

# Remote sensing & mapping of water quality in optically complex waters

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# This presentation

Introduction – water quality from remote sensing

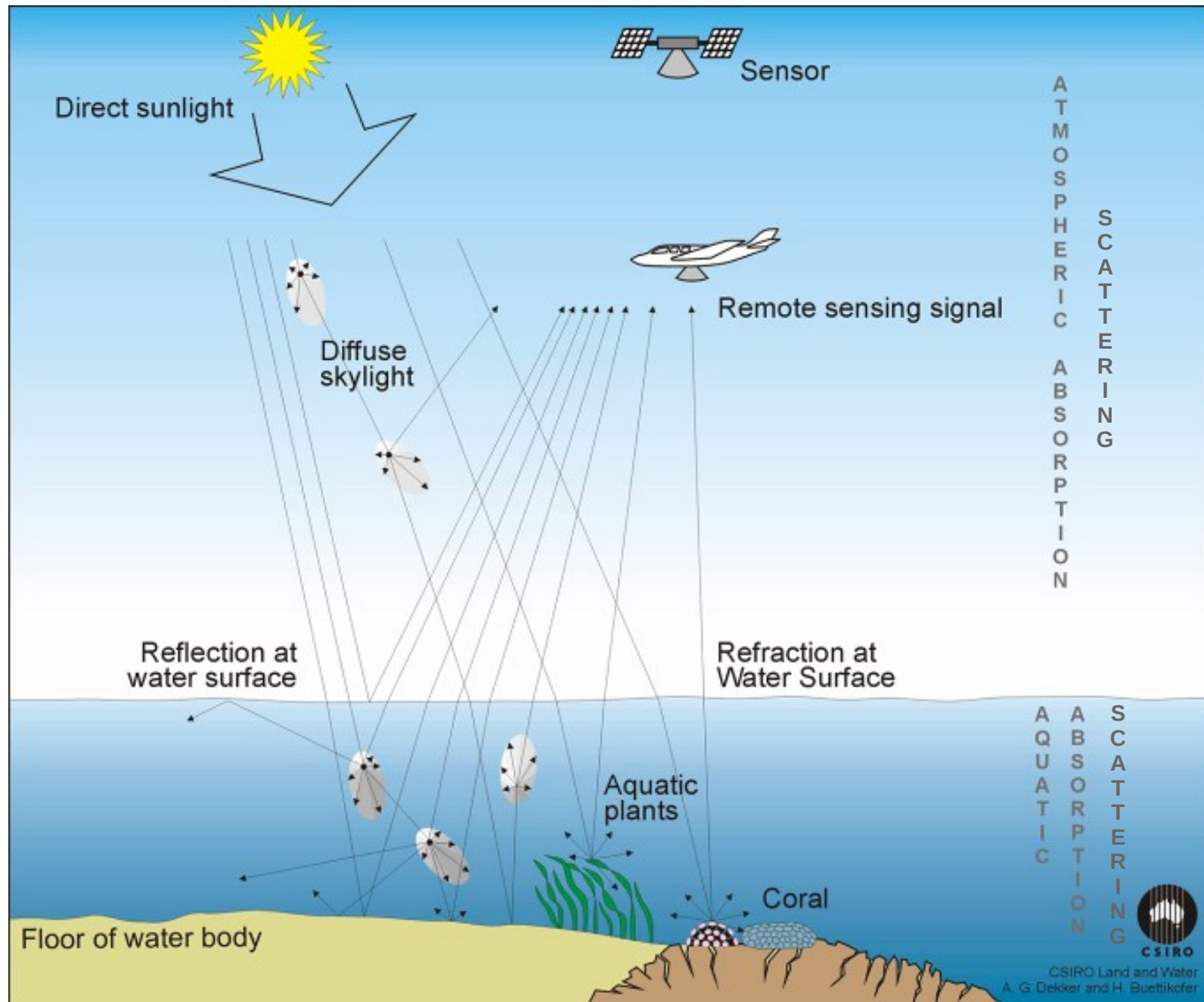
Challenges in Australian coastal waters

CSIRO's algorithm methodology for optically deep waters

Validation (opportunities)

Application – Flood plume mapping

$$L_{\text{TOA}}(\lambda) = L_{\text{Rayleigh}}(\lambda) + L_{\text{Aerosol}}(\lambda) + L_{\text{Rayleigh+Aerosol}}(\lambda) + t(\lambda) L_{\text{Water}}(\lambda)$$



# Standard atmospheric correction approach

## The “black pixel” (BP) assumption

NIR (>700 nm) open ocean is assumed to be “black” with  $L_w=0$

For turbid (coastal) waters black ocean assumption often invalid

Various approaches try to re-establish the BP assumption by removing the NIR contribution from the TOA NIR signal eg:

[Siegel et al. \(2000\)](#) via CHL estimate

[Ruddick et al. \(2000\)](#) via fixed aerosol and backscatter type

[Lavender et al. \(2005\)](#) via sediment estimate

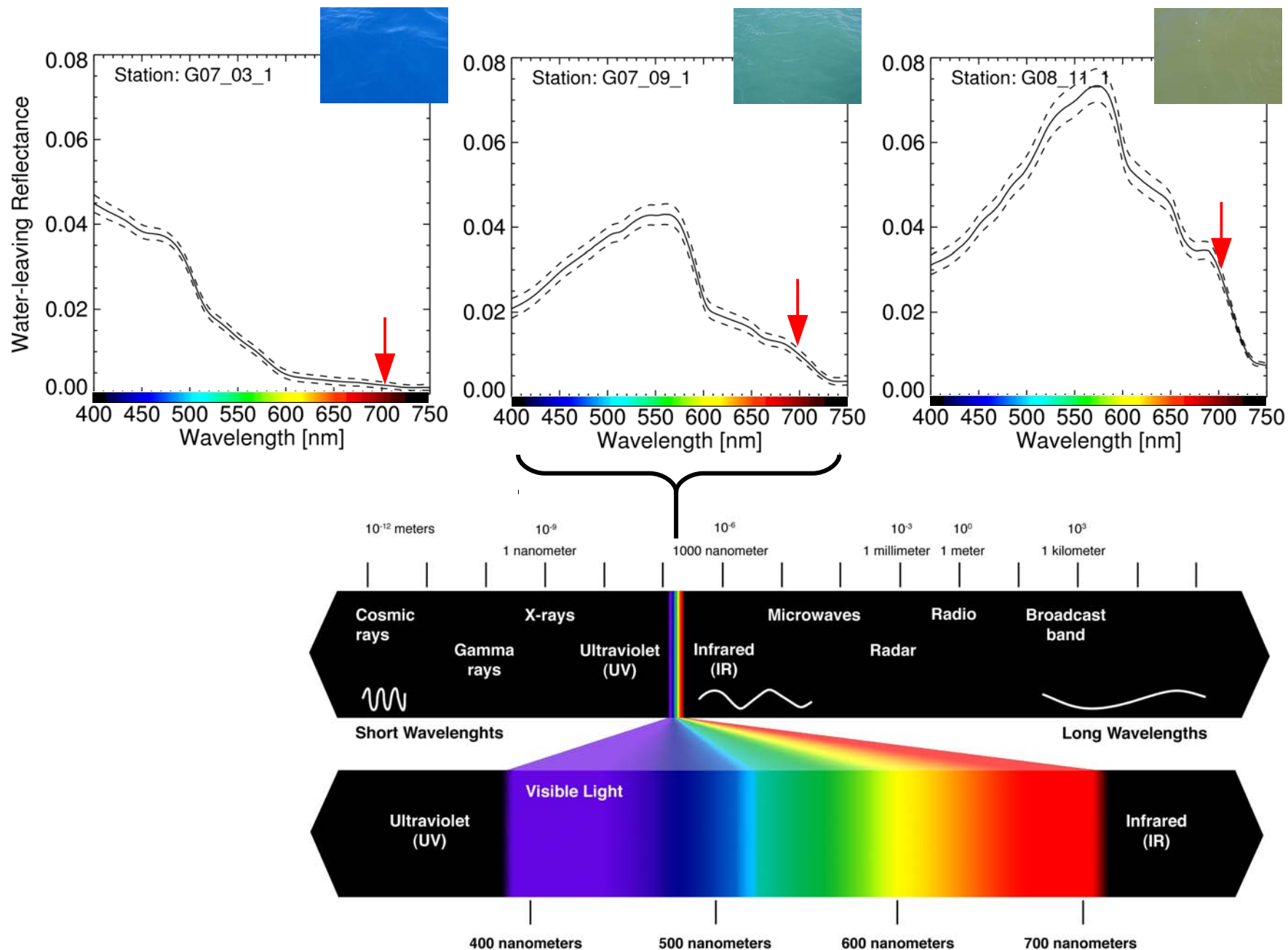
[Wang et al. \(2007\)](#) SWIR correction 1240, 1640, 2130 nm MODIS only

[Aiken and Moore \(2000\)](#) use iterative approach for bright pixels



# Ocean Colour - Reflectance

## Intro II





MODIS Aqua

10 February 2007



0 100 km



# Global algorithms & Australian coastal waters

## Problem 1

Global (empirical) algorithms are designed for open ocean (blue) waters and frequently fail when applied to optically complex coastal waters as previously shown for the Great Barrier Reef

(Qin et al. 2007, GRL)

## Problem 2

Standard atmospheric correction algorithms systematically fail for CDOM and TSS dominated coastal waters



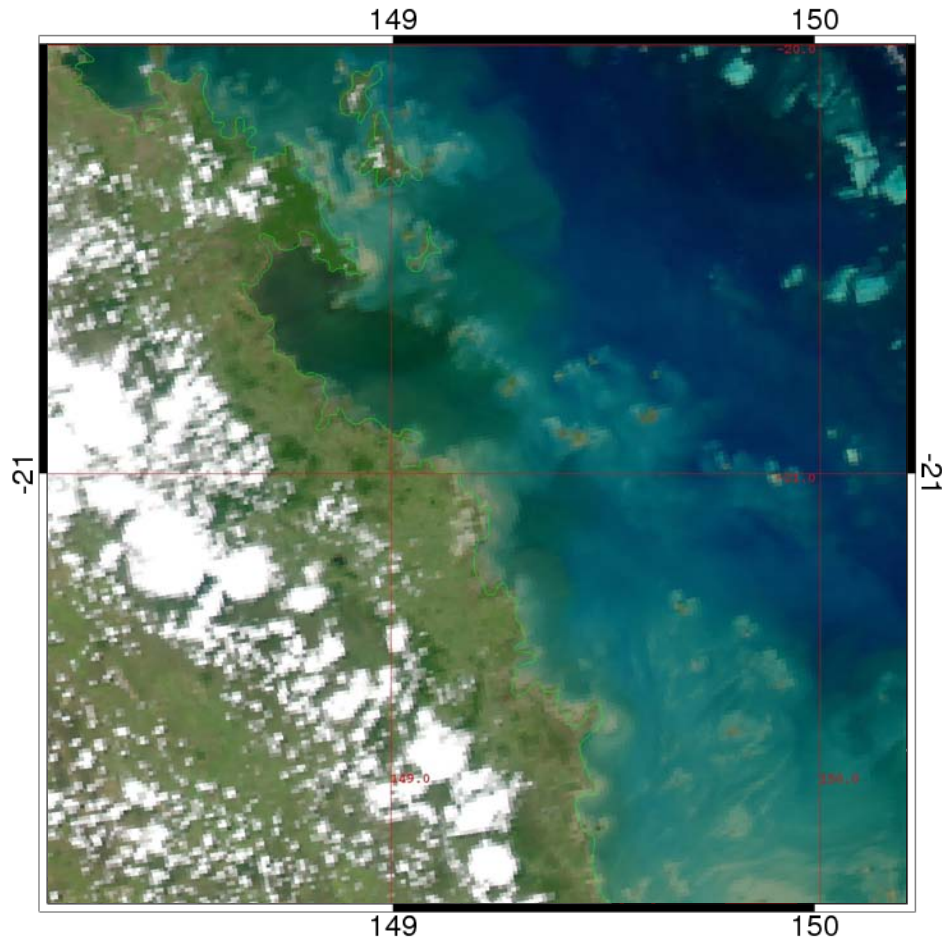
Image credit: <http://www.lib.utexas.edu>

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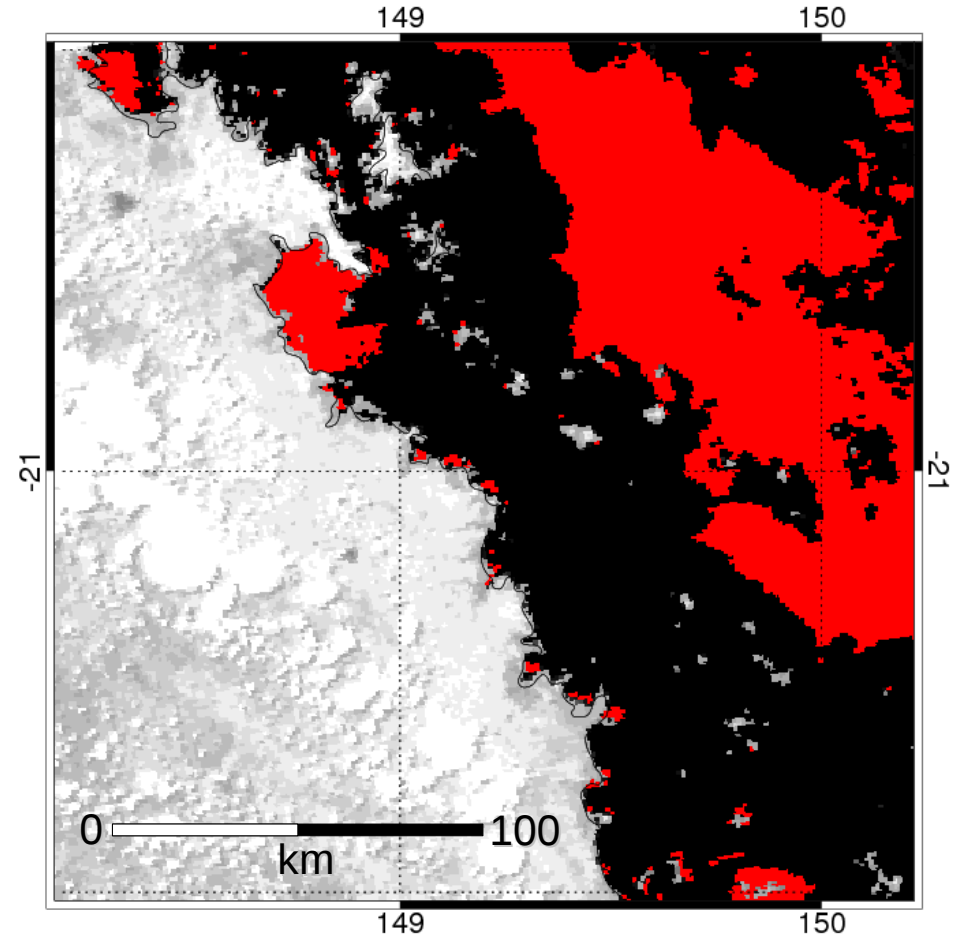


