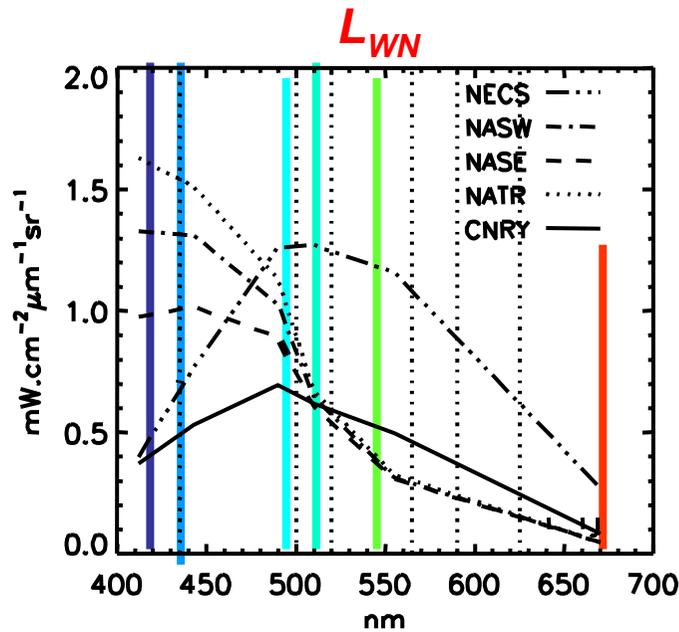
The process is much more difficult in optically complex costal waters which exhibit lack of correlation between chlorophyll-a and the other optically significant constituents (these waters are generally classified as **Case-2** and show yellow-brown colors) .



Definition and Principles

Satellite Ocean Color indicates remote sensing of the sea in the visible and near infrared with the primary objective of determining *the radiance emerging from the sea* from top-of-atmosphere signal.

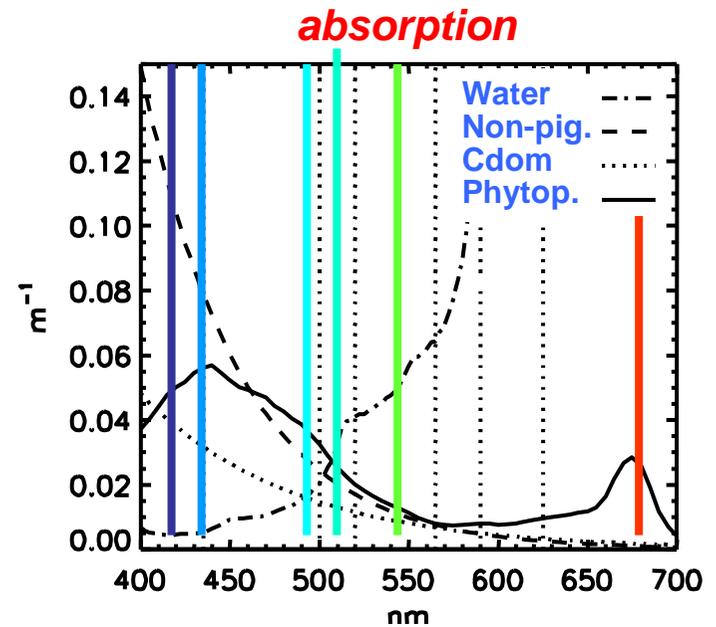
Apparent optical properties (AOP)



AOPs
vary with
IOPs



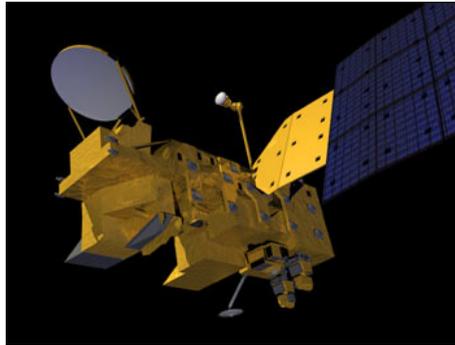
Inherent optical properties (IOP)



Derived Ocean Color products are the concentration of optically significant constituents (e.g., the *chlorophyll a* concentration, used as a proxy for the *marine phytoplankton biomass*).



AQUA and ENVISAT: Two Earth Observing Platforms



AQUA platform (NASA)

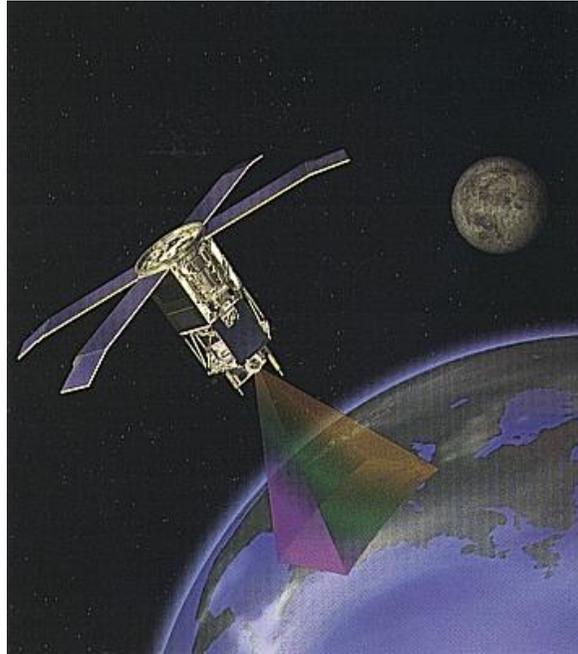


ENVISAT platform (ESA)

AQUA (NASA) and ENVISAT (ESA) are two major Earth Observing Platform to study the complex interactions amongst the land, ocean, air, ice and biological systems.



