



# Monitoring Water Quality of Inland Lakes by Remotely Sensed Observations

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# Outline

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Sect. 1

General introduction

Sect.2

**Enhanced Three-band Index** for monitoring Chlorophyll-a

Sect.3

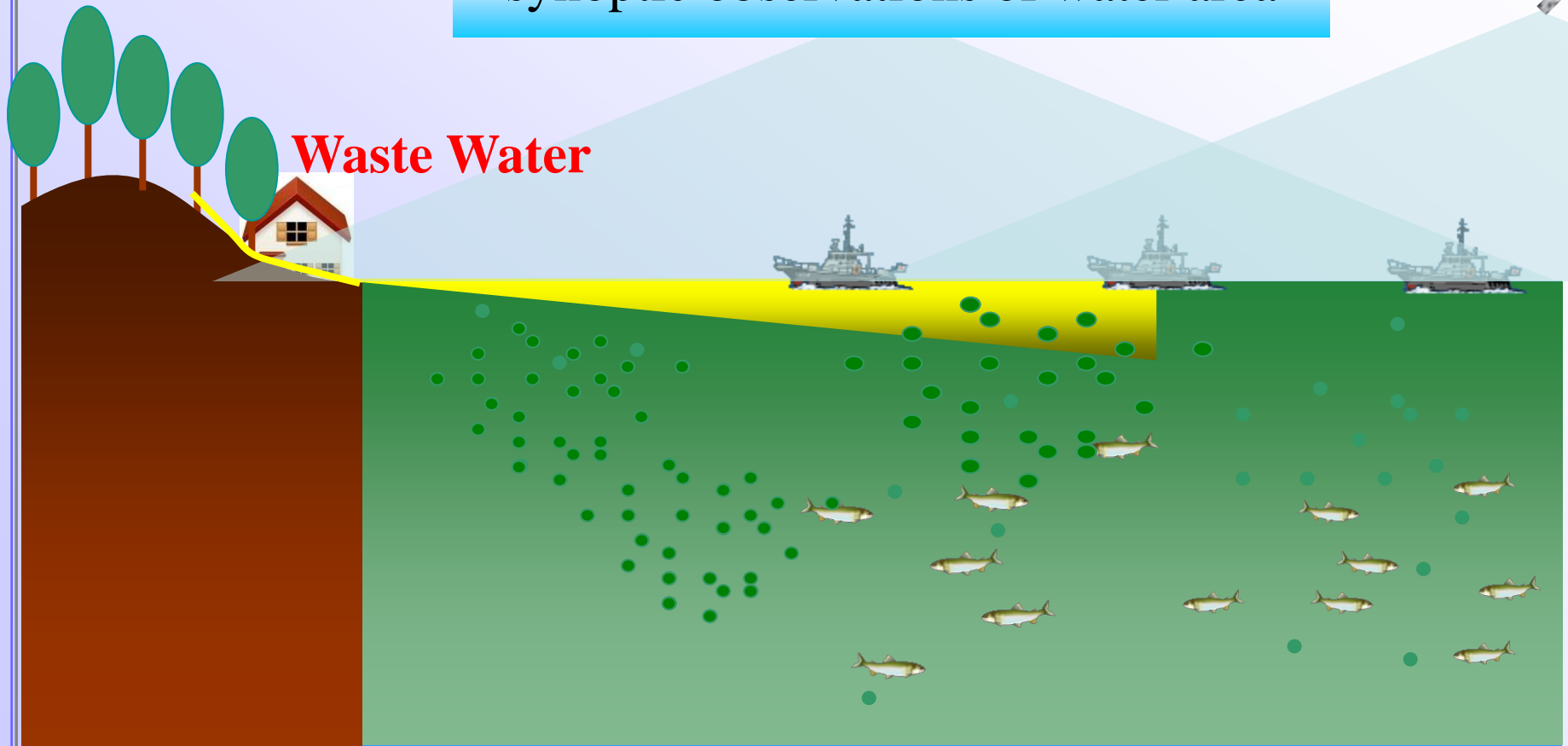
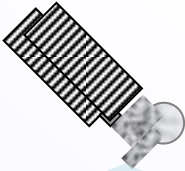
**SAMO-LUT algorithm** for monitoring water constituent concentrations

Sect. 4

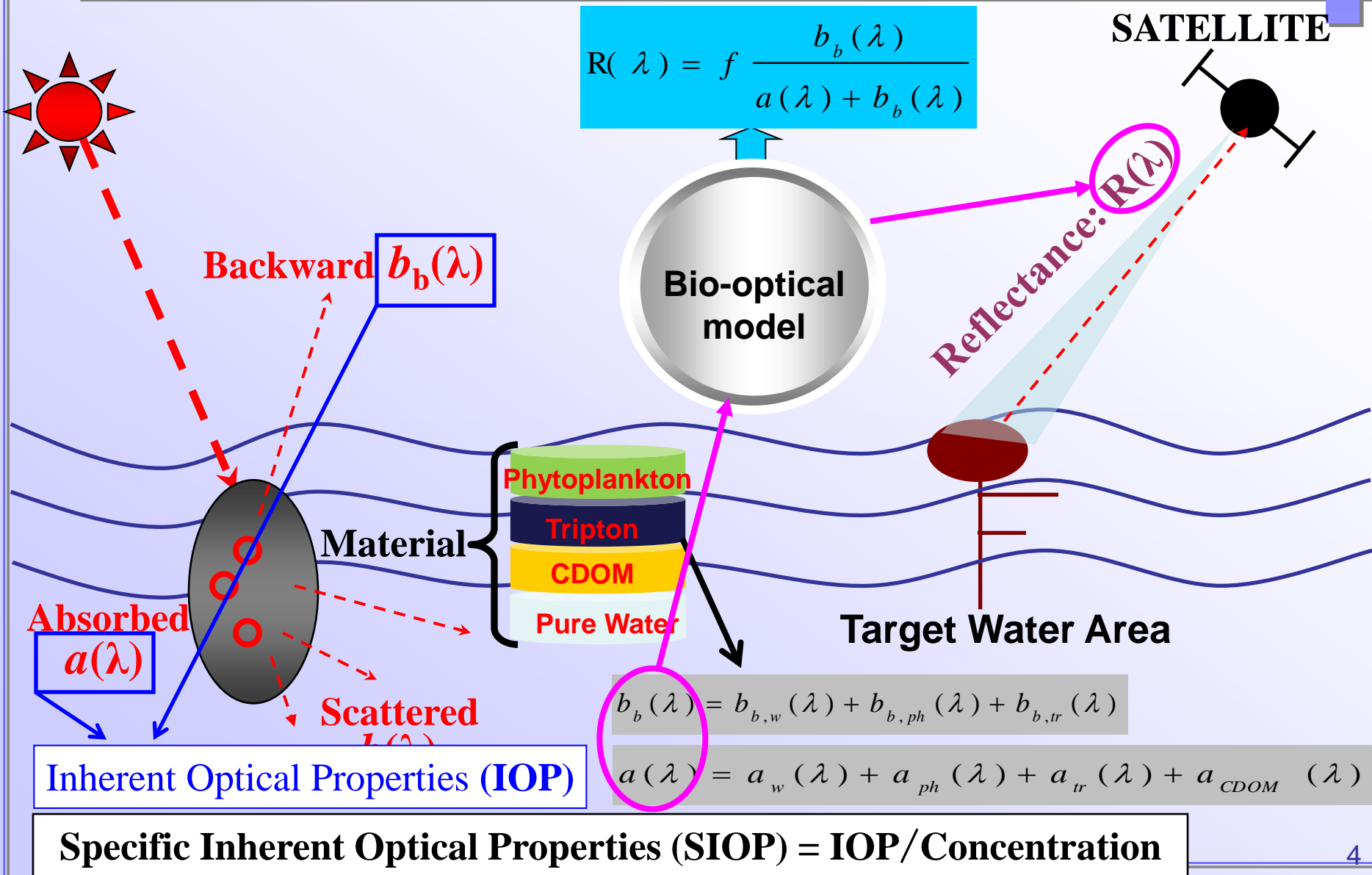
General conclusions

# Advantage of Satellite Remote Sensing

Satellite remote sensing can provide synoptic observations of water area



# Physical Principle of Remote Sensing in Water Body



# Water Types: the Perspective of Remote Sensing

- **Case I** waters : waters are those dominated by phytoplankton (e.g. open oceans).



Chl-a can be satisfactorily estimated with satellite images (Gordon & Morel, 1983)

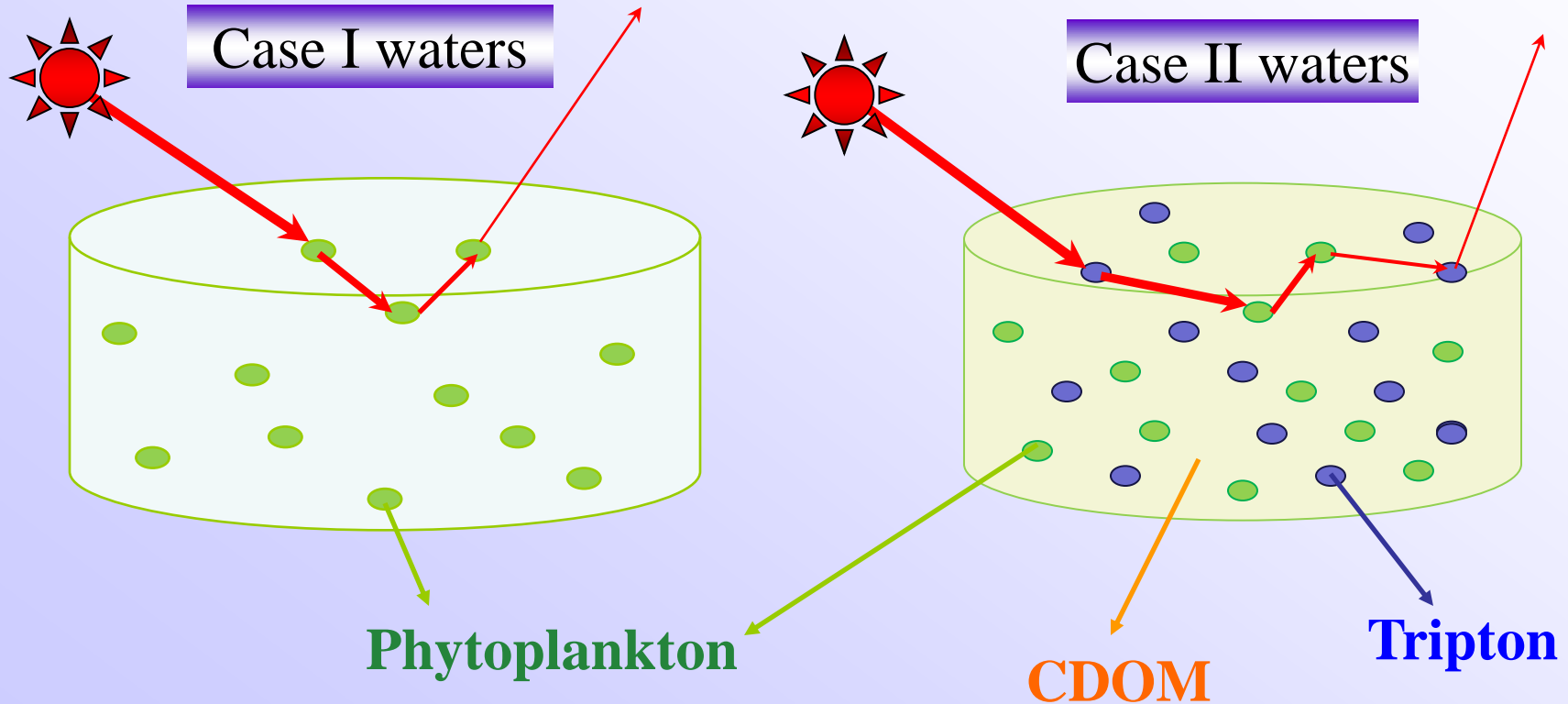
- **Case II** waters : waters are containing not only phytoplankton, but also non-phytoplankton suspended solids (i.e., tripton) and colored dissolved organic matters (CDOM) (e.g. coastal and inland waters).