# NIR atmospheric correction failure MODIS-Aqua

## Extreme 1: Mackay – Whitsundays, QLD – 22 February 2008



Spectral range 412 – 748 nm

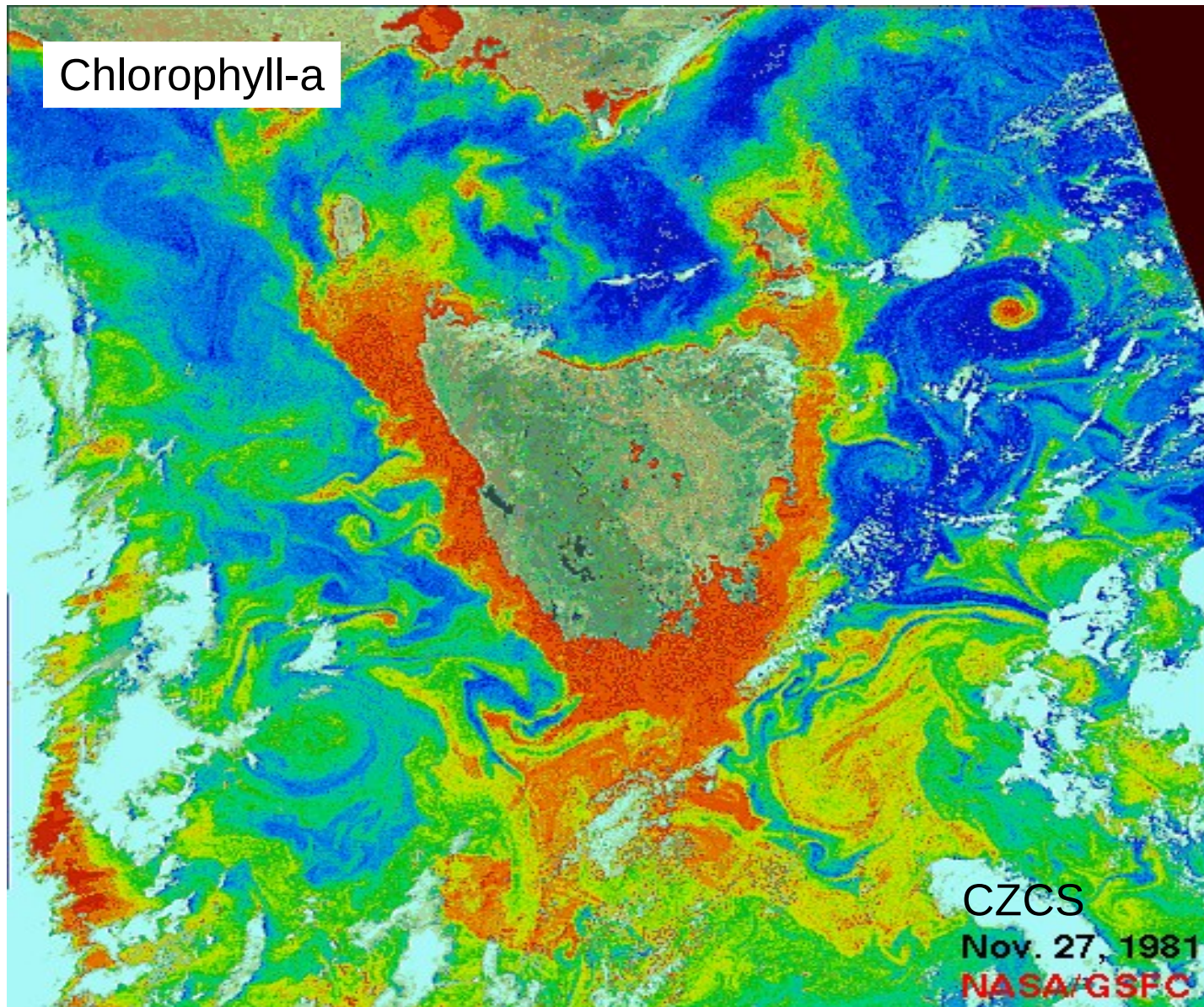


■ = Atmospheric correction failure

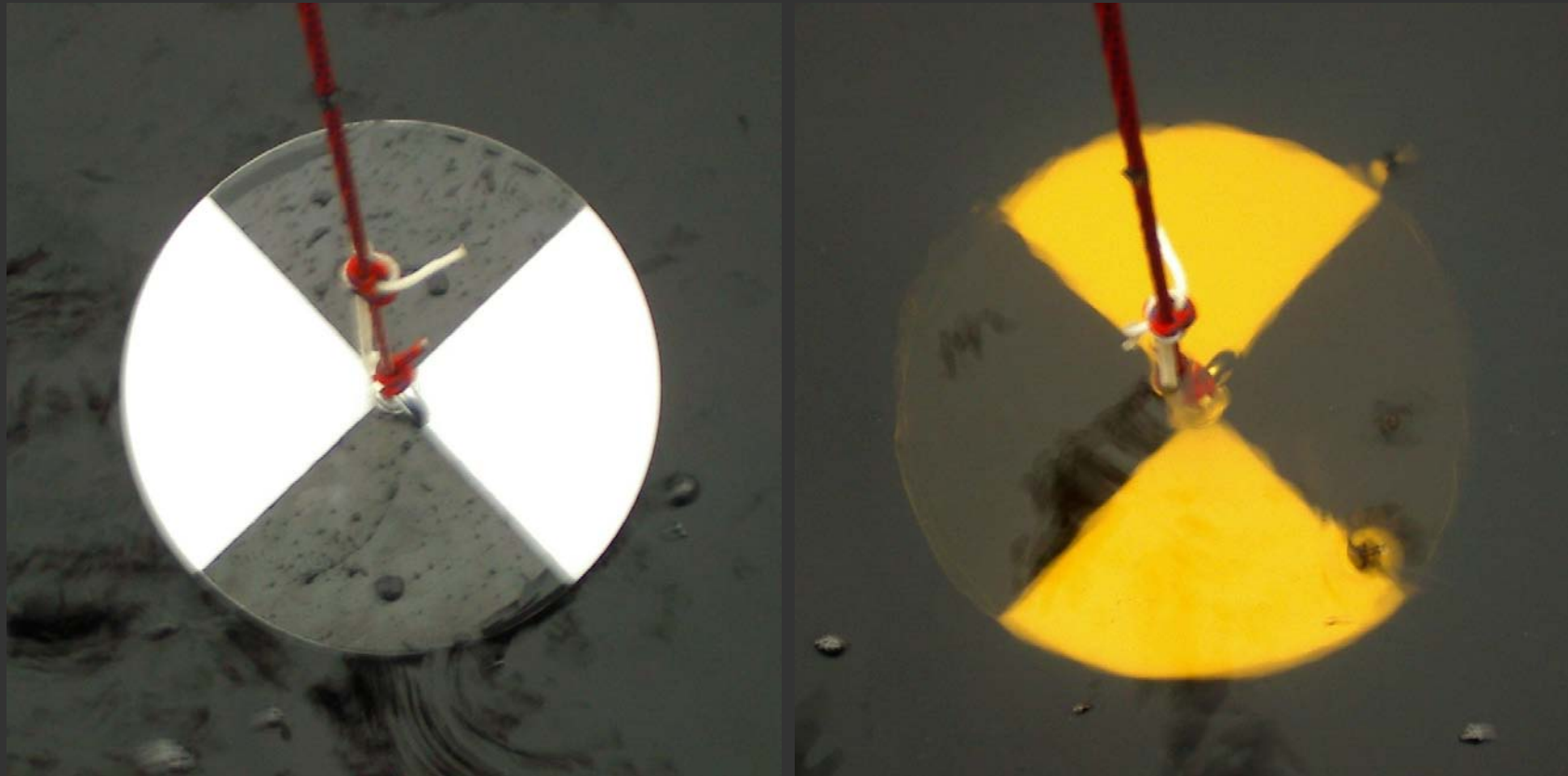




# (Global) Ocean Colour Products



CDOM dominates the absorption properties of Tasmania's coastal waters



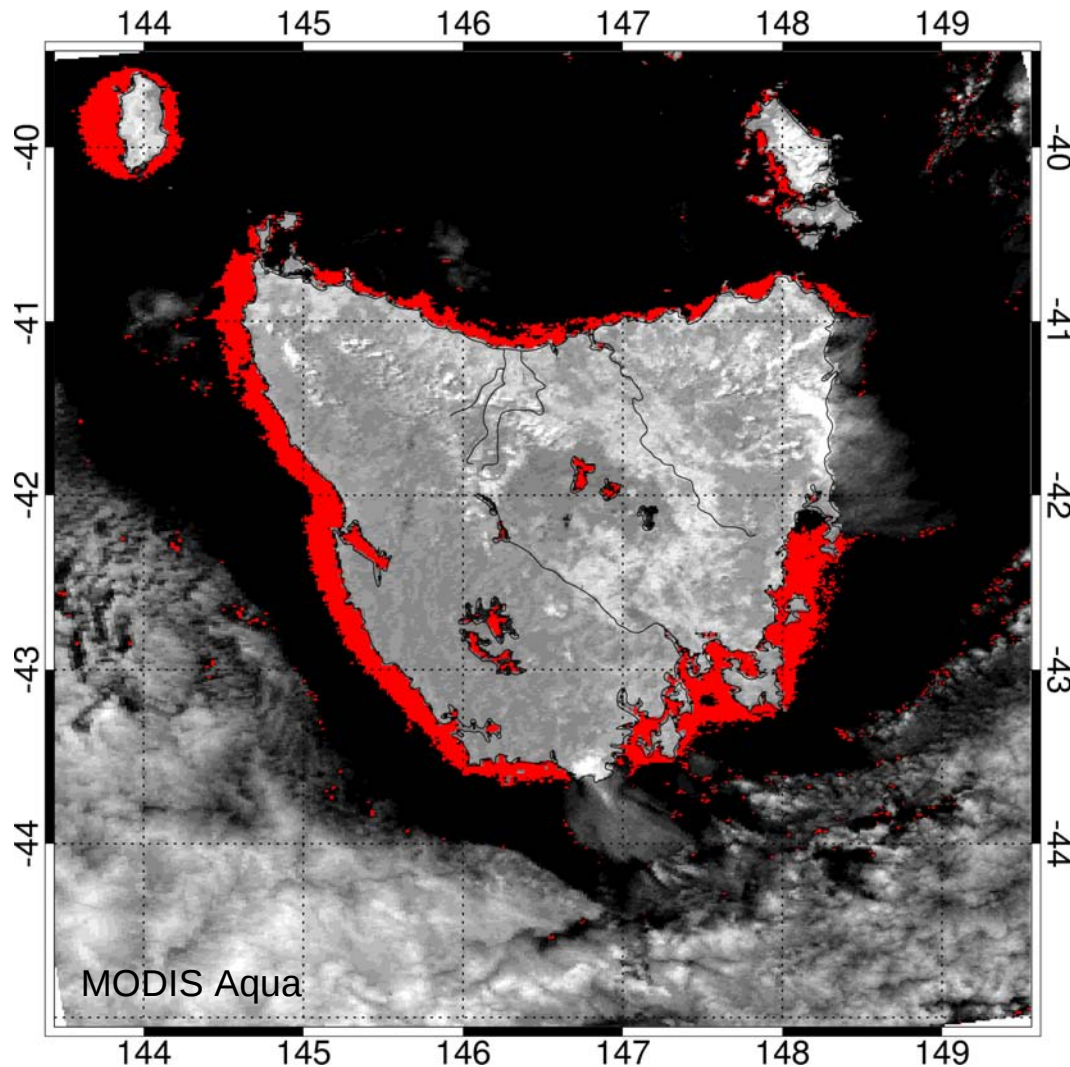
2007 field work: Absorption coefficients of up to  $5.2 \text{ m}^{-1}$  at 440 nm within several meters thick surface layers



# NIR algorithm failure for Tasmanian region

Example: Tasmania – 13 October 2003

412 – 678 nm



■ = Atmospheric correction failure

# CSIRO's algorithm approach optically deep waters

## Step 1: Atmospheric correction approach

emphasis on coastal waters - based on inverse modelling of radiative transfer simulations and Artificial Neural Network (ANN) inversion

(Schroeder et al., 2007a, 2007b, IJRS)

## Step 2: Water constituent retrieval algorithm

based on Linear Matrix Inversion (LMI) of a semi-analytical model with a variable Specific Inherent Optical Property (SIOP) parameterization