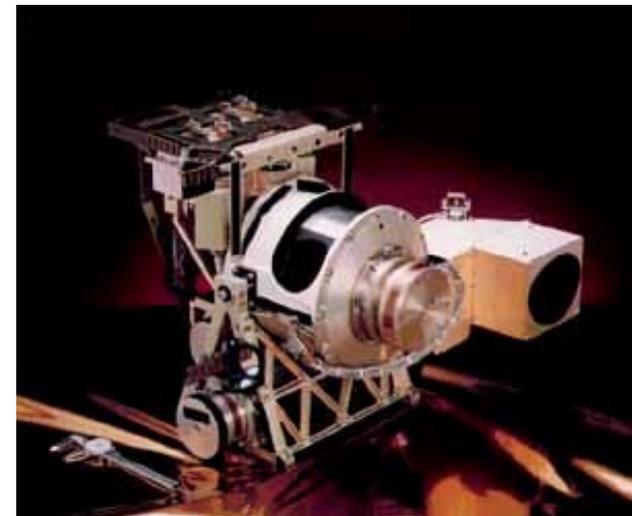
Sea-viewing Wide Field-of-view Sensor (SeaWiFS)



Band	Wavelength	
1	402-422 nm	
2	433-453 nm	
3	480-500 nm	
4	500-520 nm	
5	545-565 nm	
6	660-680 nm	
7	745-785 nm	
8	845-885 nm	

Mission Characteristics

Orbit Type	Sun Synchronous
Altitude	705 km
Equator Crossing	Noon + 20 min
Orbital Period	99 minutes
Swath Width	2,801 km (58.3 deg.)
Spatial Resolution	1.1 km LAC





