



Remotely-Sensed  
Biogeochemical Cycles  
in the Ocean

## Breakout session n°10: Joint use of Bio-Argo and Ocean Colour

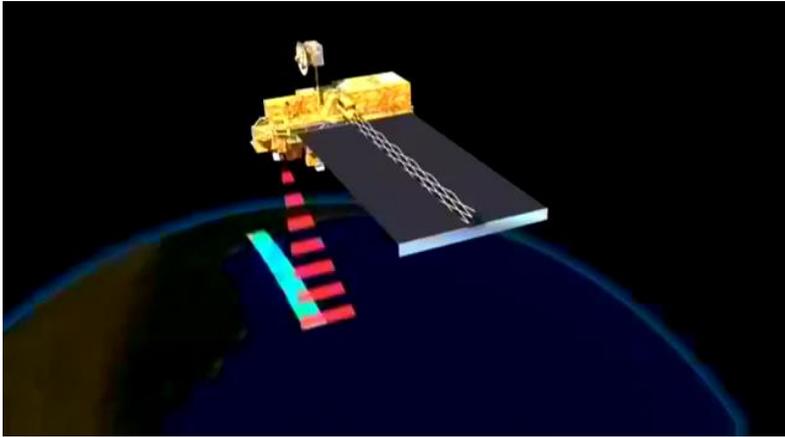
Extending surface bio-optical properties to depth:  
a neural network for merging ocean color  
and Argo data

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Fabrizio D'Ortenzio<sup>1</sup>, Bernard Gentili<sup>1</sup>, Antoine Poteau<sup>1</sup> and Catherine Schmechtig<sup>1</sup>

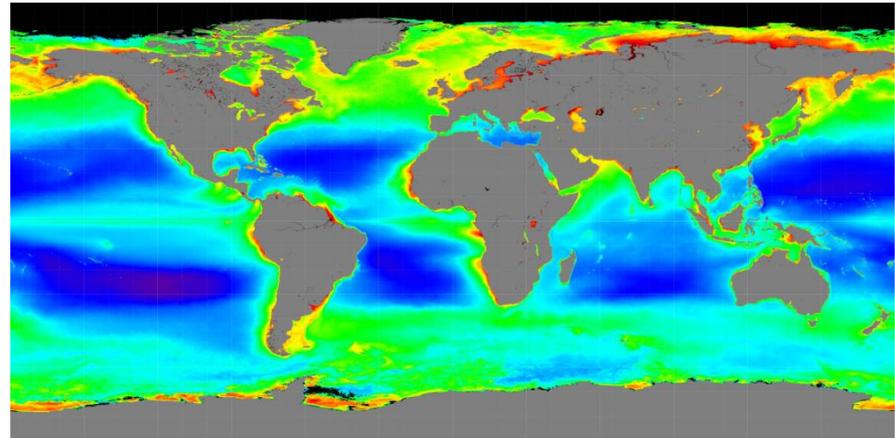
1: Laboratoire d'Océanographie de Villefranche, LOV  
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3: Plymouth Marine Laboratory, PML

The 18th of June 2015, San Francisco, CA, USA

Surface distribution of several bio-optical parameters already well known with « ocean color » satellite data (Feldman *et al.*, 1989)



BUT only 1/5<sup>th</sup> of the productive zone (Morel and Berthon, 1989)



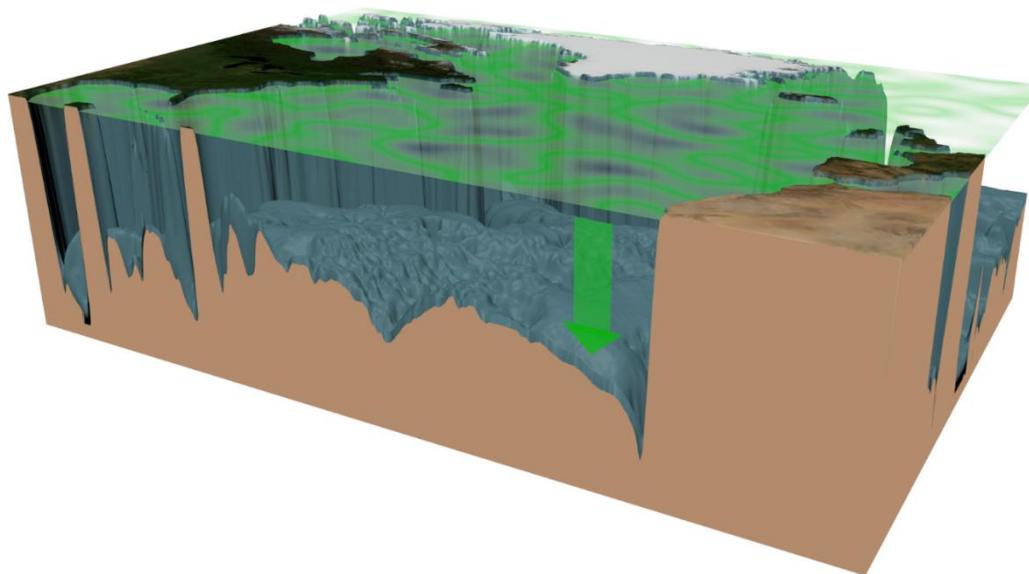
Surface [Chl] climatology by SeaWifs

## Lack of information on the vertical

→ surface estimations have to be extended to depth

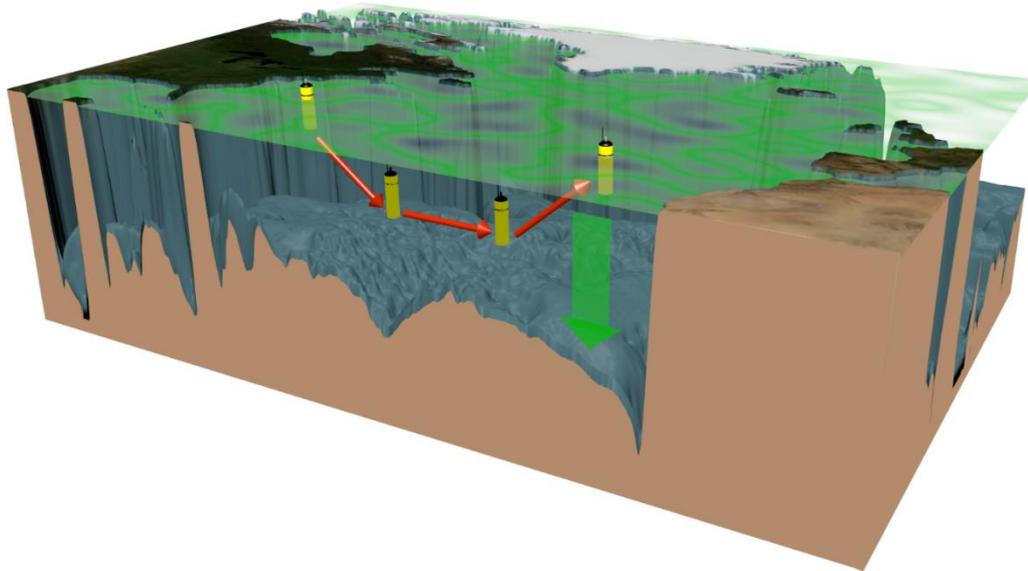
## OBJECTIVE

How to extend surface bio-optical properties to depth from ocean color data?