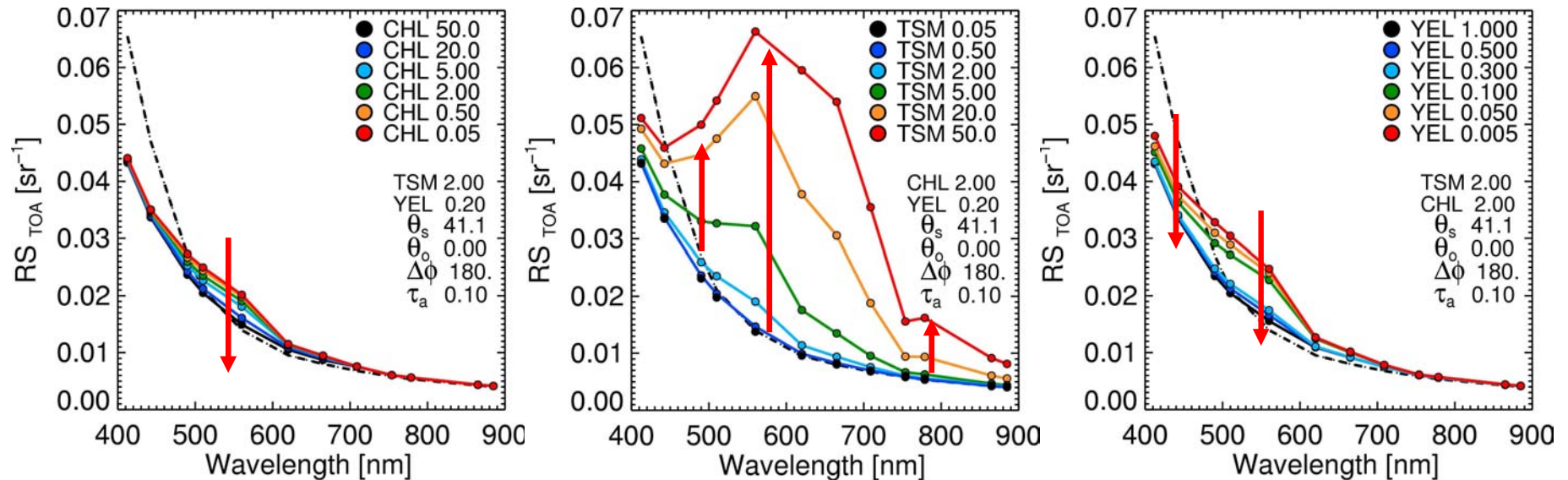
(Brando et al., 2008, Ocean Optics XIX)



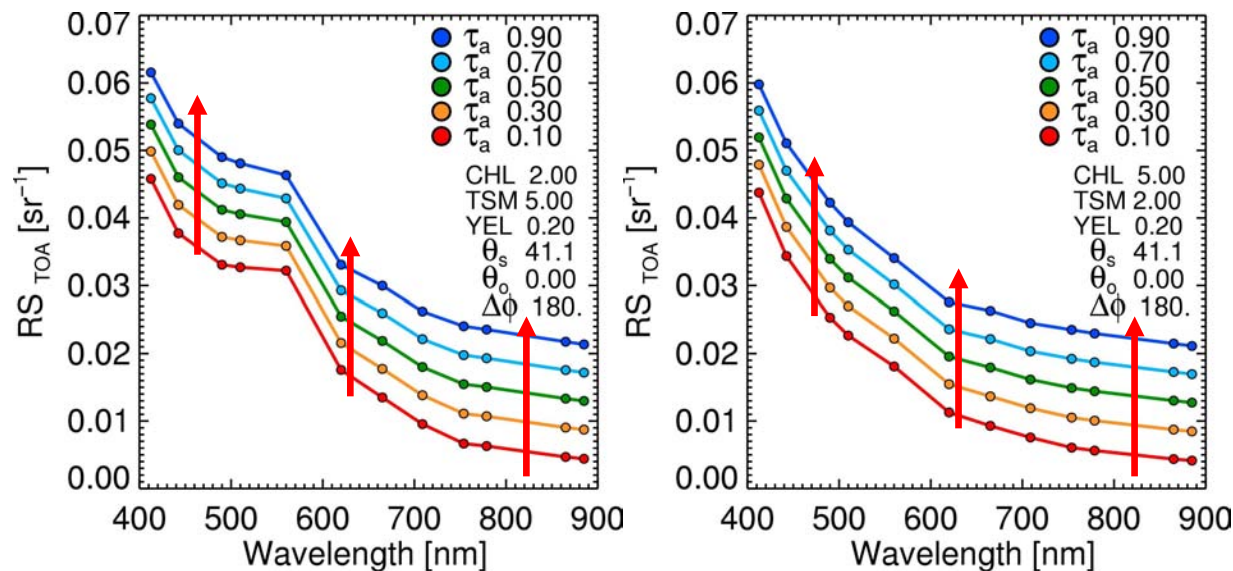
# Simulated Top-of-Atmosphere Spectra

## Coastal waters

... depending on variations of the water constituents

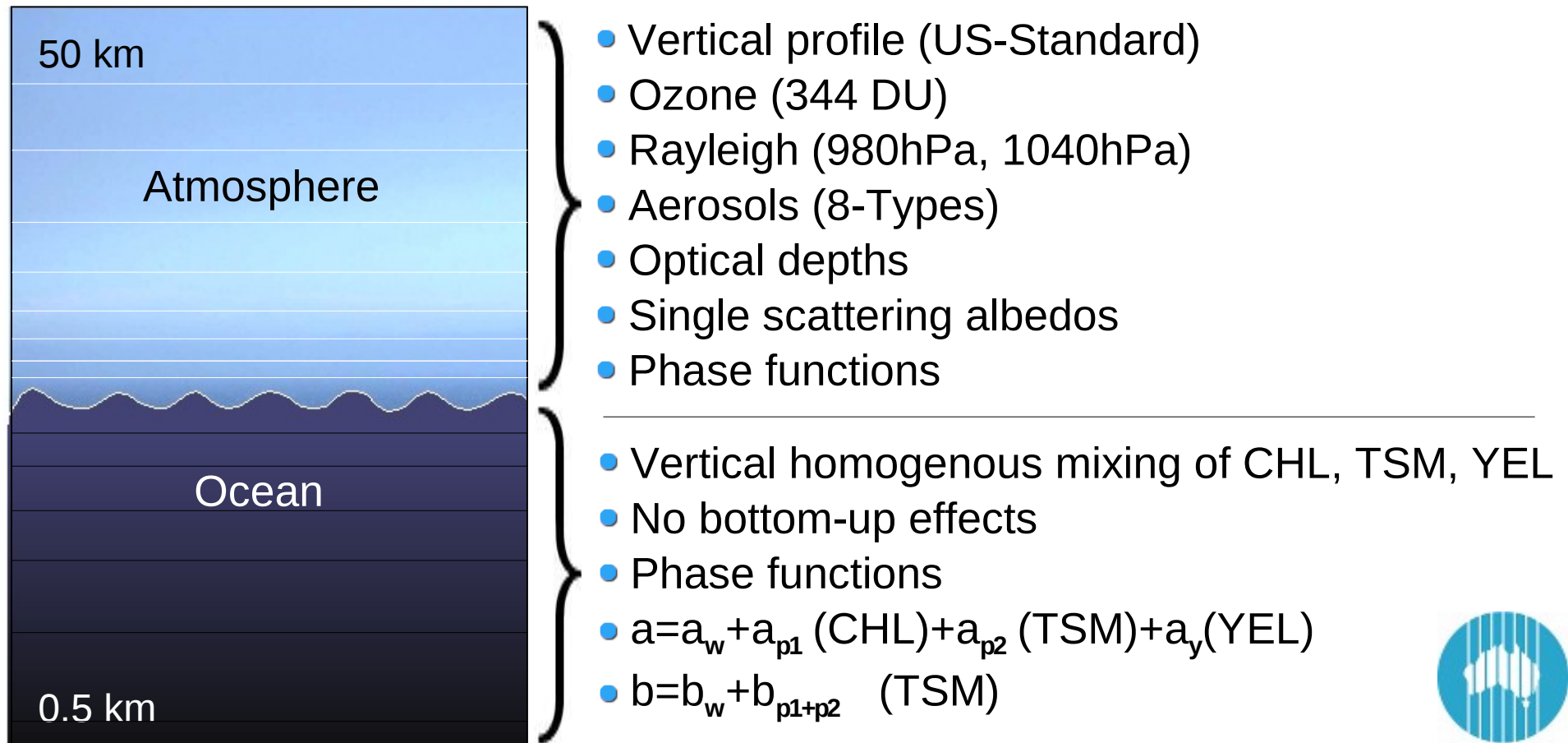


... depending on variations of the type and concentration of the aerosol



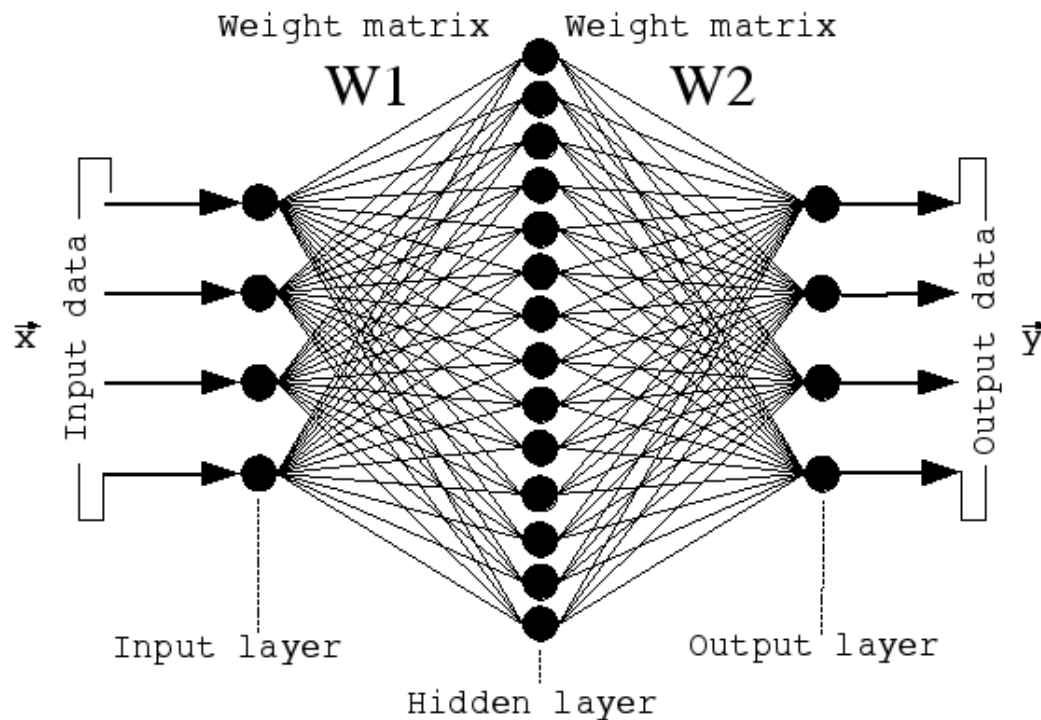
# Atmospheric correction – Forward model

- Coupled radiative transfer model based on matrix-operator method (FUB)
- Simulates the upward azimuthally resolved radiance field (TOA & BOA)
- Variety of different sun and observing geometries and as well as different types of atmospheric and oceanic constituents



# Atmospheric correction – Inverse model

- ANN: multi-layer perceptron



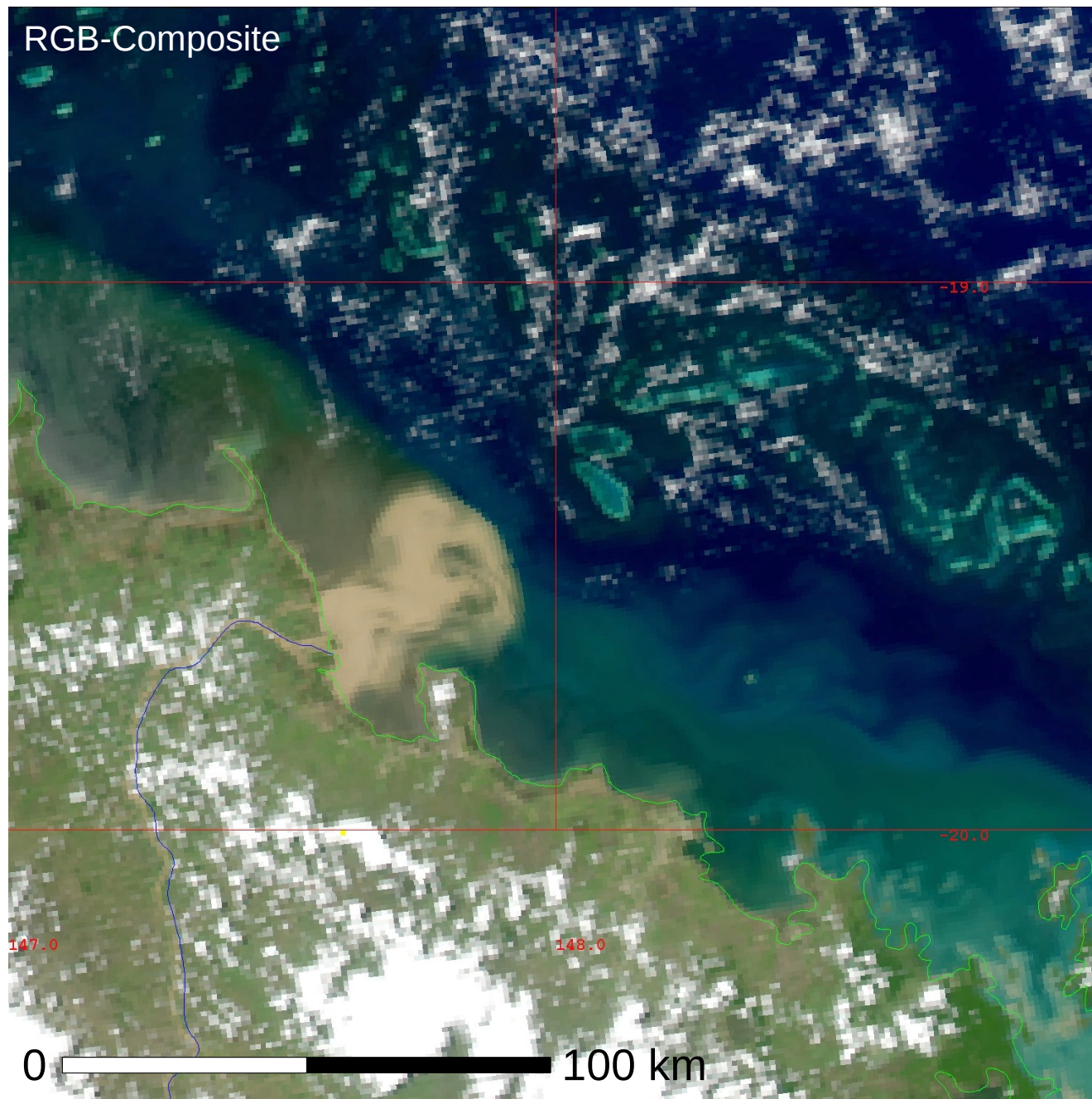
$$\vec{y} = S \{ W2 \cdot S(W1 \cdot \vec{x}) \}$$
$$S(x) = (1 + e^{-x})^{-1}$$