Ocean Color: the concept

$$L_{Toa}(\lambda) = L_{atm}(\lambda) + t_d L_w(\lambda) + t L_g(\lambda)$$

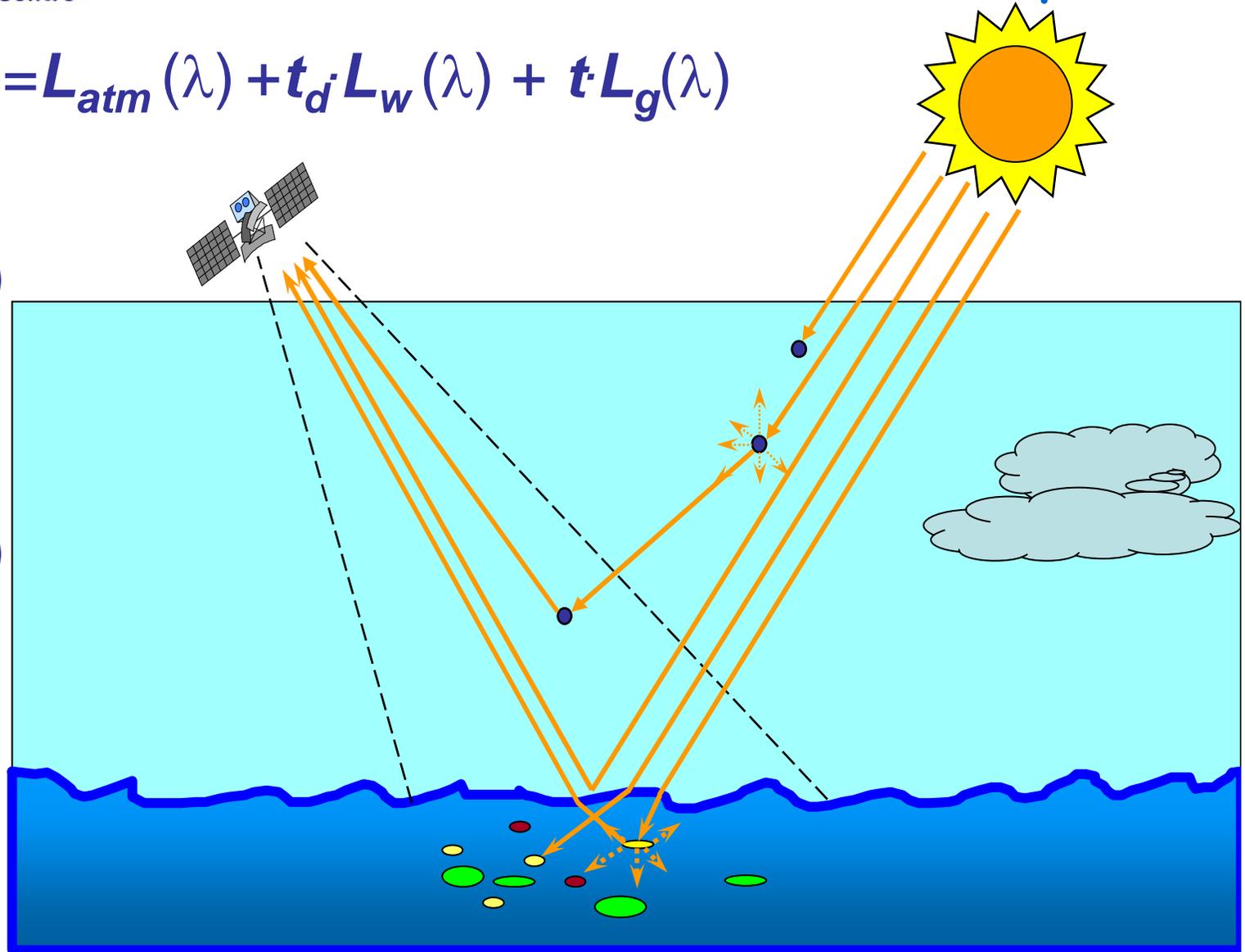
Joint Research Centre

$L_{Toa}(\lambda)$

$L_{atm}(\lambda)$

$L_g(\lambda)$

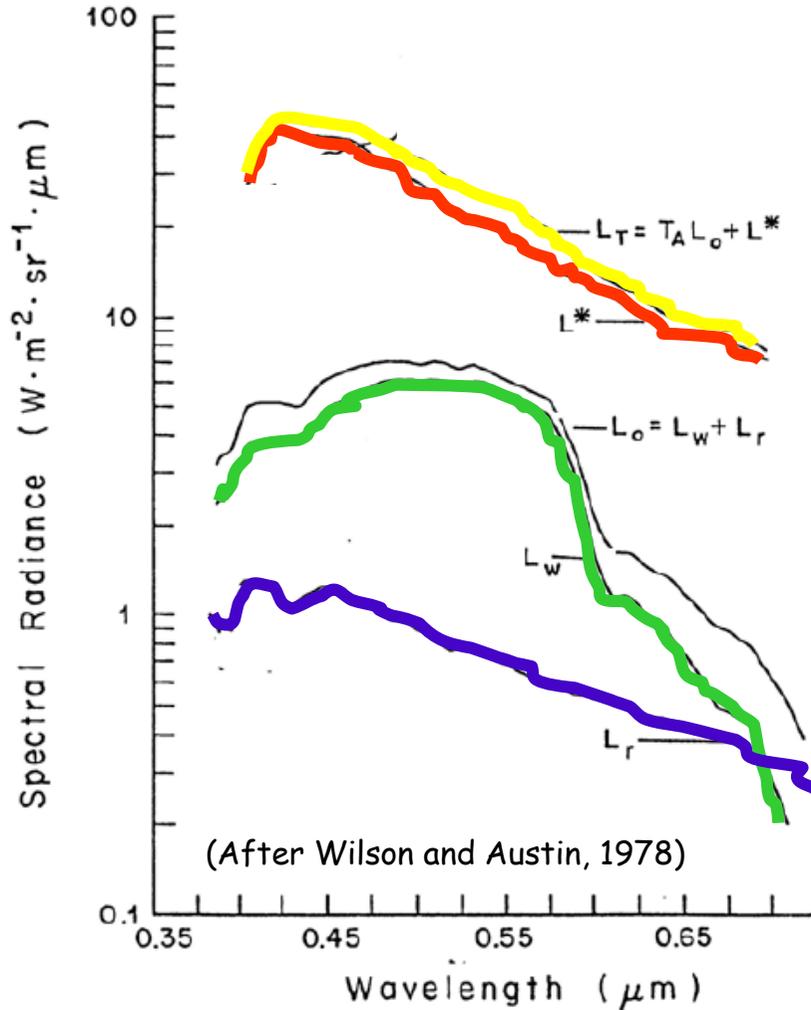
$L_w(\lambda)$



L_w is the desired quantity which carries information on the materials suspended and dissolved in the sea water (L_w is approximately 5-10% of L_{ToA})



The Water-Leaving Radiance



$$L_{ToA}(\lambda) = L_{atm}(\lambda) + t_d L_w(\lambda) + t L_g(\lambda)$$

L_{ToA} : Top of Atmosphere Radiance

L_{atm} : Atmospheric Radiance

L_g : Glint Radiance

L_w : Water Leaving Radiance

1% error in atmospheric correction or satellite calibration leads to a 10% error in L_w



Thus the need of high quality ocean color radiometric products is supported by a major effort in defining advanced atmospheric correction and calibration methods



Atmospheric correction

Atmospheric correction is the process of quantifying the atmosphere and sea-surface contributions to L_{ToA} aiming at determining the water-leaving radiance L_W .

Total radiance observed by the satellite is composed of 5-10% ocean signal (water-leaving radiance) and 90-95% atmosphere signal (atmospheric and glint radiances) mostly due to scattering by aerosols, and molecules (Rayleigh scattering).

The atmospheric contribution is computed using satellite measurements performed in the near-infrared (e.g., 780 and 870 nm). Since clear water highly absorbs in this spectral region, the radiance measured is almost entirely due to scattering by the atmosphere and surface reflectance. Thus **near-infrared satellite measurements, combined with model simulations of atmospheric scattering and absorption processes, are used to remove the contribution due to the atmosphere signal.**

Final product is the Normalized Water-Leaving radiance (L_{WN}) determined from L_W in the blue, green and red spectral regions.



Normalization of L_W

L_W varies with the bidirectional reflectance properties of seawater, which depend on illumination geometry, viewing geometry, seawater content, atmospheric content, sea surface.

The Normalized Water-Leaving radiance L_{WN} is introduced to make ocean color radiometric measurements independent of the measuring conditions: L_{WN} is the ideal radiance that would be measured if the Sun was at the zenith, in the absence of atmosphere, and with the Earth at its mean distance from the Sun.

$$L_{WN}(\lambda) = L_W(\lambda) \left(D^2 t_d(\lambda) \cos \theta_0 \right)^{-1} C_{f/Q}(\lambda, \theta_0, \tau_A, Chla)$$

where:
$$\left(D^2 t_d(\lambda) \cos \theta_0 \right)^{-1} = \frac{E_0(\lambda)}{E_d(0^+, \lambda)},$$

$$C_{f/Q}(\lambda, \theta_0, \tau_A, Chla) = \frac{f_0(\lambda, \tau_A, Chla)}{Q_0(\lambda, \tau_A, Chla)} \left(\frac{f(\lambda, \theta_0, \tau_A, Chla)}{Q_n(\lambda, \theta_0, \tau_A, Chla)} \right)^{-1}$$

Note: the remote sensing reflectance is given by $R_{rs}(\lambda) = L_{WN}(\lambda) / E_0(\lambda)$

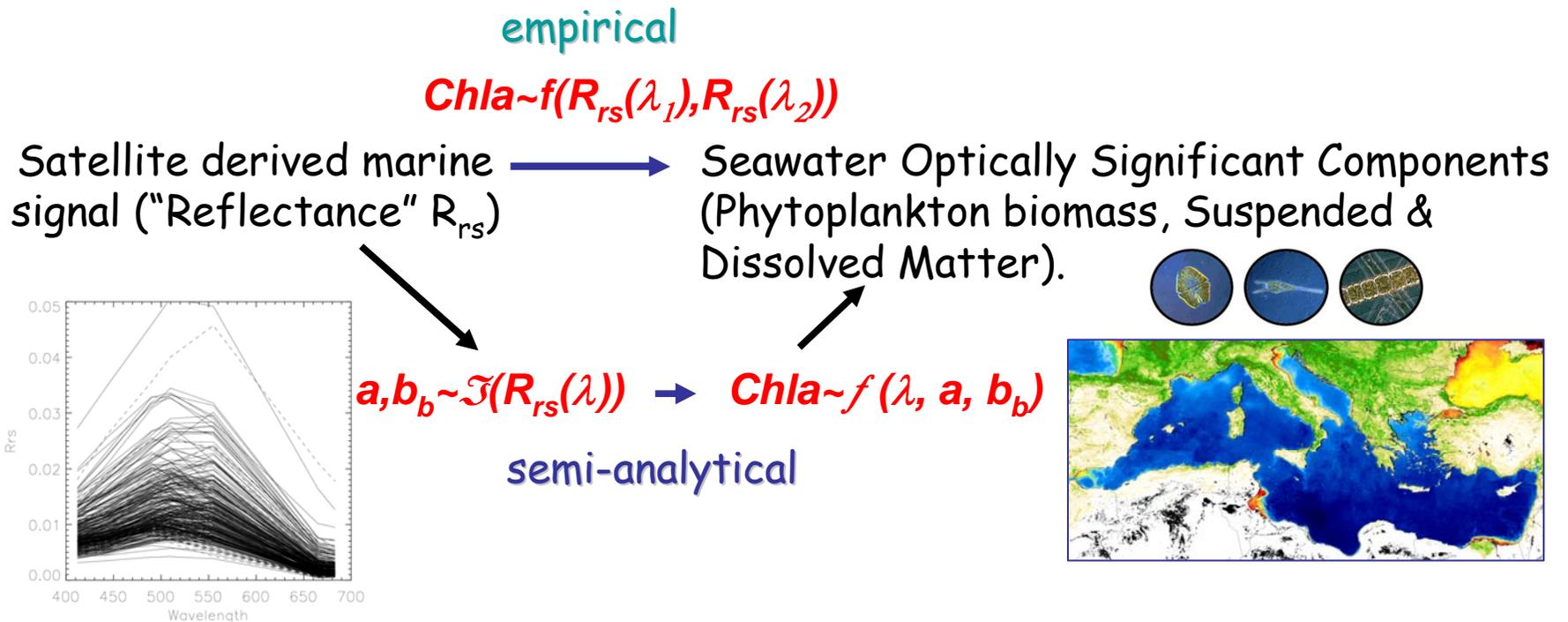


Bio-optical modeling



Bio-optical models:

the tools to convert satellite atmospherically corrected data into high-level products