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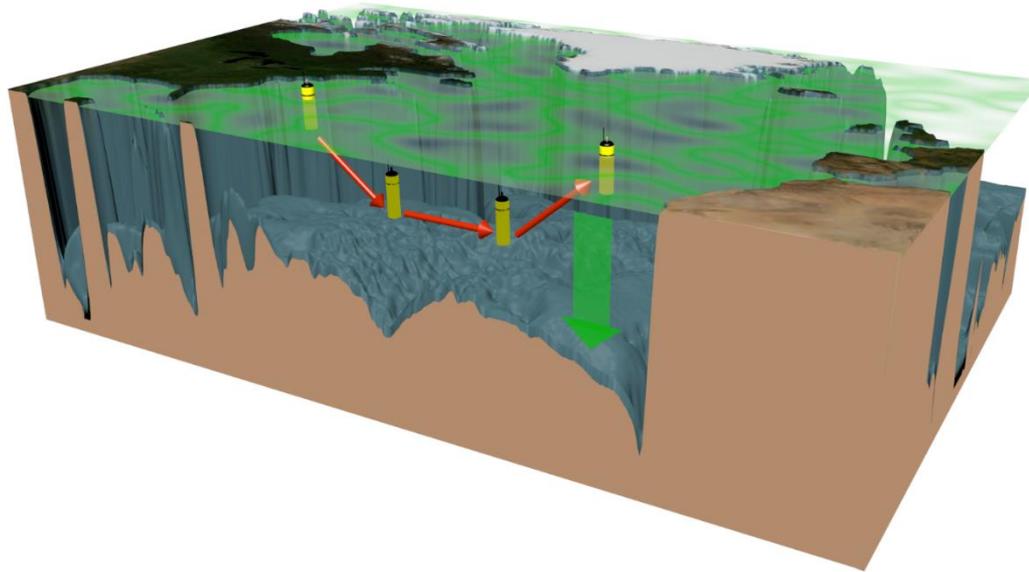
How to extend surface bio-optical properties to depth from ocean color data?



- ❑ **Physical state of the water column (from CTD profiles)**
  - Influences nutrient and light availability for phytoplankton growth
  - Available at high spatio-temporal frequency with Argo data

## OBJECTIVE

How to extend surface bio-optical properties to depth from ocean color data?



### ❑ Physical state of the water column (from CTD profiles)

- Influences nutrient and light availability for phytoplankton growth
- Available at high spatio-temporal frequency with Argo data

### ❑ Methodology: MLP for Multi-Layered Perceptron (neural network) :

- Iterative statistical method of learning
- Universal approximators
- Complex, noisy and non-linear distributions

OBJECTIVE

How to extend surface bio-optical properties to depth from ocean color data?

**OBJECTIVE**

To develop a neural network-based method using merged ocean color and Argo data to extend surface bio-optical properties to depth

Method after refer as **SOCA** (for Satellite Ocean Color and Argo data to vertical distribution of bio-optical properties)

- Better constrain the vertical distribution because production driven by nutrients availability, driven by physical properties
- Available at high spatio-temporal frequency with Argo data

- Perceptron (neural network):
- Iterative statistical method of learning
- Universal approximators
- Complex, noisy and non-linear distributions

## Development of 2 neural networks :

- Optical particulate **backscattering** coefficient ( $b_{bp}$ ) considered as a proxy of Particulate Organic Carbon (POC) (e.g. Loisel et al., 2002; Stramski et al., 2008) or phytoplankton carbon (e.g. Behrenfeld et al., 2005)
- **SOCA-BBP**: MLP linking surface  $b_{bp}(700)$  values with  $b_{bp}(700)$  vertical distribution

## Development of 2 neural networks :

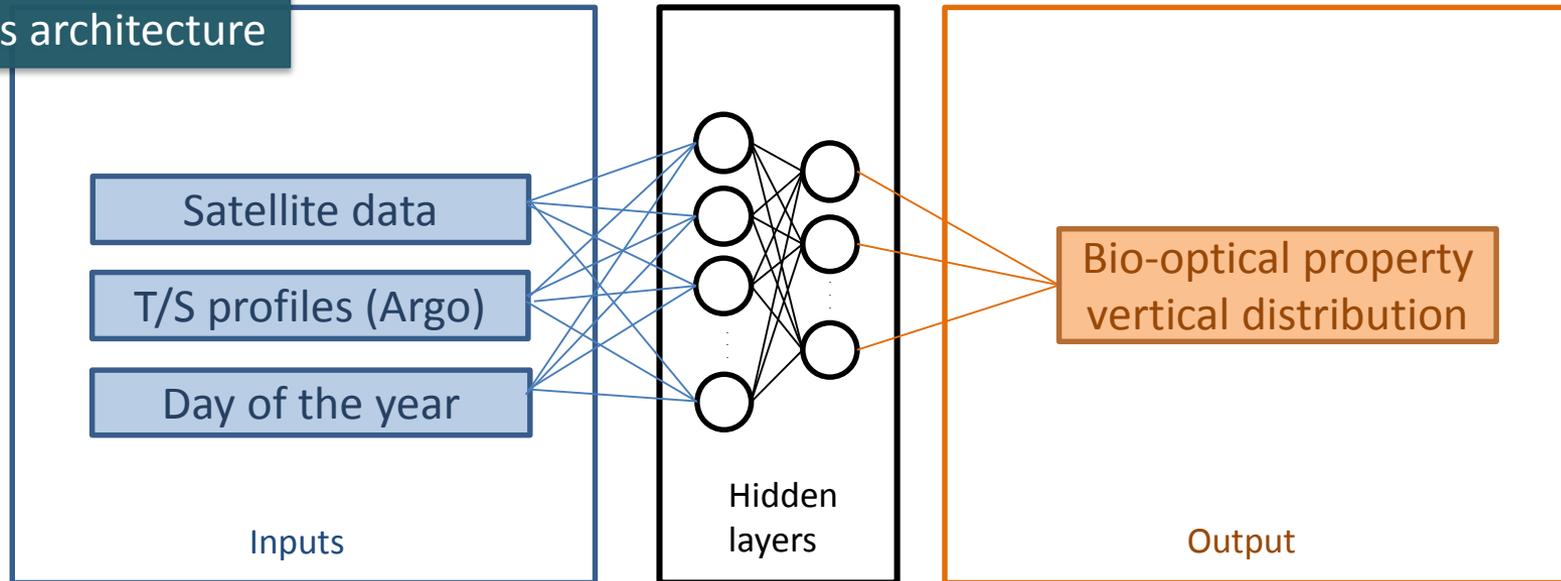
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→ **SOCA-BBP**: MLP linking surface  $b_{bp}(700)$  values with  $b_{bp}(700)$  vertical distribution
- **Chlorophyll  $a$  concentration** ([Chl]), proxy of the phytoplankton biomass (e.g. Cullen, 1982)  
→ **SOCA-CHL**: MLP linking surface [Chl] values with vertical distributions of [Chl] associated to the total phytoplankton biomass and to 3 size classes (micro-, nano- and pico-phytoplankton; Uitz *et al.*, 2006)

