



Hydrological optics and hyperspectral remote sensing of lakes

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My research interests

Environmental remote sensing

- Bio-optical and hyperspectral imaging of inland waters
- Hyperspectral vegetation classification and biodiversity assessment in peatlands and wetlands
- Remote sensing of vegetation biophysical parameters; carbon cycling in terrestrial ecosystems
- Detection and mapping of plant stress; climate-induced drought stress
- LiDAR remote sensing in fluvial geomorphology

Other

- Phytoplankton ecology and toxicology
- Water resource socio-economics

see also: www.sbes.stir.ac.uk/people/hunter

Outline

- Why lakes and why remote sensing?
- Basic theory of underwater optics
- Measuring and modelling the underwater light environment
- Remote sensing of in-water constituents and water column processes
- NERC ARSF / EUFAR project in Lake Balaton

Why lakes?

Key component of global biosphere; ecosystem services (biodiversity, biogeochemical cycling, food)

Sinks for terrestrially-derived materials and pollutants (sediments, nutrients, pesticides, toxins)

Sensitive to meteorological forcing (e.g., Livingstone & Padisák, 2007); sentinels of climate change?

Statutory monitoring requirements under EU directives

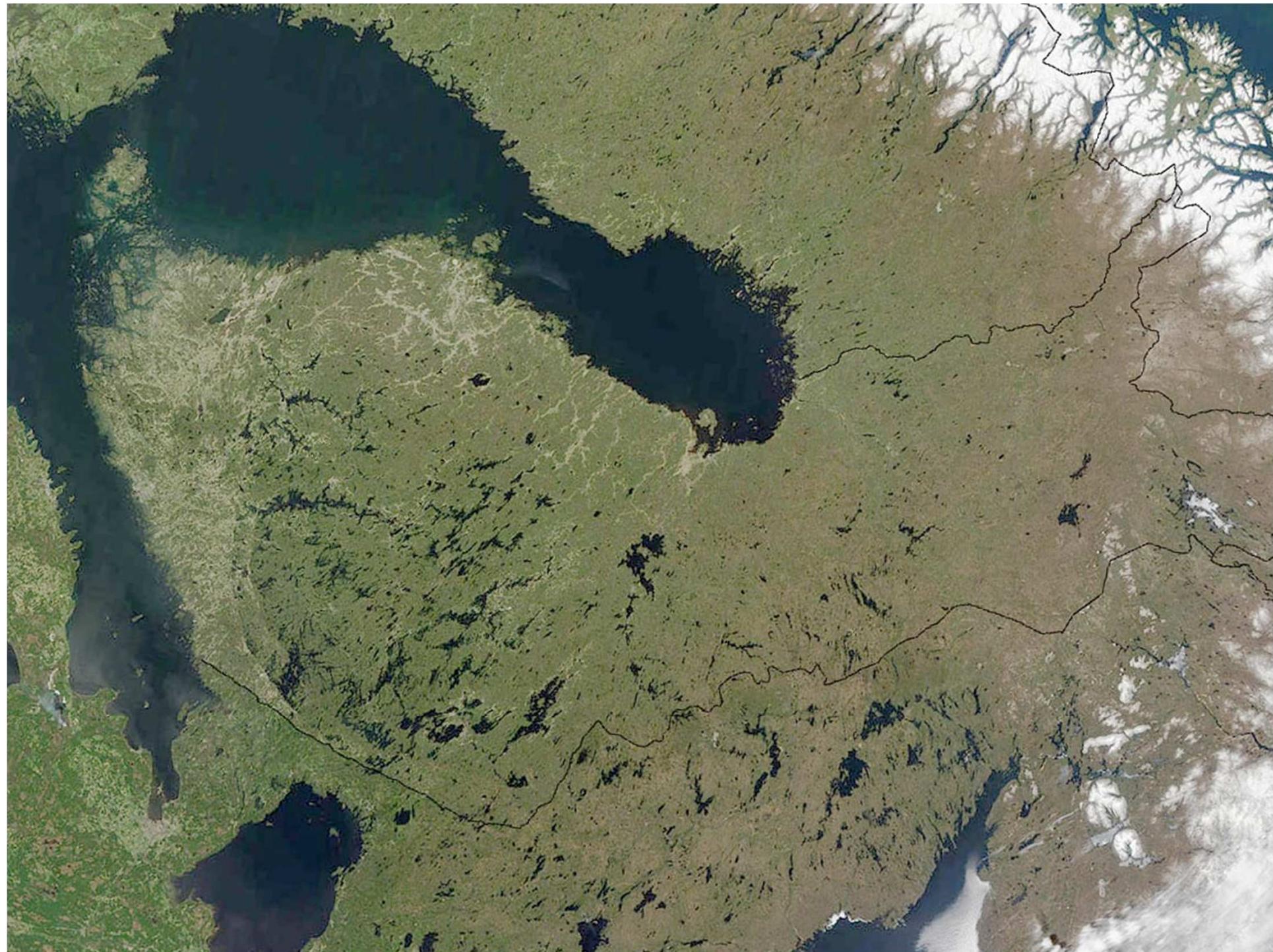
Habitats Directive (92/43/EEC)

Water Framework Directive (2000/60/EC)

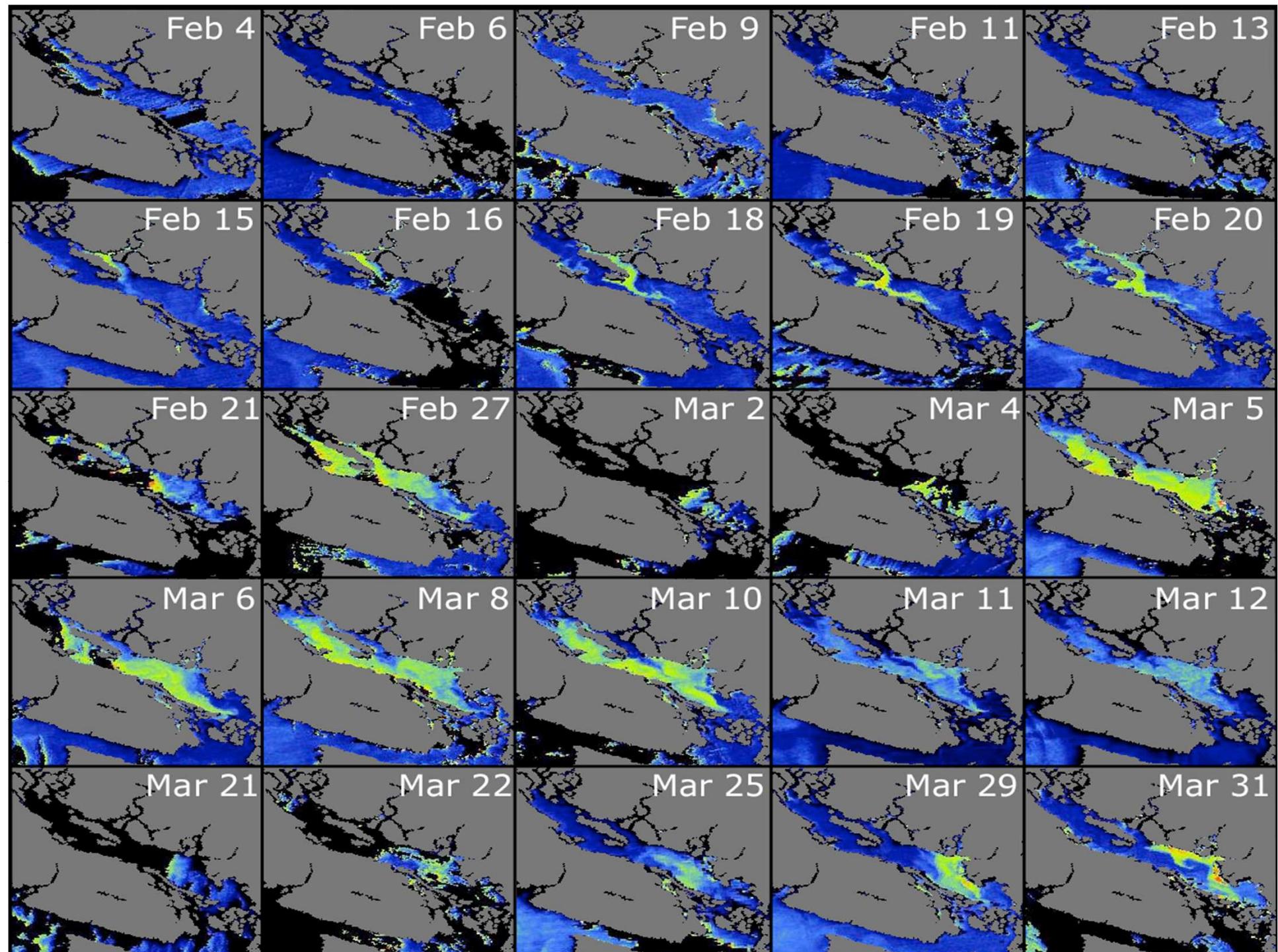
Why remote sensing?

- numbers, densities and size distribution of lakes
 - e.g., Europe: $1.5 \times 10^6 > 1 \text{ ha}$
 - $500\,000 > 10 \text{ ha}$
- marked space-time variability;
implications for
representativeness of
observations
- costs, lack of resources,
inaccessibility









The science in a nutshell...

To develop algorithms (models) to relate the optical properties (colour) of water to its biogeochemical composition...

...and use this information to study changes in lake condition and function at high temporal resolution and at global scales