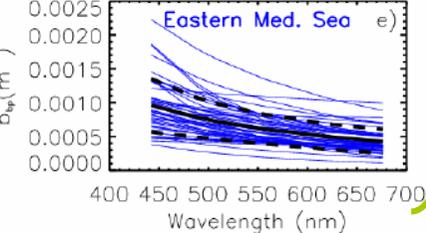
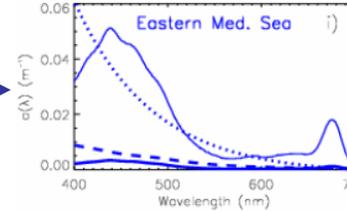
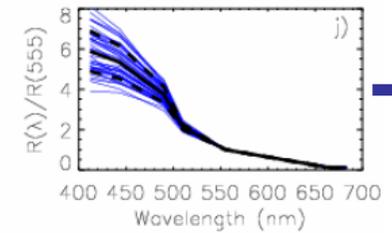
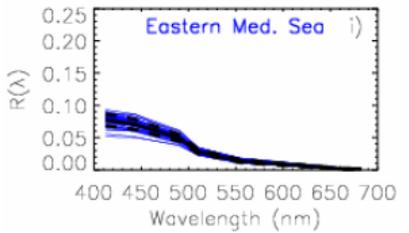
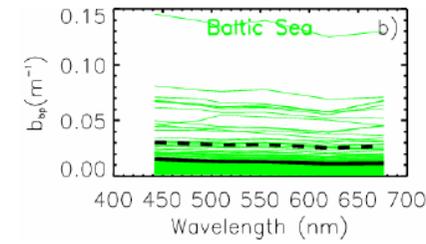
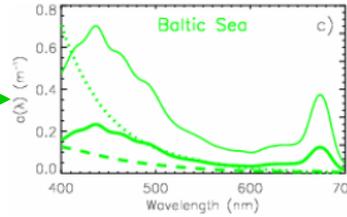
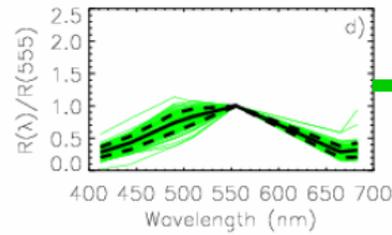
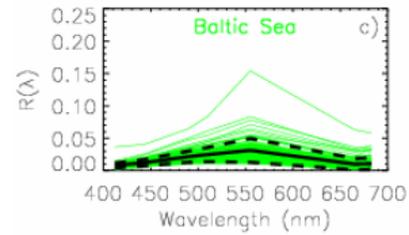
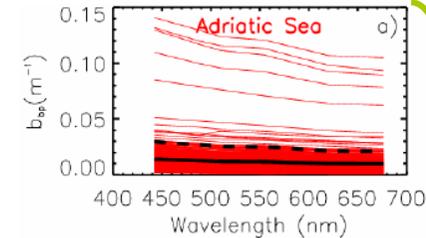
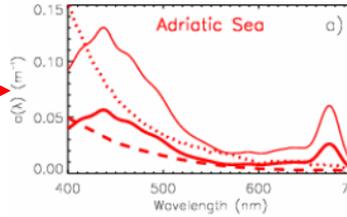
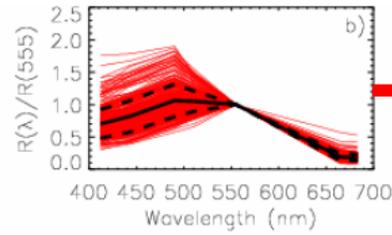
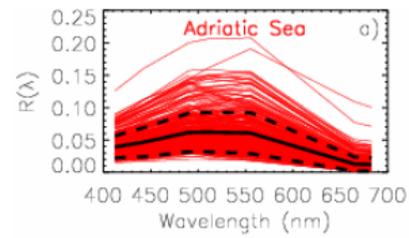


Semi-Analytical Models

Reflectance (R)

Absorption (a)

Back-scattering (b_b)