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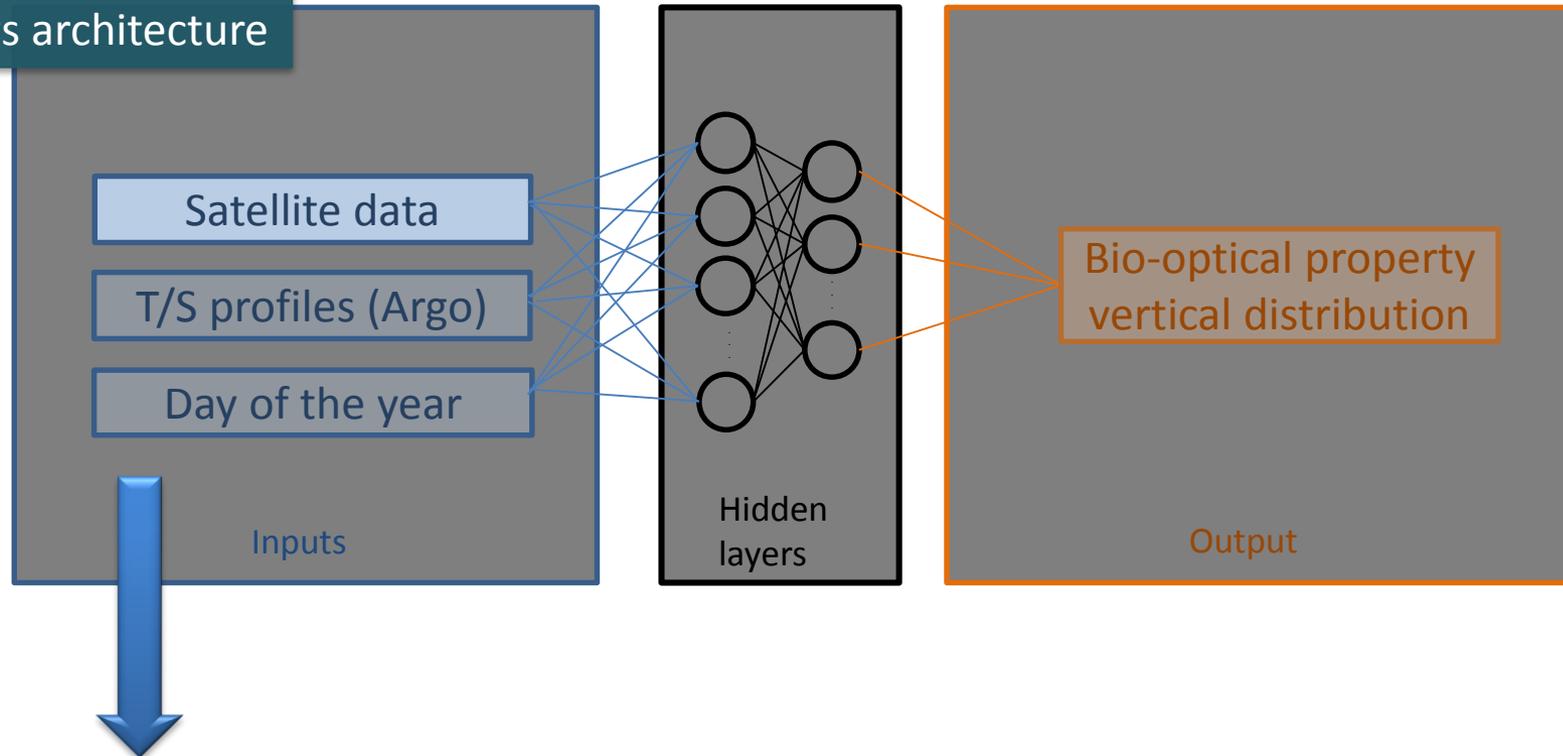
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**Argo T/S profiles and ocean color matchup**  
→ **vertical profiles of  $b_{bp}$  and [Chl]**