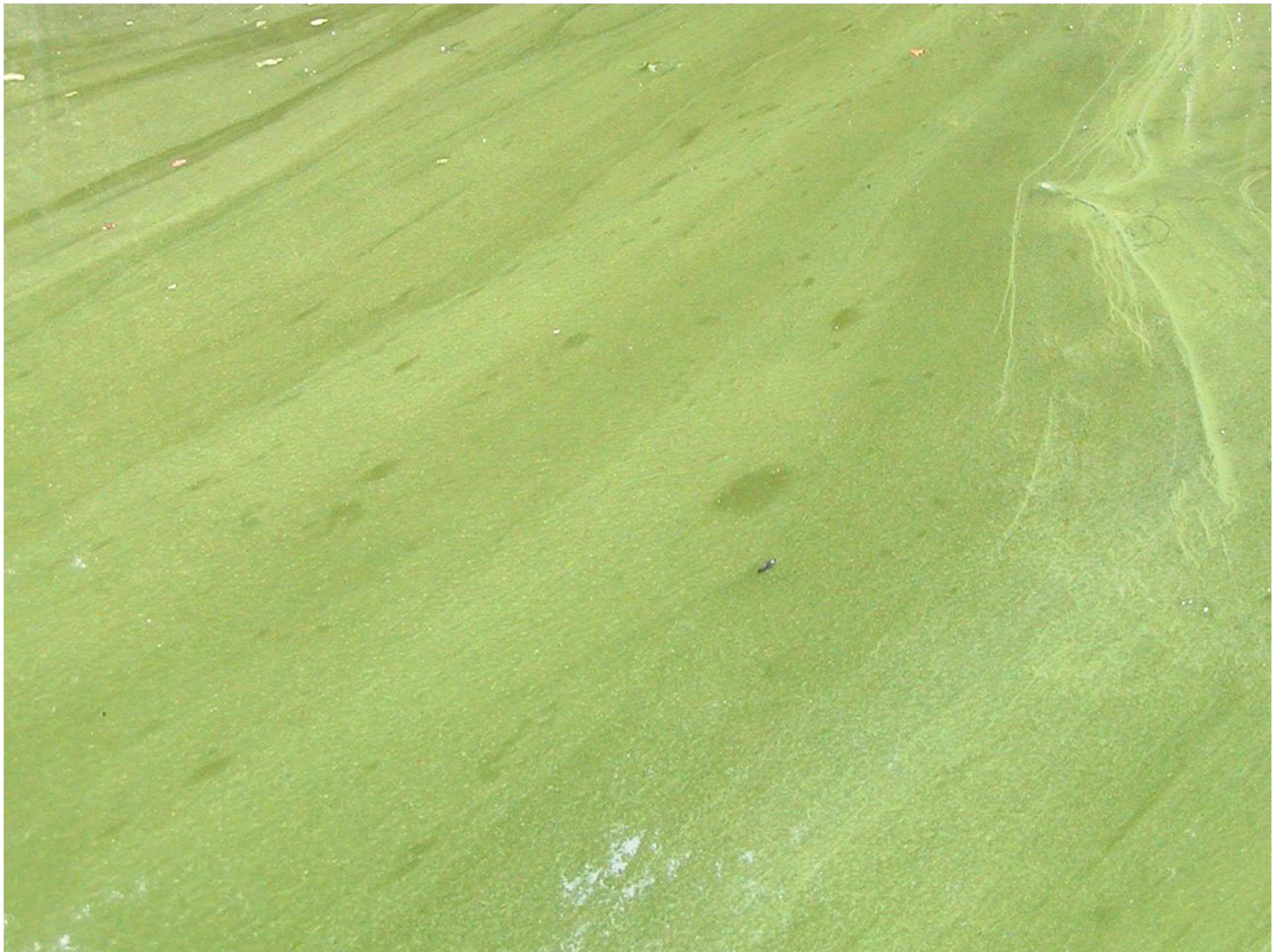
Basic theory: underwater optics



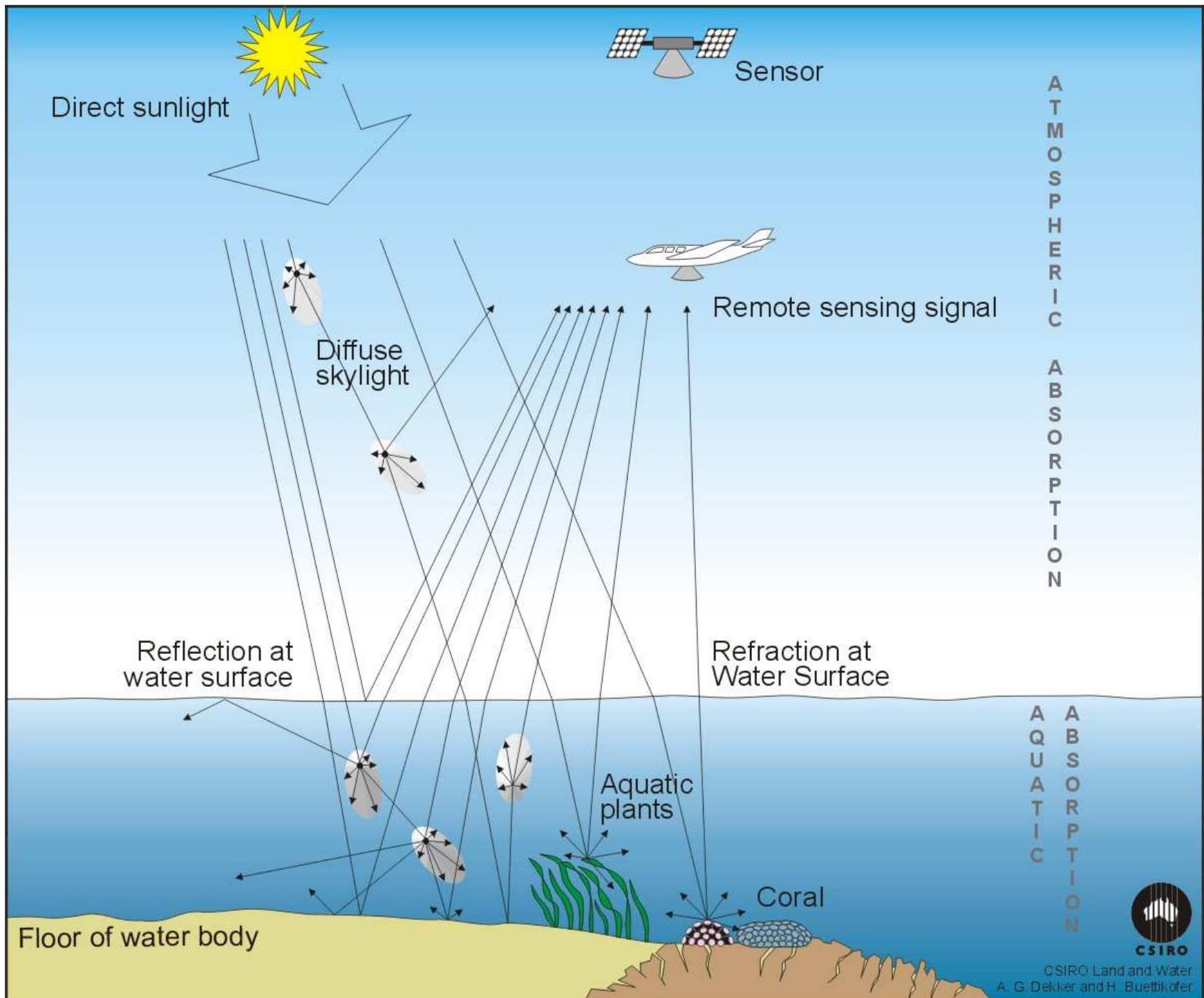




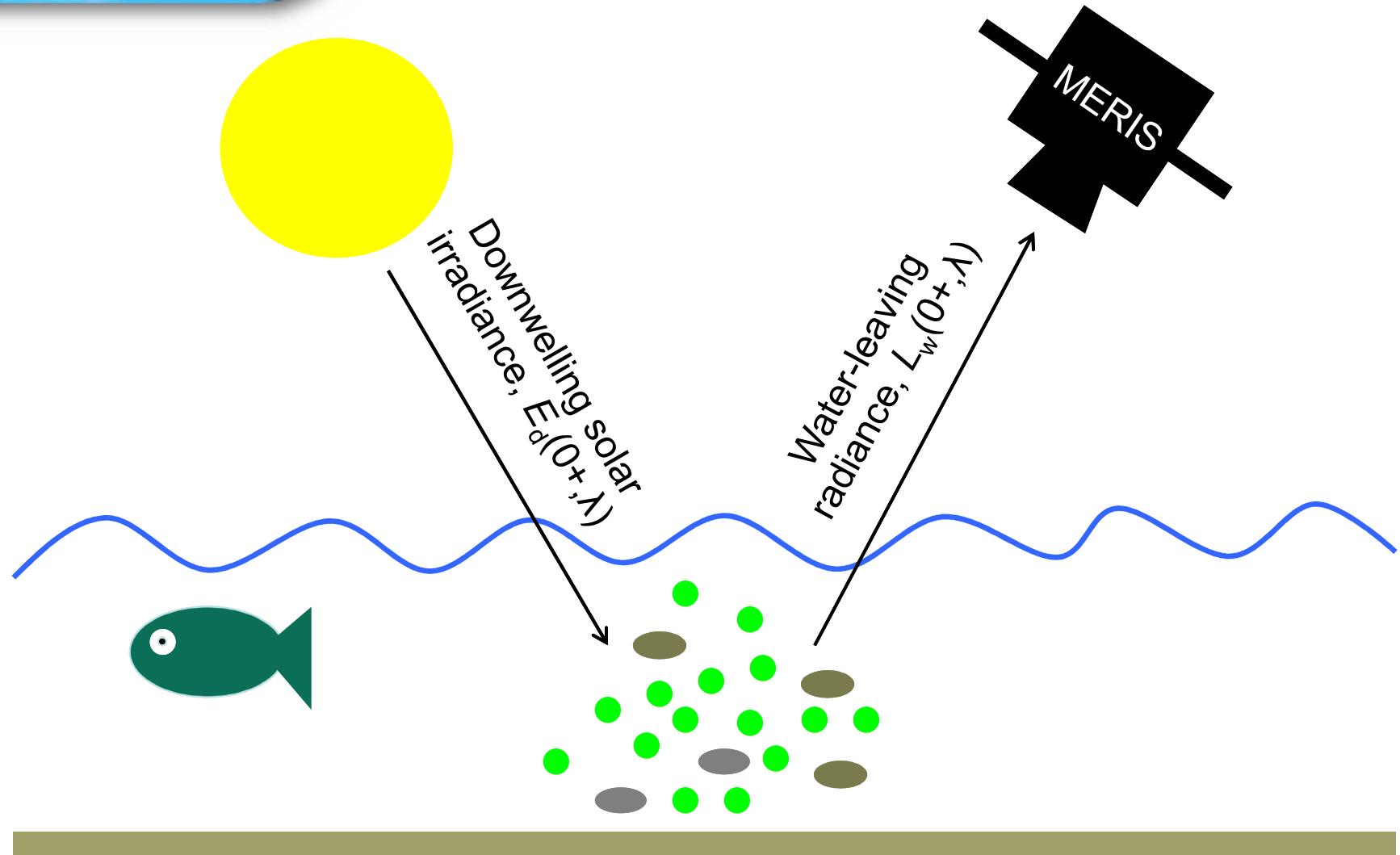








Optical properties of waters



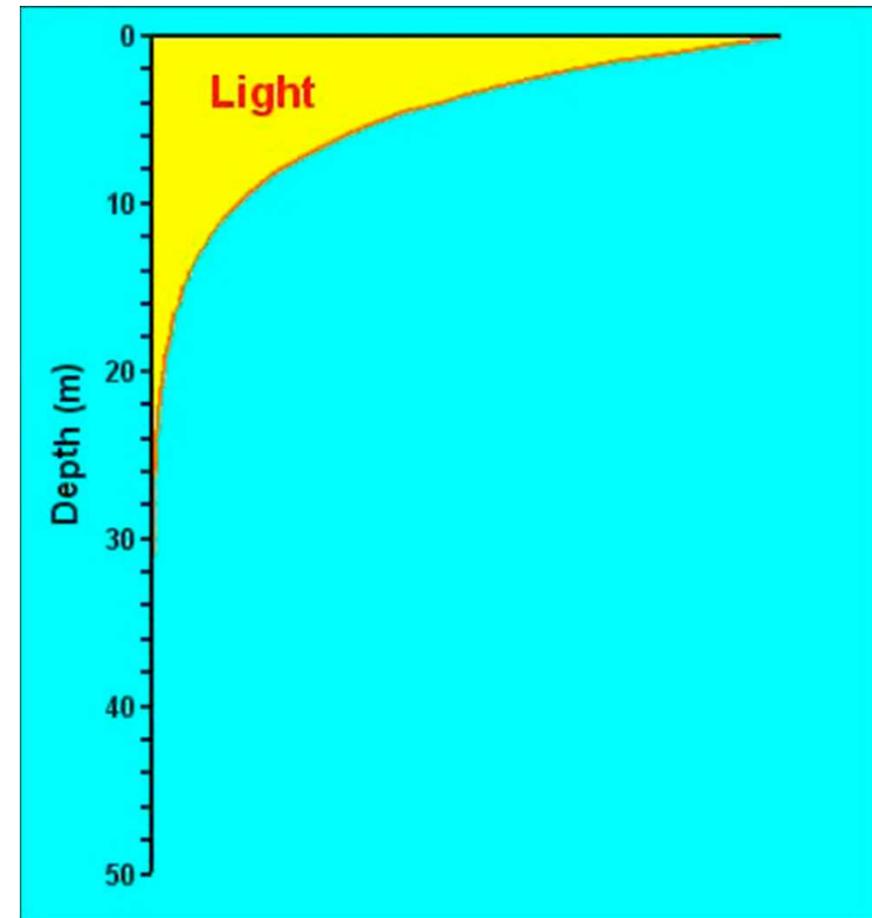
Light penetration

Coefficient of attenuation (K_d , m $^{-1}$)

Rule of thumb: 90% of signal
comes from first optical depth

[Optical depth = $1/K_d$ (m)]

Optical depths in Lake Balaton
(June 2010) = 0.16-5.95 m



Optical properties of waters

(Specific) Inherent optical properties (IOPs/SIOPs)

- absorption (a , m^{-1})
- scattering (b , m^{-1})
- backscattering (b_b , m^{-1})
- beam attenuation (c , m^{-1}) ($c = a + b$)
- volume scattering function [$\beta(0)$]

Describe the properties of the medium itself; invariant of the ambient light field

Apparent optical properties (AOPs)

- subsurface irradiance reflectance [$R(0-)$]^{*}
- remote sensing reflectance [$R_{rs}(0+)$]
- coefficient of light attenuation (K_d)
- transparency (e.g., Secchi disc depth)
- colour

Radiometric quantities that are dependent on the ambient light field and viewing geometry

*considered a quasi-IOP

Optically-active constituents

Absorption (a , m⁻¹) - removal of light by:

- water itself
- coloured dissolved organic matter
- phytoplankton
- mineral particles and detritus

Scattering (b , m⁻¹) - molecular and particulate; directional:

- water itself
- phytoplankton
- suspended sediments

Beam attenuation (c , m⁻¹) – ($c = a + b$)

Volume scattering function (β) – directionality of light scattering

Optical properties of waters

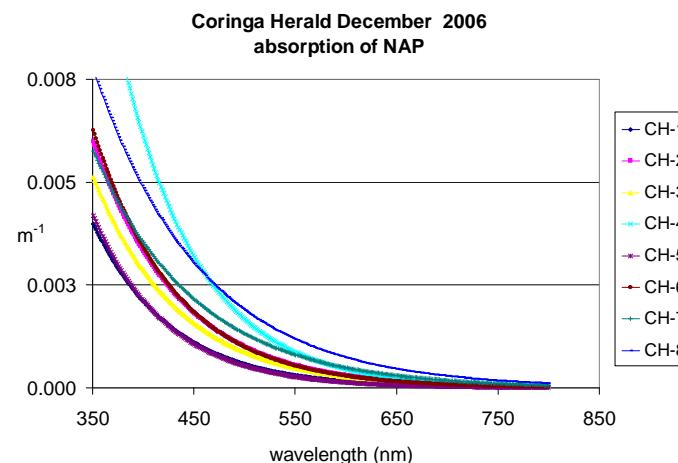
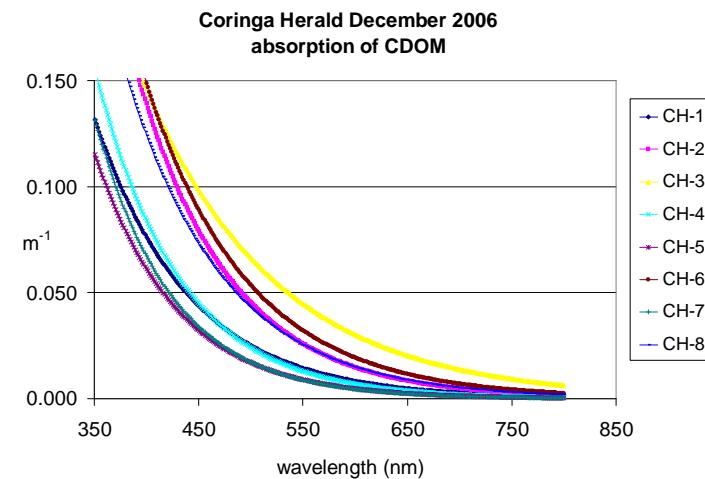
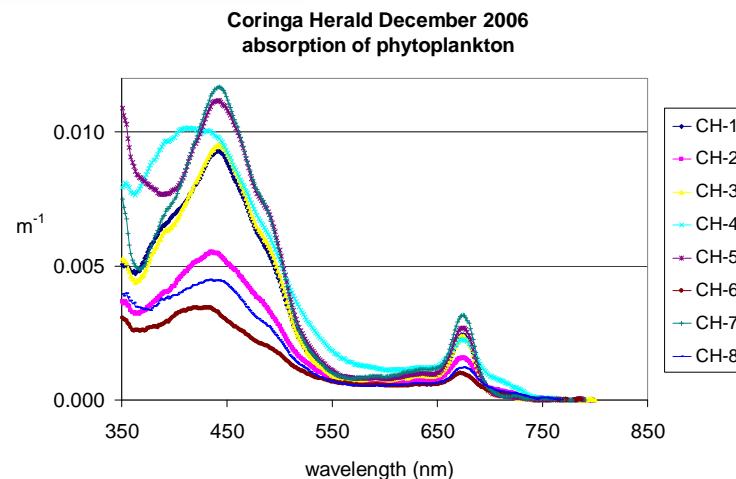
absorption peak for particles, detritus, and dissolved substances absorption peak for photosynthetic pigment (low- medium concentrations) accessory photosynthetic pigments absorption and fluorescence peak for chlorophyll



coloured dissolved organic matter,
mineral particles
sediment

atmospheric correction

Optical properties of waters



Classification of water types

Case I (open oceans)

3 optically-active components (OACs):
(1) pure water; (2) phytoplankton;
(3) coloured dissolved organic matter.



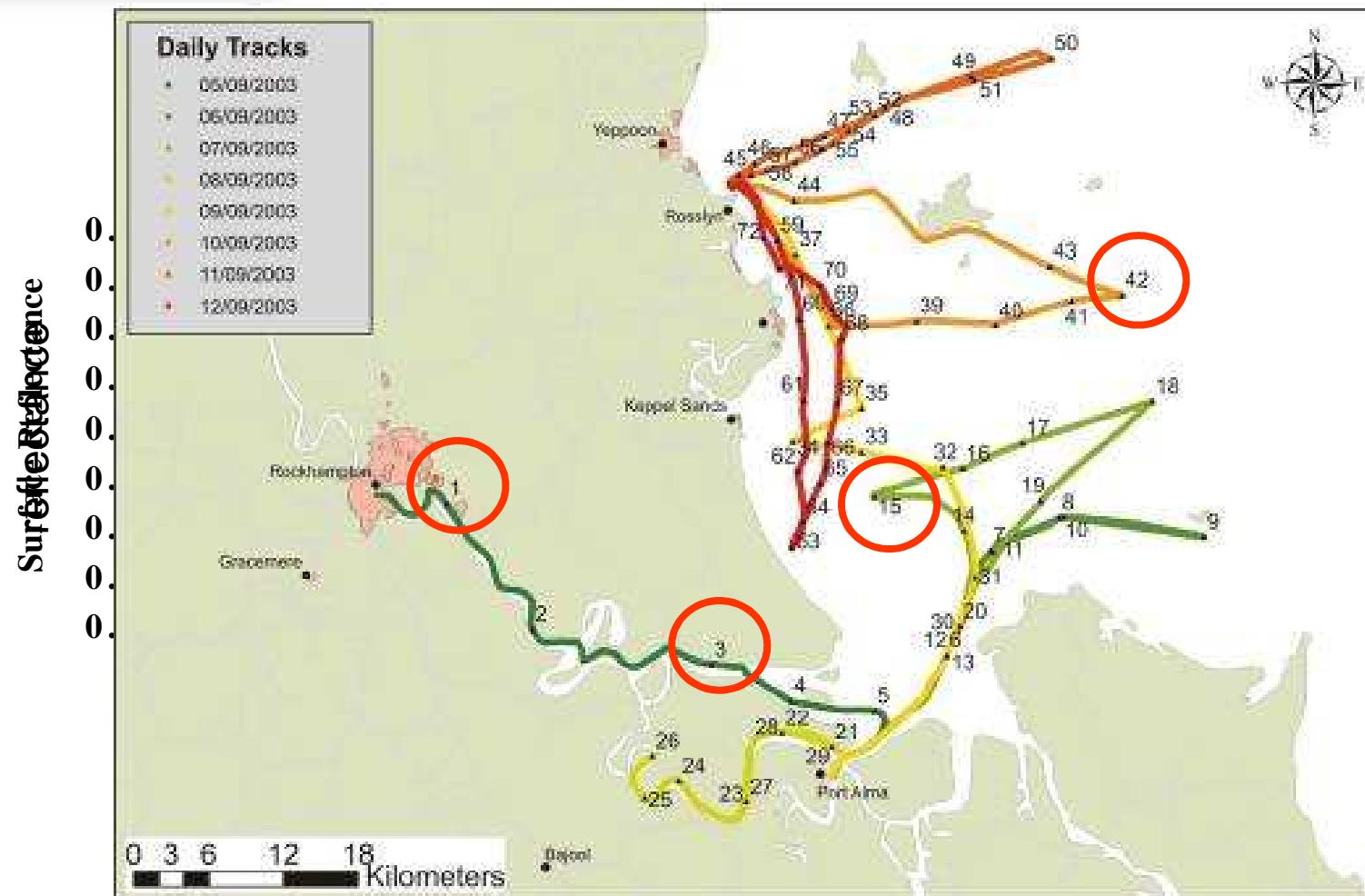
Case II (coastal and inland waters)

4 *non-covarying* OACs:
(1) pure water; (2) phytoplankton;
(3) non-algal particles;
(4) coloured dissolved organic matter.