Chla

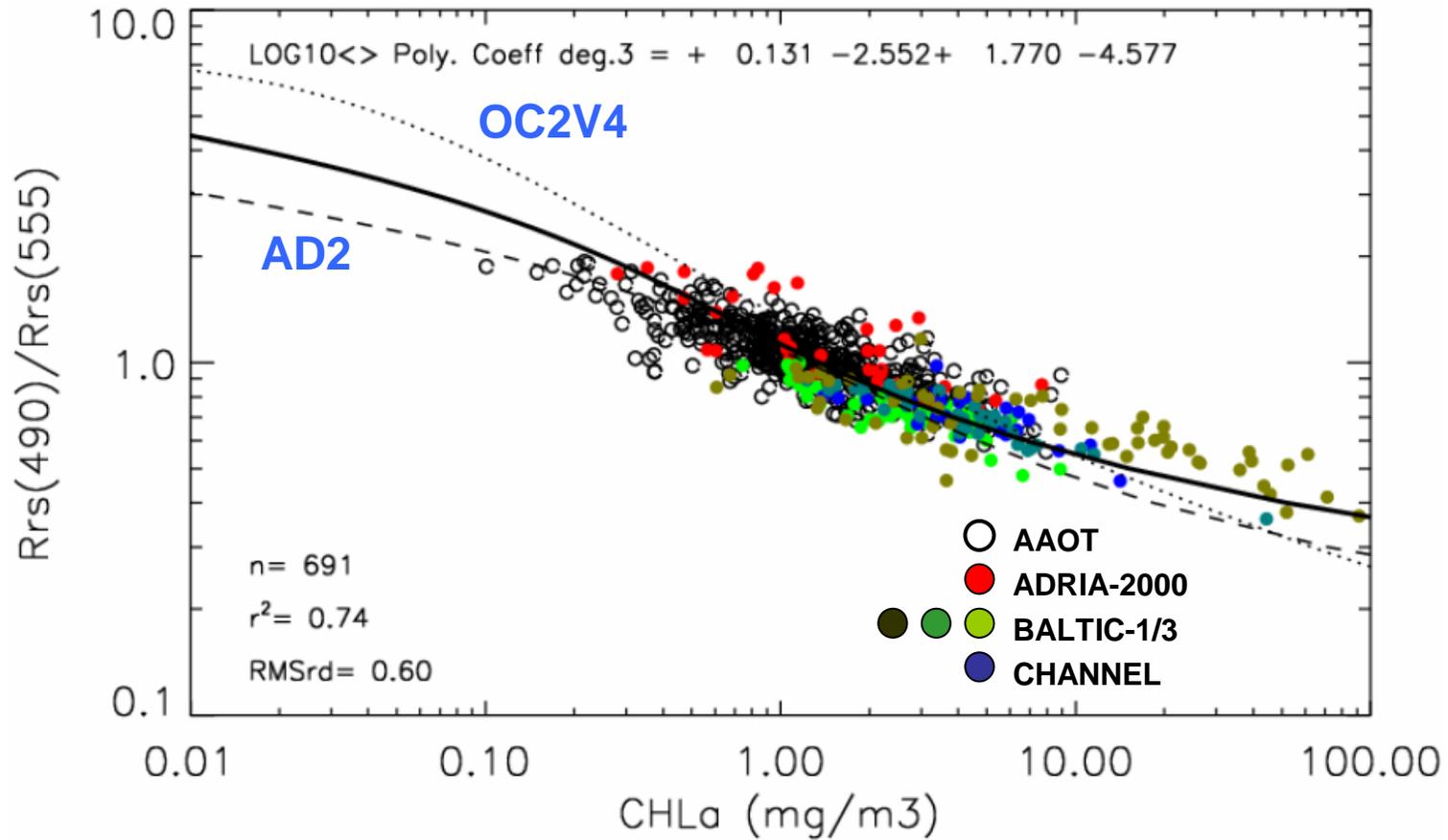
$$a, b_b \sim \mathfrak{F}(R_{rs})$$

$$Chla \sim f(a, b_b)$$

J.F. Berthon, F. Mélin and G. Zibordi (2007), Ocean Colour Remote Sensing of the Optically Complex European Seas, in: V. Barale and M. Gade ed.s, "Remote Sensing of the European Seas", Springer, Dordrecht (NL), pp. 35-52.



Empirical Models (Algorithms)



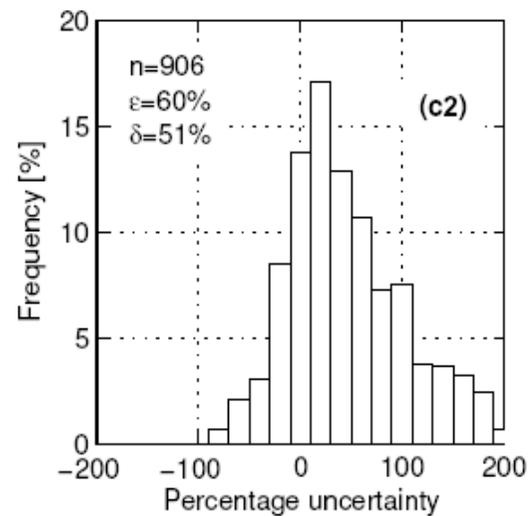
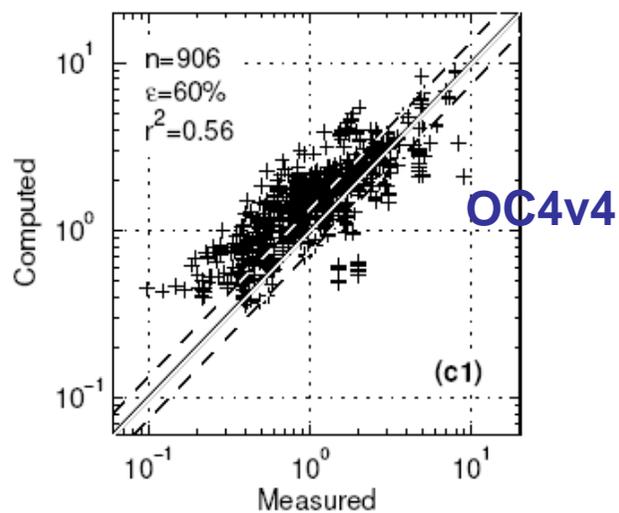
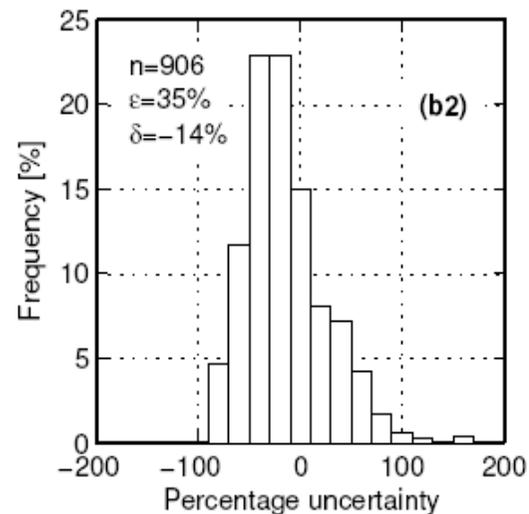
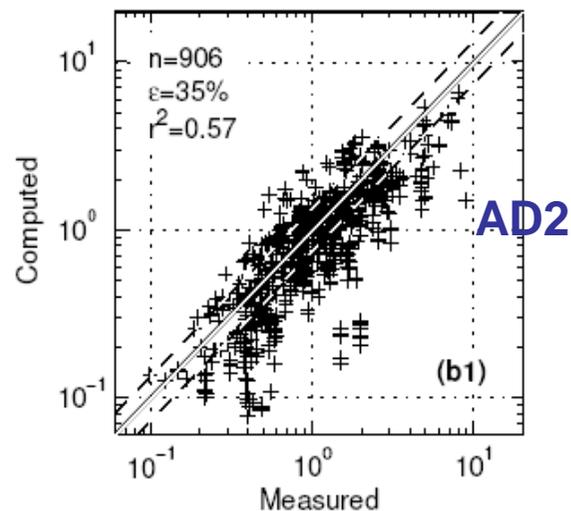
J.-F., Berthon, G. Zibordi, D. van der Linde, E. Canuti, E. Eker-Develi. Regional bio-optical relationships and algorithms for the Adriatic Sea, the Baltic Sea and the English Channel/North Sea suitable for ocean colour sensors. *EUR 22143EN*, 2006.