Still challenging

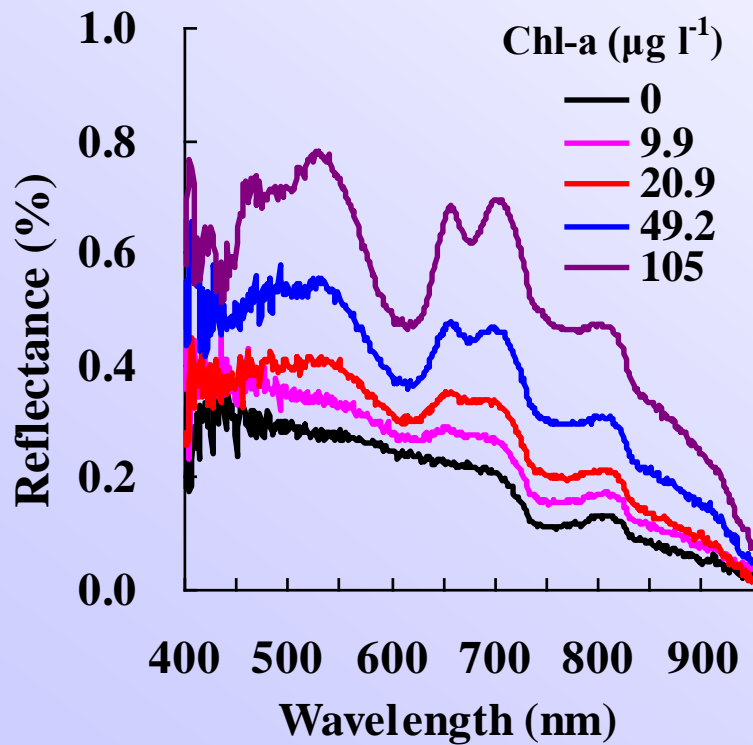
# Difficulty for Monitoring Case II Waters

## Interactions between Different Components

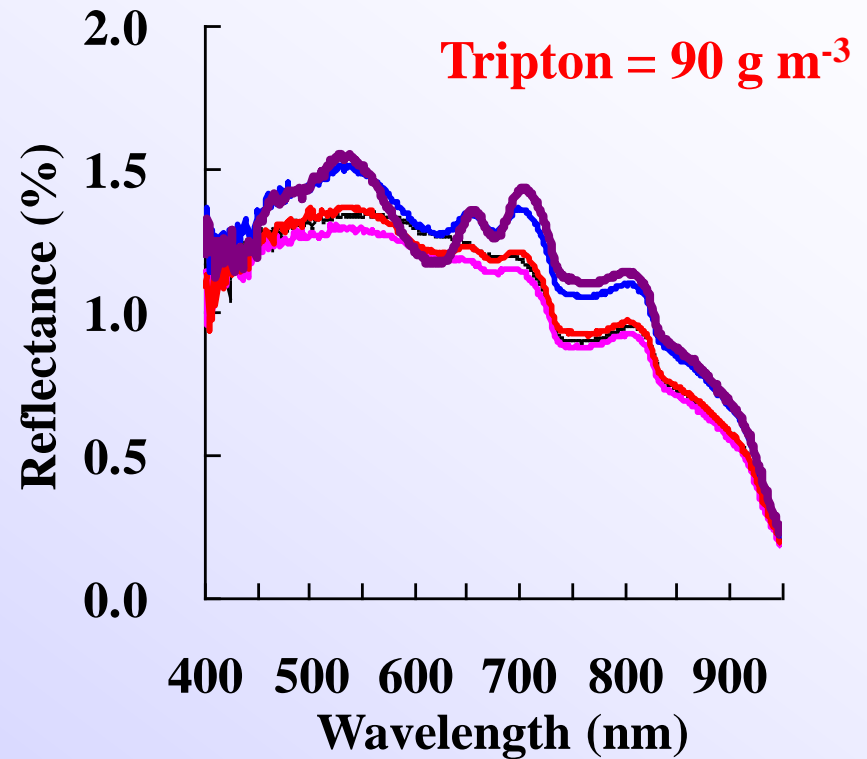


# Reflectance spectra for case I and II waters

Case I waters



Case II waters



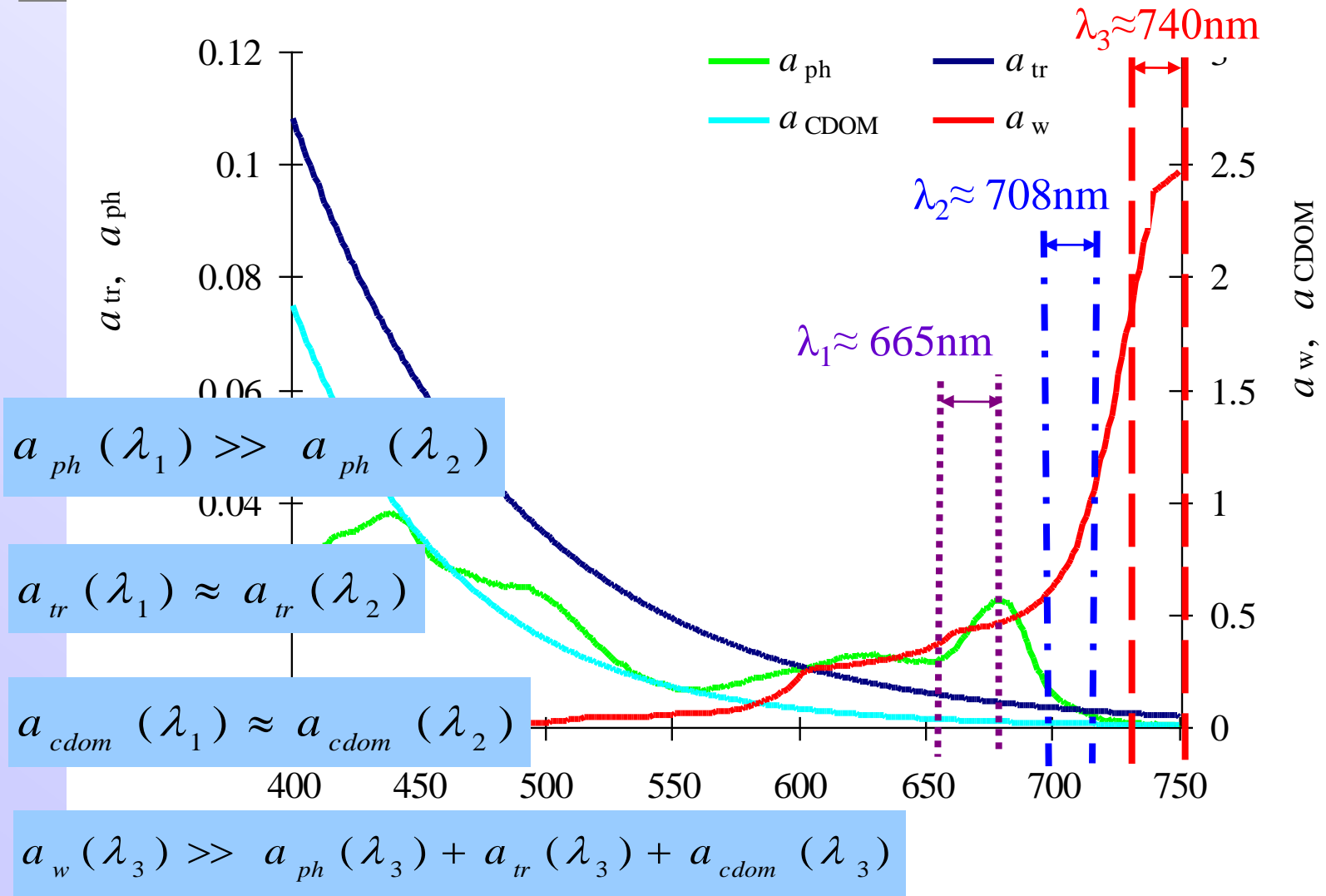
# Review of Previous Studies for Remote Sensing in Case II Waters

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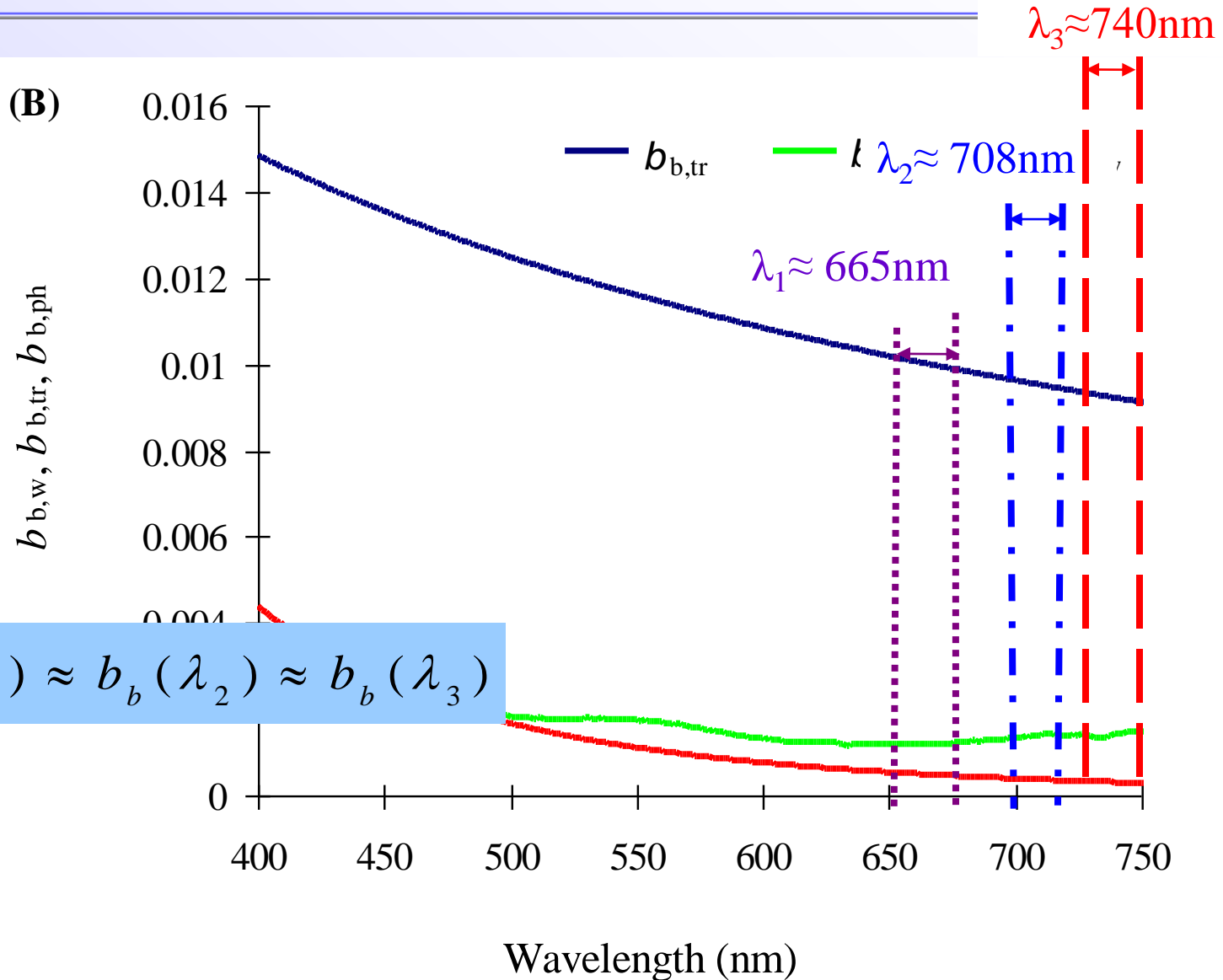
- A number of methods have been proposed, e.g., Matrix Inversion Method (Hoogenboom *et al.*, 1998), Band-ratio index (Gons, 1999; Ammenberg *et al.*, 2002), Three-band index (Gitelson *et al.*, 2008), and so on.
- The basic idea for all of the methods is **to isolate the interactions** between different components.
- Among these methods, the **3-band index** significantly outperformed other indices (Gitelson *et al.*, 2008).



# Principle of the 3-band index: Absorption



# Principle of the 3-band index: Backscattering



# Principle of the 3-band index: Reflectance

$$\left(\frac{1}{R_1} - \frac{1}{R_2}\right) \times R_3$$

$$= \frac{1}{j} \left( \frac{a_1 + b_{b,1}}{a_3 + b_{b,3}} - \frac{a_2 + b_{b,2}}{a_3 + b_{b,3}} \right) \times \frac{b_{b,3}}{a_3 + b_{b,3}}$$

$$\approx \frac{a_1 + b_{b,1} - (a_2 + b_{b,2})}{a_3 + b_{b,3}} \times \frac{b_{b,3}}{a_3 + b_{b,3}}$$

$$\approx \frac{a_1 - a_2}{a_3 + b_{b,3}}$$

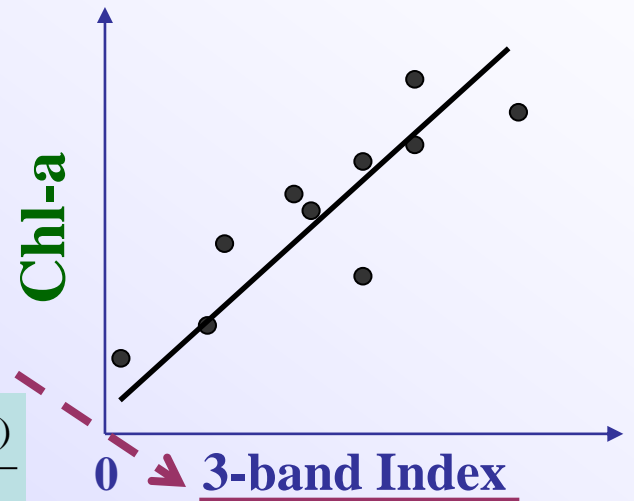
$\propto \text{Chl} - a$

$$= \frac{(a_{ph,1} + a_{tr,1} + a_{CDOM,1} + a_{w,1}) - (a_{ph,2} + a_{tr,2} + a_{CDOM,2} + a_{w,2})}{a_{w,3} + a_{ph,3} + a_{tr,3} + a_{CDOM,3} + b_{b,3}}$$

$$\approx \frac{a_{ph,1} + a_{w,1} - a_{w,2}}{a_{w,3} + a_{w,3}}$$

Constants

$$\text{Chl} - a = m \cdot \left(\frac{1}{R_1} - \frac{1}{R_2}\right) \times R_3 + n$$



# Problem of the 3-band index

For waters with **high turbidity** (e.g. some Asian Lakes), the assumption

~~$$a_{\text{water}}(\lambda_3) \gg a_{\text{ph}}(\lambda_3) + a_{\text{tripton}}(\lambda_3) + a_{\text{CDOM}}(\lambda_3)$$~~

is not reasonable anymore (Le *et al.*, 2009).