Variability in water colour



Optical models

Total absorption and backscattering

$$a_t(\lambda) = a_w(\lambda) + a_{ph}(\lambda) + a_{NAP}(\lambda) + a_{CDOM}(\lambda)$$

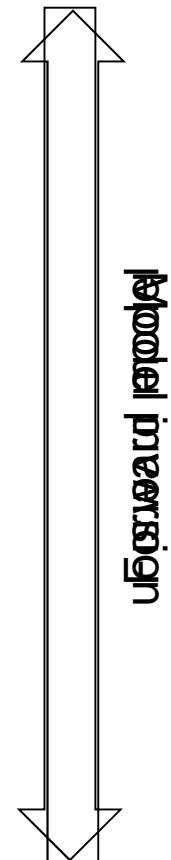
$$b_{b,t}(\lambda) = b_{b,w}(\lambda) + b_{b,ph}(\lambda) + b_{b,NAP}(\lambda)$$

Subsurface irradiance reflectance (Gordon et al. 1975)

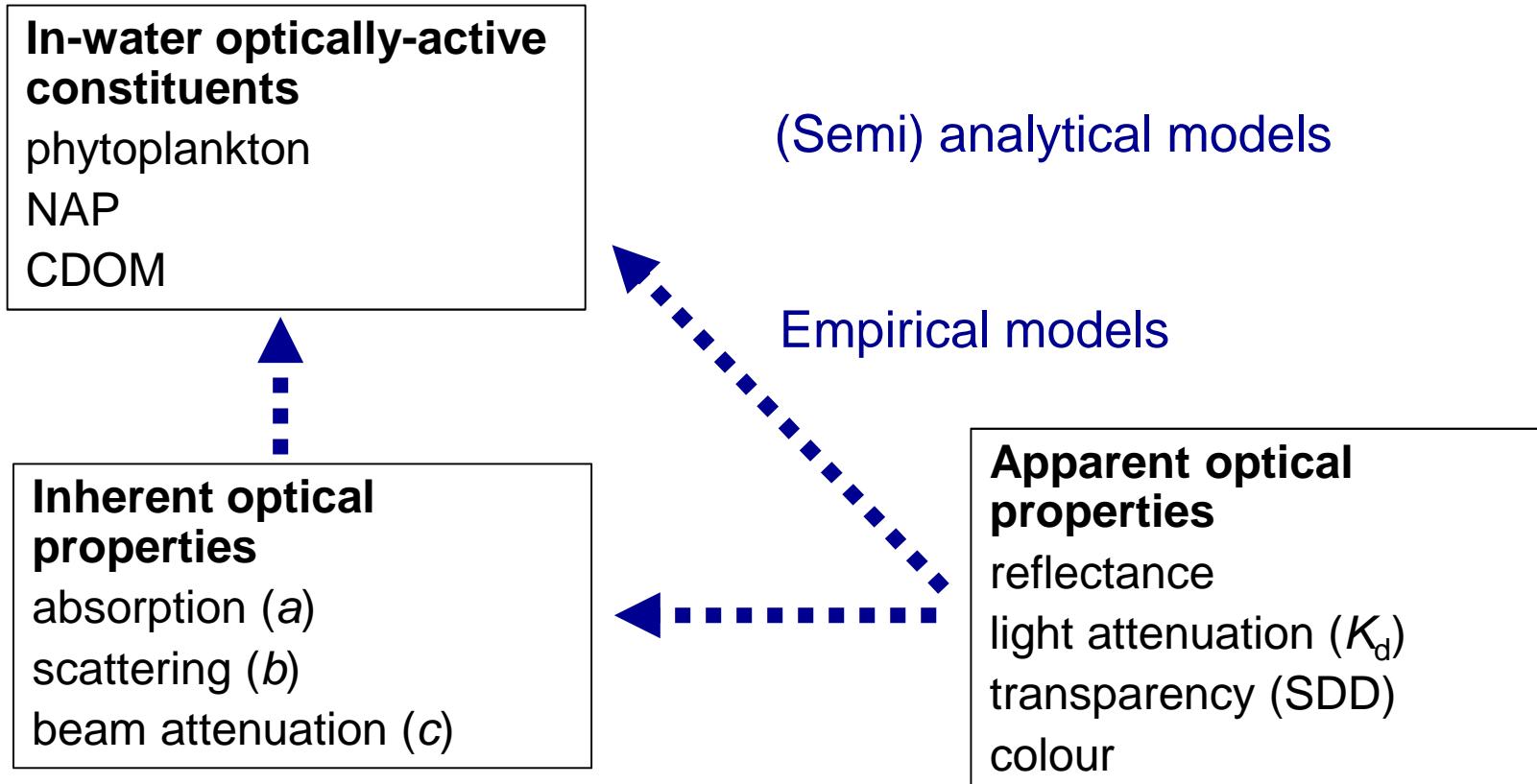
$$R(0-) = E_u/E_d = f * (b_b/a + b_b) \quad \text{where } f \sim 0.33$$

Remote sensing-reflectance

$$R_{rs}(0+) = L_w/E_d = (T * f/Q) * (b_b/a + b_b) \quad \text{where } T \sim 0.54; f/Q \sim 0.0949$$



Models for constituent retrieval

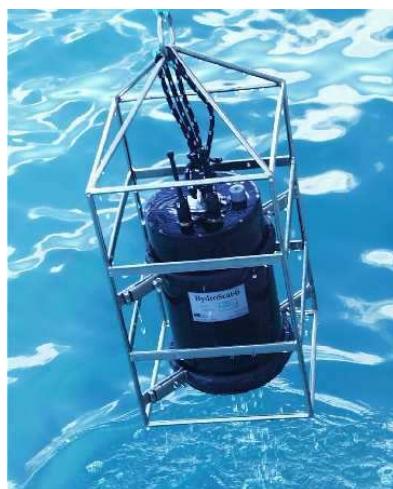


In- and above-water optical instruments and measurements

IOPs: active sensors

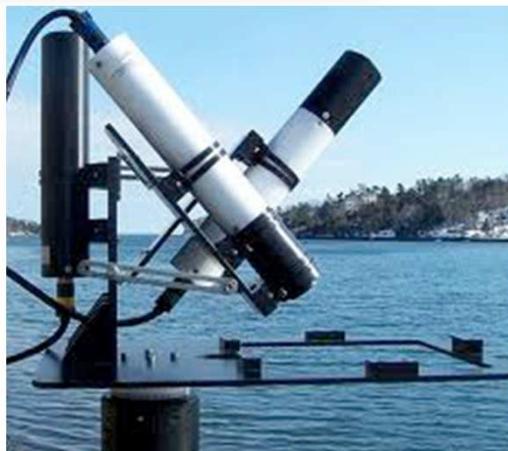


Wetlabs AC-S: hyperspectral absorption (a)
and attenuation (c) [$b = c - a$]



Wetlabs BB3: spectral backscattering (b_b)

AOPs: passive sensors



Satlantic HyperSAS: above surface water-leaving radiance (L_w) or reflectance (R_{rs})



Satlantic HyperOCRs: subsurface downwelling irradiance and upwelling radiance (subsurface irradiance reflectance)

IOP and AOP rigs





Remote sensing platforms

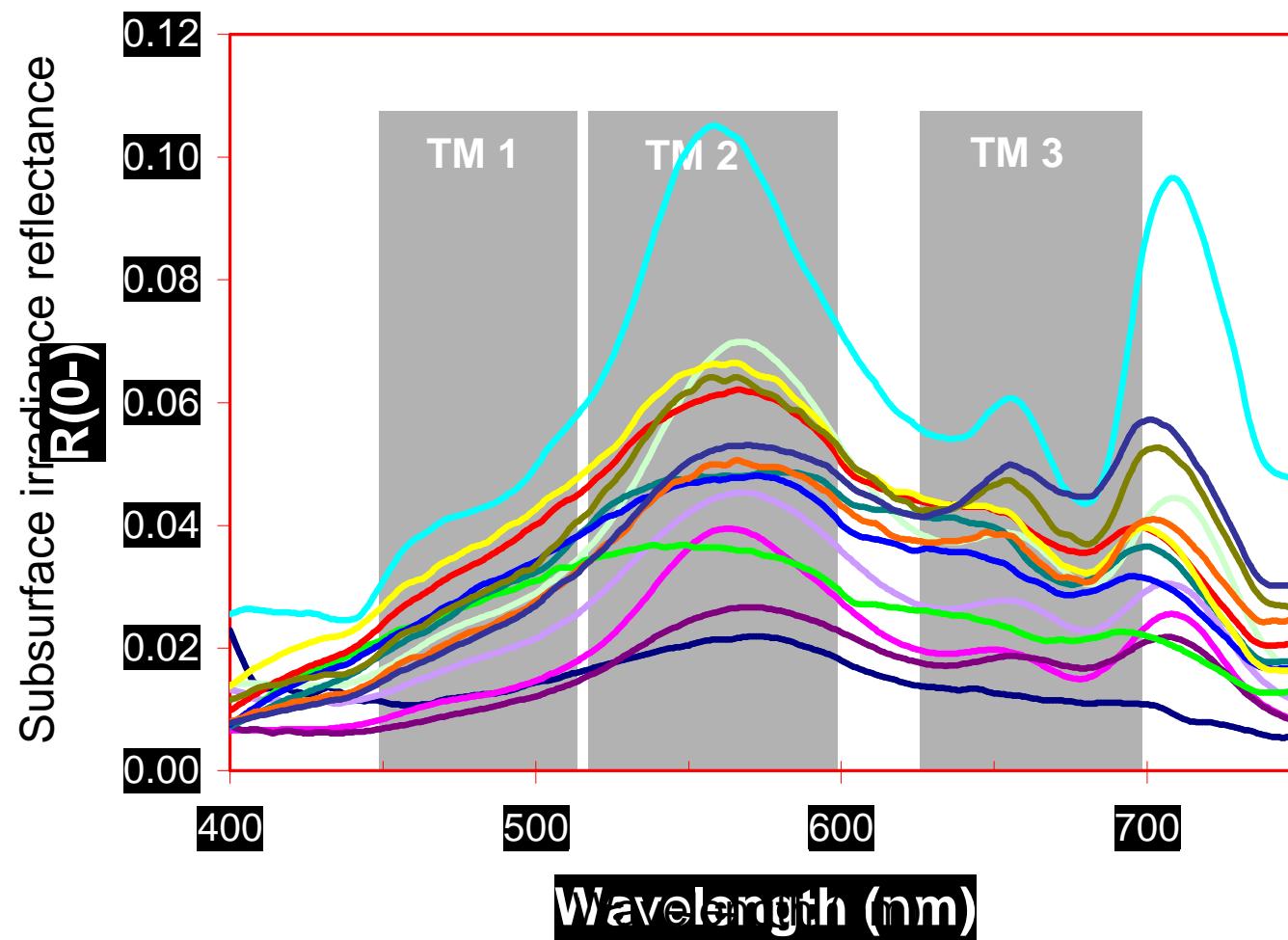


AISA Eagle-Hawk on NERC ARSF Dornier



MERIS on ESA Envisat

Why hyperspectral?



Remote sensing applications

Parameters & processes of interest

Water transparency; light availability (SDD, K_d)
Total (inorganic / organic) suspended solids (TSS)
concentration
size spectra
Coloured dissolved organic matter (CDOM)
Phytoplankton
chlorophyll-a
phycocyanin & other accessory pigments
size spectra & functional groups
harmful algal blooms
primary productivity
Bottom conditions (if observable)
depth
submerged habitat