- The network's free parameters (weights) are adapted during a supervised learning procedure
- Training data set: 200.000 randomly selected vectors from the simulated data base (Gaussian signal-dependent noise, PCA)



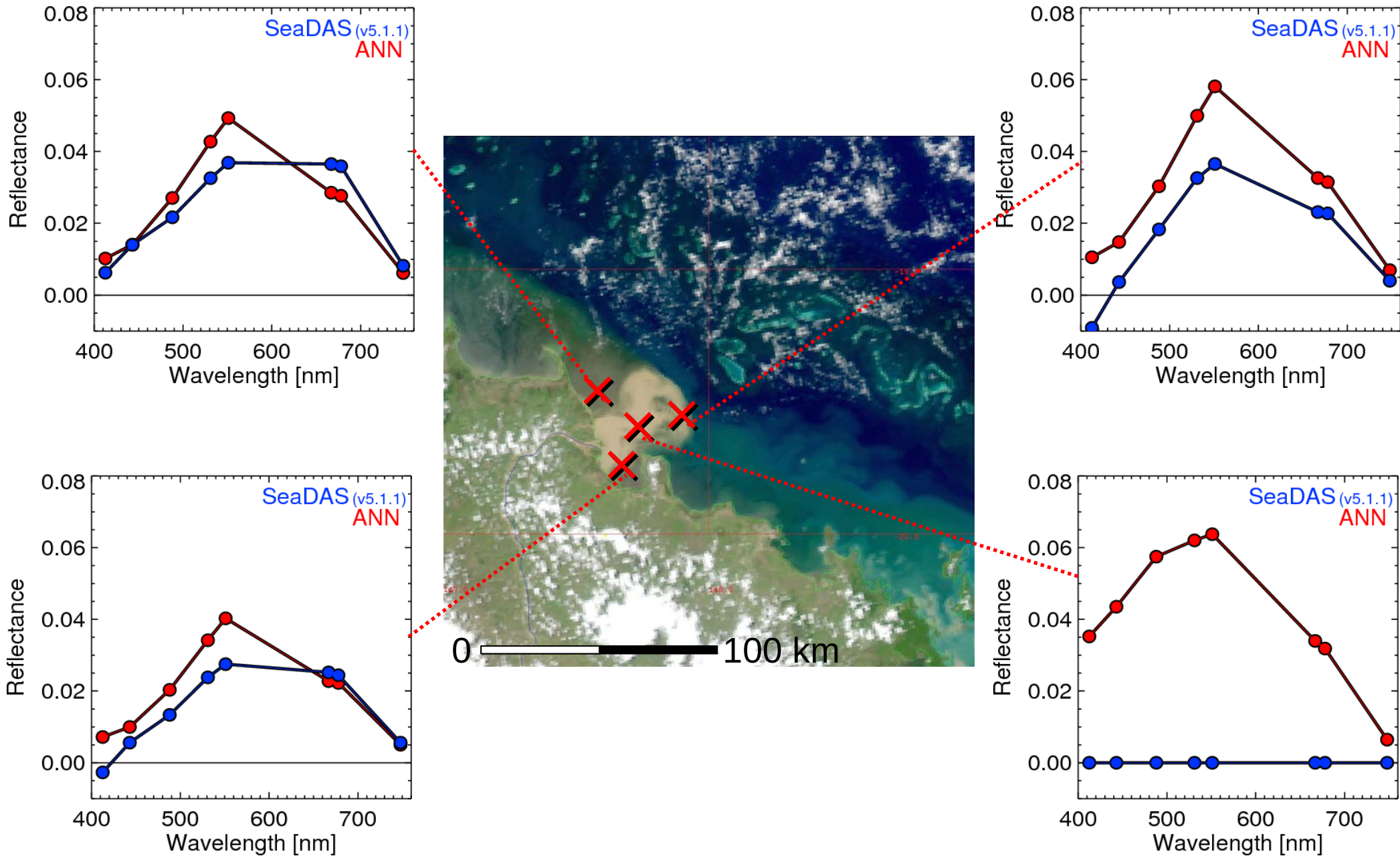
# Application atmospheric correction – MODIS-Aqua

## Extreme 1: Burdekin, QLD – 22 February 2008



# Application atmospheric correction MODIS

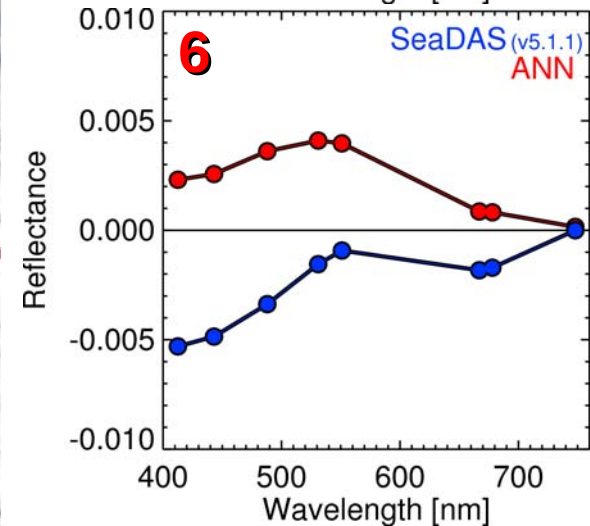
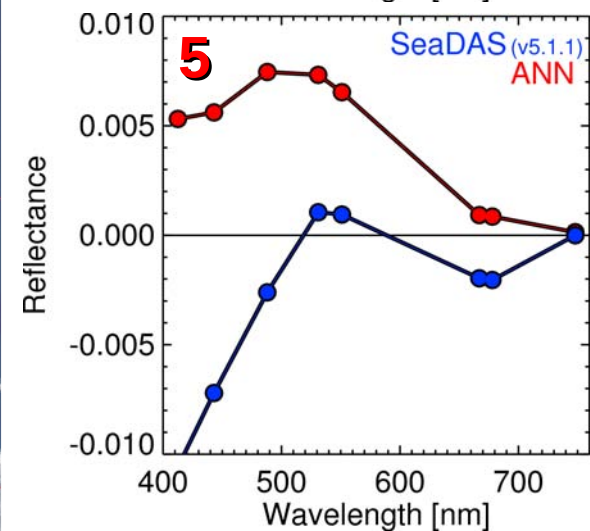
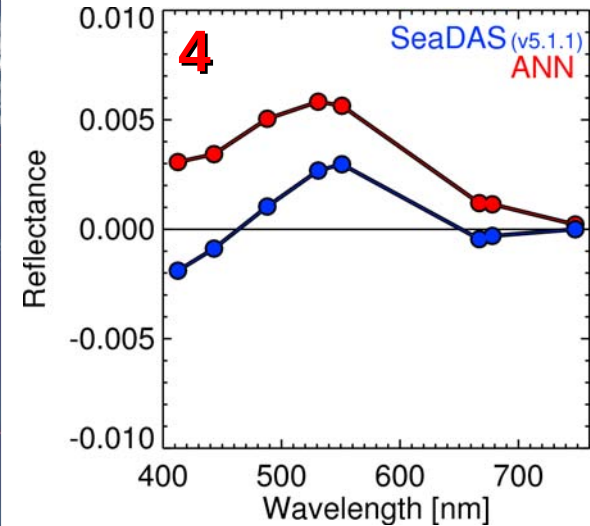
## Extreme 1: Burdekin, QLD – 22 February 2008





# Reflectance spectra

Example: Tasmania – 13 October 2003





# Atmospheric correction validation MODIS

Match-up data base with more than 130 in-situ reflectance measurements (Trios RAMSES, Simbada)

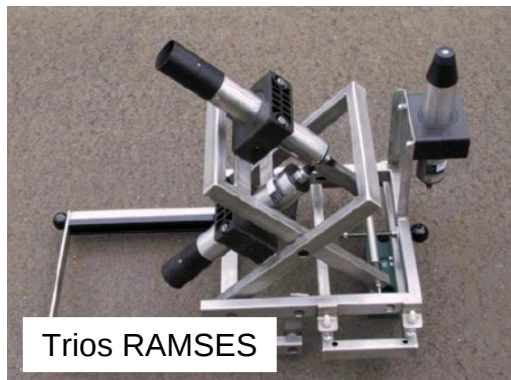


Image: Revamp protocols



Image: LOA Université de Lille

CSIRO, GKSS, MUMM – mainly coastal waters

Data quality requirements:

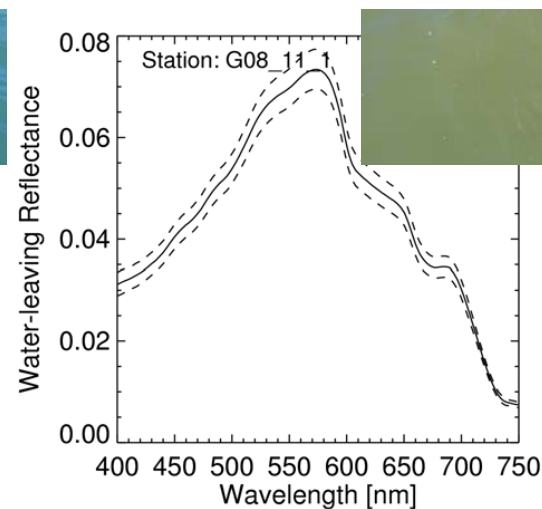
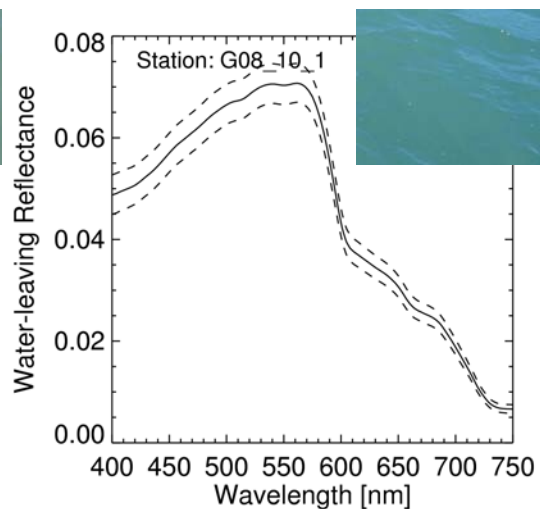
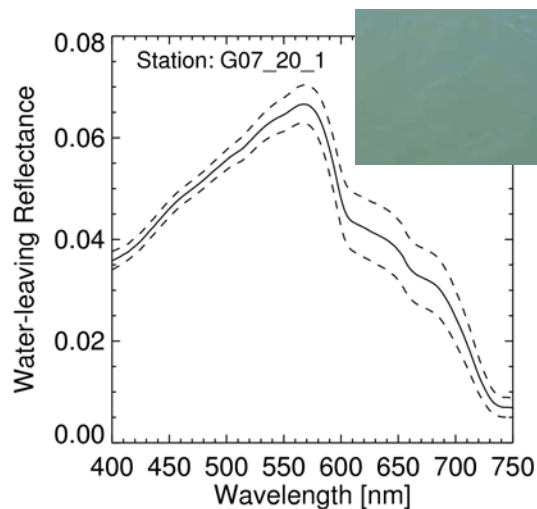
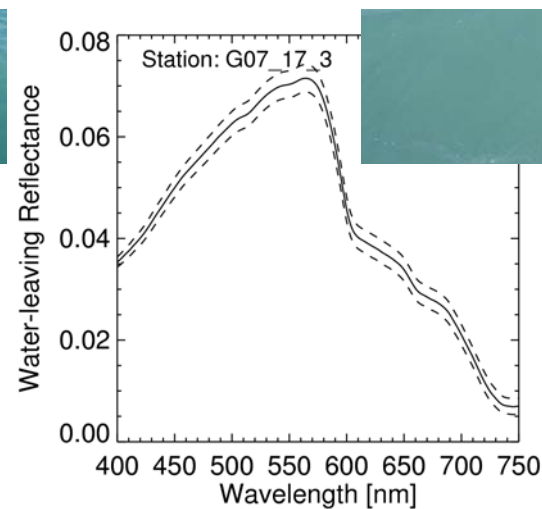
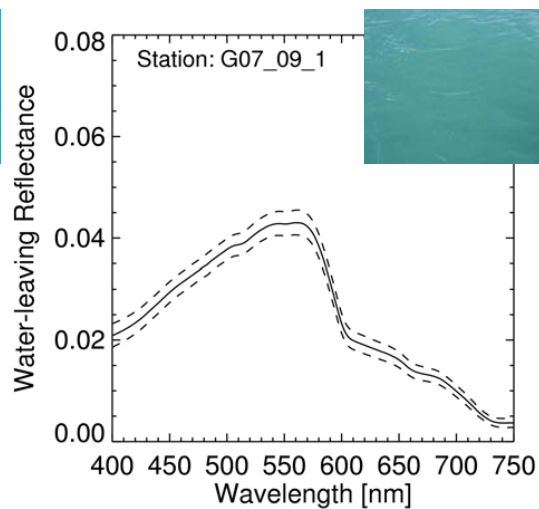
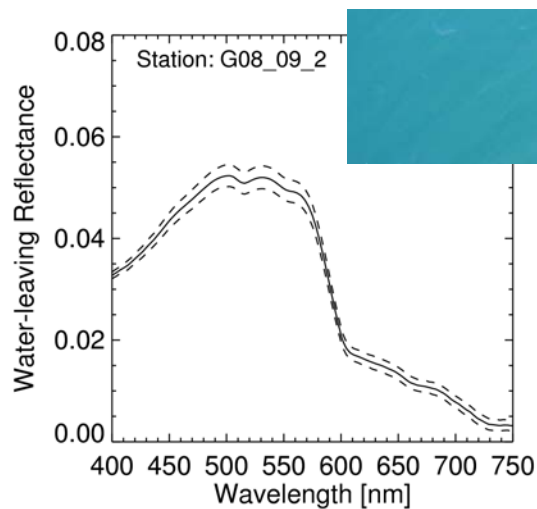
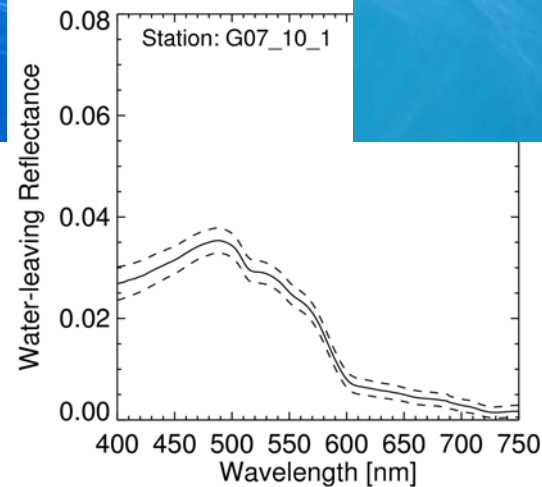
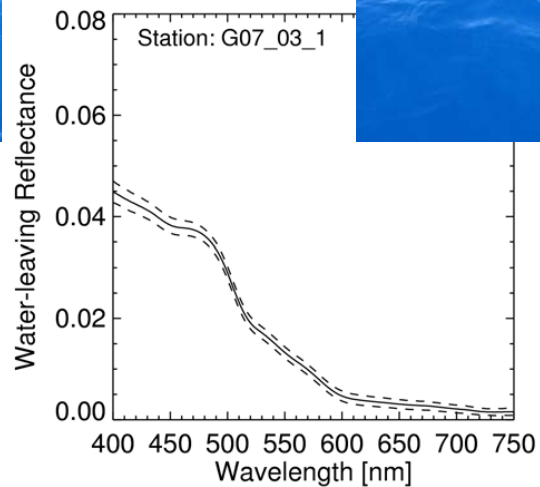
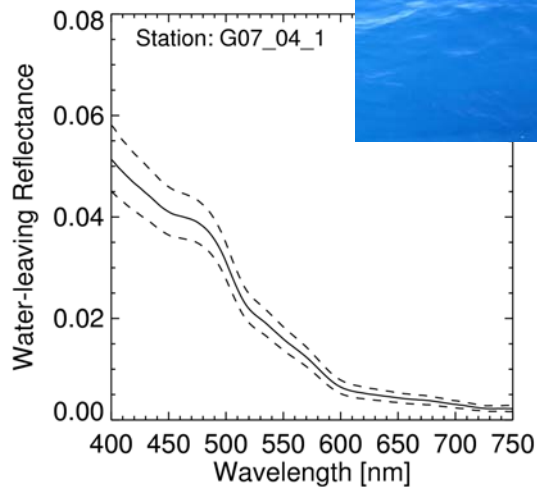
Maximum of  $\pm 60$  min to satellite over pass