In these cases, the **3-band index** becomes:

$$\frac{[1/R_{rs}(\lambda_1) - 1/R_{rs}(\lambda_2)] \times R_{rs}(\lambda_3)}{a_w(\lambda_3) + a_{\text{ph}}(\lambda_3) + a_{\text{tr}}(\lambda_3) + a_{\text{CDOM}}(\lambda_3)}$$



Chl - a

Denominator includes the absorption not only phytoplankton, but also tripton and CDOM

**The existing methods still cannot provide satisfactory estimations for turbid case II waters.**

# Main objectives of my research

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To develop **more effective methods** to isolate the interactions between water components



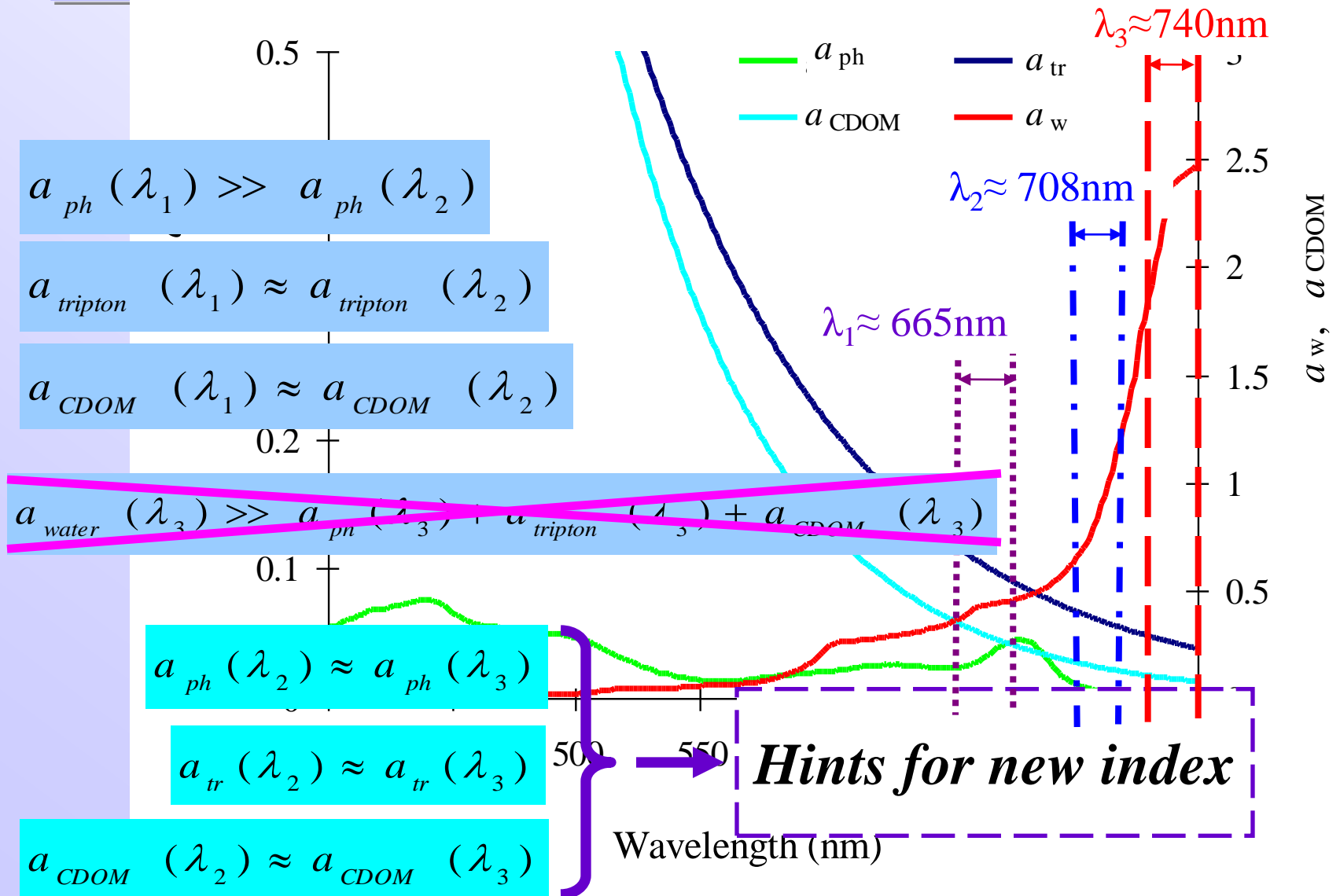
To validate their **performances** using simulation, *in situ* and satellite datasets

# Section 2



**An Enhanced 3-band Index for  
estimating Chl-a in turbid Case II waters**

# Absorption properties for turbid water case (e.g. TSS=20 g m<sup>-3</sup>)



# Development of the Enhanced 3-band Index

$$\frac{[1/R_{rs}(\lambda_1) - 1/R_{rs}(\lambda_2)]}{[1/R_{rs}(\lambda_3) - 1/R_{rs}(\lambda_2)]}$$

$$= \frac{\frac{1}{f} \left( \frac{a_1 + b_{b,1}}{b_{b,1}} - \frac{a_2 + b_{b,2}}{b_{b,2}} \right)}{\frac{1}{f} \left( \frac{a_3 + b_{b,3}}{b_{b,3}} - \frac{a_2 + b_{b,2}}{b_{b,2}} \right)}$$

$$\approx \frac{a_1 - a_2}{a_3 - a_2}$$

$\propto [\text{Chl} - a]$

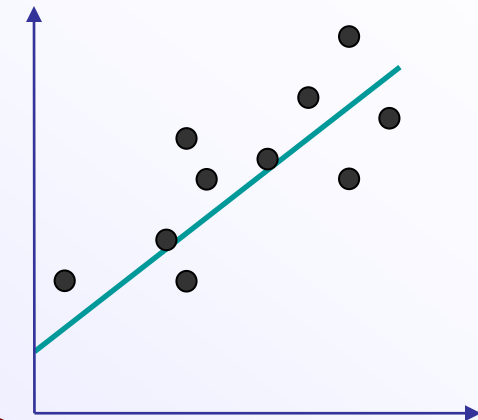
$$\text{Chl-a} = l \times \text{New Index} + m$$

$$= \frac{(a_{ph,1} + a_{tr,1} + a_{CDOM,1} + a_{w,1}) - (a_{ph,2} + a_{tr,2} + a_{CDOM,2} + a_{w,2})}{(a_{ph,3} + a_{tr,3} + a_{CDOM,3} + a_{w,3}) - (a_{ph,2} + a_{tr,2} + a_{CDOM,2} + a_{w,2})}$$

$$\approx a_{ph}(\lambda_1) [a_w(\lambda_1) - a_w(\lambda_2)] / [a_w(\lambda_3) - a_w(\lambda_2)]$$

Constants

Chl-a



New Index



# Study Areas

## Lake Kasumigaura.

Japan

Location: Ibaraki



## Lake Dianchi, China

Location: Yunnan Province, China

Area: 300 km<sup>2</sup>

Average depth: 6 m

Case II water:

- mixture of phytoplankton, tripton, and CDOM
- dominant species cyanobacteria

Field survey: 4 times

(Oct. 23, 2007; Jul. 15, 2008; Mar. 12, 2009; Jul.-Aug., 2009)

Total 53 sites

# Data Collections

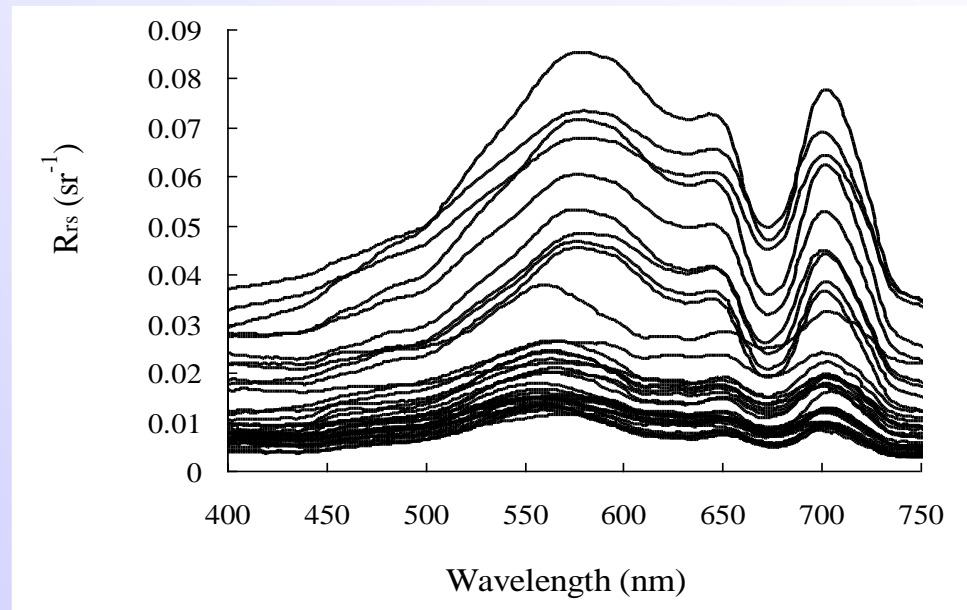
## Phytoplankton (Chl-a)

Samples taken to lab;  
SCOR-UNESCO equations.