## MLPs architecture



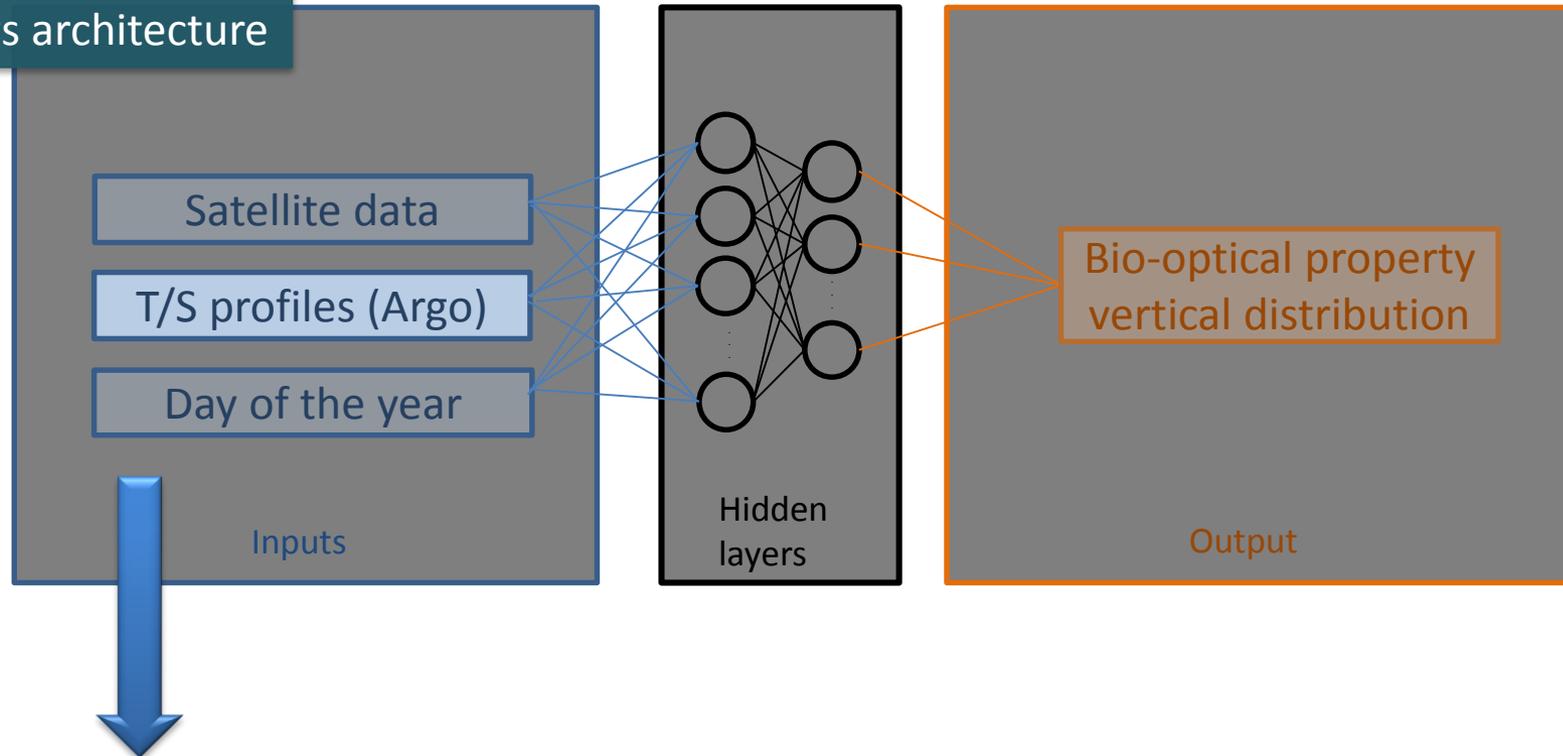
## MLPs architecture



**SOCA-BBP:**  $\log(\mathbf{b}_{bp}(700))$  derived from QAA (Lee et al., 2002,2009) and  $\log([\text{Chl}]_{sat})$

**SOCA-CHL:**  $\log([\text{Chl}]_{sat})$  and **PAR**

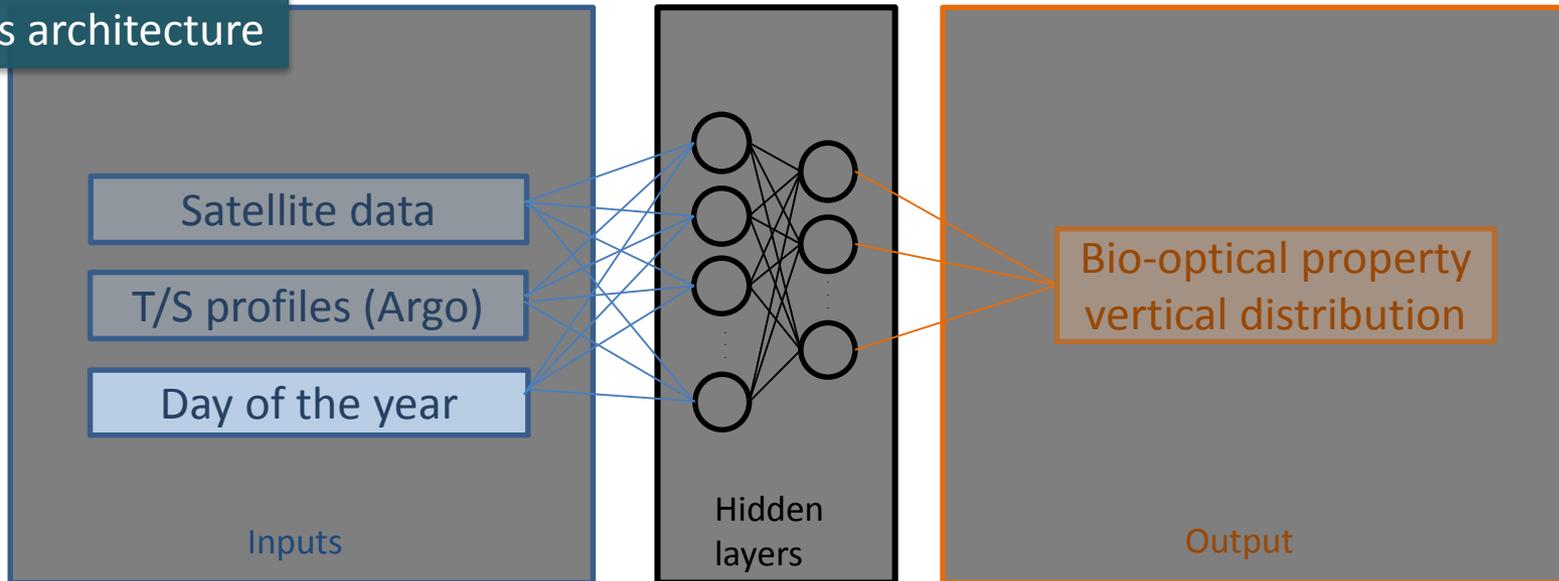
## MLPs architecture



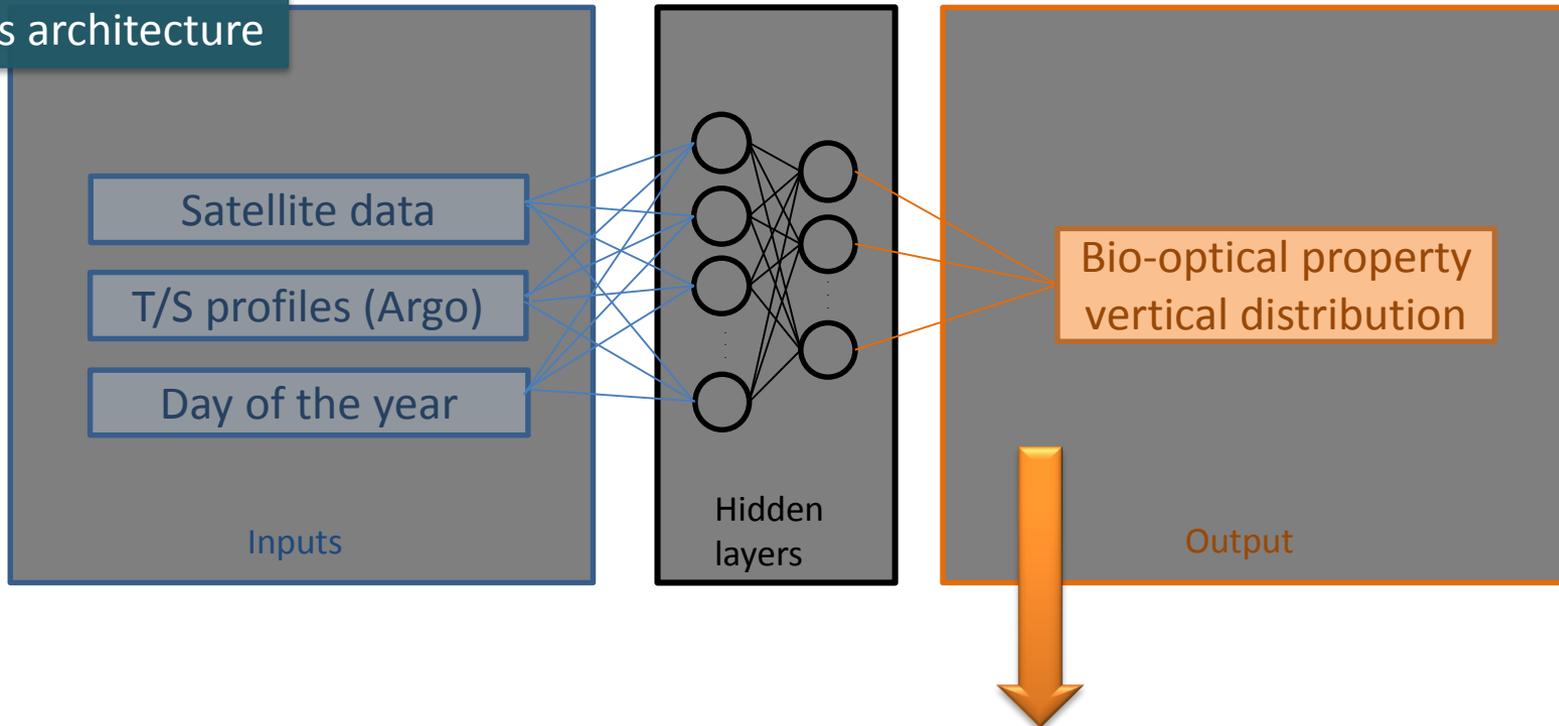
## Vertically-resolved physical properties derived from the Argo T/S profiles:

- Mixed layer depth  $Z_m$  (de Boyer Montégut *et al.*, 2004)
- 4 **density** values
- $Z_{norm}$  a normalization depth defined as the depth at which the Chl profile returns to a background value: computed empirically from  $Z_m$  and  $Z_e$  (the euphotic depth)