Median from an unflagged (land, sun glint, cloud) 3x3 pixel box

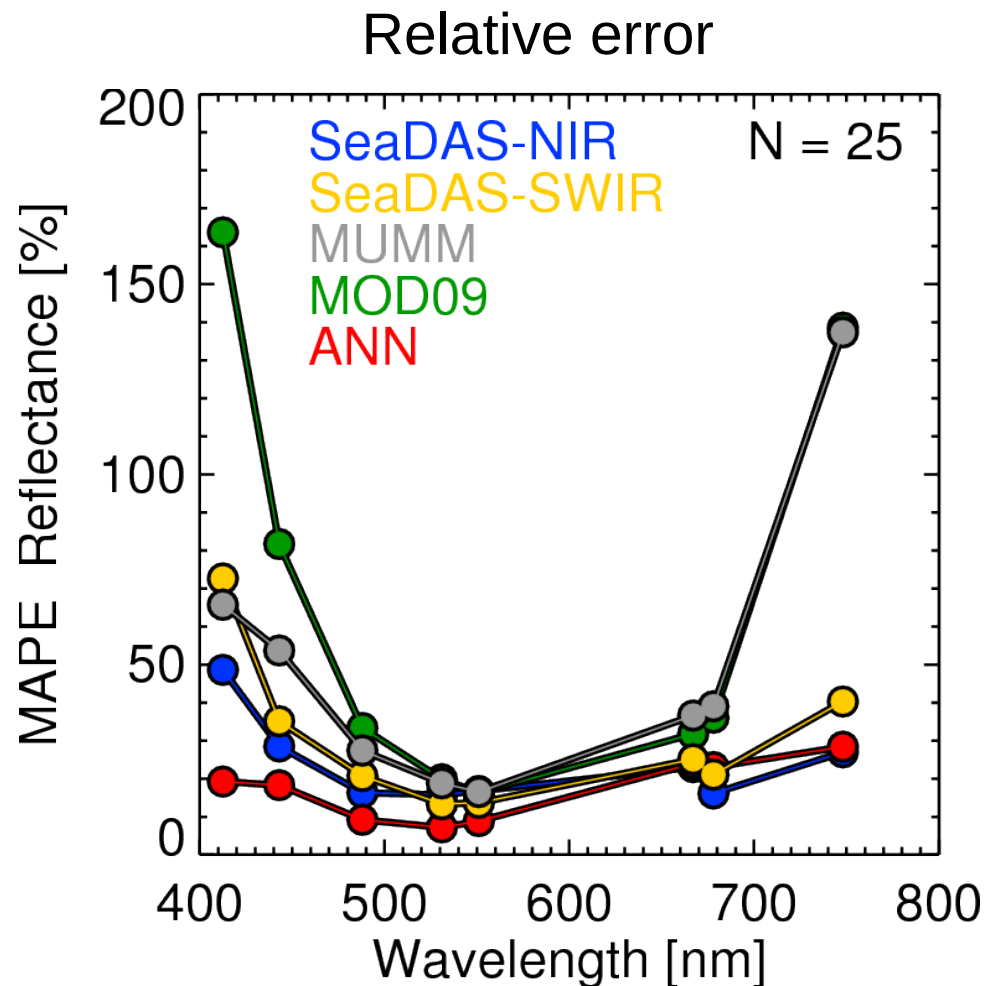
Inputs checked with respect to simulation ranges



# Reflectance Spectra Fieldwork GBR 2007/08



# Atmospheric correction inter-comparison vs in-situ SeaDAS-NIR, -SWIR, MOD09, ANN, MUMM



Spectral range [nm]

[412] [412-748]

**MOD09**

**163%**

**65%**

**MUMM**

**73%**

**49%**

**SeaDAS-SWIR**

**66%**

**30%**

**SeaDAS-NIR**

**48%**

**24%**

**ANN**

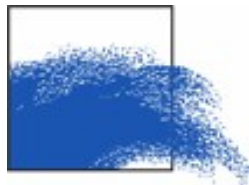
**19%**

**17%**

Analysis ongoing of MUMM's full reflectance  
data base >750 spectra

Match-ups:  $71 \pm 60$  min,  $50 \pm 30$  min

(Schroeder et al, 2010 in prep.)





# Overview water constituents retrieval

## Linear Matrix Inversion (LMI)

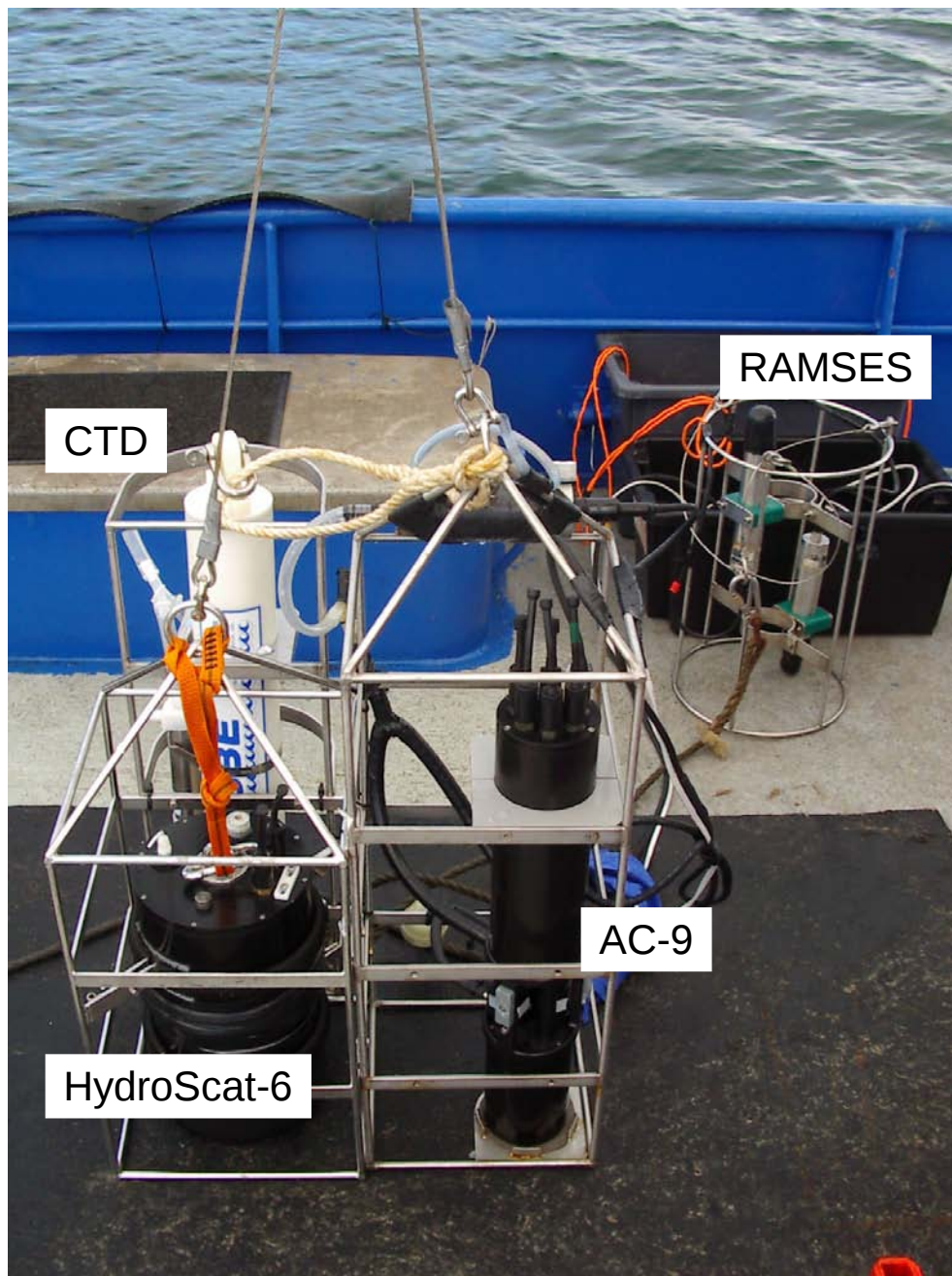
$$R \simeq g_1 \left( \frac{b_b}{(b_b + a)} \right) + g_2 \left( \frac{b_b}{(b_b + a)} \right)^2$$

(Gordon 1988)

$$\begin{aligned} a_{\text{tot}}(\lambda) &= a_w(\lambda) + a_{\text{PHY}}(\lambda) + a_{\text{NAP}}(\lambda) + a_{\text{CDOM}}(\lambda) \\ &= a_w(\lambda) + \sum a_j^*(\lambda) C_j \end{aligned}$$

$$\begin{aligned} b_{b, \text{tot}}(\lambda) &= b_{b, w}(\lambda) + b_{b, \text{PHY}}(\lambda) + b_{b, \text{NAP}}(\lambda) \\ &= b_{b, w}(\lambda) + \sum b_{b, j}^*(\lambda) C_j \end{aligned}$$