Comparison of Algorithms

AD2 $Chl-a = 10.0^{(0.091 - 1.620R_{2S} - 1.148R_{2S}^2 - 4.949R_{2S}^3)}$

OC4v4 $Chl-a = 10.0^{(0.366 - 3.067R_{4S} + 1.930R_{4S}^2 + 0.649R_{4S}^3 - 1.532R_{4S}^4)}$





Algorithms applicability

Novelty Detection provides a mean to identify those inputs (reflectance spectra) for which the performance of an algorithm is likely to be unreliable

The algorithms perform correctly when the satellite data are represented in the training data sets



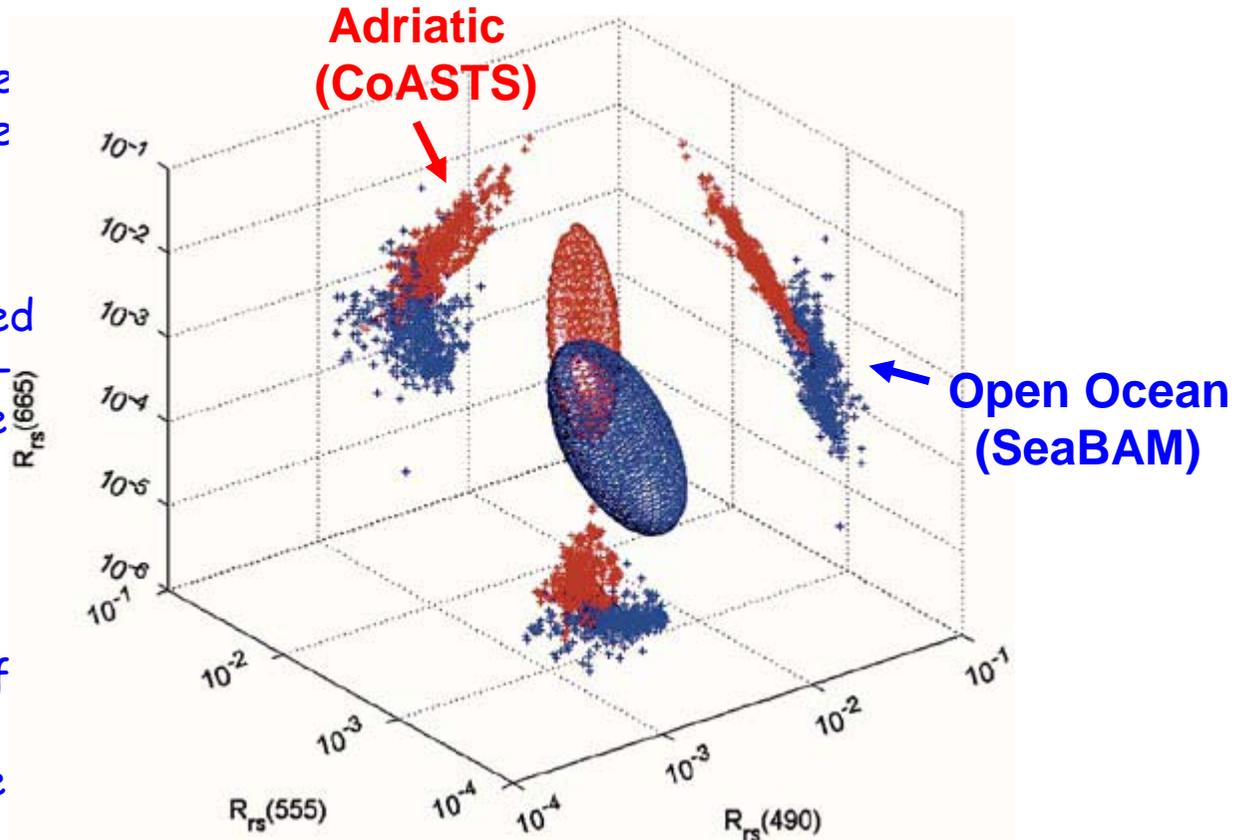
The training data sets applied for developing the bio-optical algorithms, are used to determine

their *probability density function*

which defines the range of applicability of algorithms to satellite data



hyper-ellipsoid
(example for a 3D space)



D.D'Alimonte, F.Mélin, G.Zibordi and J.F.Berthon. Use of the novelty detection technique to identify the range of applicability of empirical ocean color algorithms. *IEEE Transactions in Geoscience and Remote Sensing*, 41:2833-2843, 2003.

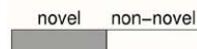
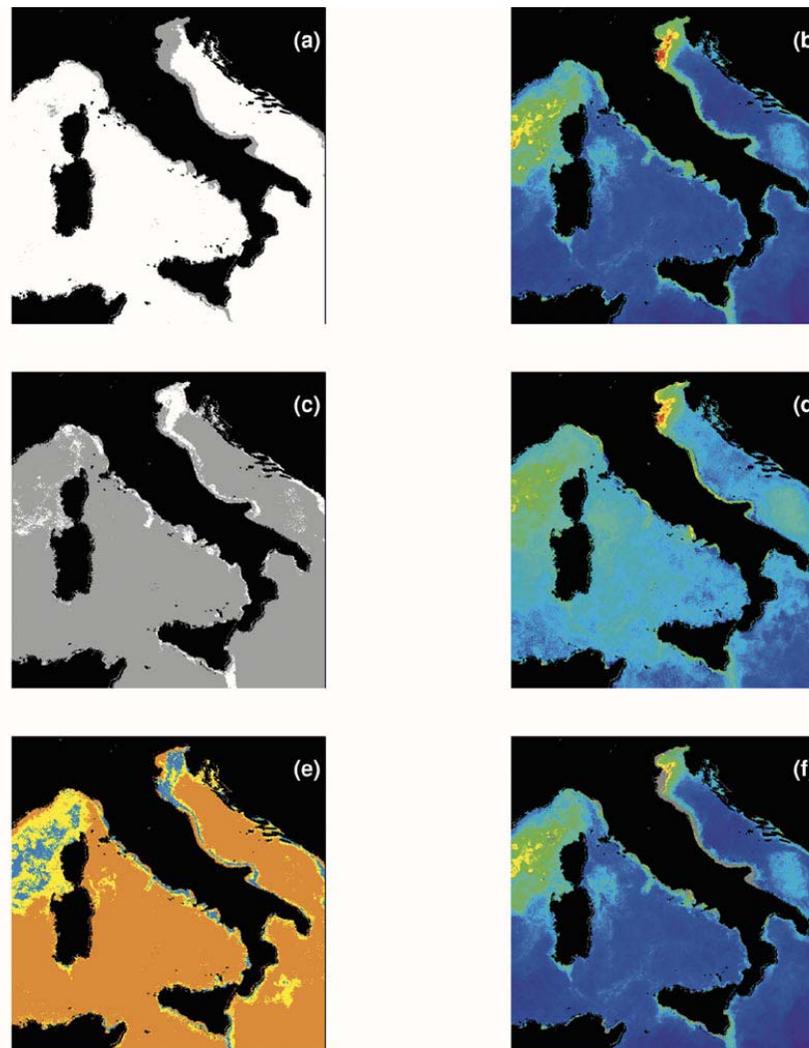
Dynamic Algorithms Application