## MLPs architecture



## MLPs architecture



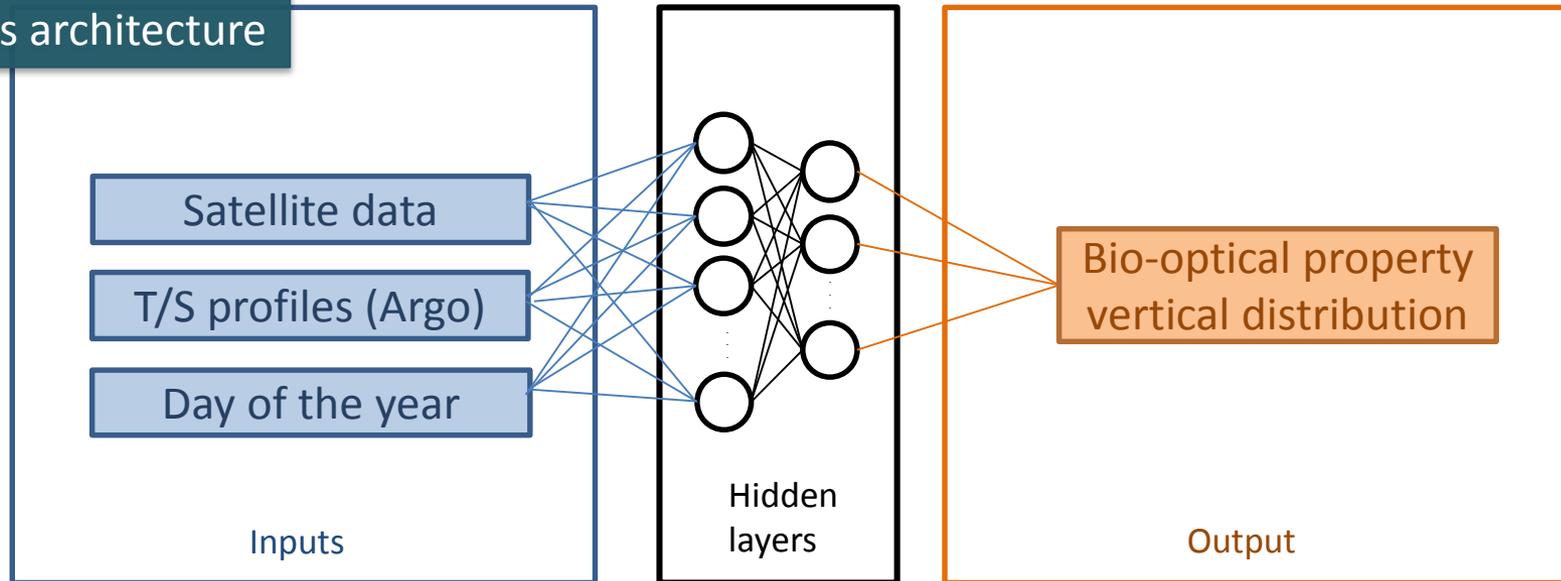
- **SOCA-BBP:**

→ 10  $b_{bp}$  values for the water column simultaneously

- **SOCA-CHL:**

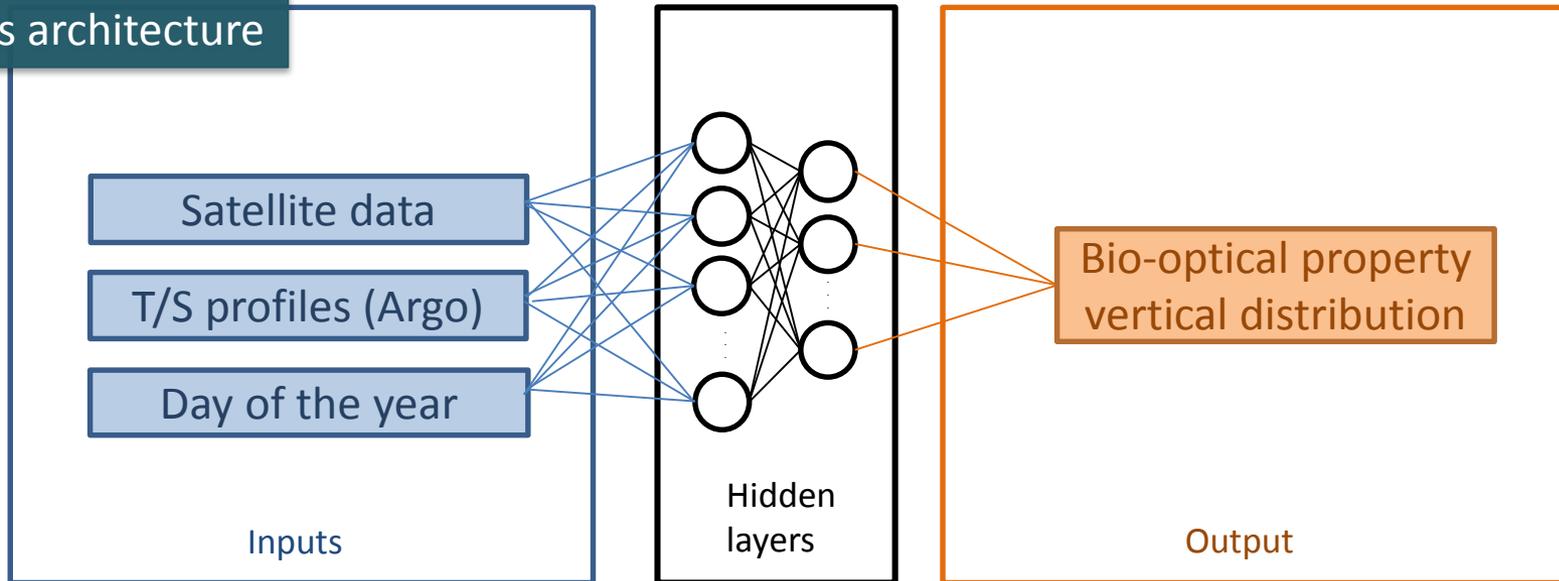
→ Chl profile shape (10 normalized Chl values) used to retrieve the Chl and the assemblage of phytoplankton communities using FLAVOR method (Sauzède et al., 2015, JGR)

## MLPs architecture



MLPs trained and validated using essentially Bio-Argo floats data.

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Be careful!

**Argo** data (only T/S) used as inputs.  
Bio-Argo floats data are only used as reference values to train and validate the MLPs.

## MLPs training and validation databases:

Concurrent profiles of density and bio-optical properties especially collected by **Bio-Argo floats**, concomitant with **satellite products** (i.e. matchups from 9-km/8-day MODIS Aqua)

MLPs **training** using 80% of the database and **validation** using 20% of the database

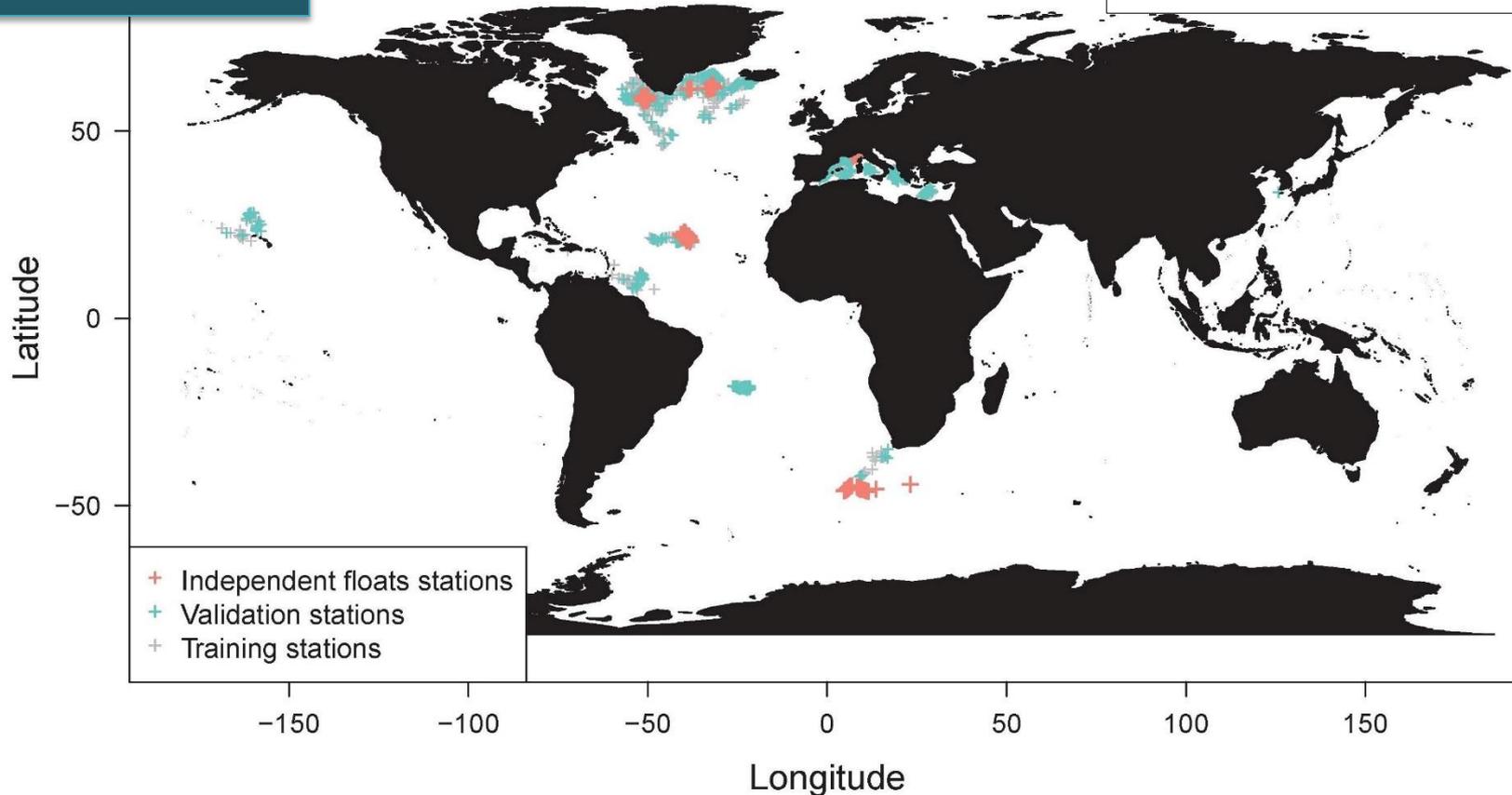
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### SOCA-BBP