Chlorophyll: Case I waters

SeaWiFS OC4 algorithm

$$R_{\text{MAX}}$$

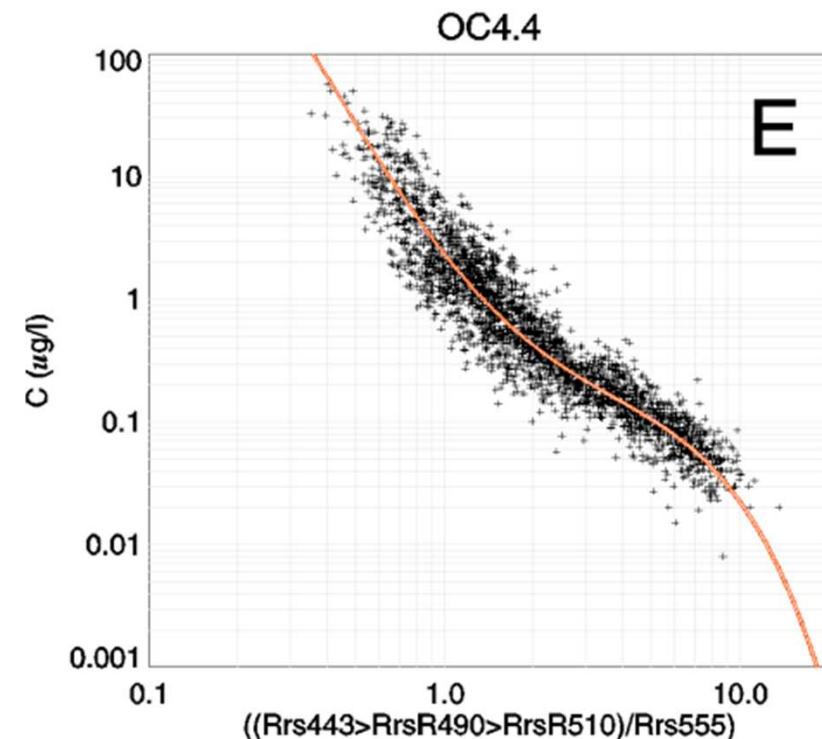
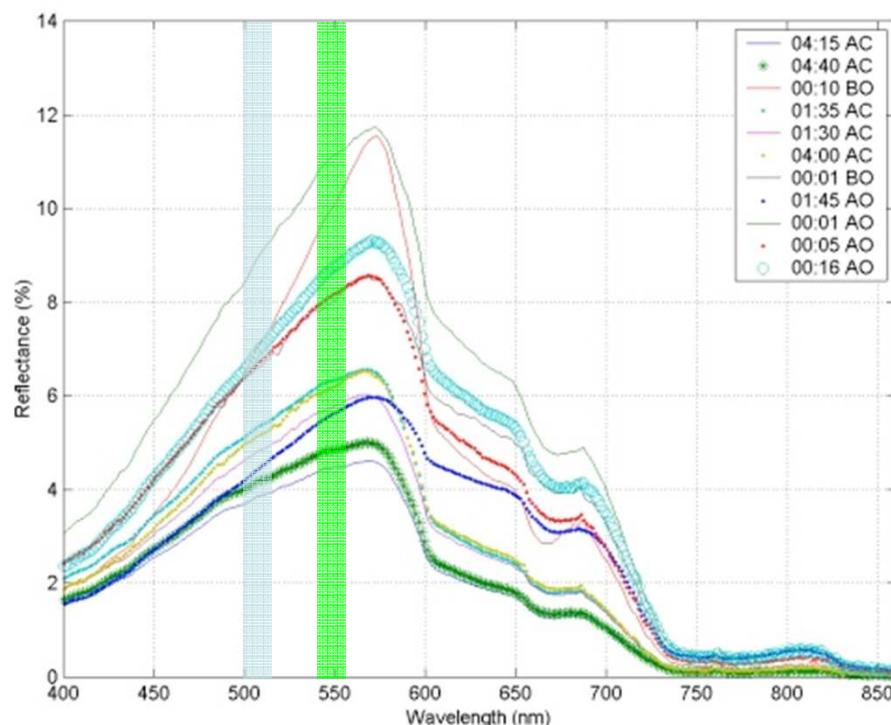
= Maximum of [R_{rs} -ratio(443/555, 490/555, 510/555)]

$$R_L$$

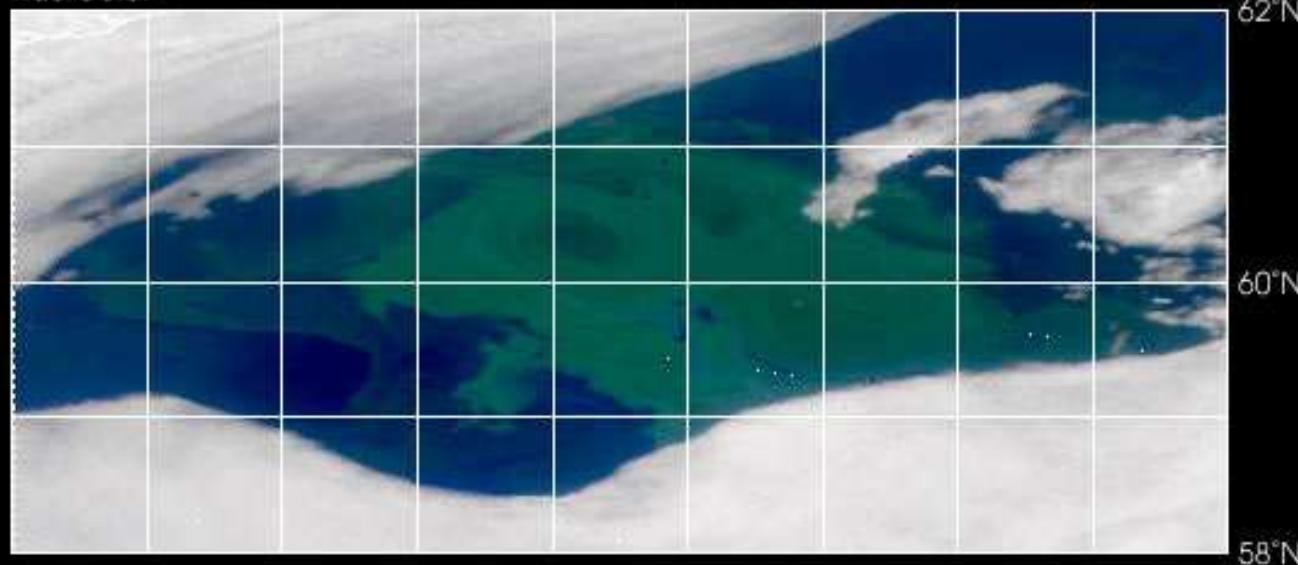
$$= \log_{10}(R_{\text{MAX}})$$

$$\text{Log}_{10}(\text{Chl-a})$$

$$= 0.366 - 3.067R_L + 1.930R_L^2 + 0.649R_L^3 - 1.532R_L^4$$



True Color

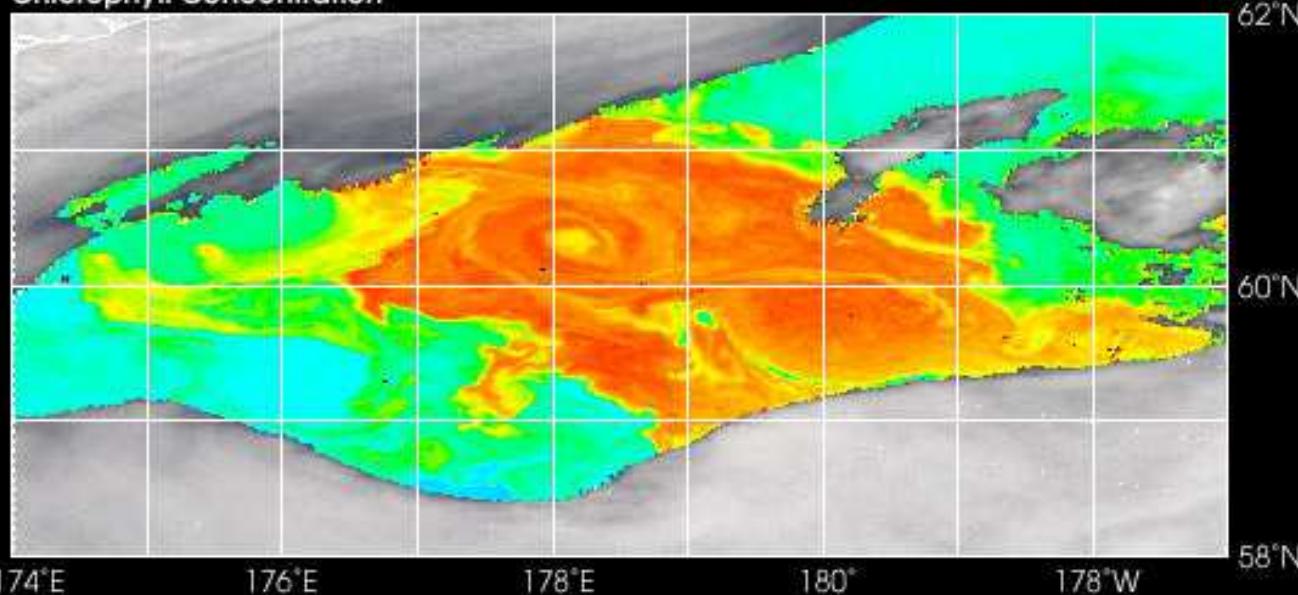


62°N

60°N

58°N

Chlorophyll Concentration



62°N

60°N

58°N

174°E

176°E

178°E

180°

178°W

Chlorophyll Concentration (mg/m^3)