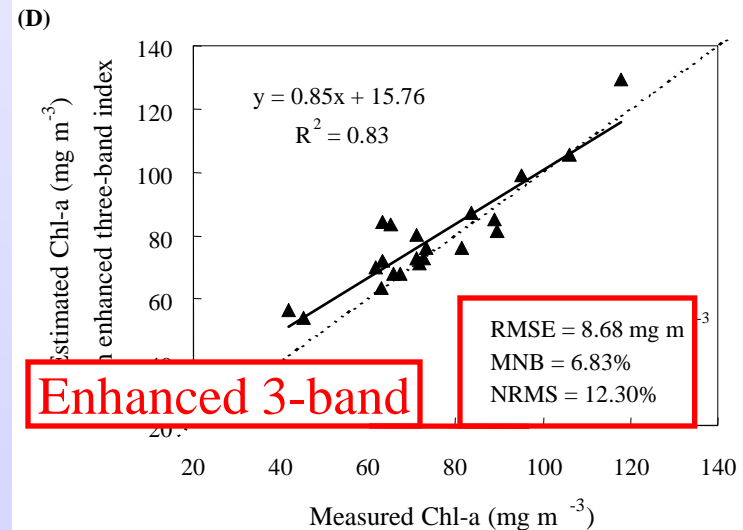
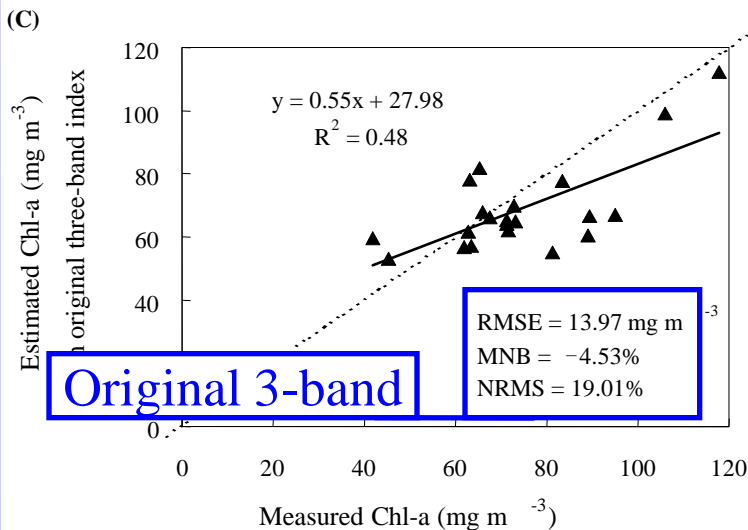
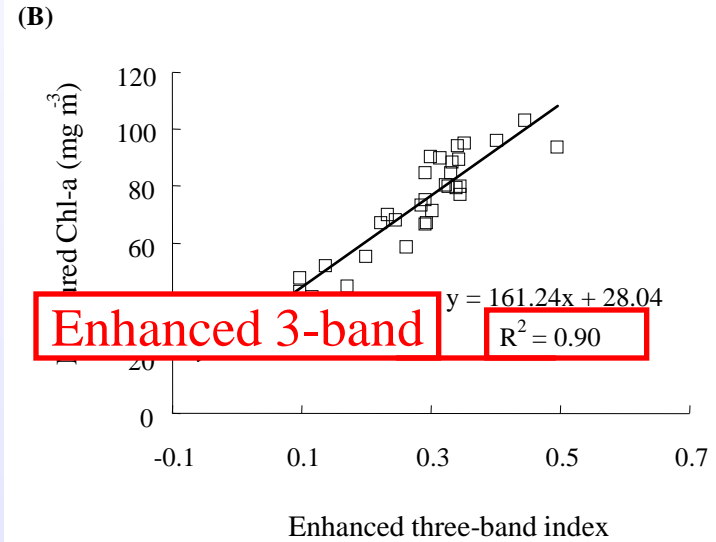
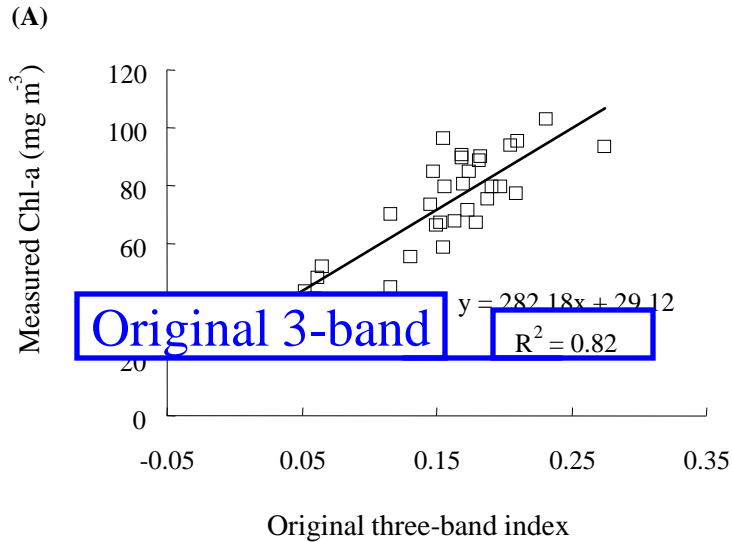


## Remote sensing reflectance

Water-leaving radiance;  
Downward irradiance;  
Downward radiance of  
skylight.



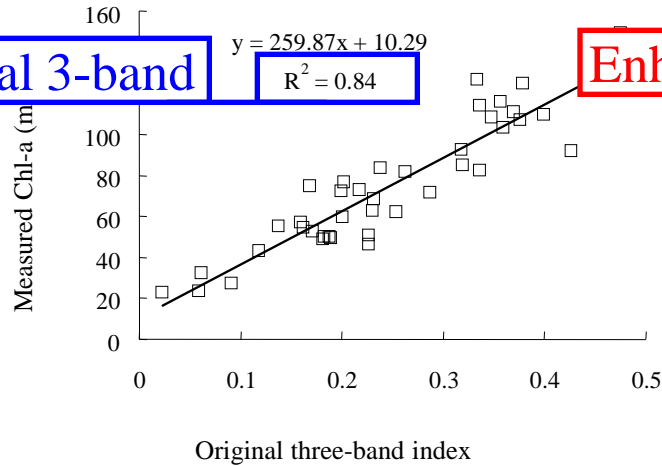
# Results for Lake Kasumigaura, Japan



# Results for Lake Dianchi, China

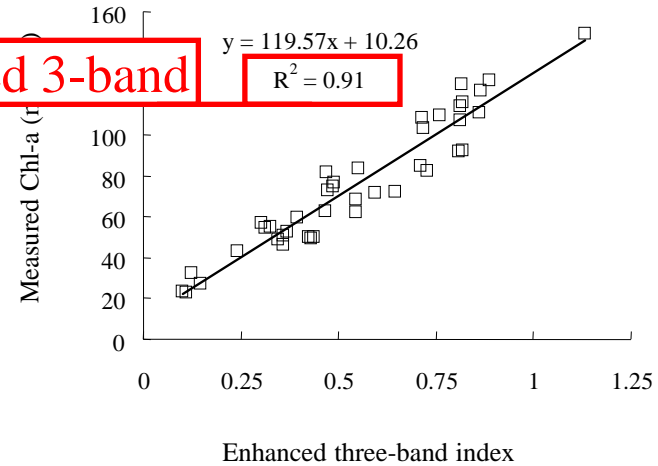
(A)

Original 3-band



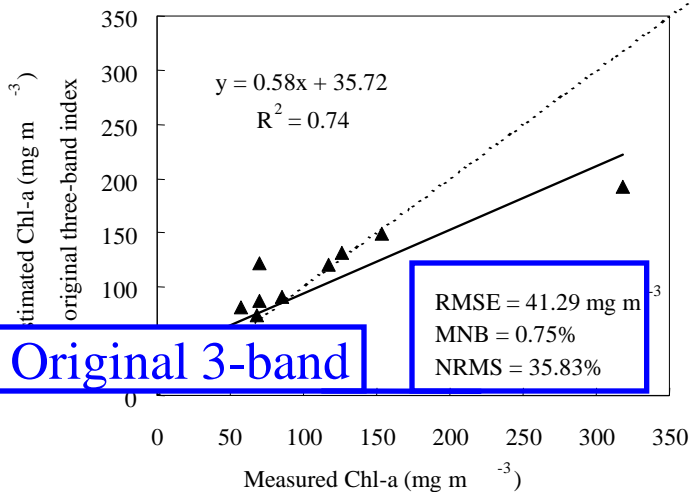
(B)

Enhanced 3-band



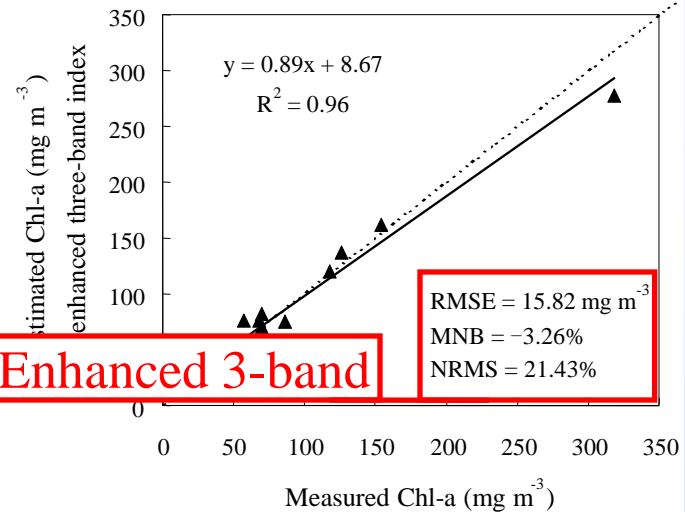
(C)

Original 3-band



(D)

Enhanced 3-band



# Section 3



Estimation of water constituent concentrations in case II waters by **semi-analytical model-optimizing and look-up-tables (SAMO-LUT)**

*Objective:*

To provide information of not only Chl-a, but also tripton and CDOM.

# Selection of Three Previous Semi-analytical Models

- Chl-a estimation model (3-band index):

$$Chla = p \times [R_{rs}^{-1}(B_7) - R_{rs}^{-1}(B_9)] \times R_{rs}(B_{10}) + q$$

(Gitelson *et al.*, 2008)

- Tripton estimation model:

$$Tr = m \times R_{rs}(B_{10}) + n \quad (\text{Ammenberg } et al., 2002)$$

- CDOM estimation model:

$$a_{CDOM}(440) = a \times [R_{rs}(B_7) / R_{rs}(B_5)] + b$$

(Ammenberg *et al.*, 2002)

$B_5$ ,  $B_7$ ,  $B_9$ , and  $B_{10}$  denote MERIS bands 5 (555-565 nm), 7 (660-670 nm), 9 (703-713 nm), and 10 (750-757 nm), respectively

# Simulation test for re-analyzing previous models

- Remote-sensing reflectance generated from:

$$R_{rs}(\lambda) = f \frac{b_b(\lambda)}{a(\lambda) + b_b(\lambda)}$$

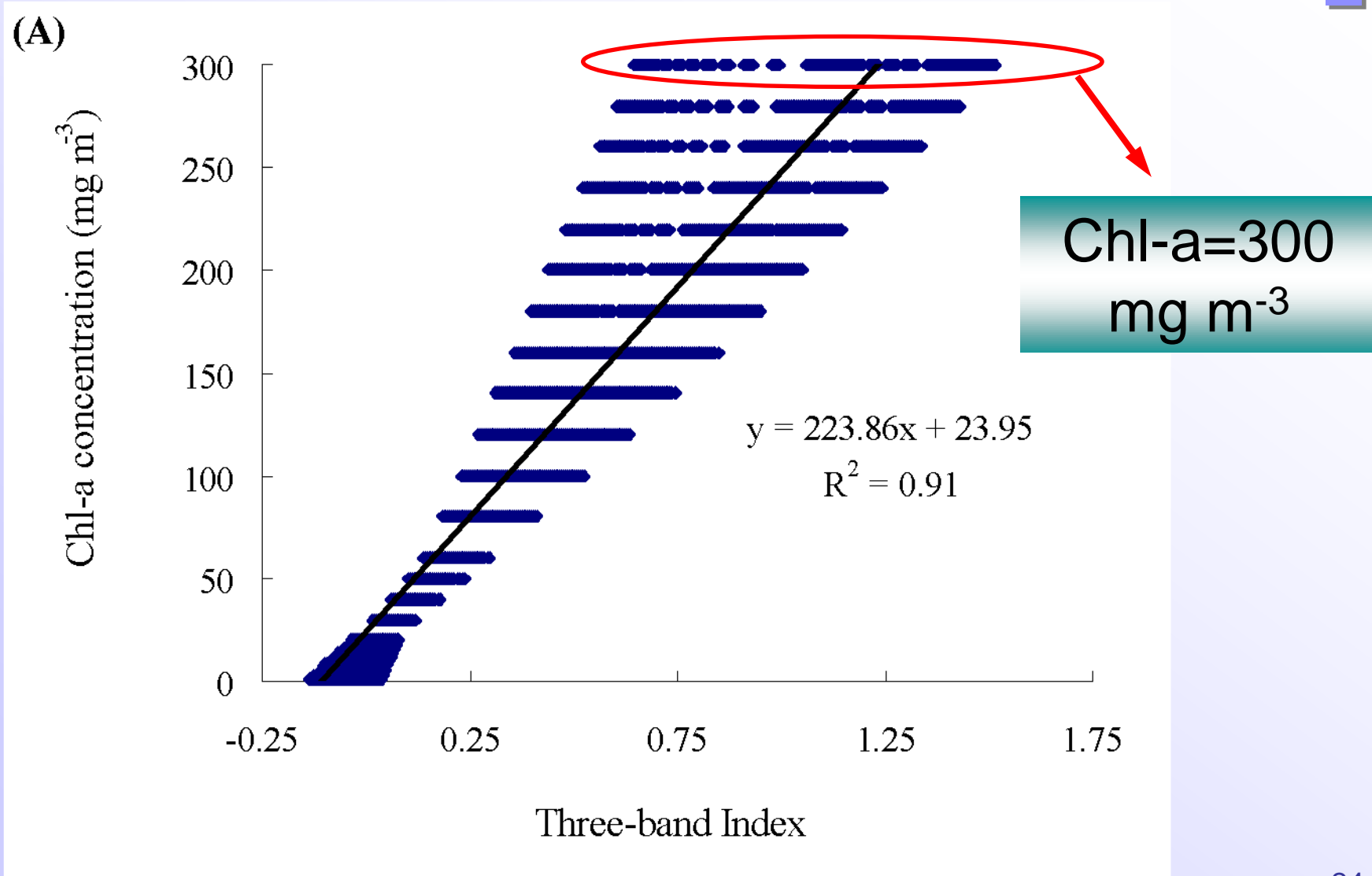
$$a = [\text{Chl} - a] a_{ph}^* + [\text{Tr}] a_{tr}^* + [\text{CDOM}] a_{CDOM}^* + a_w$$

$$b_b = [\text{Chl} - a] b_{b,ph}^* + [\text{Tr}] b_{b,tr}^* + b_{b,w}$$

- [Chl-a] ranges: 1-300 mg m<sup>-3</sup>;
- [Tr] ranges: 1-250 g m<sup>-3</sup>;
- [CDOM] ranges: 0.1-10 m<sup>-1</sup>.
- In total, 19,964 samples for regression analysis