3,190 satellite/Bio-Argo matchup



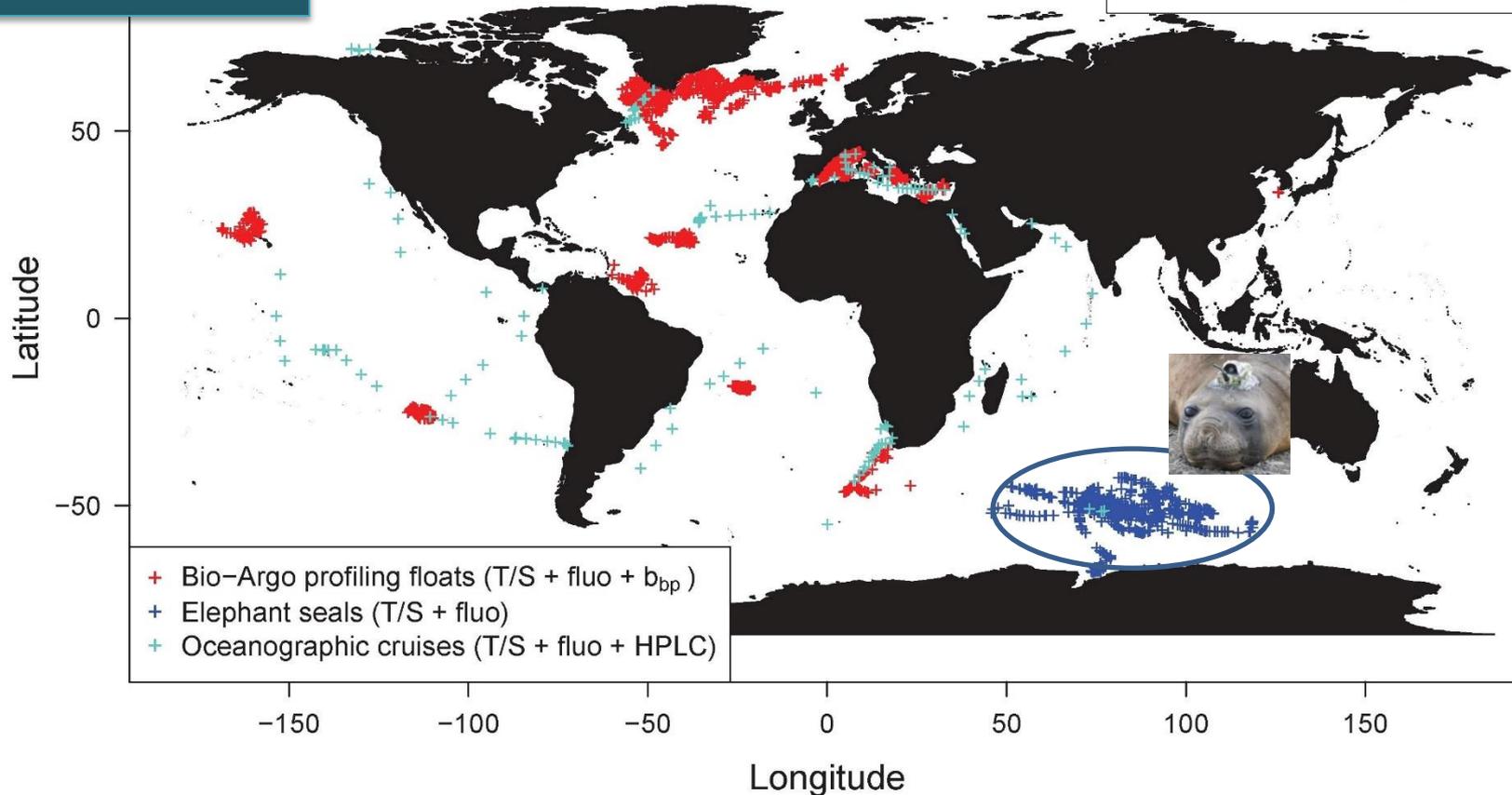
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## SOCA-CHL