>0.01

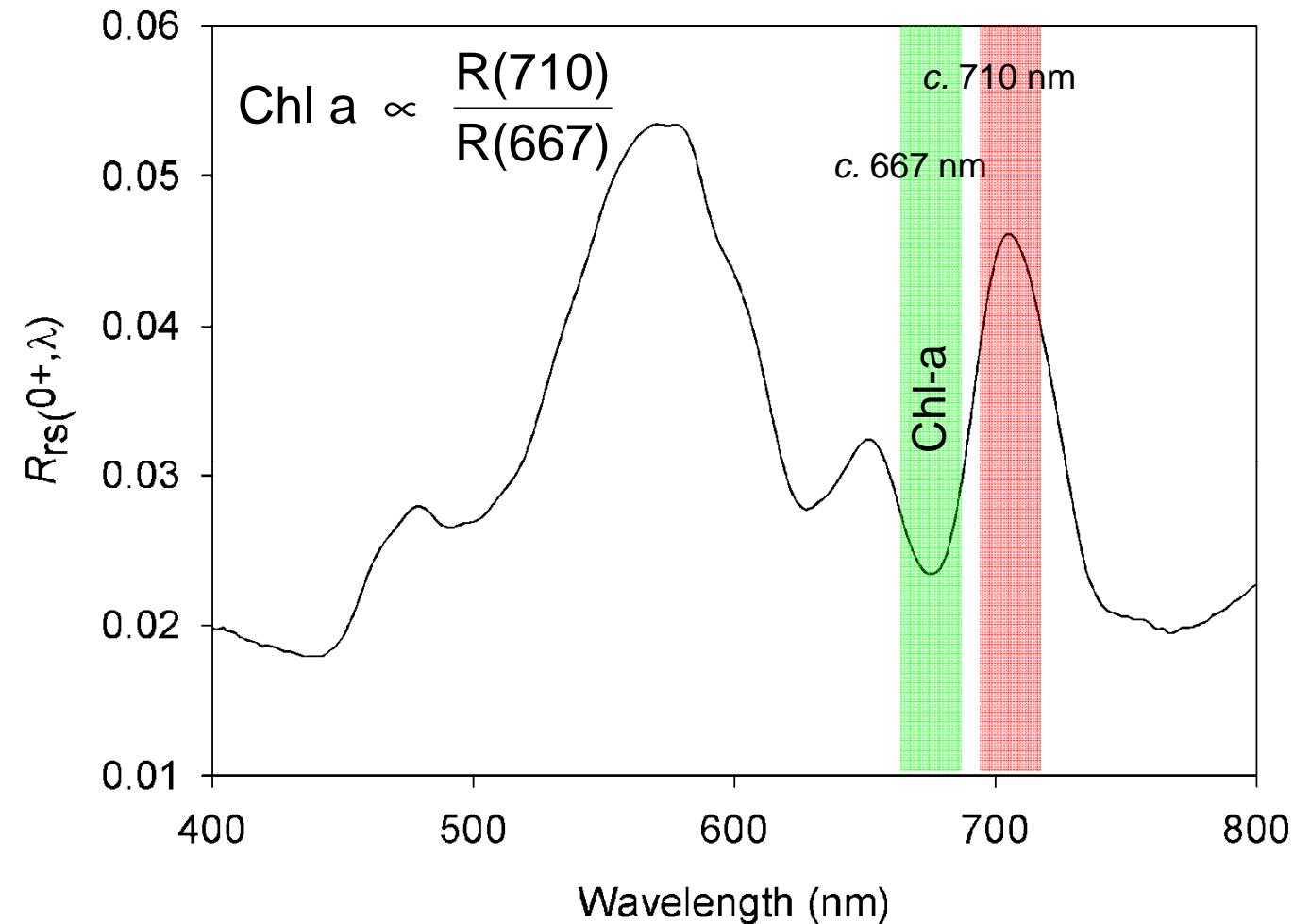
0.1

1.0

10

50

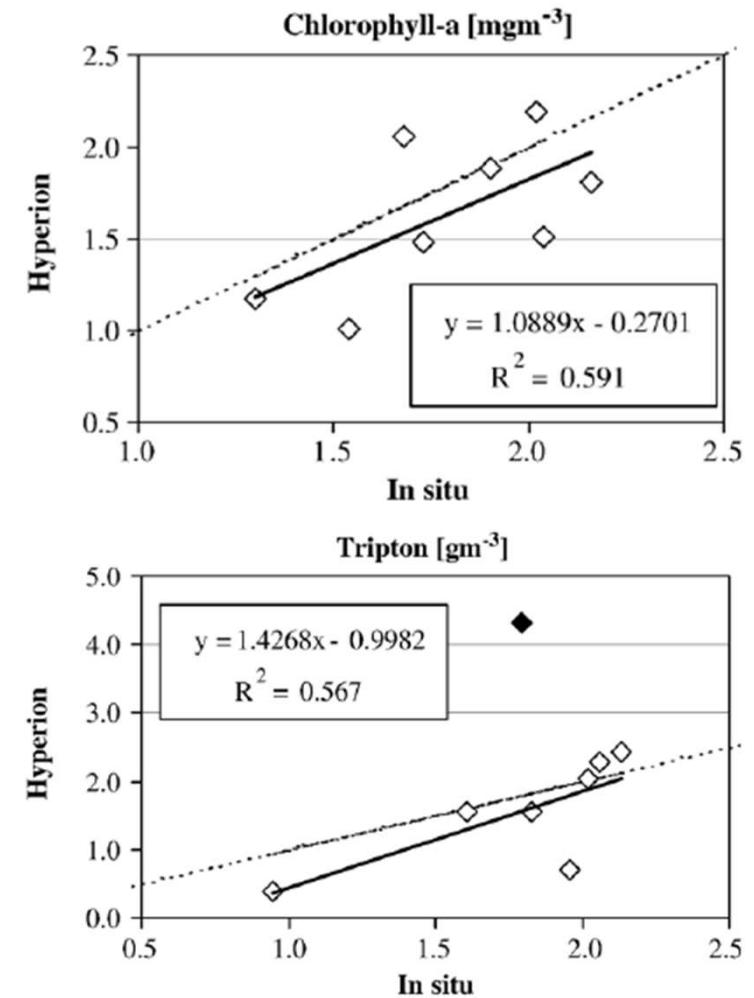
Chlorophyll: Case II waters



Lake Garda: Hyperion

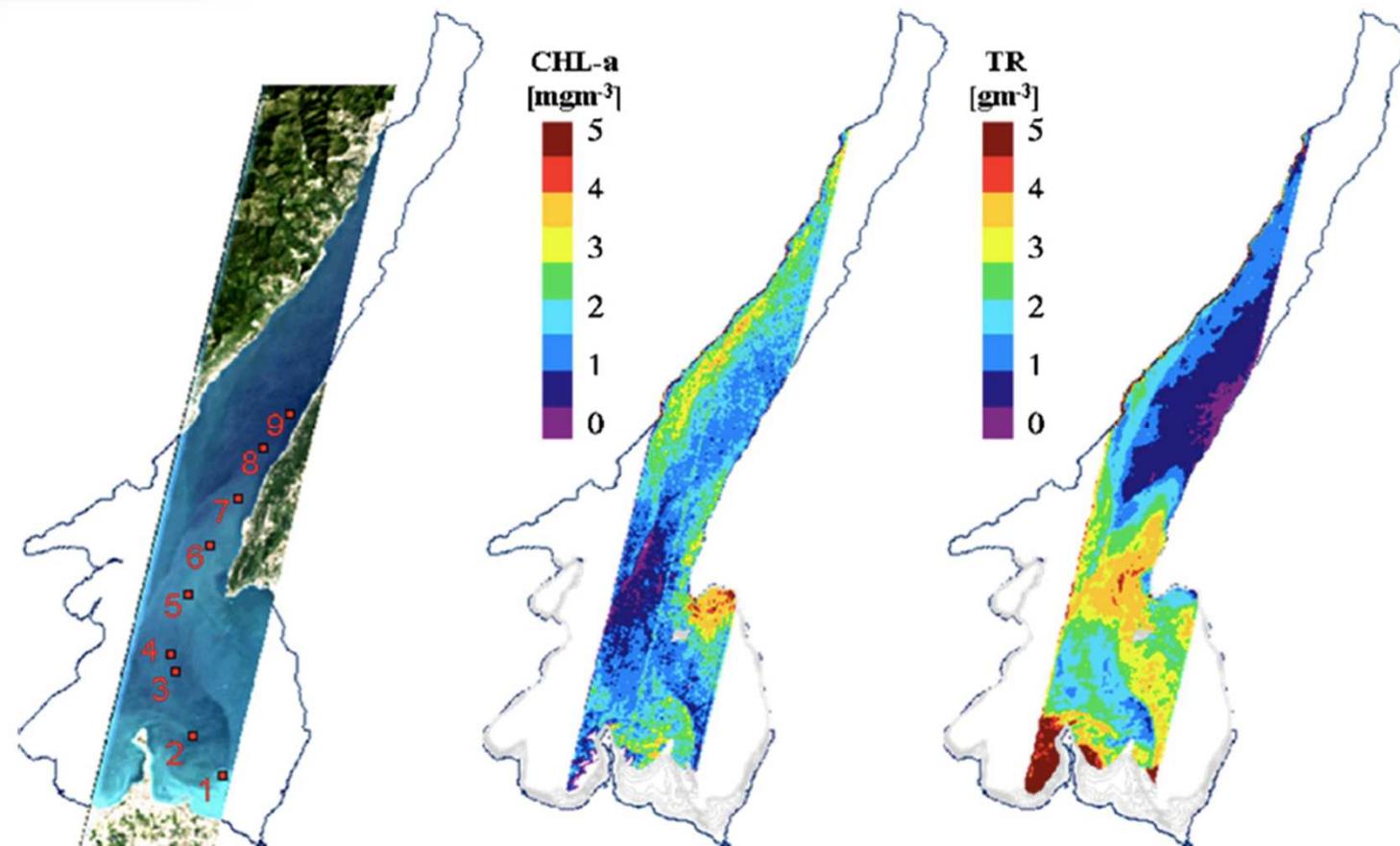
$$\begin{aligned}
 & [\text{CHL} - a] \left(a_{\text{ph}}^*(\lambda_i) + b_{\text{bph}}^*(\lambda_i) \left(1 - \frac{1 + \bar{\mu}_d(\lambda_i)/\bar{\mu}_u(\lambda_i)}{R(0-, \lambda_i)} \right) \right) \\
 & + [\text{TR}] \left(a_{\text{TR}}^*(\lambda_i) + b_{\text{bTR}}^*(\lambda_i) \left(1 - \frac{1 + \bar{\mu}_d(\lambda_i)/\bar{\mu}_u(\lambda_i)}{R(0-, \lambda_i)} \right) \right) \\
 & = a_w(\lambda_i) - b_{\text{bw}}(\lambda_i) \left(1 - \frac{1 + \bar{\mu}_d(\lambda_i)/\bar{\mu}_u(\lambda_i)}{R(0-, \lambda_i)} \right) \\
 & - a_{\text{CDOM}}(440) e^{-S_{\text{CDOM}}(\lambda_i - 440)}
 \end{aligned}$$

Bio-optical matrix inversion method (MIM)



Giardino et al. (2007) *Remote Sens Environ.* **109**:183-195

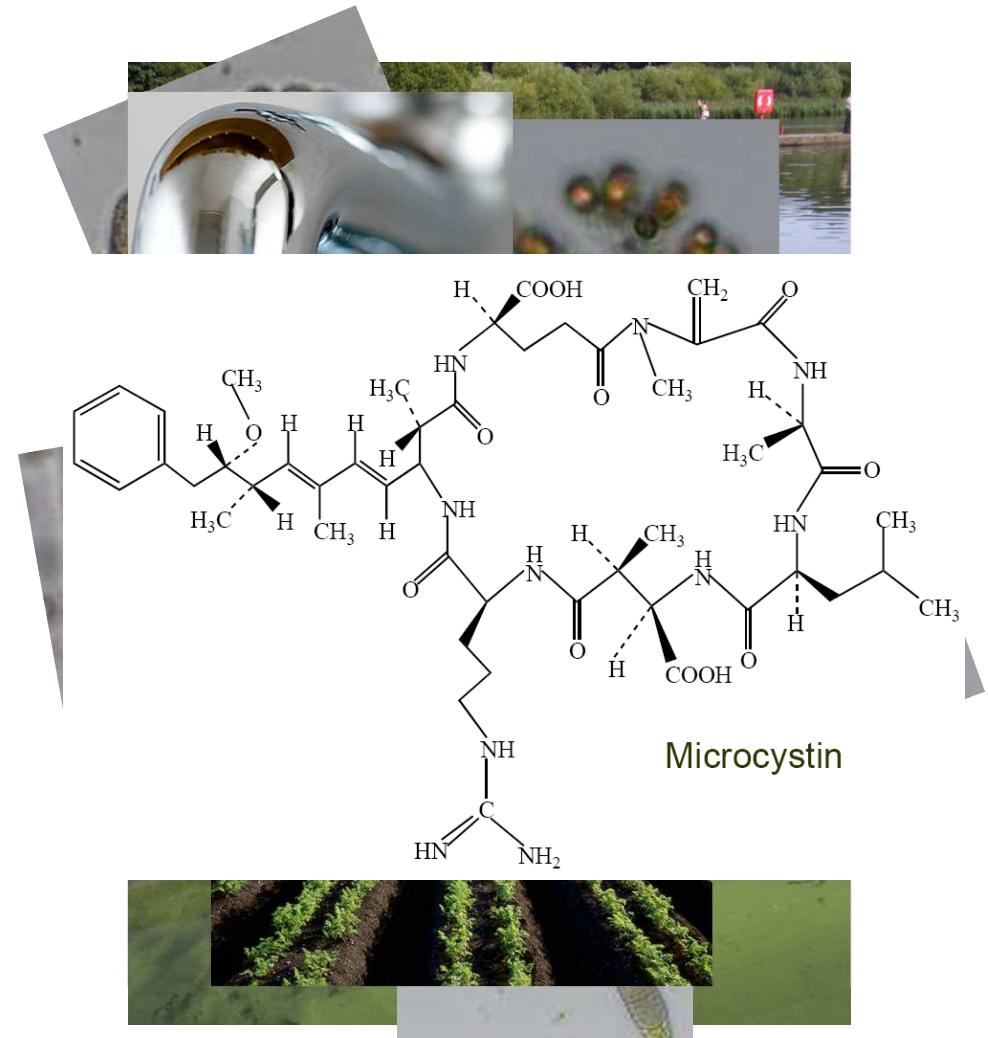
Lake Garda: Hyperion



Giardino et al. (2007) *Remote Sens. Environ.* **109**:183-195

Cyanobacteria

- Natural, ancient and cosmopolitan inhabitants of freshwaters
- Mass populations
- Numerous potent toxins
- Multiple exposure routes

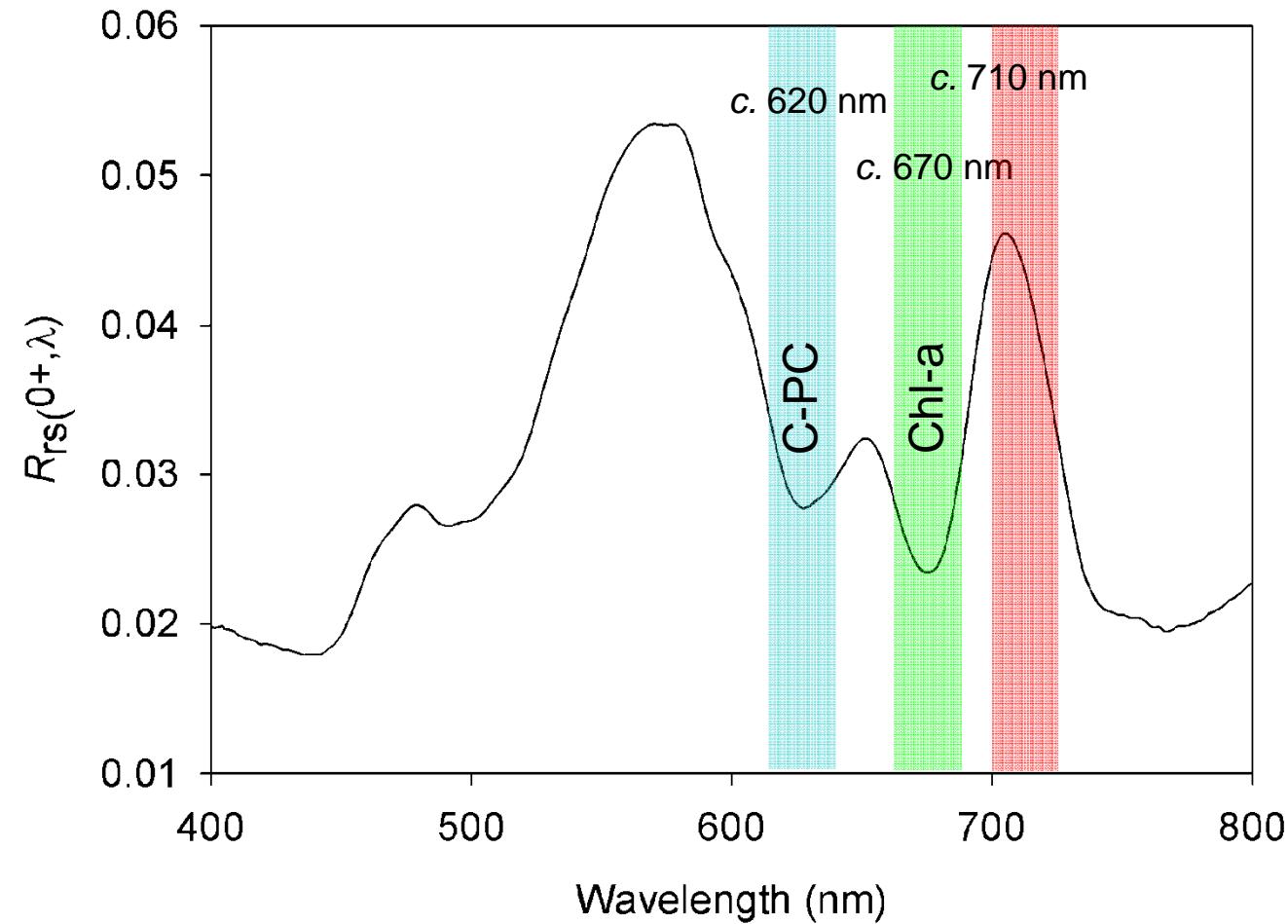


Cyanobacteria

- Inversion algorithms for chlorophyll-a (Chl-a) (total biomass)
- Cannot differentiate cyanobacteria
- New inversion algorithms for C-phycocyanin (C-PC)



C-phycocyanin



Colour groups

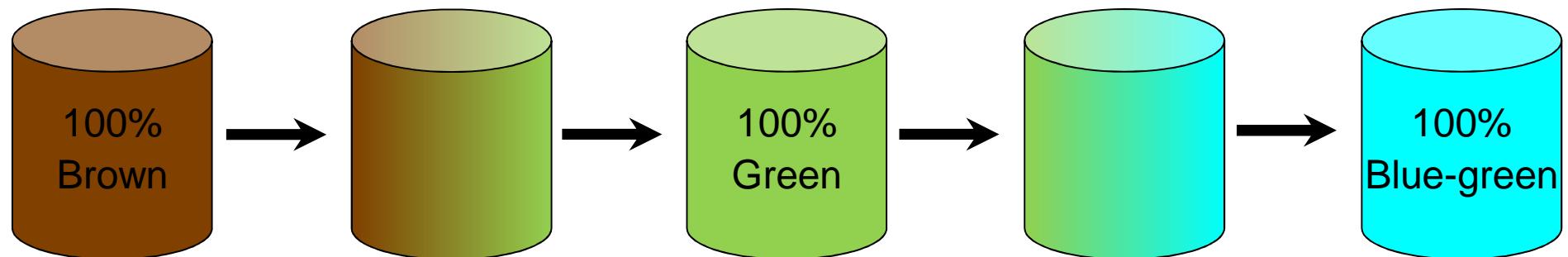
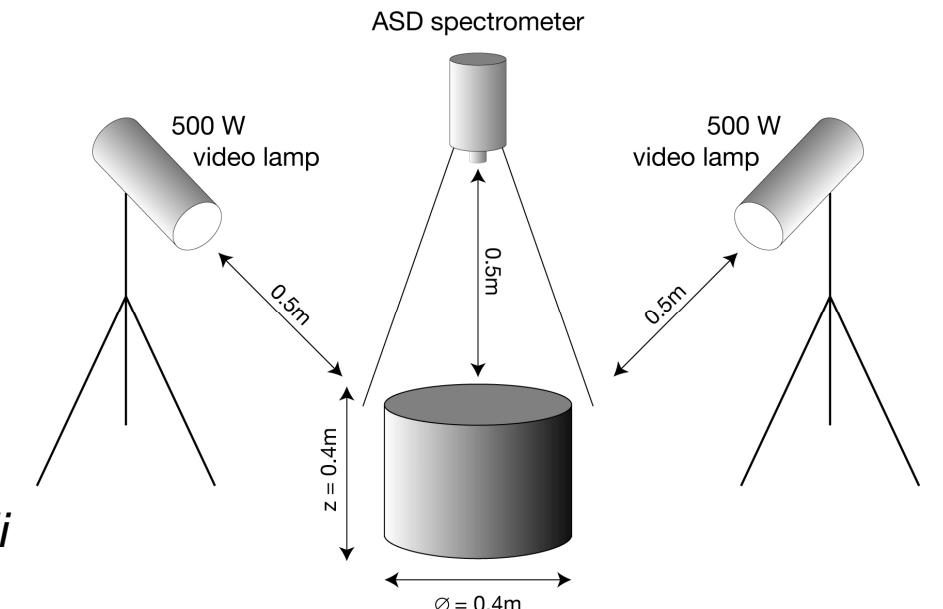
Tank experiments