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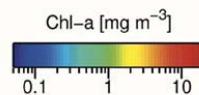
Open Ocean
(SeaBAM)

Adriatic
(CoASTS)

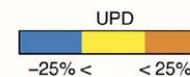
Mix Adriatic /
Open Ocean



For panels a) and c)



For panels b), d) and f)



For panel e)

↓
Identification of

“novel”

and

“non novel”

data

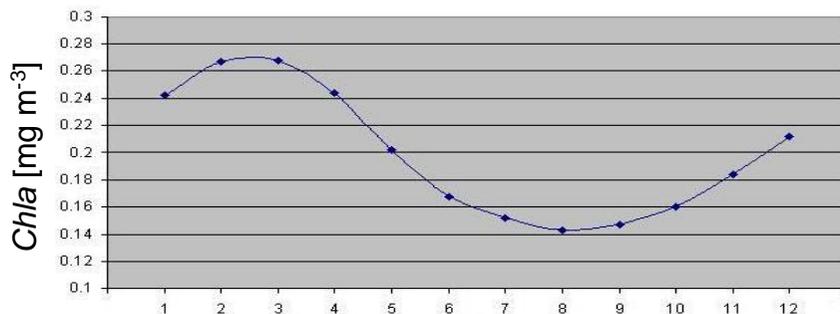
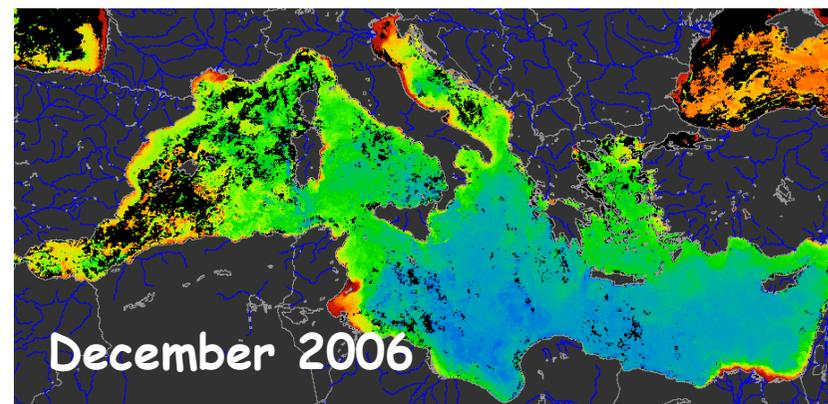
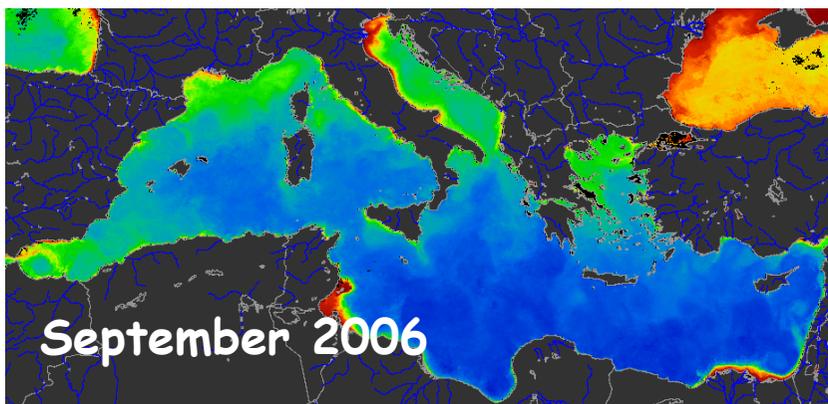
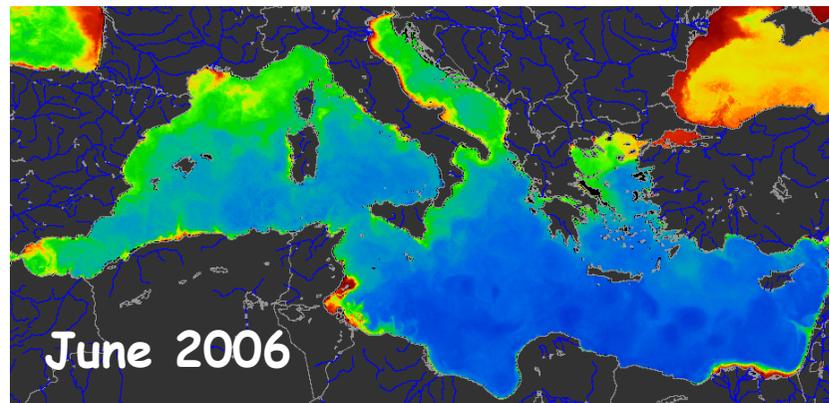
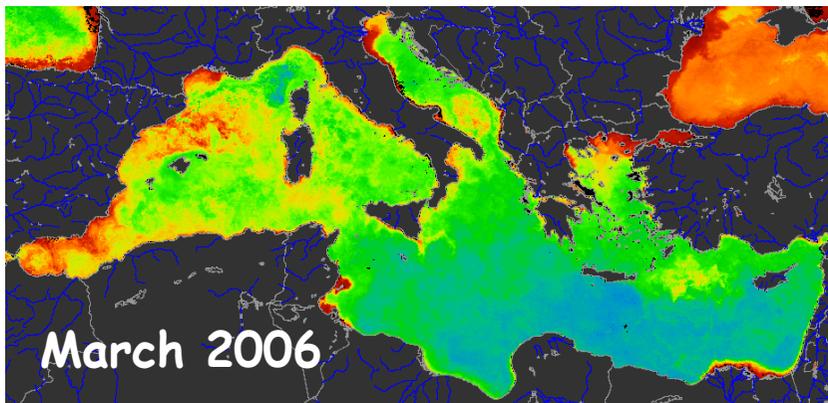
↓
Mixing
algorithms