# In-situ collection of optical properties

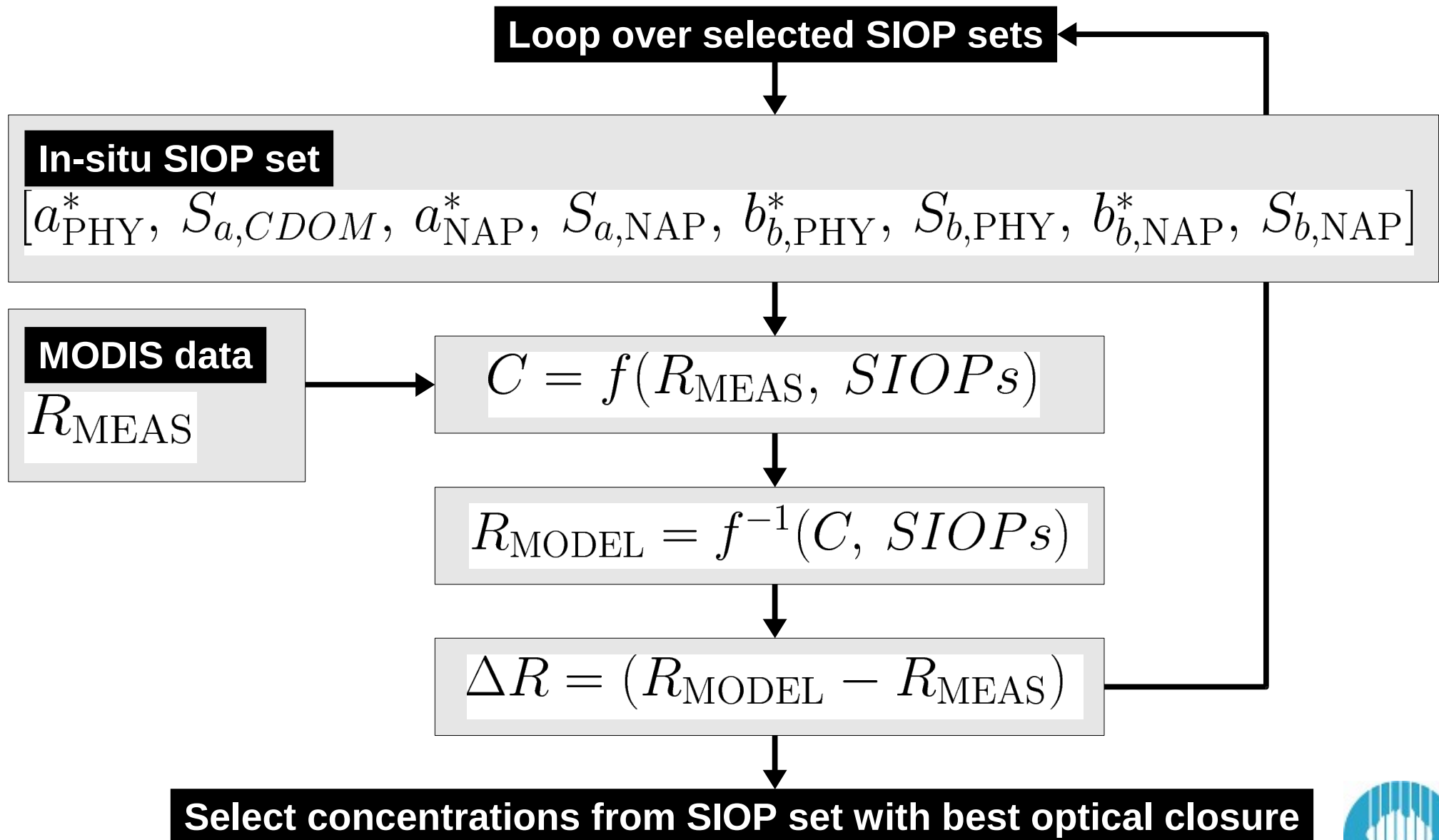


- Absorption of phytoplankton (Lab CMAR)
- Absorption of non-algal particulate matter (Lab CMAR)
- Absorption of CDOM (Lab CMAR)
- TSM, CHL (Lab CMAR)
- Particulate back-scattering (HydroScat-6)
- Dissolved plus particulate absorption and attenuation (AC-9)
- Temperature, salinity (CTD)
- Optics (RAMSES-Trios)



# Principle water constituents retrieval

## Linear Matrix Inversion (LMI)



# Validation summary in-water algorithm

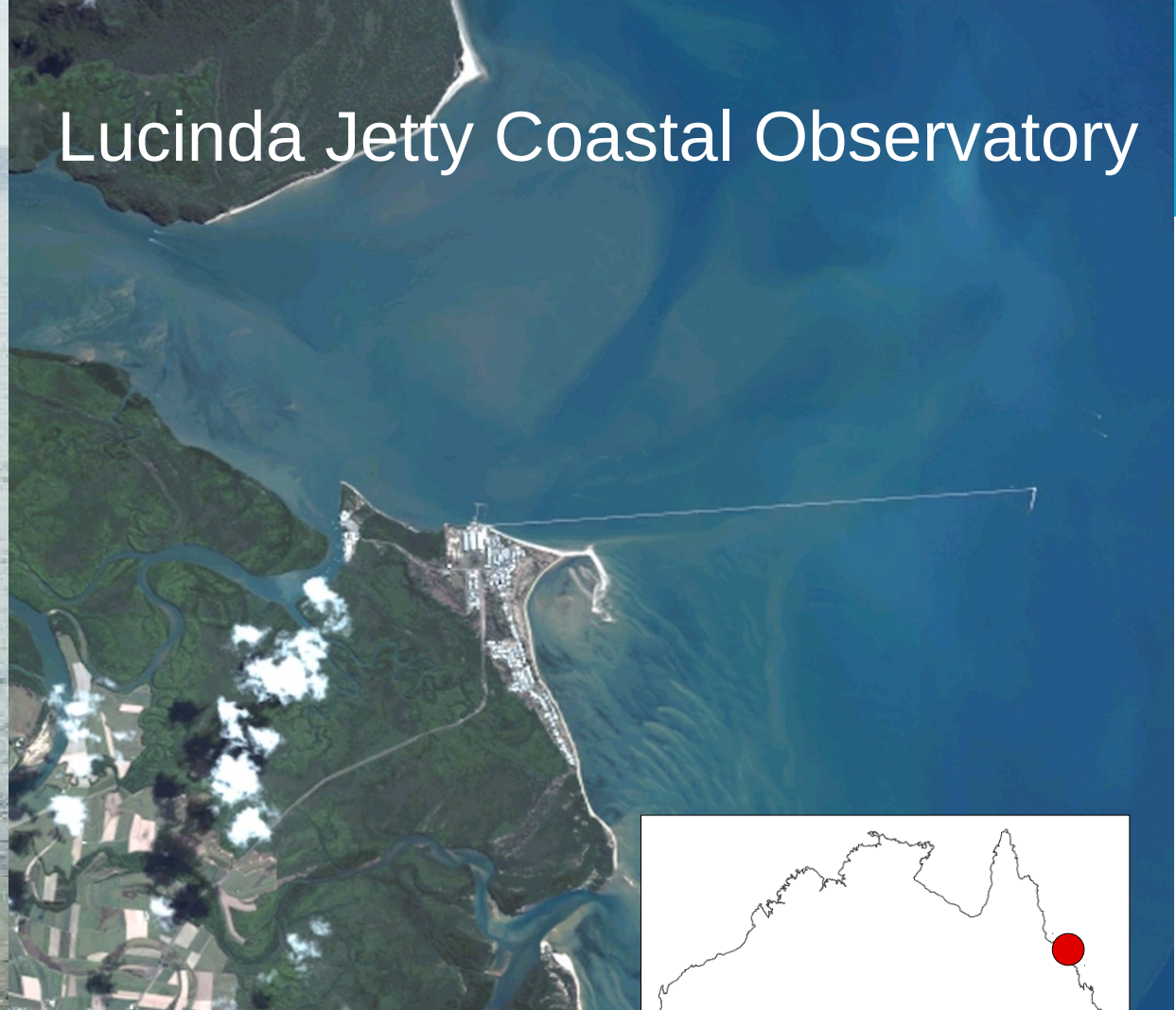
CHL	CSIRO-LMI 59%	vs	SeaDAS-GSM 90%	N=110
TSM	CSIRO-LMI 57%	vs	SeaDAS-Clark 61%	N= 33
CDOM	CSIRO-LMI 67%	vs	SeaDAS-QAA 216% (Incl. detritus)	N= 27



# Error sources

- Mixing ratio of the water constituents
- Temporal difference between sampling and satellite overpass
- Inhomogenous horizontal and vertical distribution of the water constituents within a satellite pixel (Comparison  $\sim 1\text{km}^2$  pixel with a sample of a few liters)
- Remaining errors of the atmospheric correction  $\sim 20\%$
- Errors associated with the in-situ measurements
- Reproduction of measurements, differences between the Labs.  
(Revamp Inter-Calibration Report, 2002)  
CHL 10%, TSM 50%, YEL 50%
- Not considered physical effects
- Natural variability of the water constituents in the framework of an average bio-optical model

# Lucinda Jetty Coastal Observatory

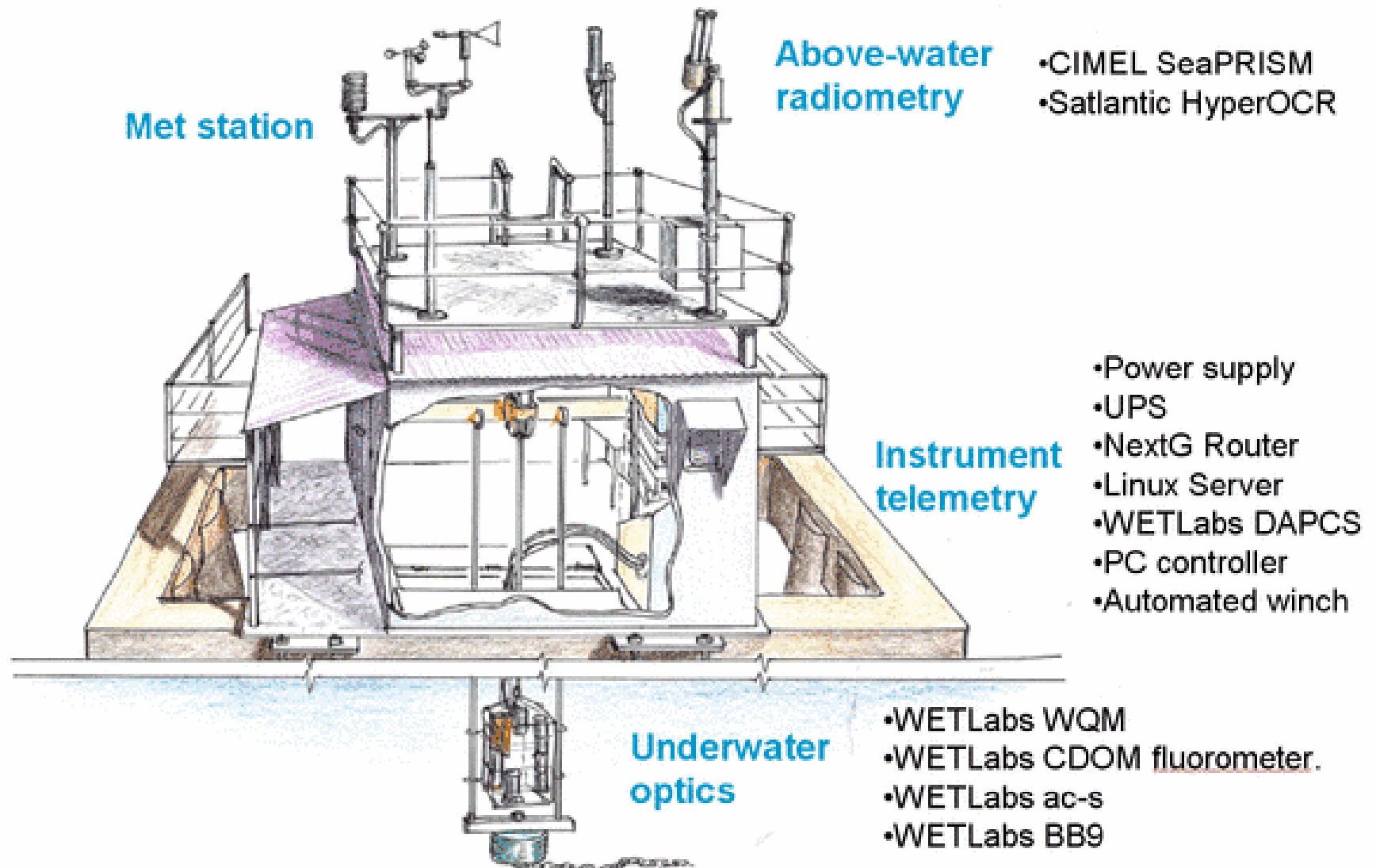


*(PI Vittorio Brando)*





# Lucinda Jetty Coastal Observatory



(PI Vittorio Brando)

# Remote sensing applications

Water quality monitoring, trend analysis & reporting

Compliance monitoring (Marine Monitoring Program)