Total Chl *a* ~ 200 mg m⁻³

Brown: diatom

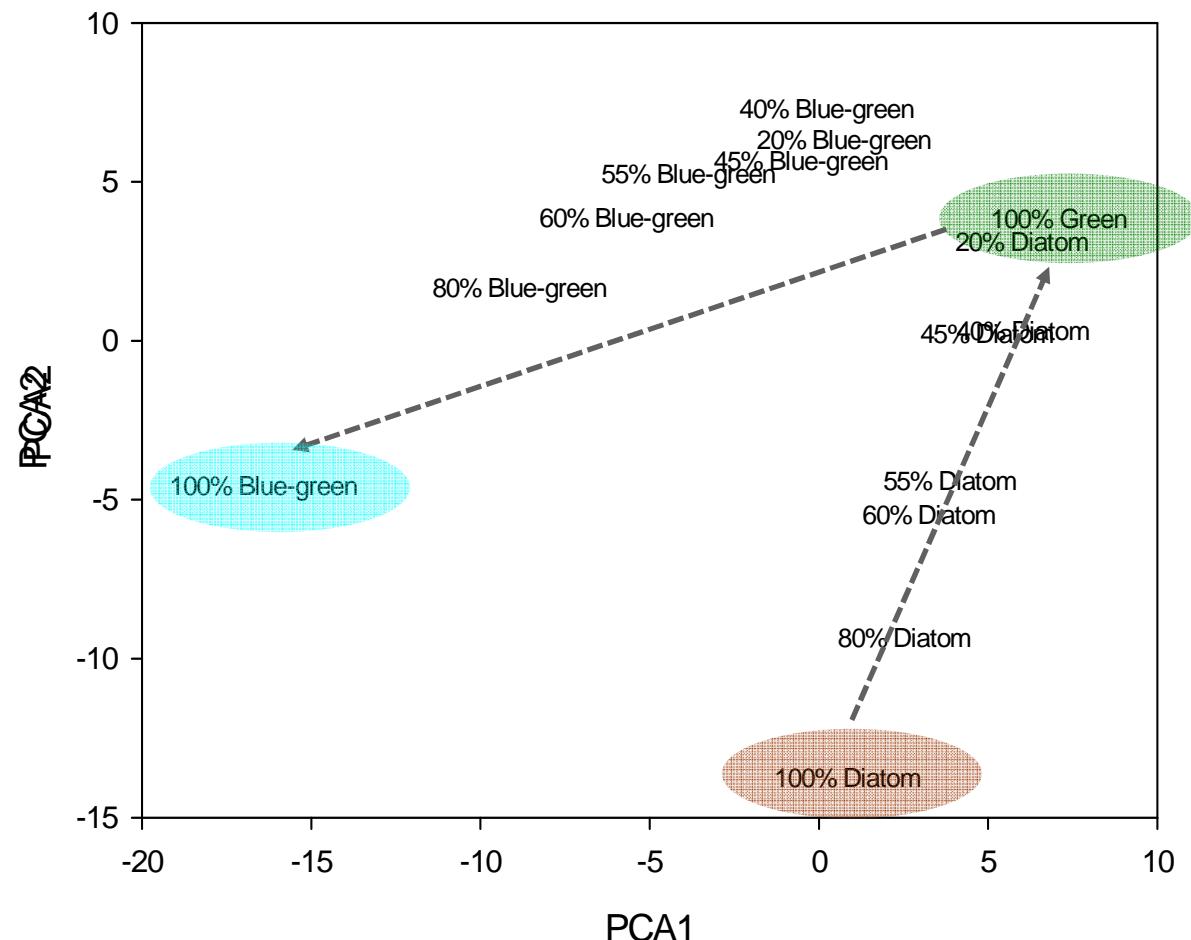
Green: *Scenedesmus armatus*

Blue-green: *Cylindrospermopsis raciborskii*



Hunter et al. (2008) *Remote Sens. Environ.* **112**:1527-1544

Colour groups

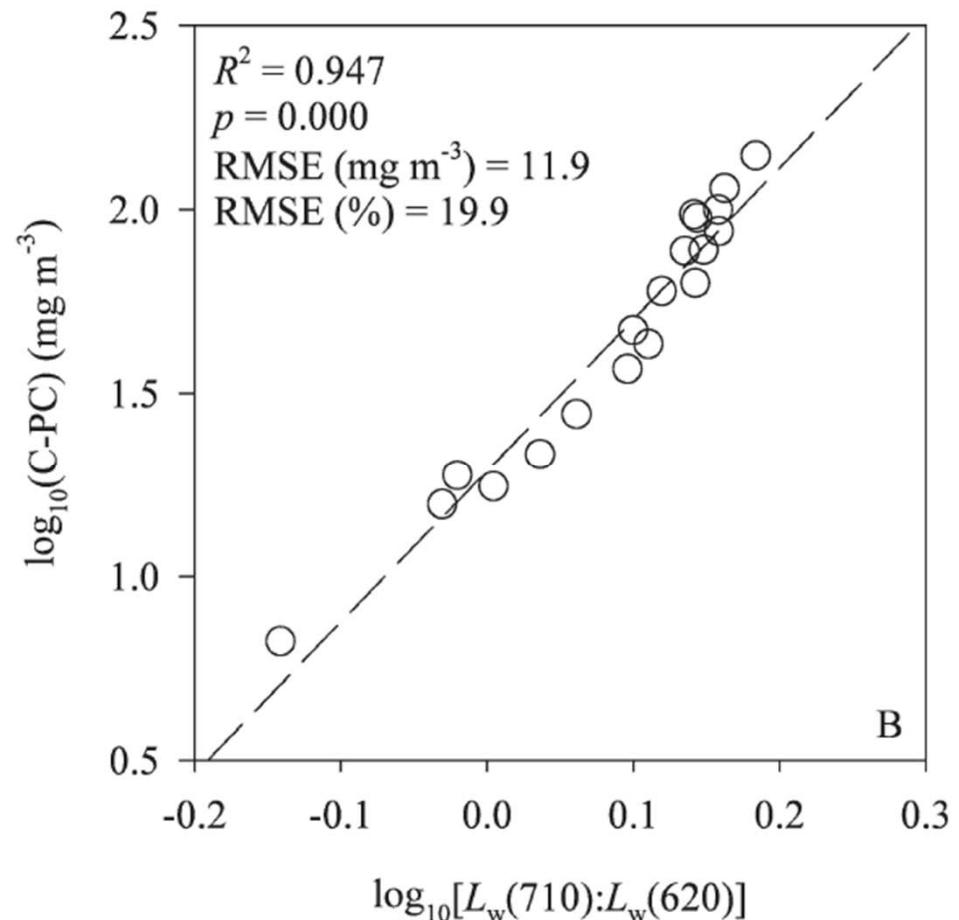


Hunter et al. (2008) *Remote Sens. Environ.* **112**:1527-1544

C-phycocyanin

Empirical algorithm

$$\log_{10}(\text{CPC}) = 1.29 + 4.12 \times \log_{10}(L_w(710):L_w(620))$$



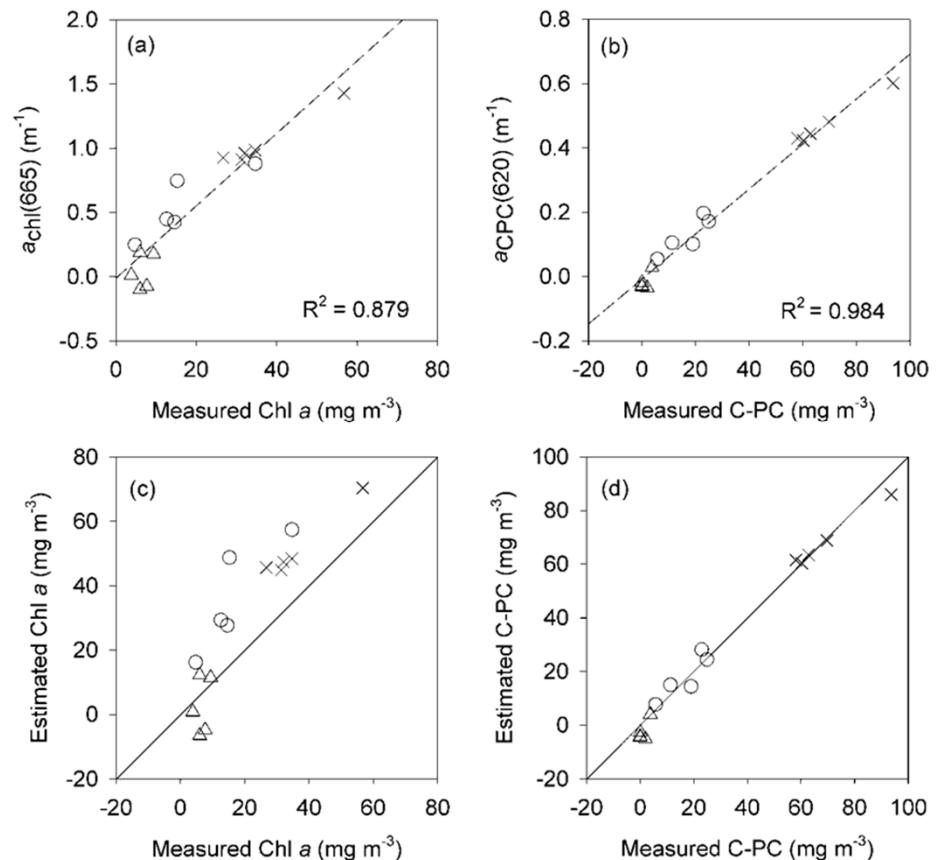
C-phycocyanin

$$a_{\text{chl}}(665) = \left(\left\{ [a_w(709) + b_b] \times \left[\frac{R(709)}{R(665)} \right] \right\} - b_b - a_w(665) \right) \times \gamma^{-1}$$

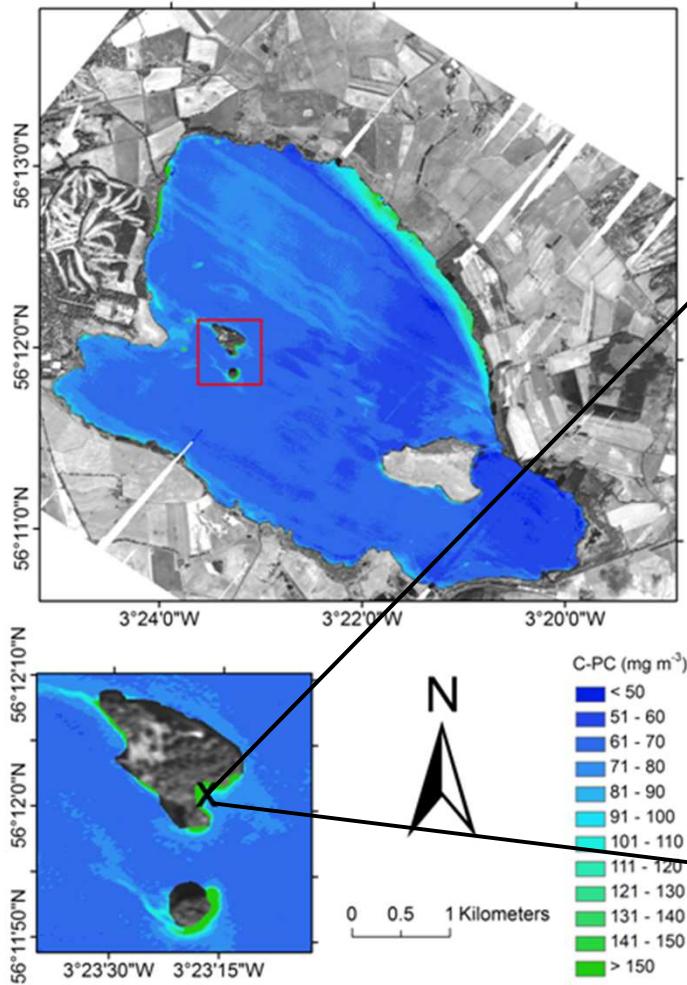
$$a_{\text{CPC}}(620) = \left(\left\{ [a_w(709) + b_b] \times \left[\frac{R(709)}{R(620)} \right] \right\} - b_b - a_w(620) \right) \times \delta^{-1} - [\varepsilon \times a_{\text{chl}}(665)]$$

where

$$b_b(779) = \frac{1.61 \times R(779)}{0.082 - 0.6 \times R(779)}.$$

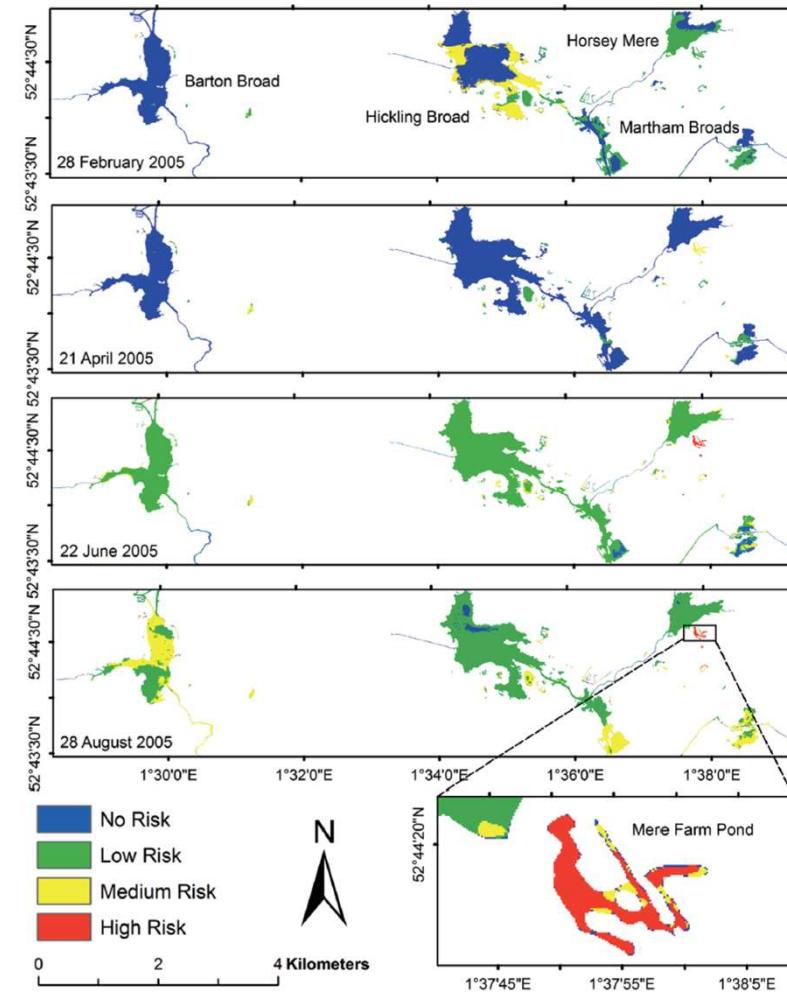
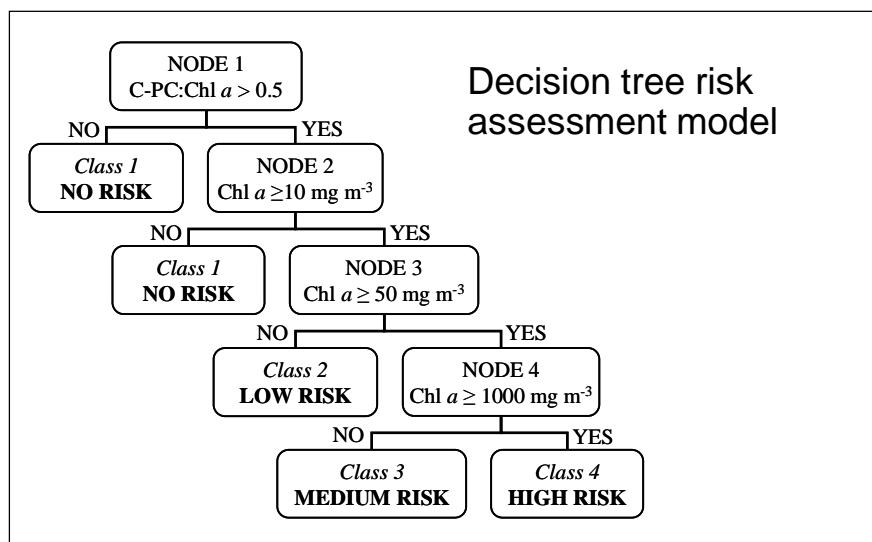