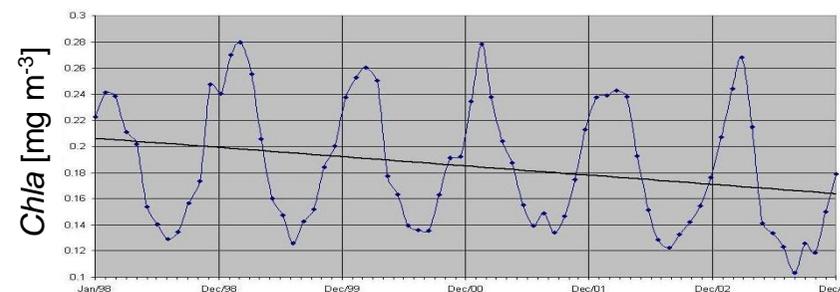
Products and Applications



Mediterranean Sea: Trends



Chla Average Basin Value (98-03)
from 12 Climatological Monthly Means



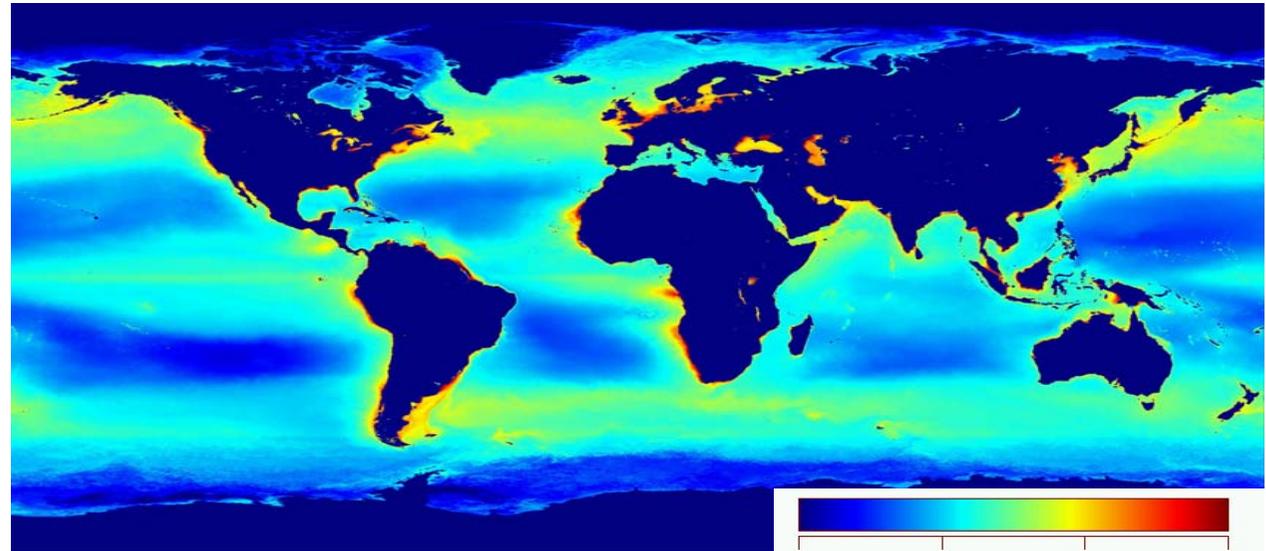
Chla Average Basin Value (98-03)
from 72 Monthly Means



SeaWiFS-derived 5-year average (1998-2002)

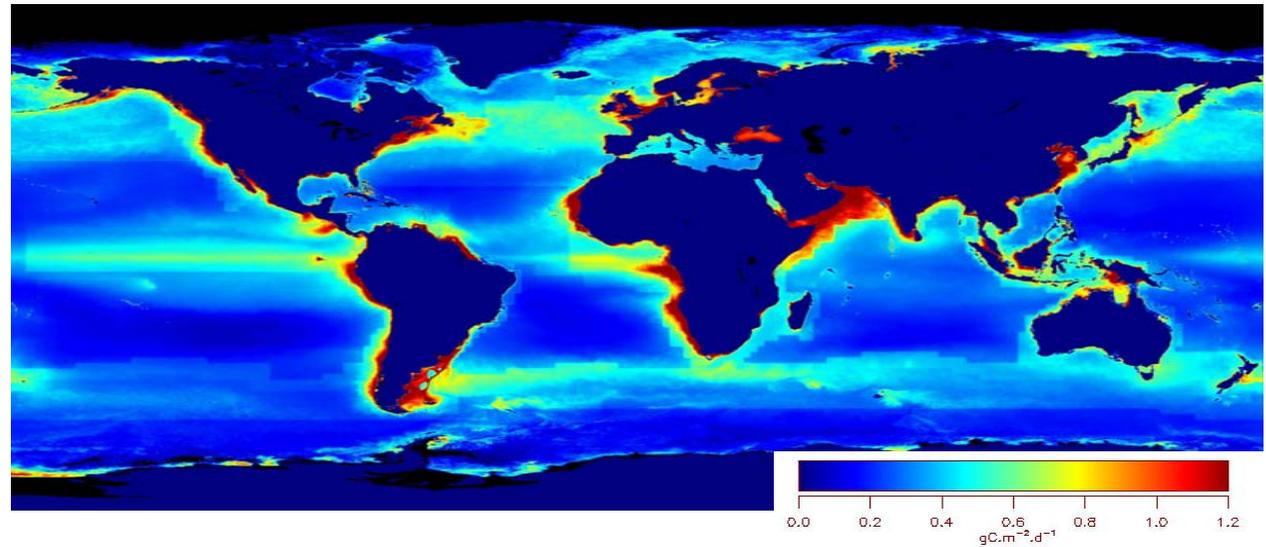
Phytoplankton
standing stock

Chl a



Primary production
(i.e., the production of
organic compounds
from atmospheric and
aquatic carbon dioxide)

48 GtC·yr⁻¹



F.Mélin and N.Hoepffner, Global Marine Primary Production: A Satellite View. *EUR 21084EN*, 2004



Concluding remark

Marine ecosystems are particularly sensitive to climate change.

Future alterations in marine ecosystems will depend as much on global climate change as on our ability to regulate exploitation pressure at sustainable levels. Such a regulation will require high ecological certainty.

Within such a framework ocean color remote sensing is an invaluable technology to support ecosystem studies through synoptic quantification of key environmental parameters.

Any Question ?