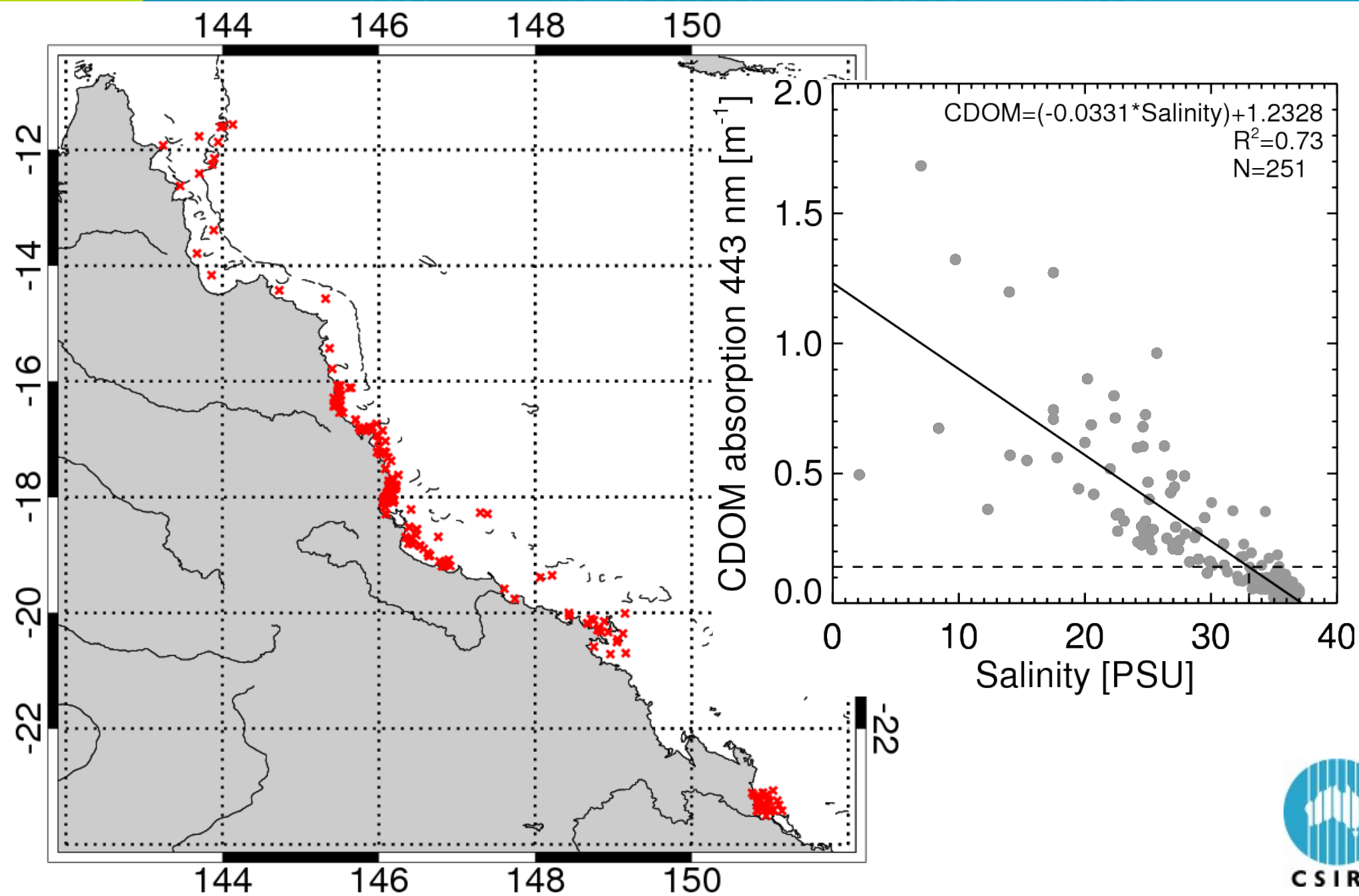
Algal bloom detection

Provide data assimilation inputs bio-geochemical models

Flood (fresh water) plume mapping



# Application – mapping fresh water plumes GBR





Photos courtesy Dr Michelle Devlin

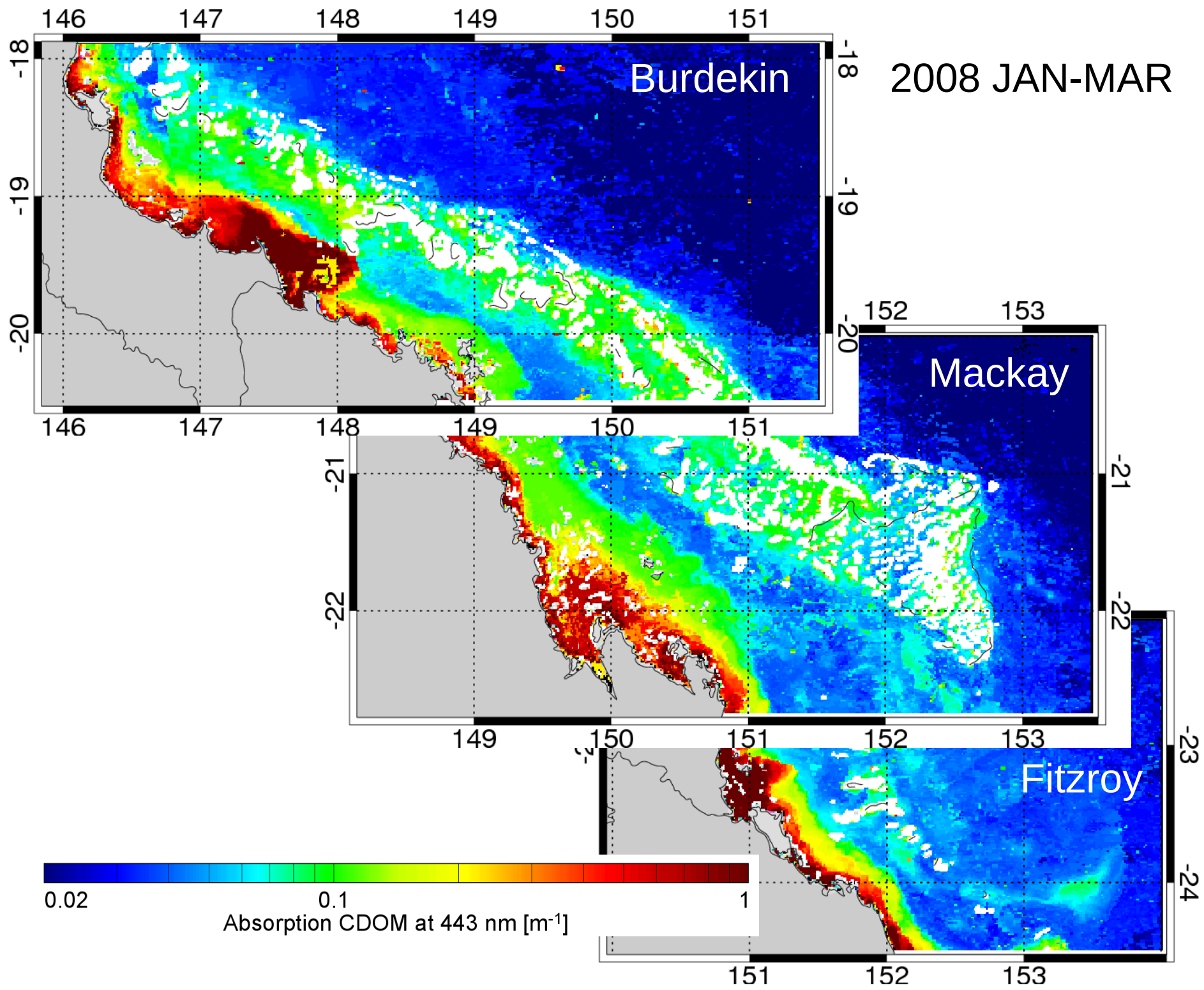


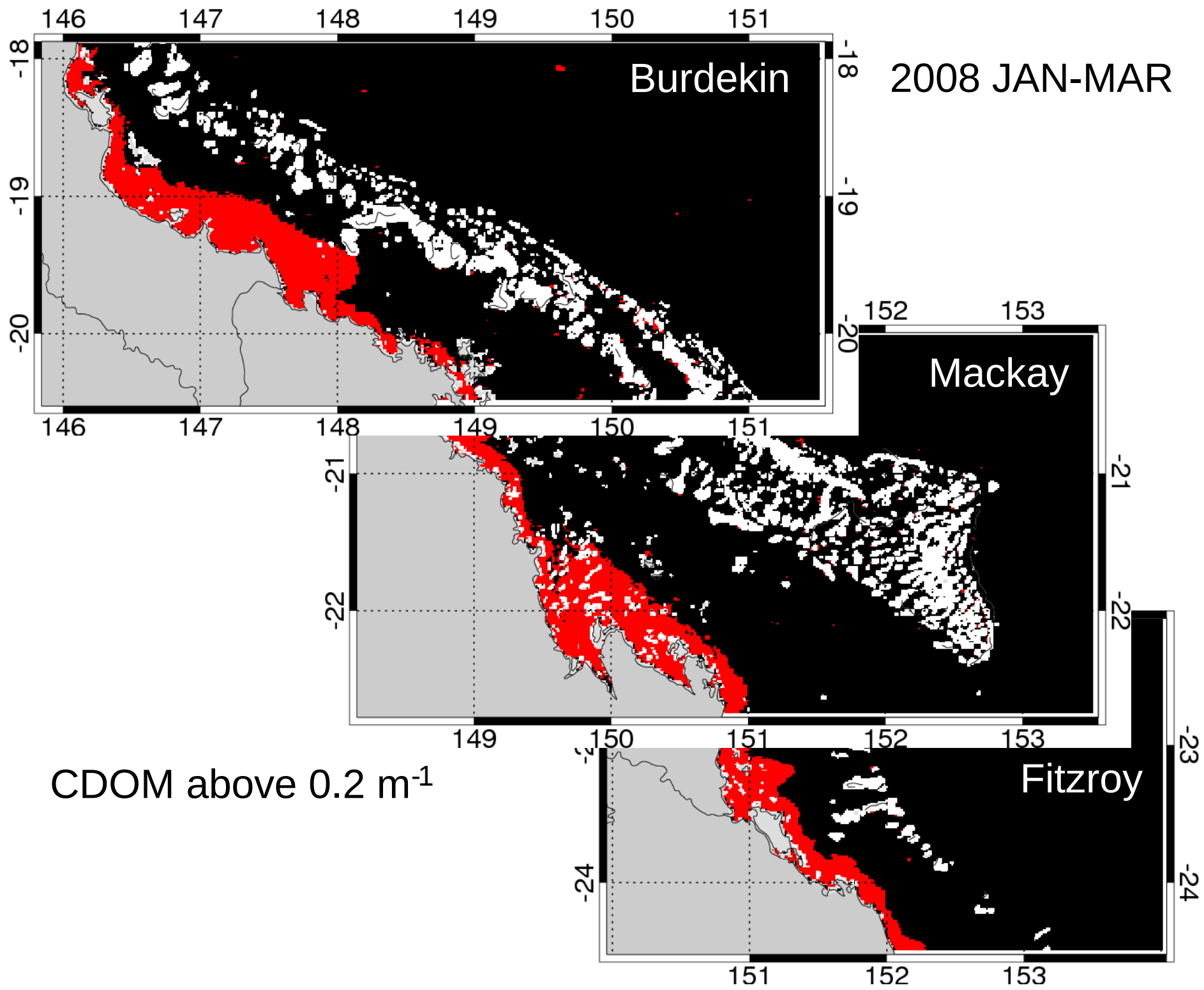


Image credit: <http://www.lib.utexas.edu>

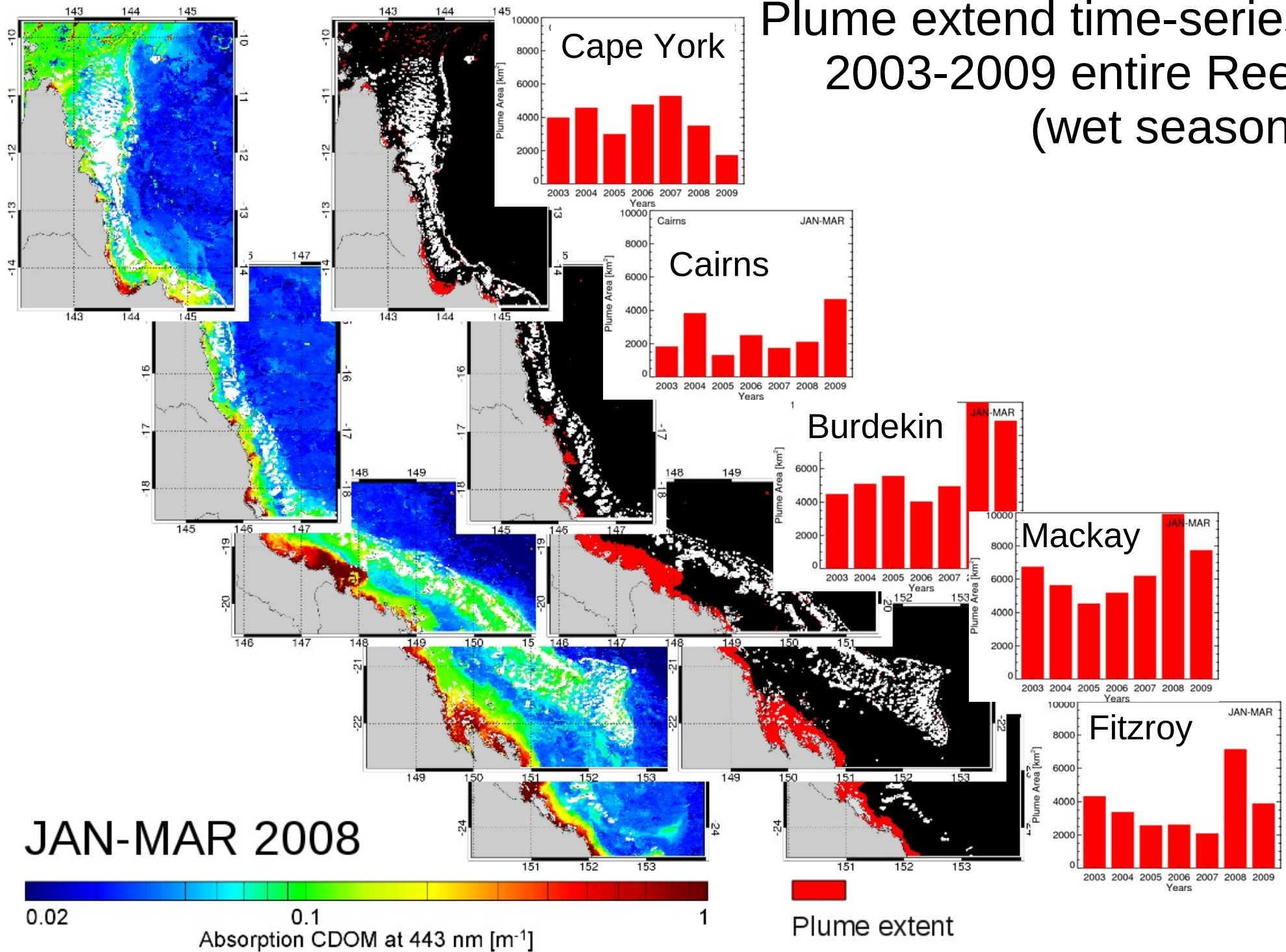
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# Plume extend time-series 2003-2009 entire Reef (wet season)





# Acknowledgements

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Dr Michelle Devlin, James Cook University, Townsville, Australia