Loch Leven, Scotland

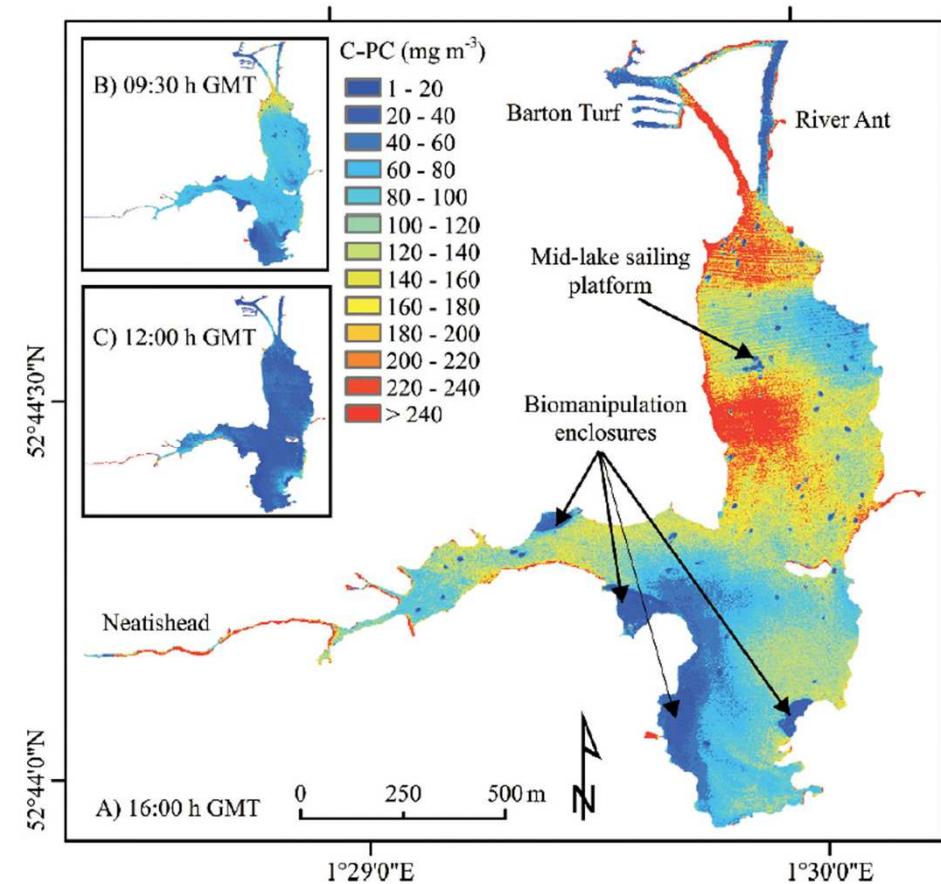
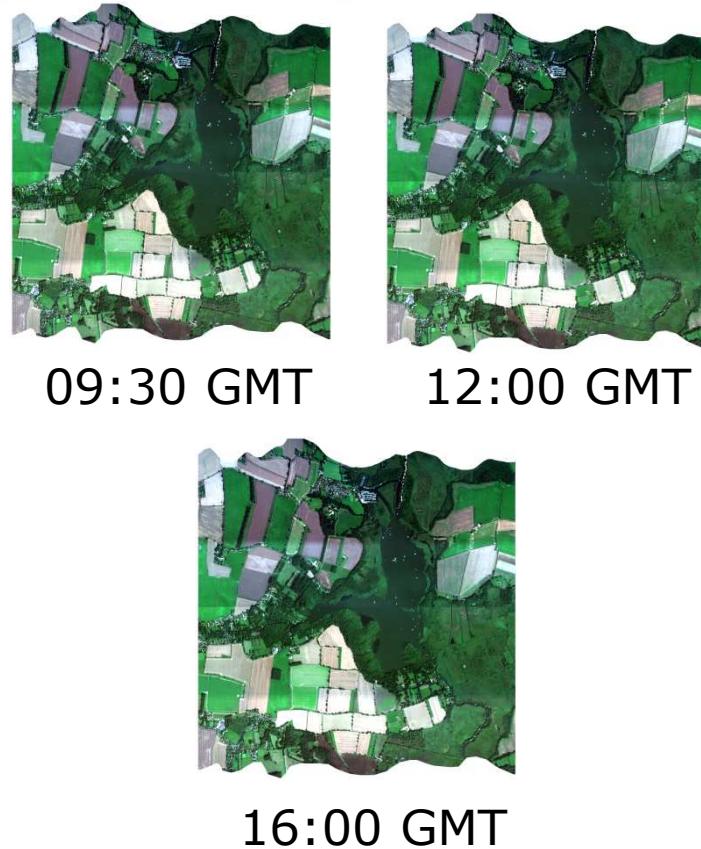


Hunter et al. (2010; in press) *Remote Sens. Environ.*

Health risk monitoring



Vertical migration



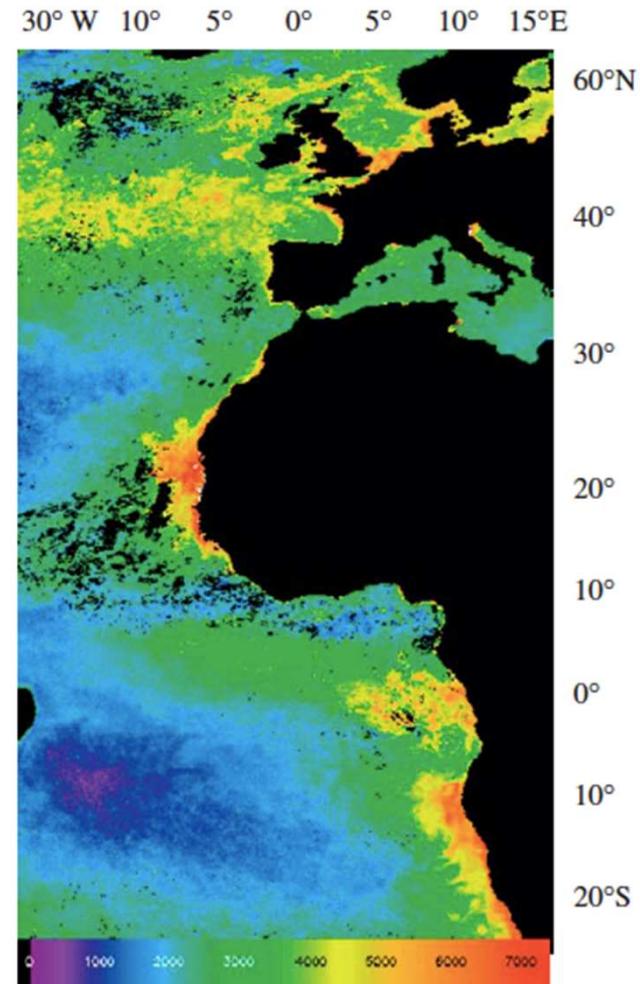
Primary production

$$\log_{10}\text{PP} = 0.599 \times \log_{10}\text{Chl-a} + 2.739$$

(Behrenfeld et al.
1998)

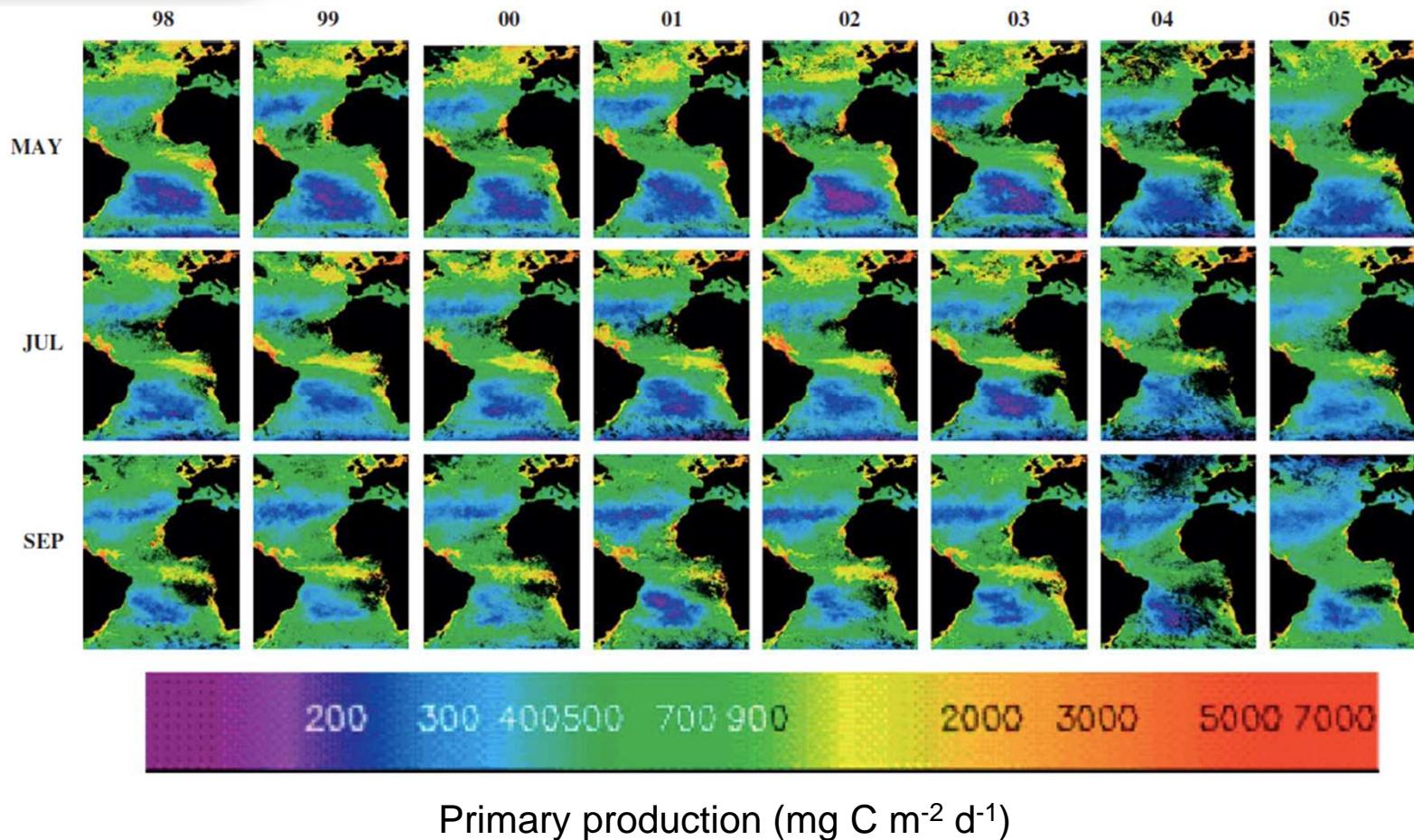
$$\text{PP} = f(C_{\text{sat}}, Z_{\text{eu}}, P_{\text{opt}}^b)$$

(Behrenfeld & Falkowski 1997)



Tilstone et al. (2009) *Deep Sea Res. II*. **56**: 918-930

Primary production



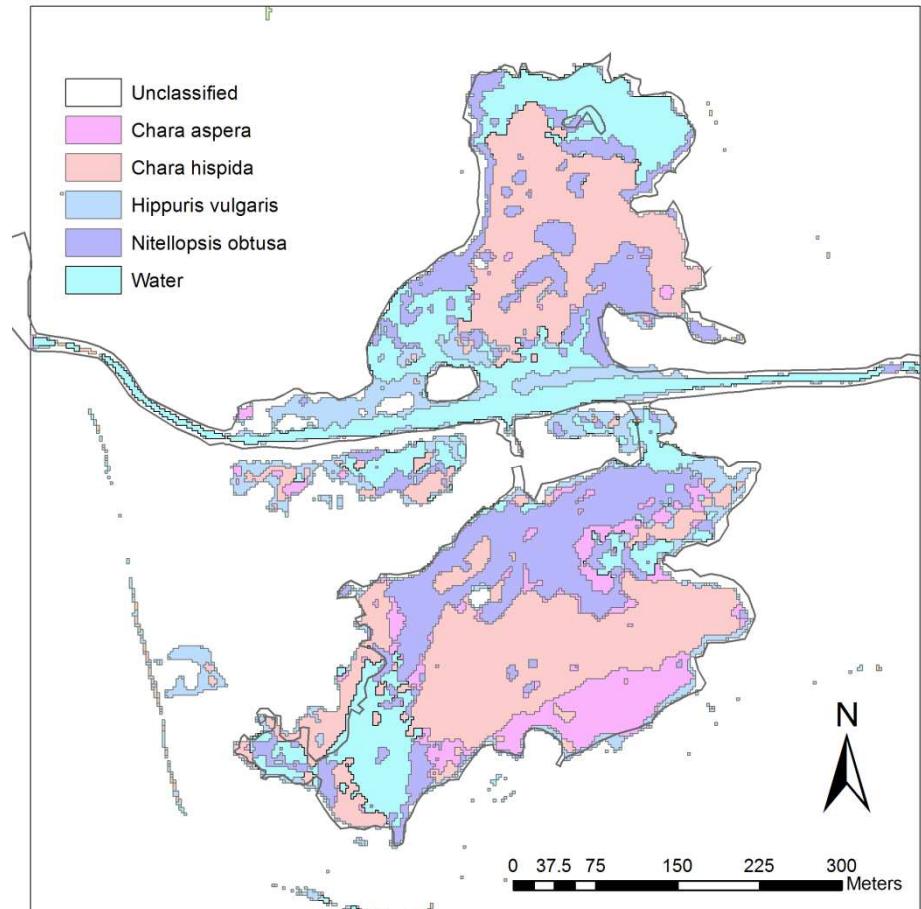
Tilstone et al. (2009) *Deep Sea Res. II*. **56**: 918-930

Submerged vegetation

CASI-2 imagery

Support Vector Machine
classification of submerged
vegetation in the Norfolk Broads