# Problem for previous model

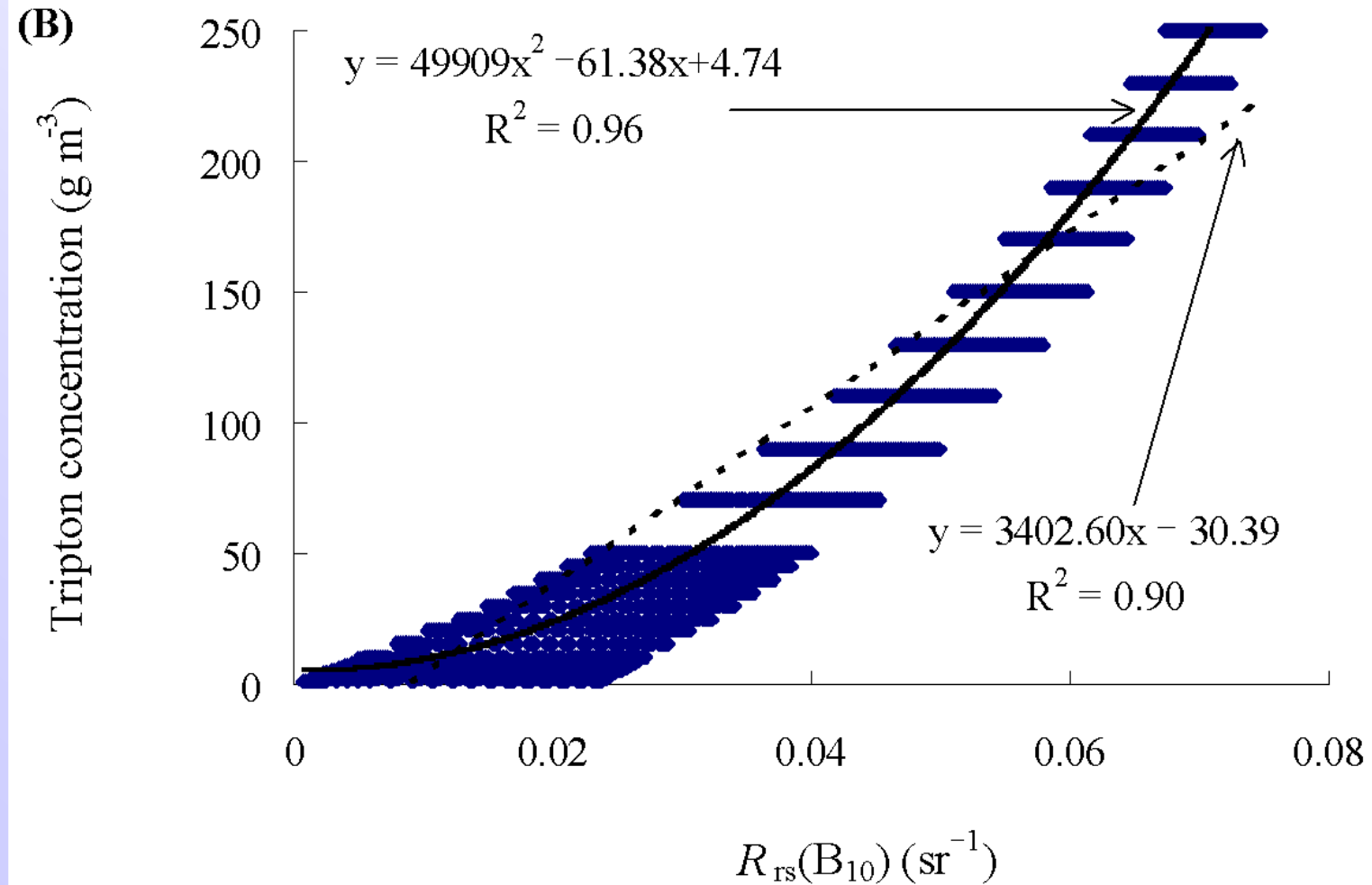
(Chl-a vs. 3-band index)





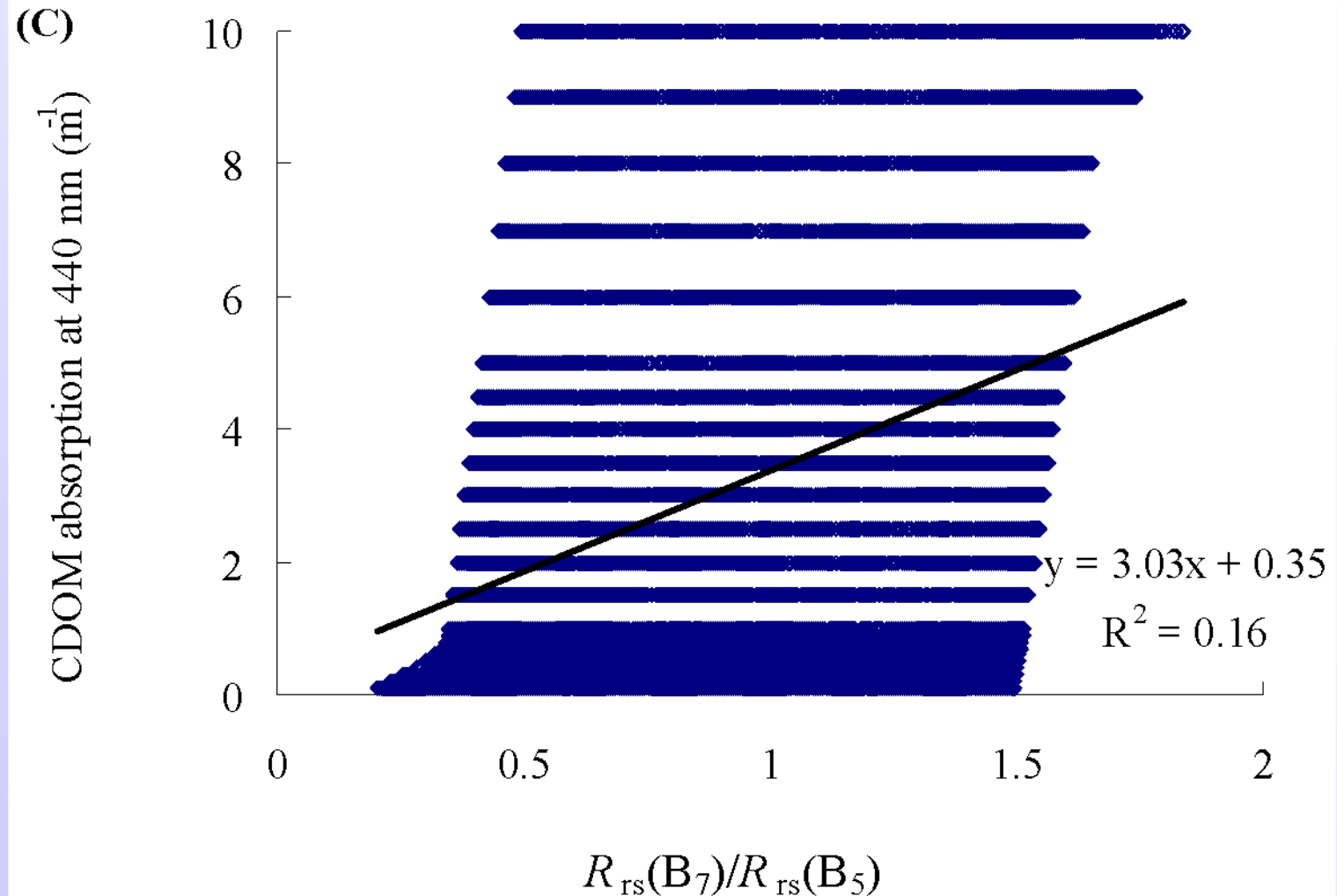
# Problem for previous model

(Tripton vs.  $R_{rs}(B_{10})$ )



# Problem for previous model

(CDOM Vs.  $R_{rs}(b_7)/R_{rs}(b_5)$ )



# Development of the new algorithm

## Semi-Analytical Model-Optimizing and Look-Up Table (SAMO-LUT)

- Basic idea of this new method is **an imaginary case II water**, in which only one component changes while the other two are controlled as constants.
- E.g. Tr and CDOM are constants; only Chl-a is changing :

$$a = [Chl - a] a_{ph}^* + [Tr] a_{tr}^* + [CDOM] a_{CDOM}^* + a_w$$

$$b_b = [Chl - a] b_{b,ph}^* + [Tr] b_{b,tr}^* + b_{b,w}$$

$$R_{rs}(\lambda) = f \frac{b_b(\lambda)}{a(\lambda) + b_b(\lambda)}$$

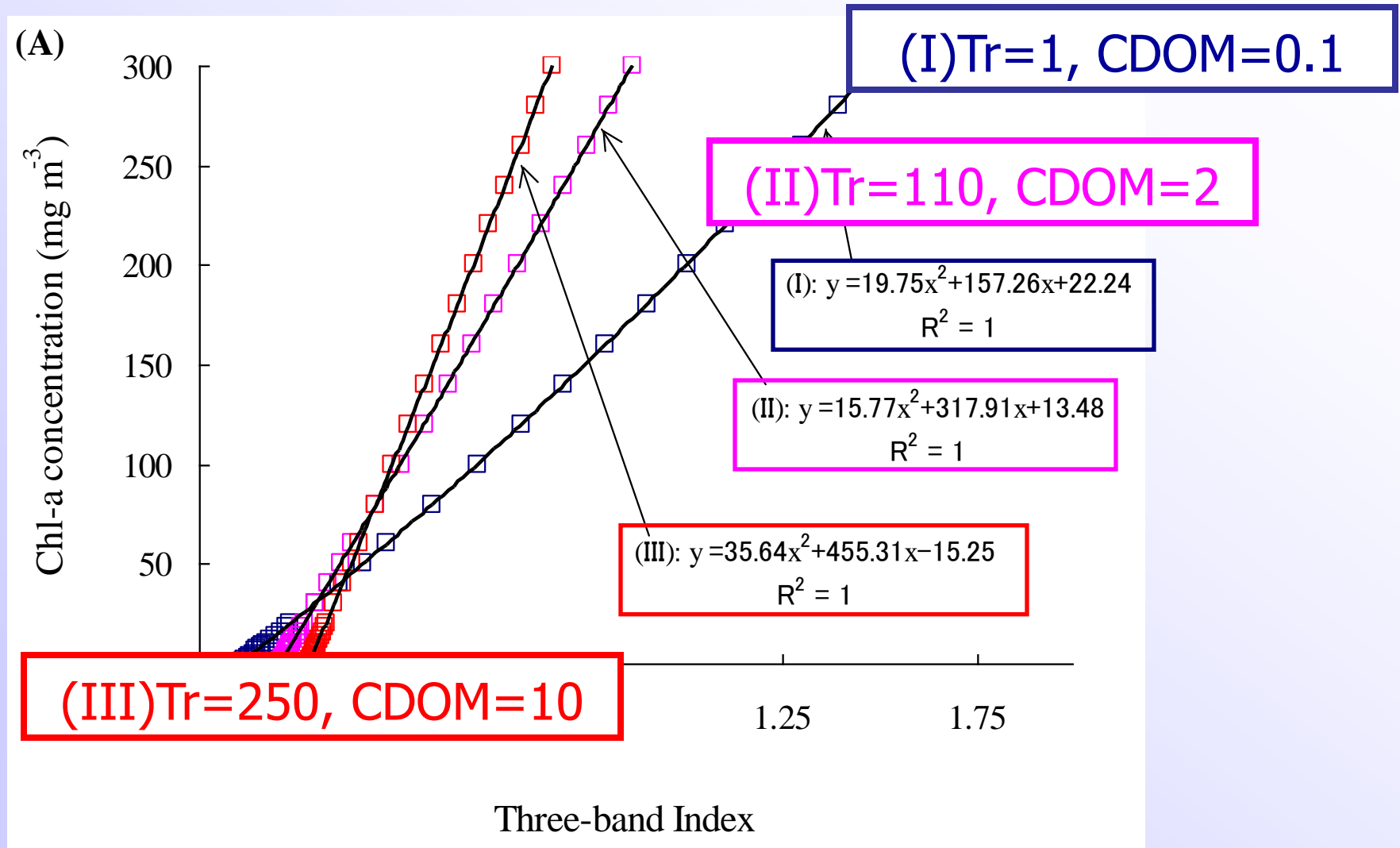
**Constants**

$R_{rs}$  is changing  
only with Chl-a.