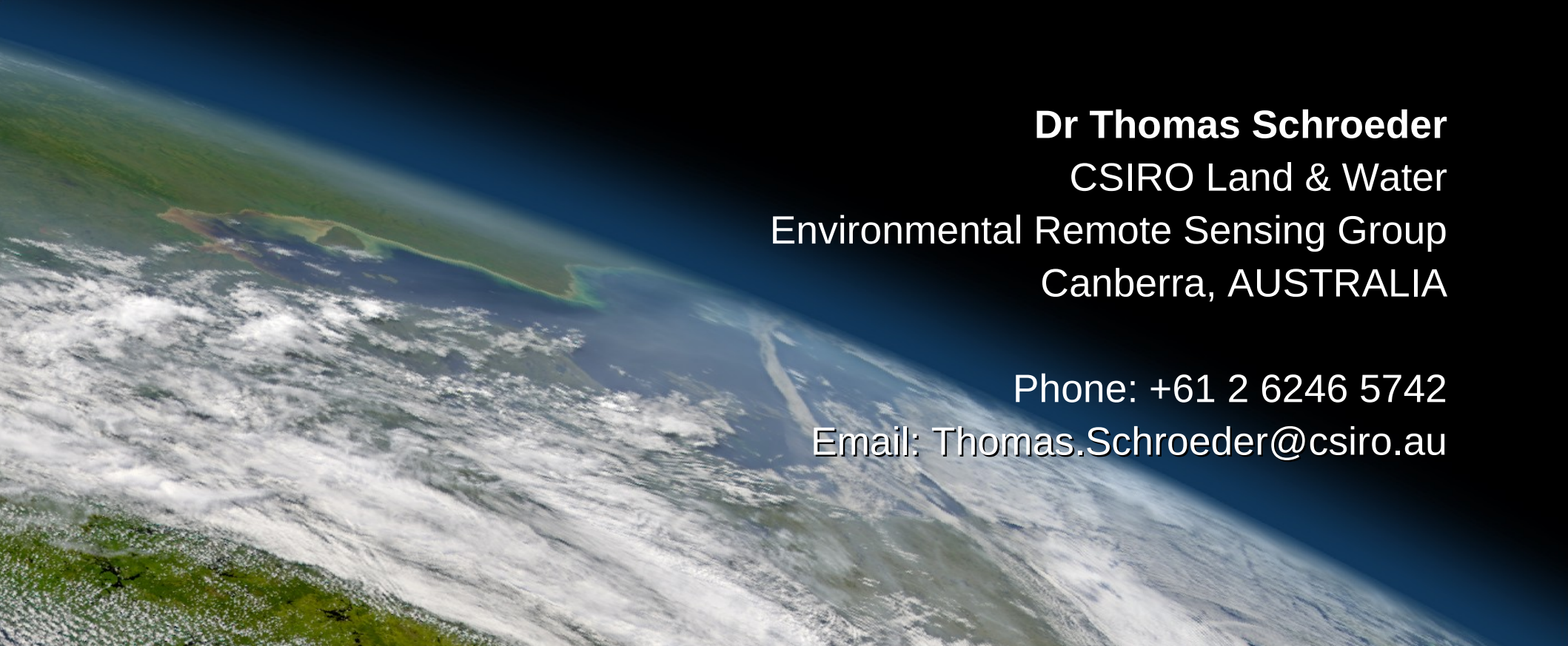
Dr Kadija Oubelkheir, CSIRO, Brisbane, Australia

Prof Juergen Fischer, Free University Berlin, Germany

Integrated Marine Observing System (IMOS)

CSIRO Wealth from Oceans Flagship

NASA & ESA



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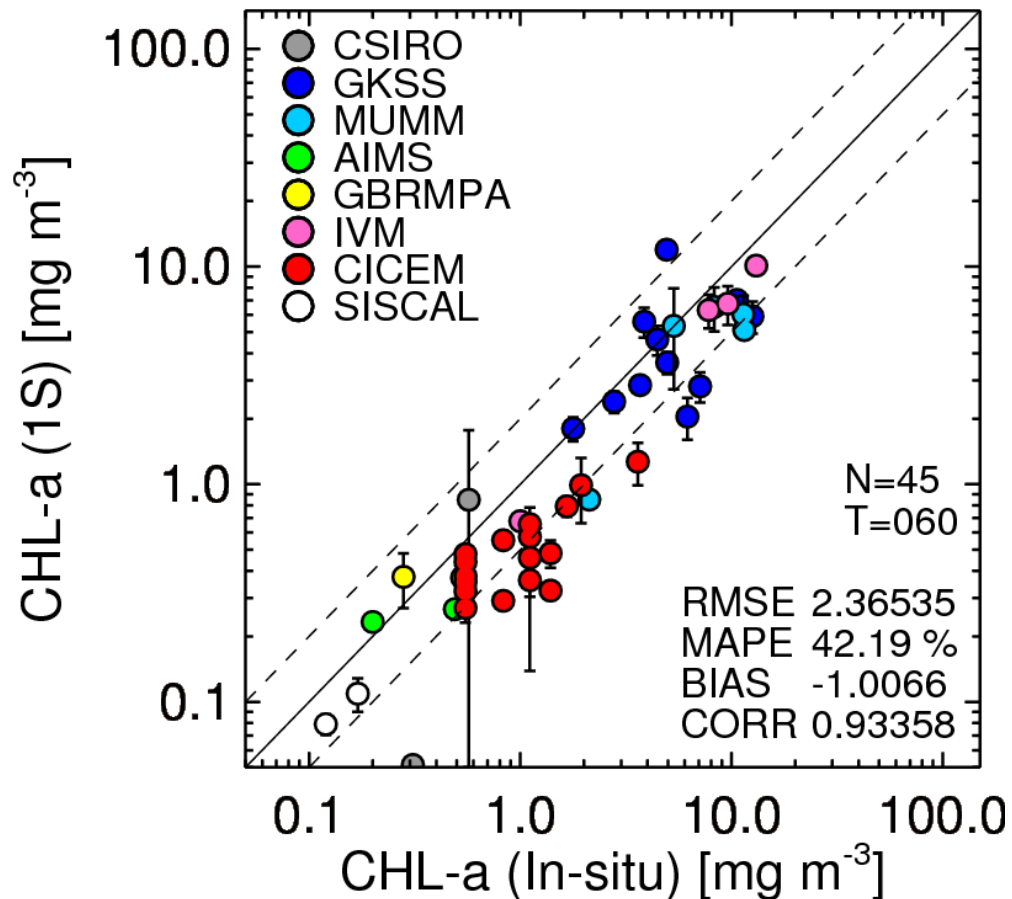
Thank you



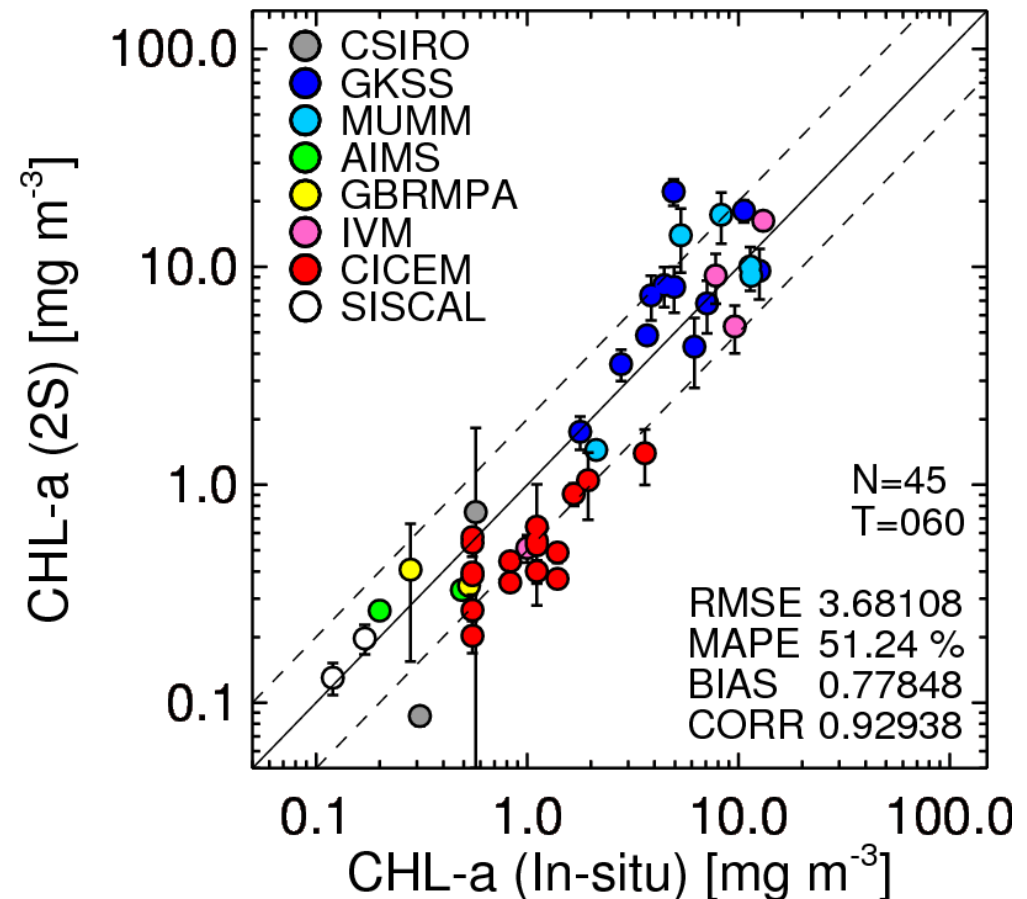
# Chlorophyll-a retrieval - MERIS

Data quality IPF 5.05 (MEGS 7.4.1)

1-step inversion  
(implicit atmospheric correction)



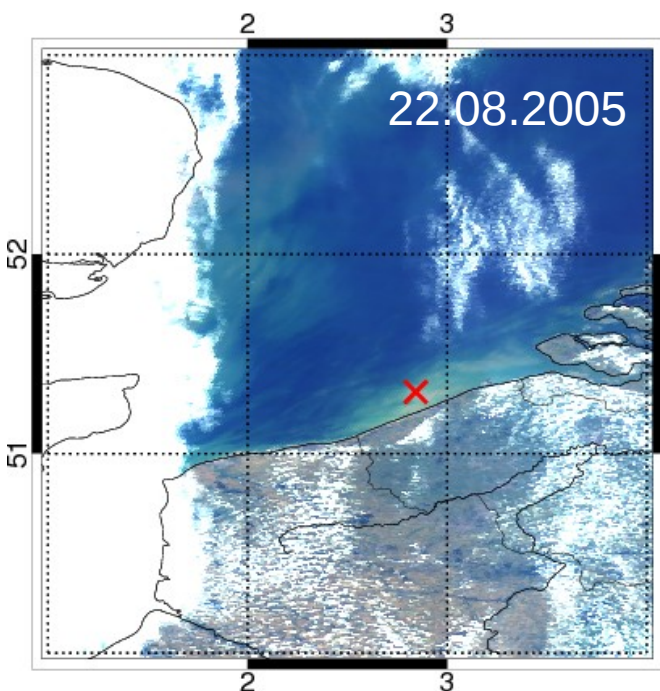
2-step inversion  
(explicit atmospheric correction)



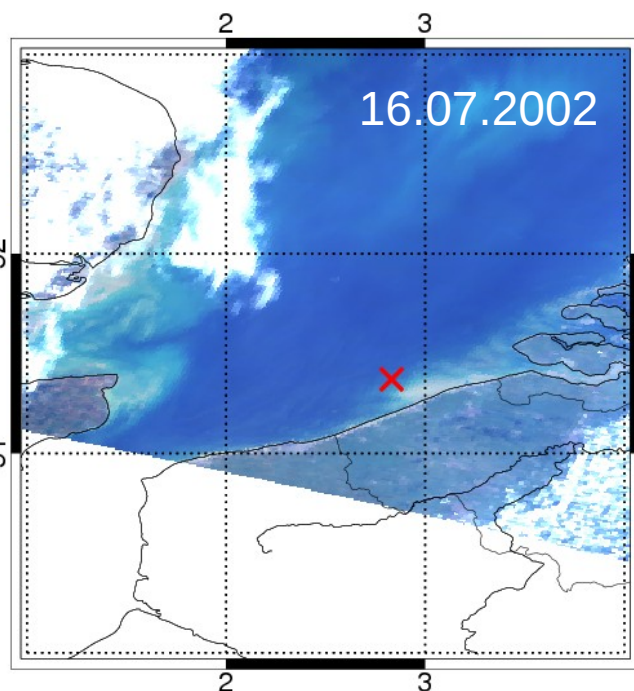


# MODIS match-up examples

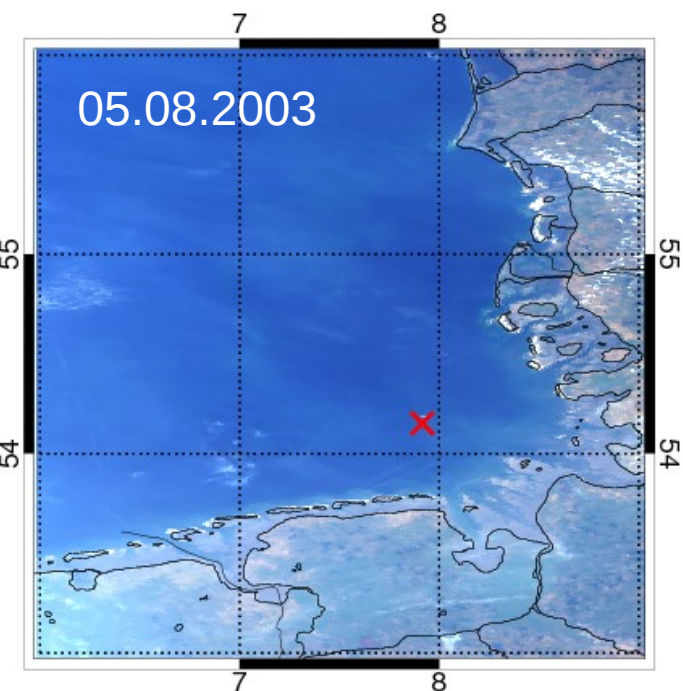
## North Sea



MODIS 2 A2005234.1055 - 20050822 - LAT: 51.309 LON: 2.842 - MUMM



MODIS 2 A2002197.1125 - 20020716 - LAT: 51.371 LON: 2.831 - MUMM



MODIS 2 A2003217.1030 - 20030805 - LAT: 54.152 LON: 7.915 - GKSS