Classification accuracy = 87.3%
Kappa statistic = 0.84



NERC ARSF & EUFAR: Lake Balaton 2010

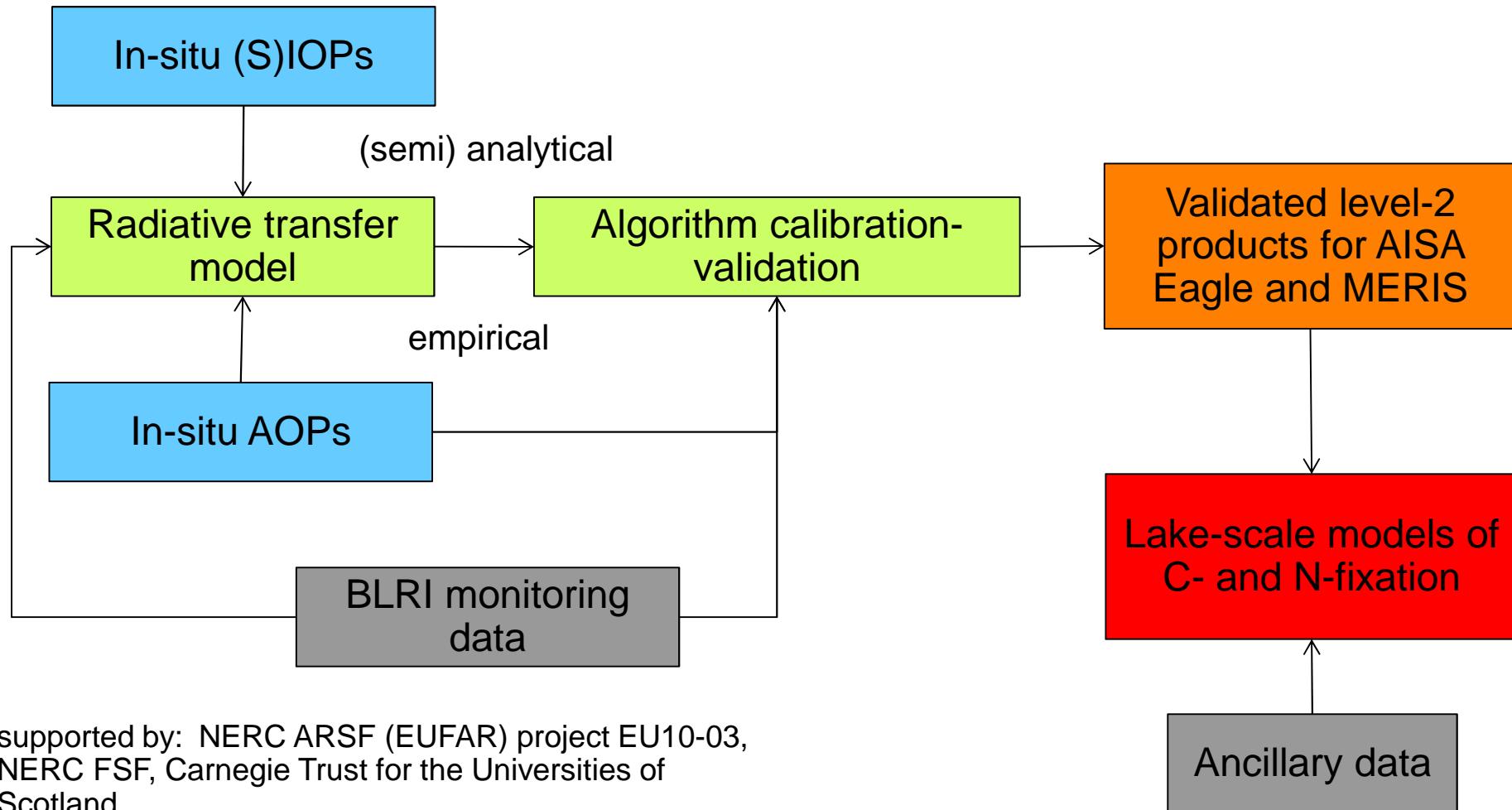
Peter Hunter, Caitlin Lyman & Andrew Tyler
University of Stirling

Steve Groom & Victor Martinez Vicente
Plymouth Marine Laboratory

Attila Kovacs & Matyas Presing
Balaton Limnological Research Institute

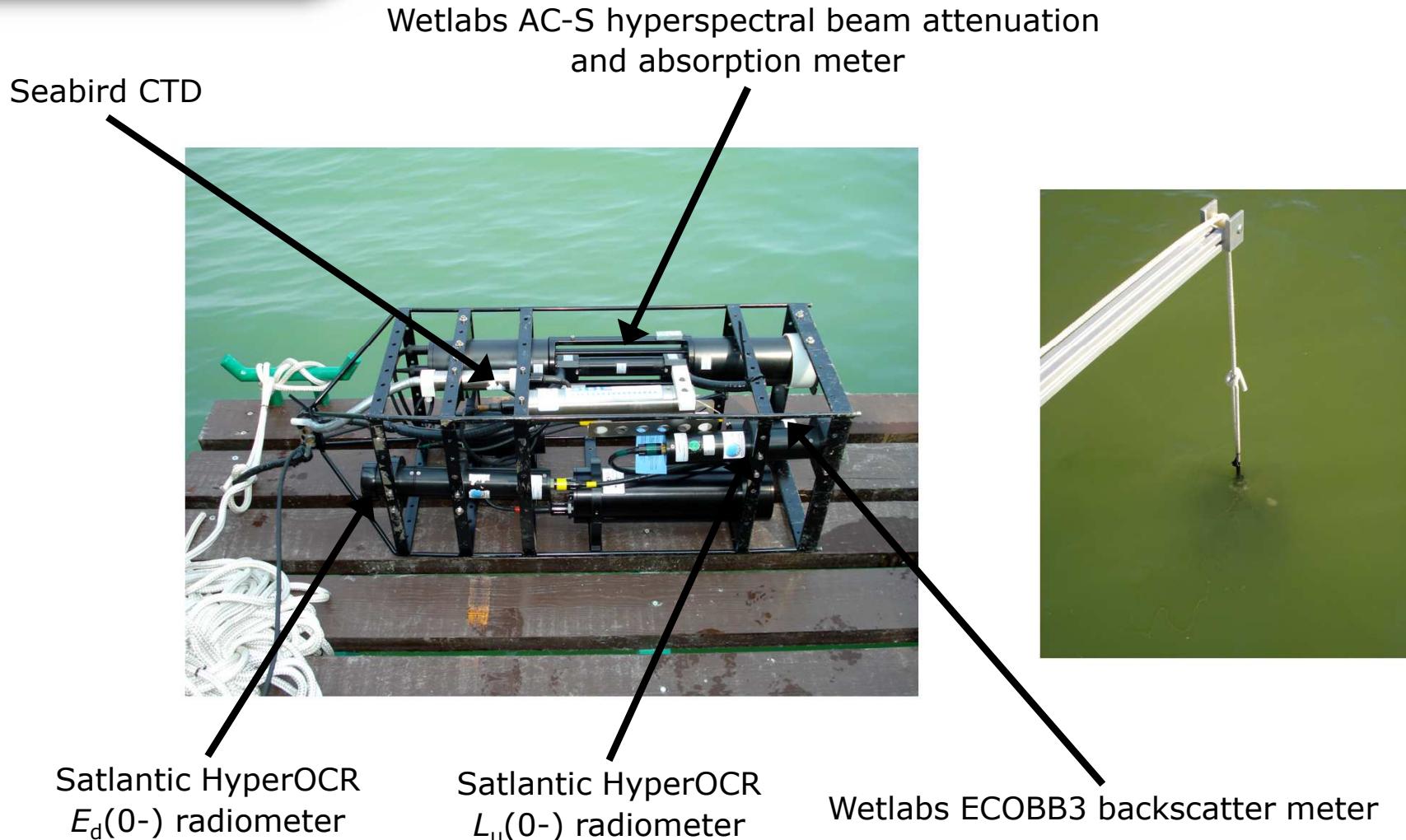
Tom Preston
Scottish Universities Environmental Research Centre

NERC ARSF Lake Balaton 2010

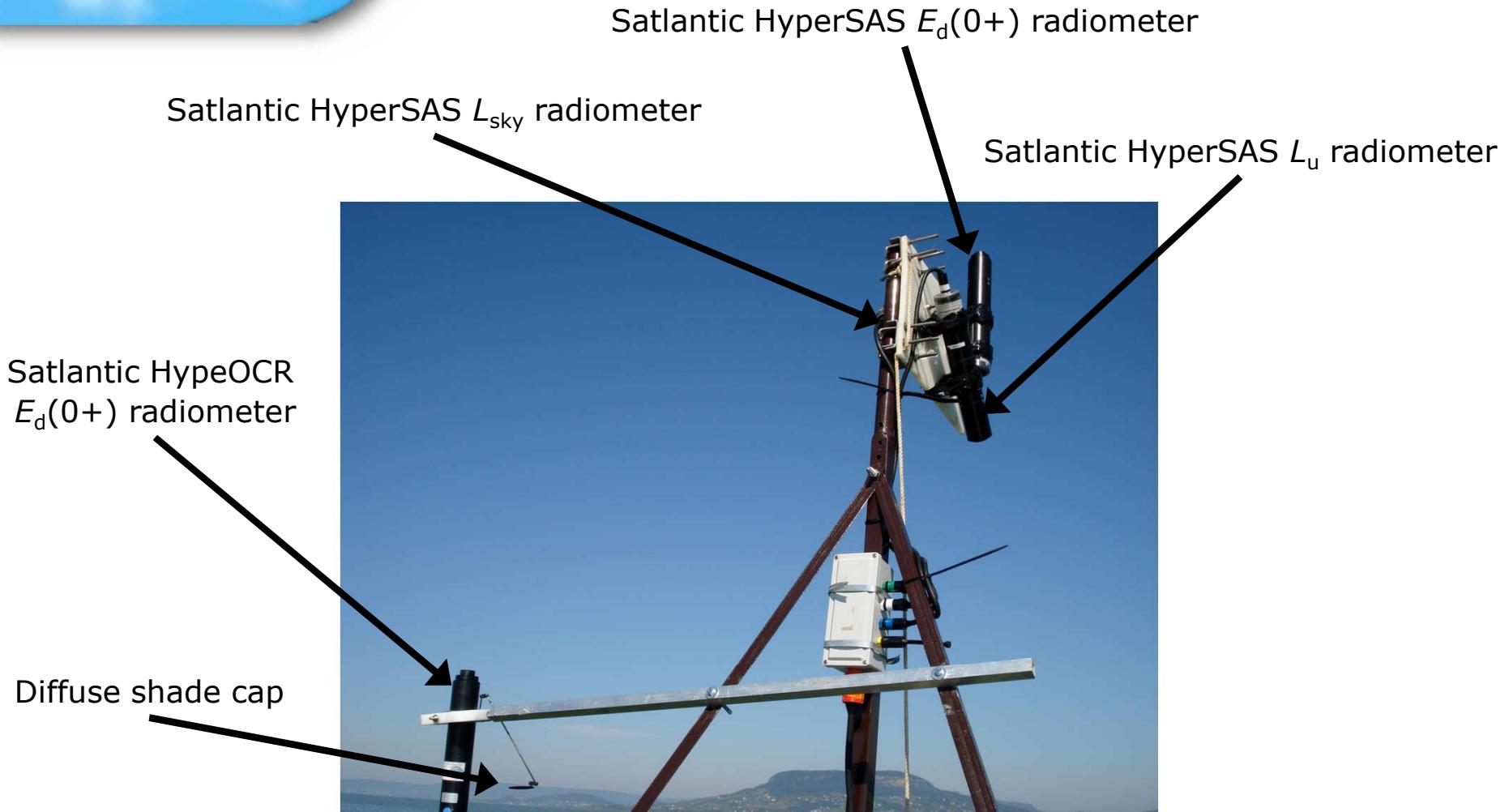


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NERC ARSF Lake Balaton 2010



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Ship's winch

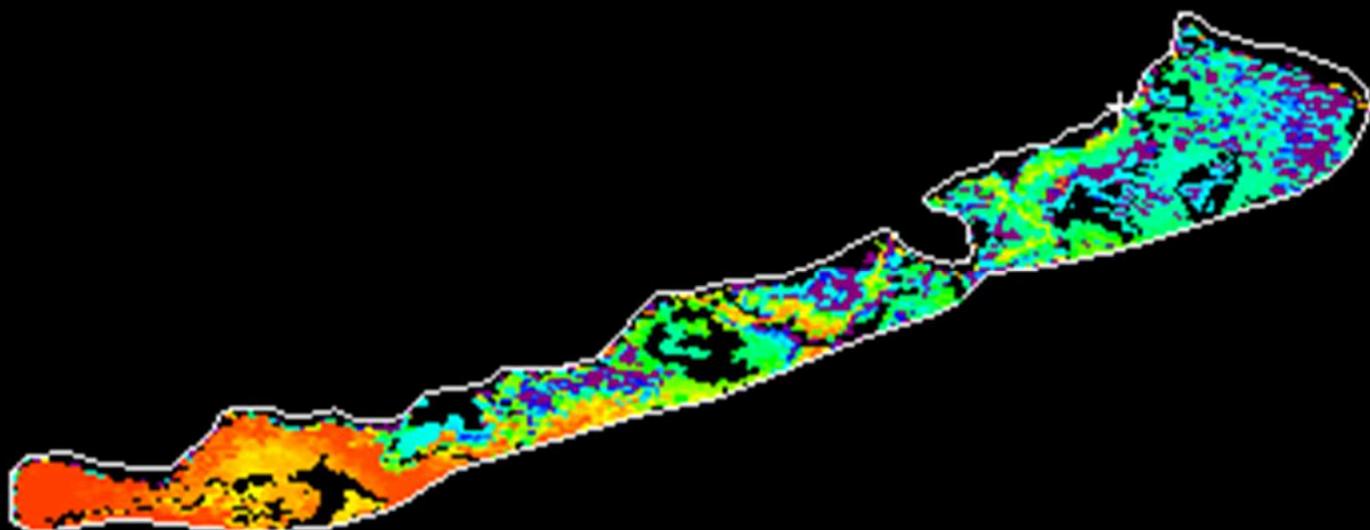


Onboard laboratory



NERC ARSF

MERIS algal_2
Lake Balaton 11 August 2010



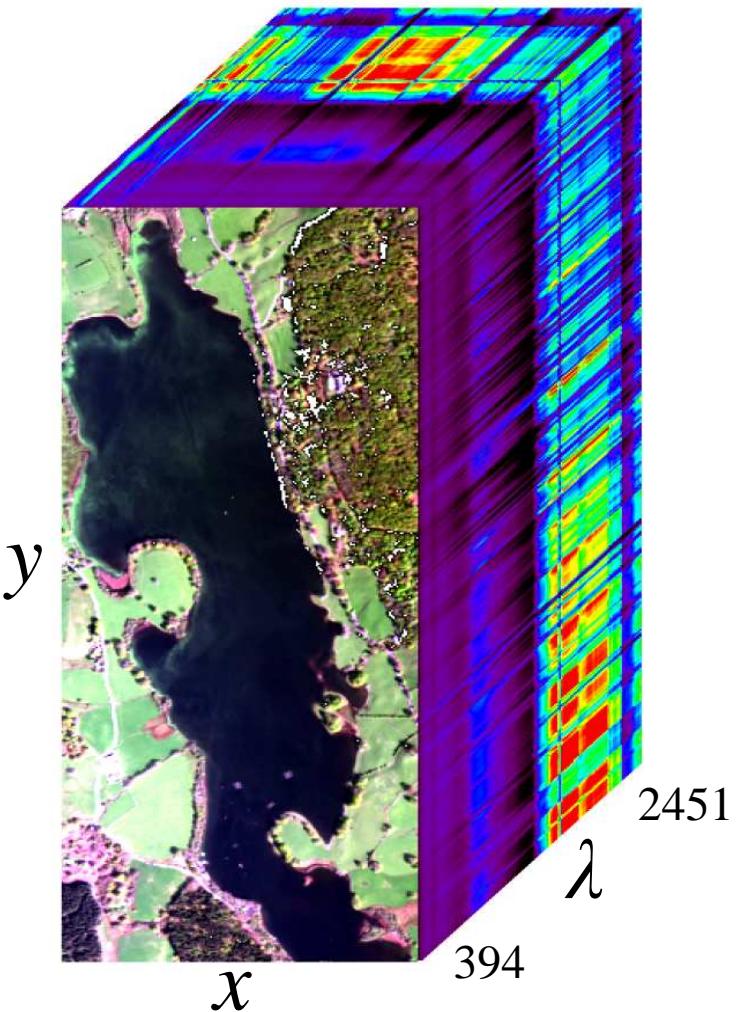
PML Remote Sensing Group

0.01 0.02 0.04 0.09 0.2 0.4 0.7 2 3 4 6 8 20 40

Hands-on practical

Hyperspectral AISA Eagle and Hawk imagery collected over Esthwaite Water in April 2007

- Test an existing empirical Chl-a retrieval algorithm
- Design and apply an empirical algorithm for C-phycocyanin retrieval
- Test a semi-analytical algorithm for Chl-a retrieval (if time permits)



A few acknowledgements...

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NERC Field Spectroscopy Facility

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