**Possibility for improving  
three-band index**

# Example of tripton and CDOM controlled as constants (e.g. Chl-a vs. 3-band index)



# Construction of Look-up Tables (LUT) using simulation data

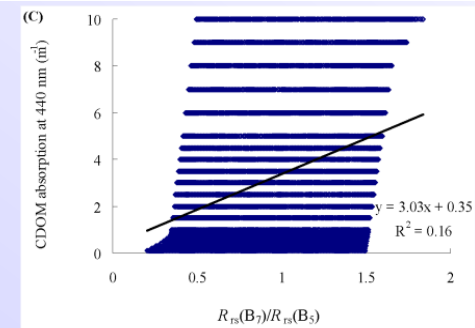
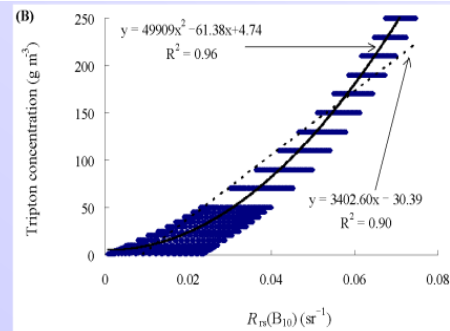
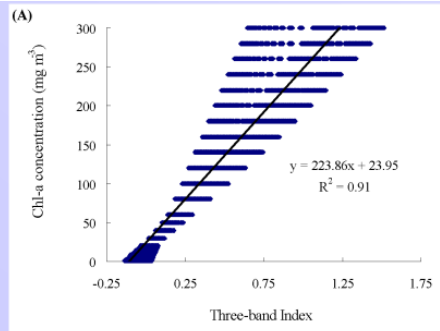
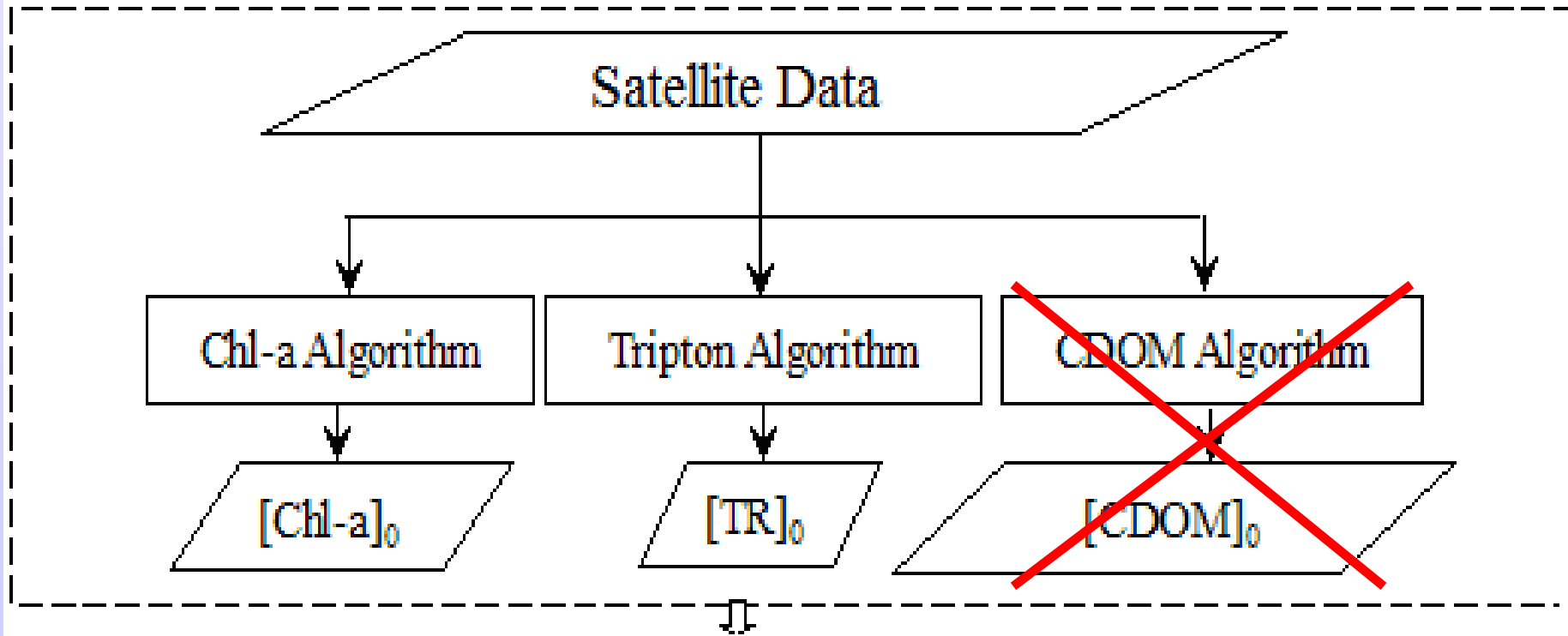
- The coefficients of Chl-a model for each combination of Tripton and CDOM are stored in the LUT.

		Tripton ( $\text{g m}^{-3}$ )					
		1	2	...	...	249	250
CDOM ( $\text{m}^{-1}$ )	0.1	$(p_{1,1}; q_{1,1}; r_{1,1})$	$(p_{1,2}; q_{1,2}; r_{1,2})$	...	...	$(p_{1,249}; q_{1,249}; r_{1,249})$	$(p_{1,250}; q_{1,250}; r_{1,250})$
	0.2	$(p_{2,1}; q_{2,1}; r_{2,1})$	$(p_{2,2}; q_{2,2}; r_{2,2})$	...	...	$(p_{2,249}; q_{2,249}; r_{2,249})$	$(p_{2,250}; q_{2,250}; r_{2,250})$
	0.3	$(p_{3,1}; q_{3,1}; r_{3,1})$	$(p_{3,2}; q_{3,2}; r_{3,2})$	...	...	$(p_{3,249}; q_{3,249}; r_{3,249})$	$(p_{3,250}; q_{3,250}; r_{3,250})$
	...	...	...	...	...	...	...
	...	...	...	...	...	...	...
	9.9	$(p_{99,1}; q_{99,1}; r_{99,1})$	$(p_{99,2}; q_{99,2}; r_{99,2})$	...	...	$(p_{99,249}; q_{99,249}; r_{99,249})$	$(p_{99,250}; q_{99,250}; r_{99,250})$
	10	$(p_{100,1}; q_{100,1}; r_{100,1})$	$(p_{100,2}; q_{100,2}; r_{100,2})$	...	...	$(p_{100,249}; q_{100,249}; r_{100,249})$	$(p_{100,250}; q_{100,250}; r_{100,250})$

The practical problem is how to find the optimal models for a given pixel?

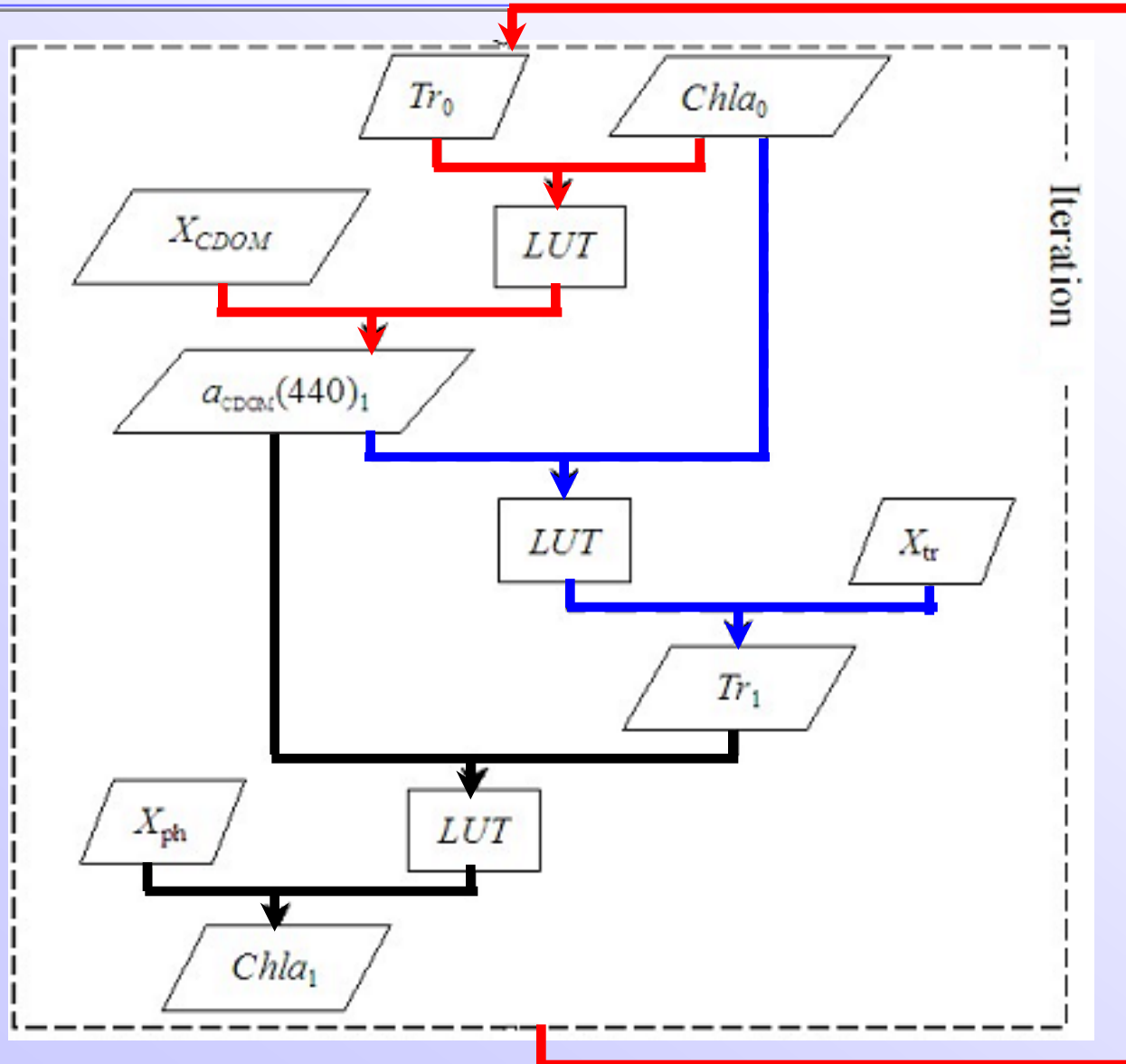
# Iteration strategy in the SAMO-LUT method

## (Step 1: getting initial values)



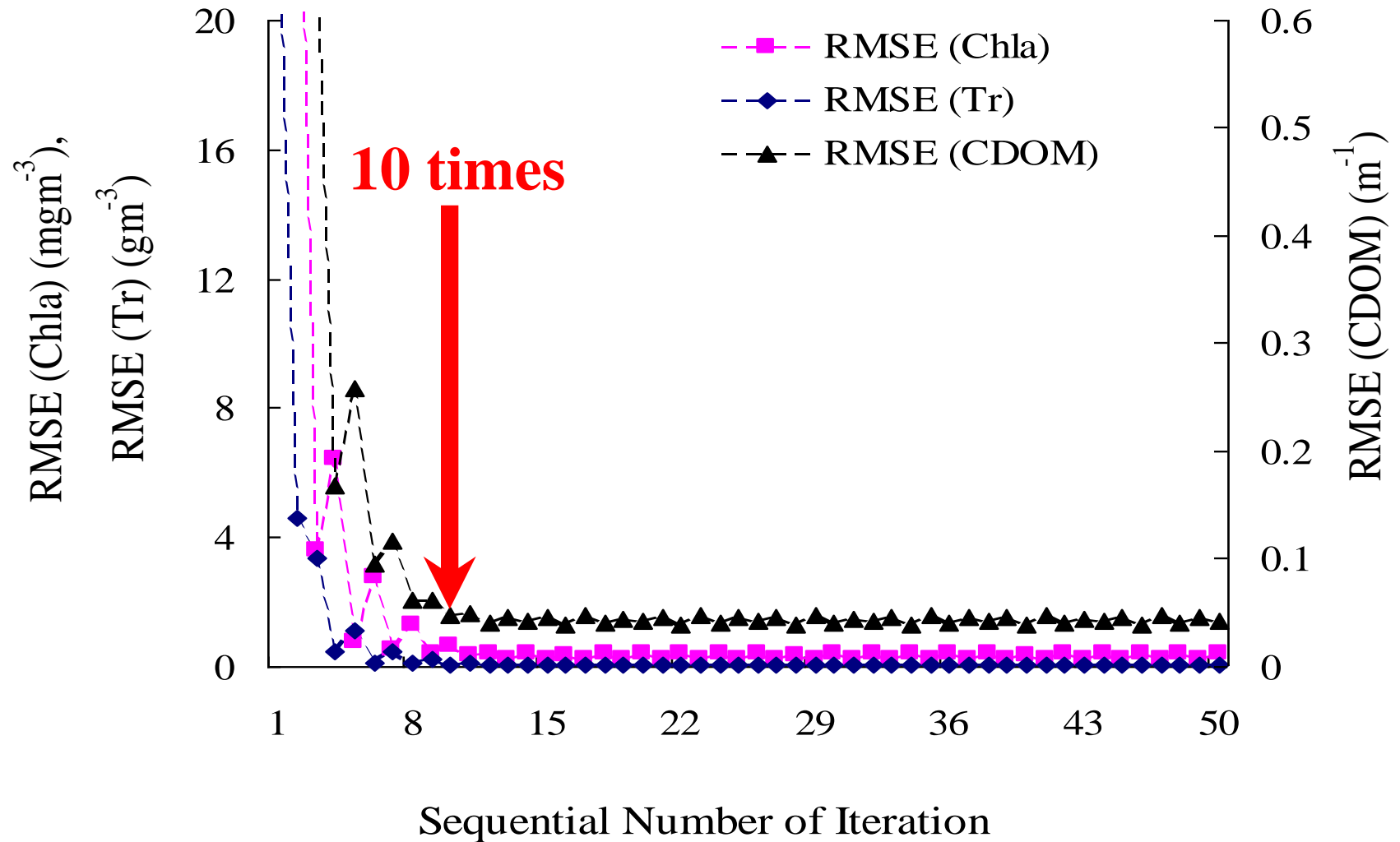
# Iteration strategy in the SAMO-LUT method

(Step 2: optimizing estimation model)



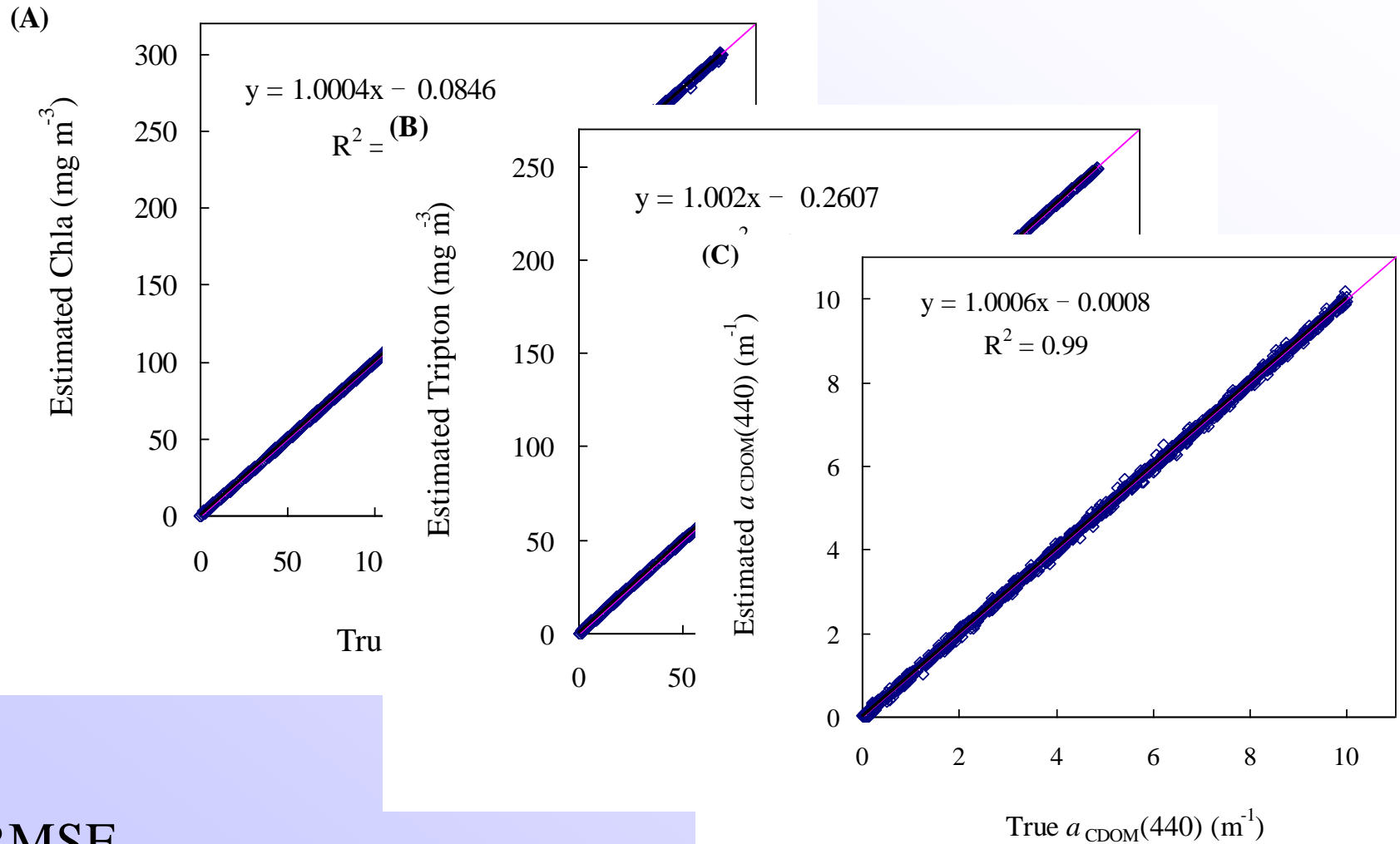
**Repeat !**  
 $Chla_0 = Chla_1$   
 $Tr_0 = Tr_1$

# Ending the iteration by difference between the **current and last iterations**





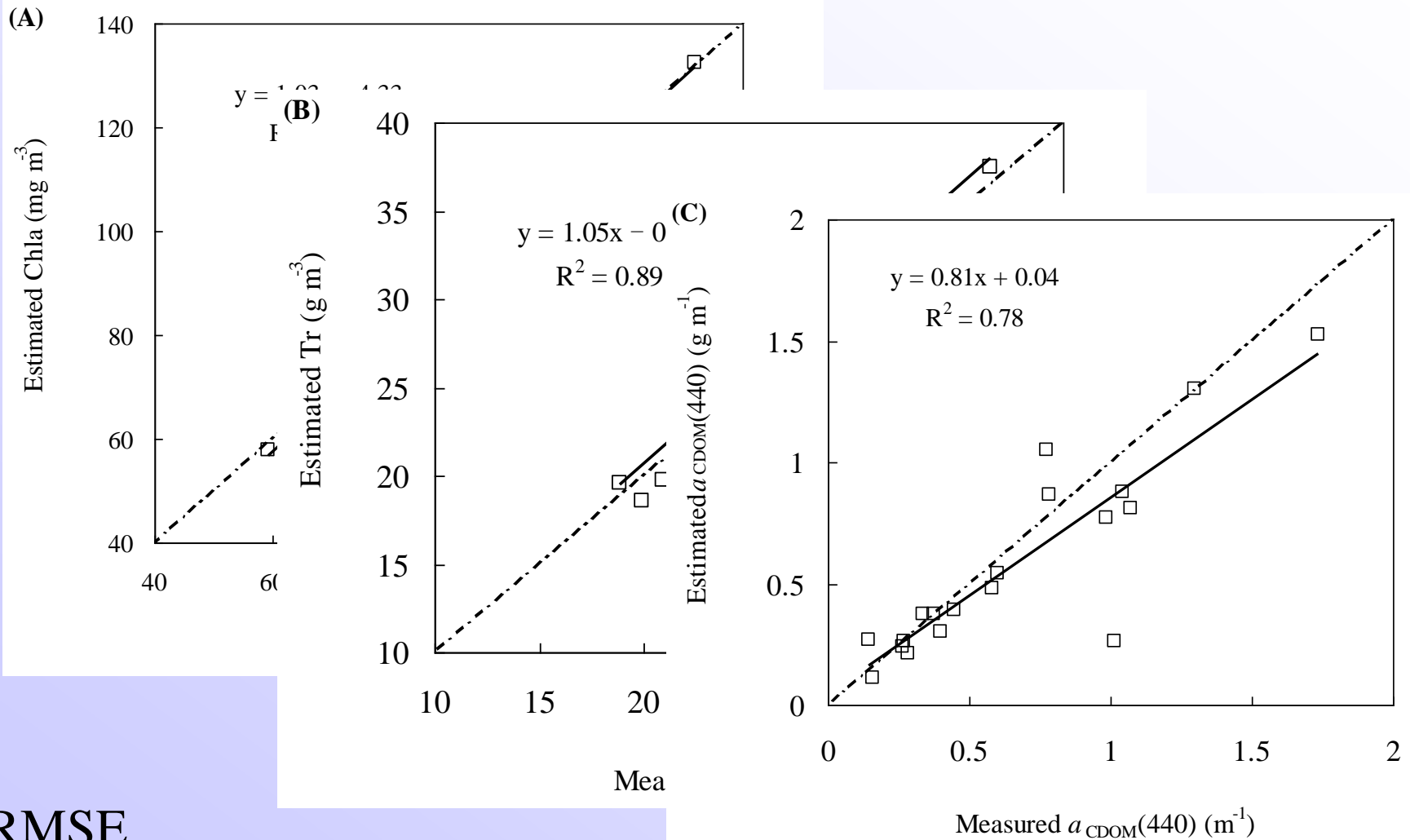
# Validation results: By simulated data (noise-free)



RMSE

(0.43, 0.42, 0.06)

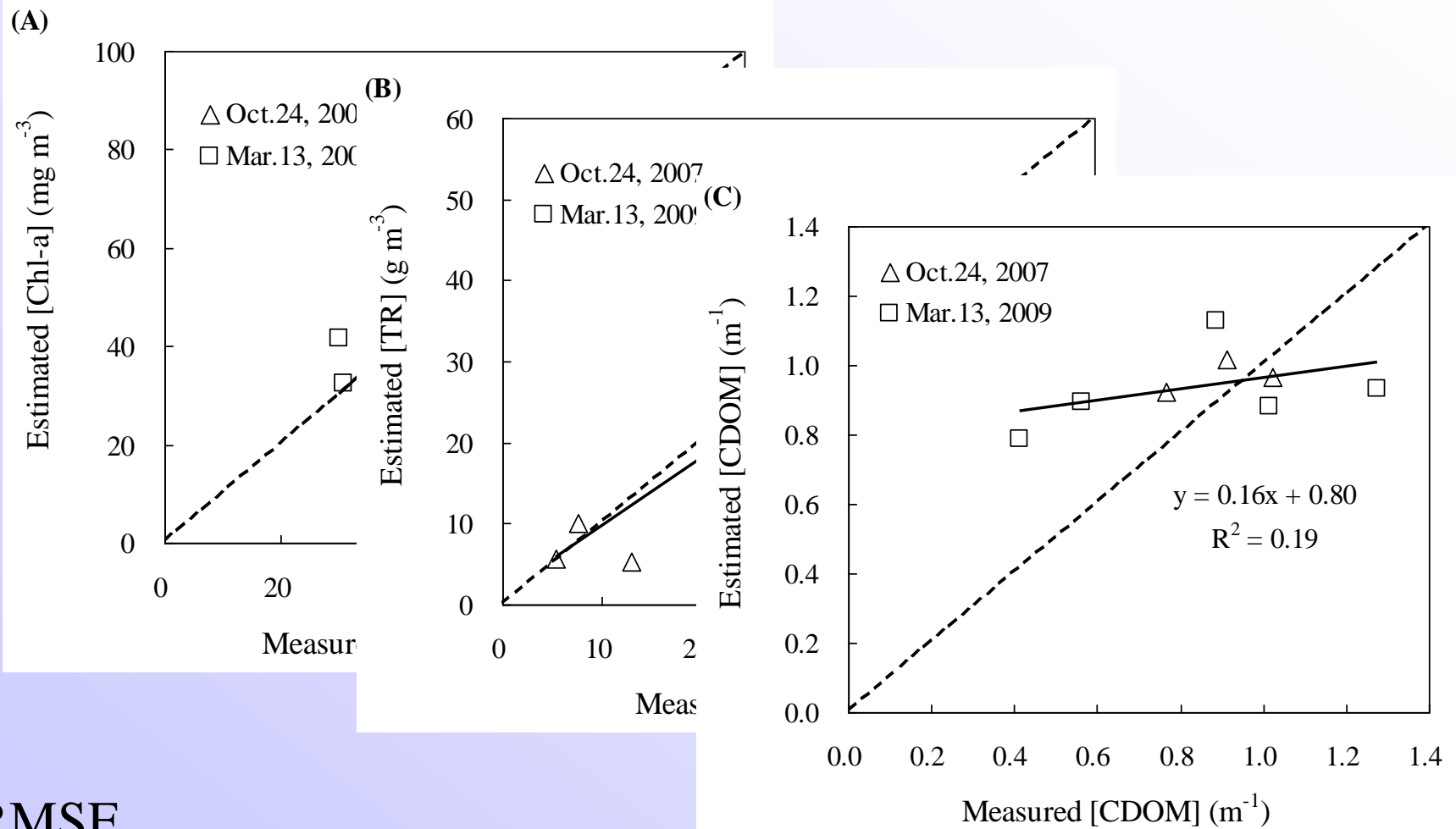
# Validation results: By *in situ* data ( low noise )



RMSE

(3.4,1.8,0.2)

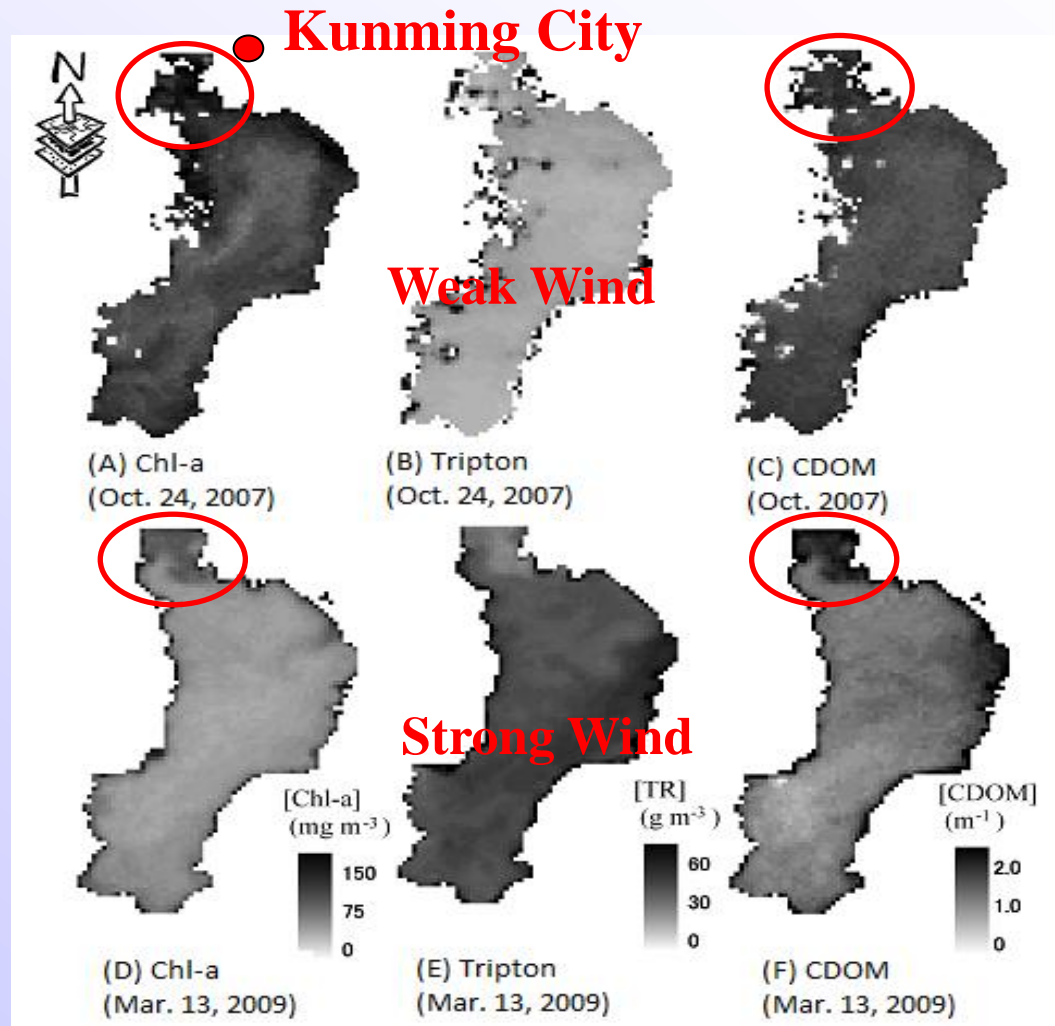
# Validation results: by MERIS data ( high noise)



RMSE

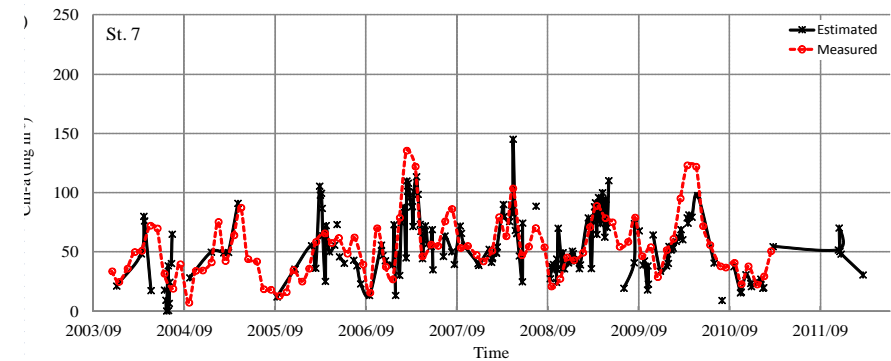
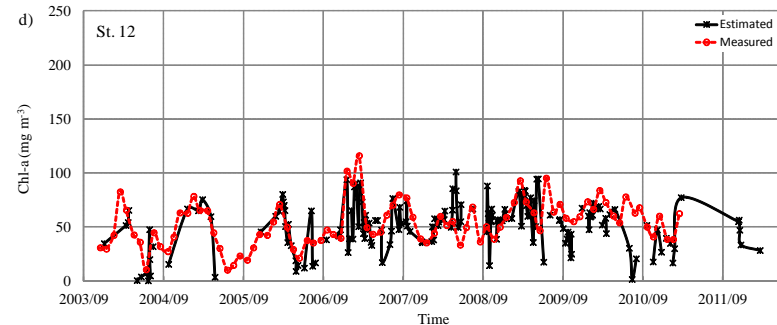
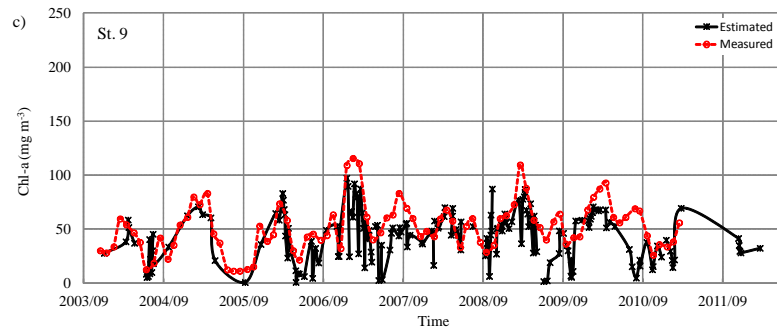
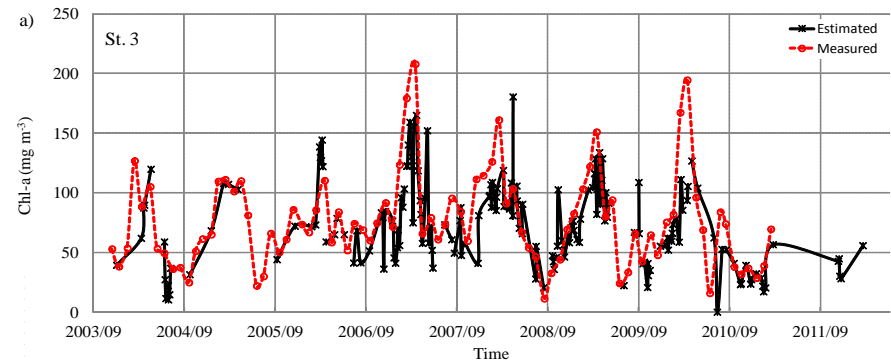
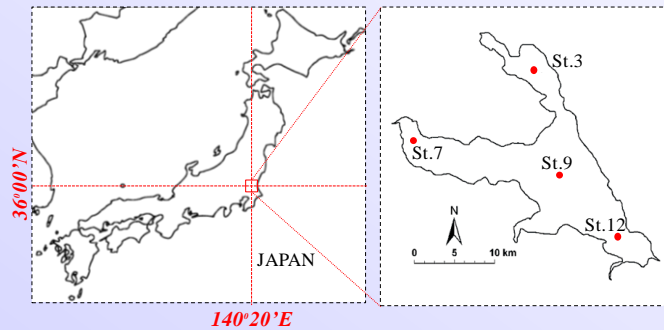
(8.8, 6.9, 0.25)

# Spatial Distribution of Water Quality



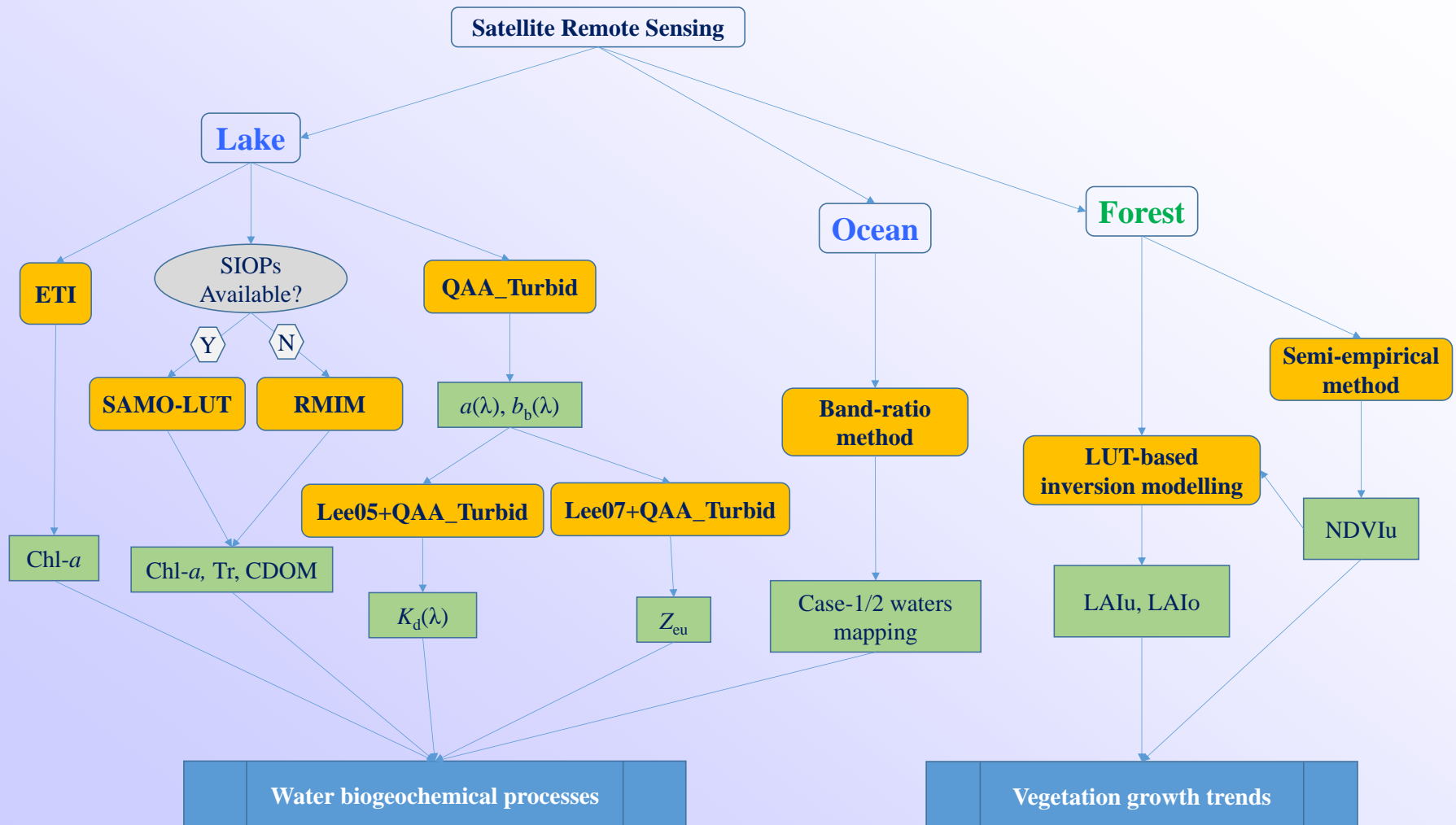
# Long-term Monitoring in Lake Kasumigaura

(Matsushita, Jaelania, Yang et al., 2015, RSSJ)



Time-series of satellite estimation show high agreements with field measurements.

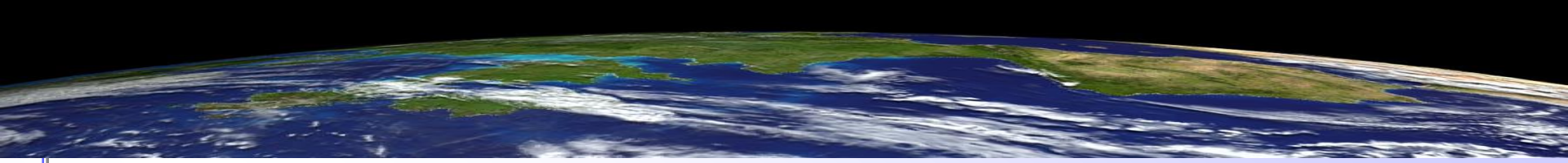
# Conclusions: my proposed algorithms



# Take-home Messages

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- Scientists in Chiba University, Japan are working on satellite monitoring of water quality.
- Our proposed algorithm worked well in long-term monitoring of water quality parameters in Japan's lakes.
- We are challenging the continental/global monitoring of inland lakes, to aid the conventional investigations.



Thank you!