4,202 satellite/profiles matchup



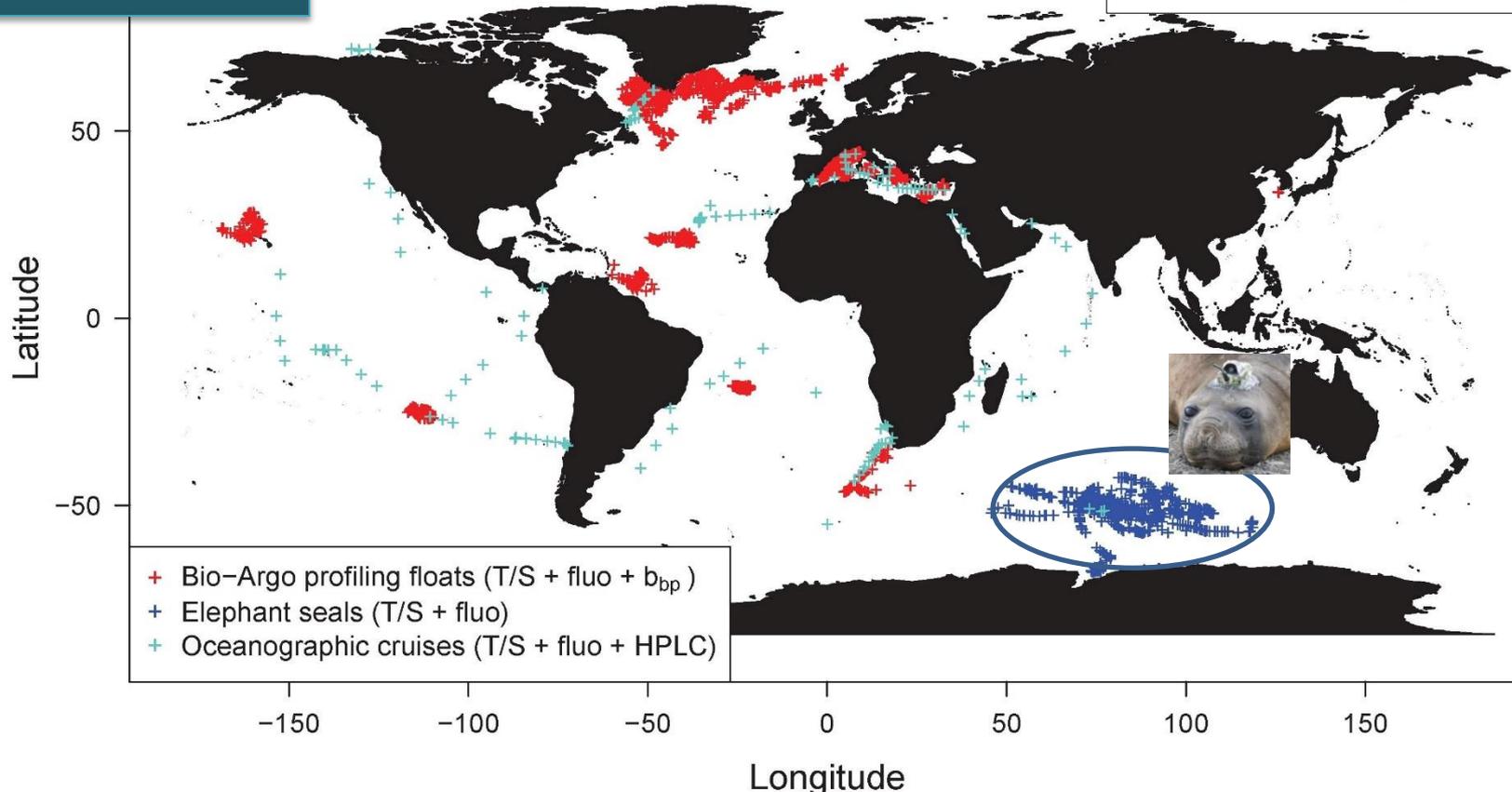
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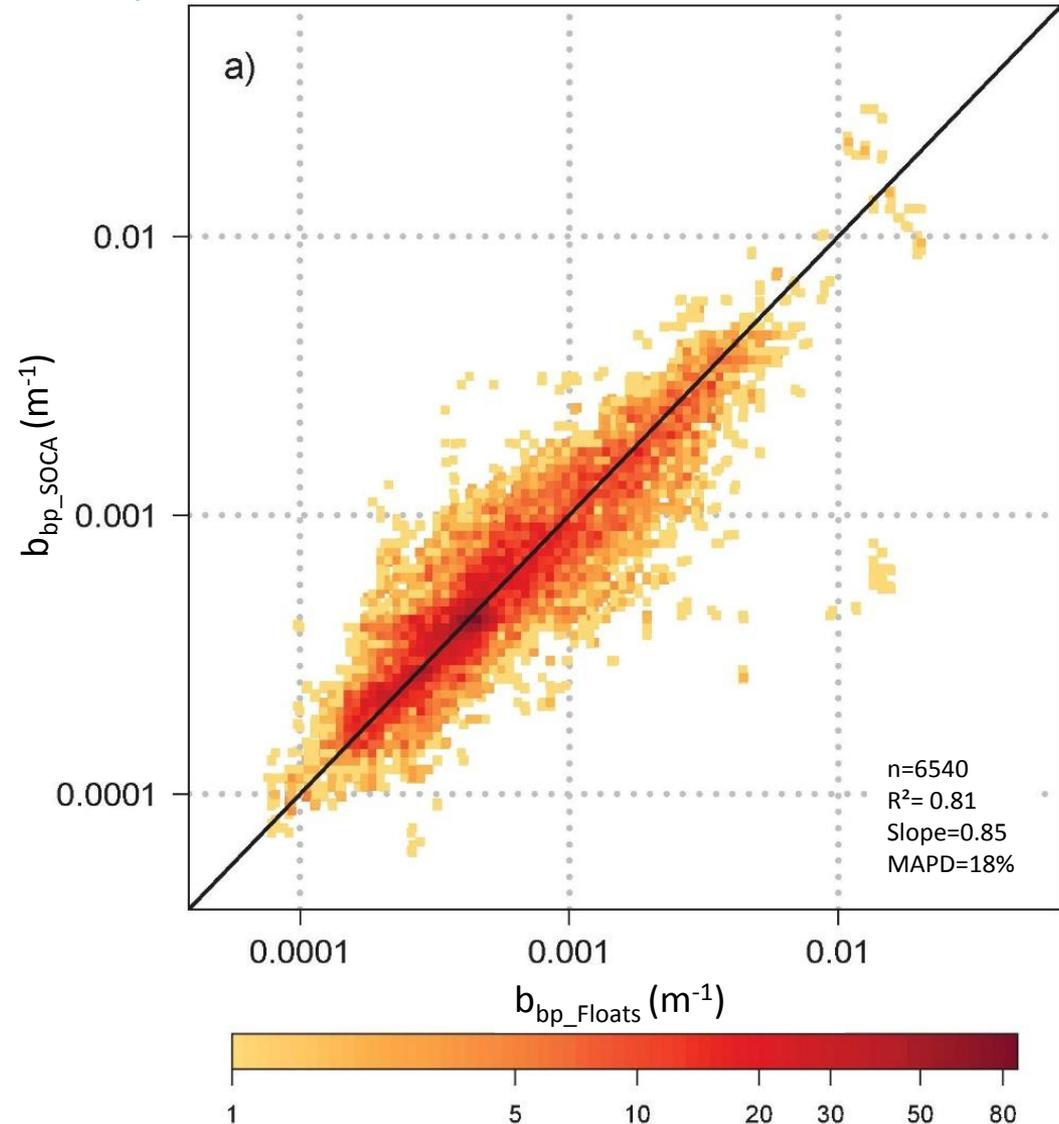
Databases representative of the global ocean → SOCA method applicable at a global scale

## SOCA-BBP

performance for  $b_{bp}$  retrieval

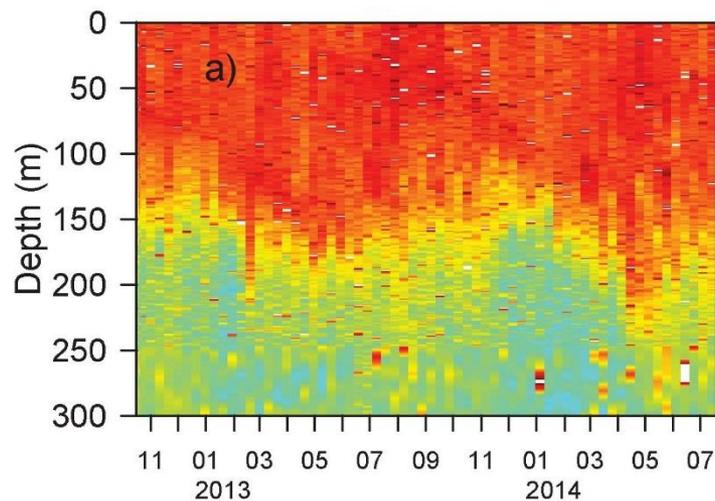
## Validation:

→ 654 profiles from the 20% of the Bio-Argo database

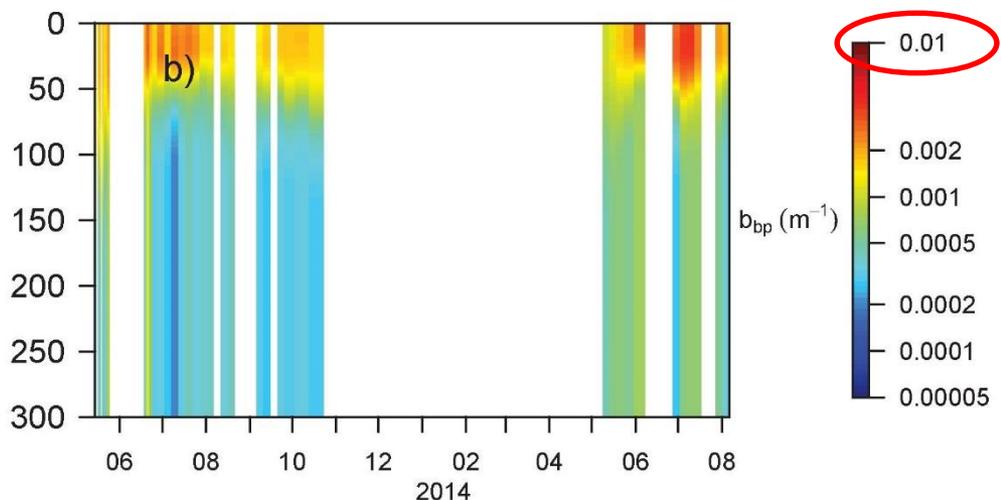
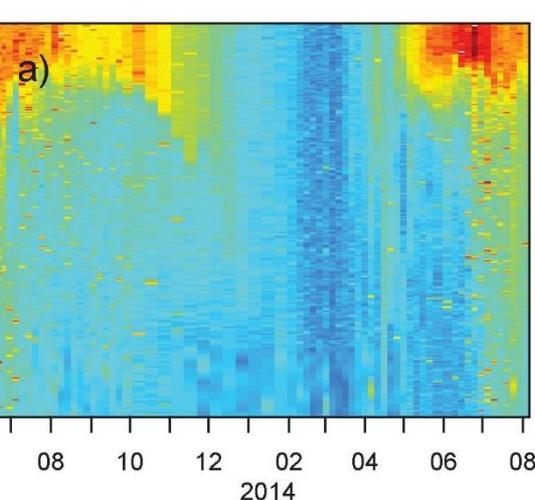
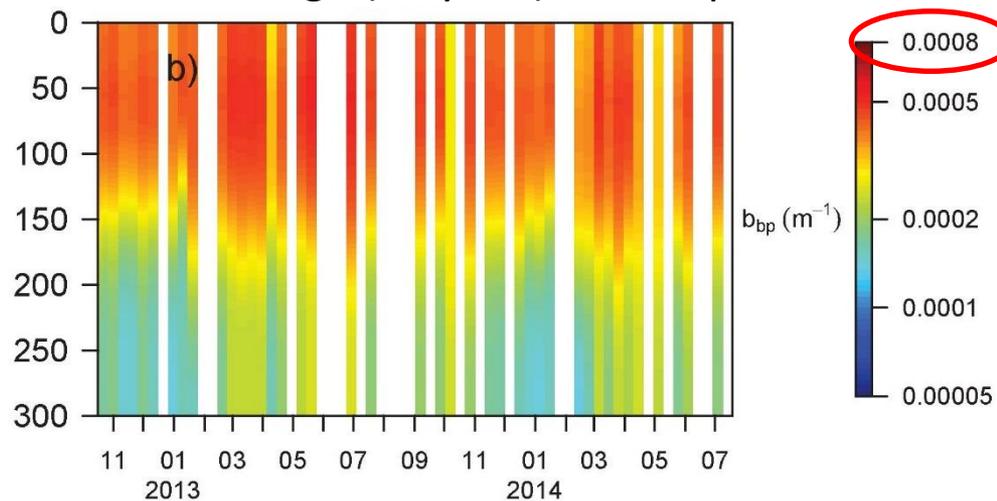


## SOCA-BBP

validation using 2 time series from the « independent » floats

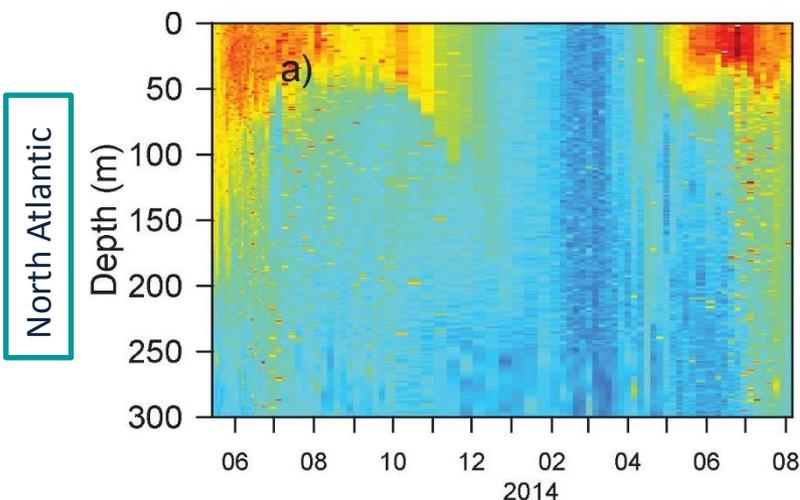
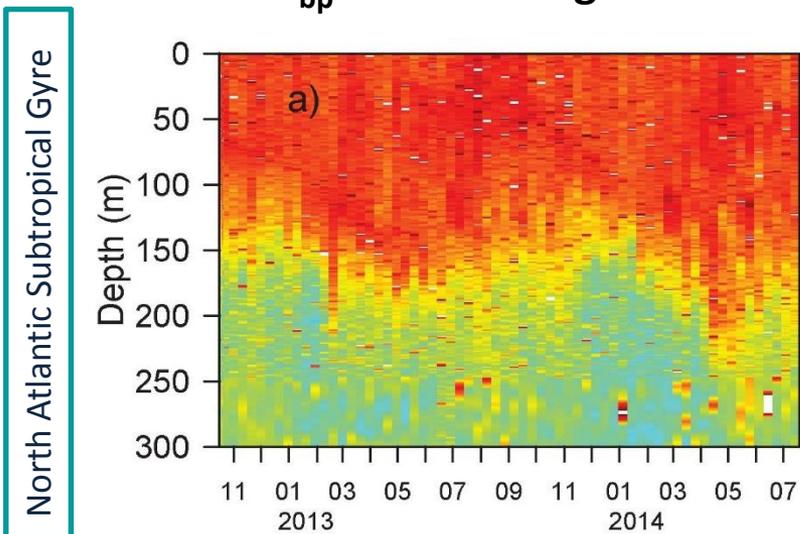
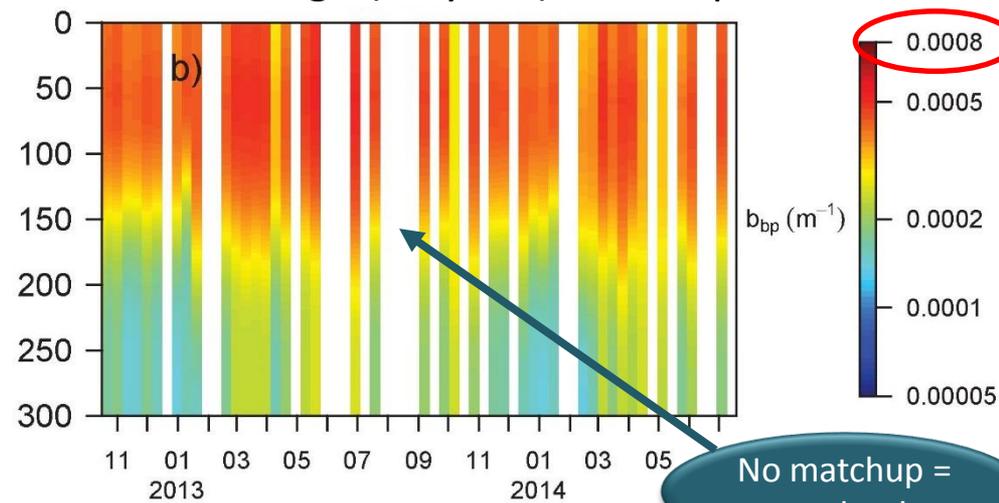
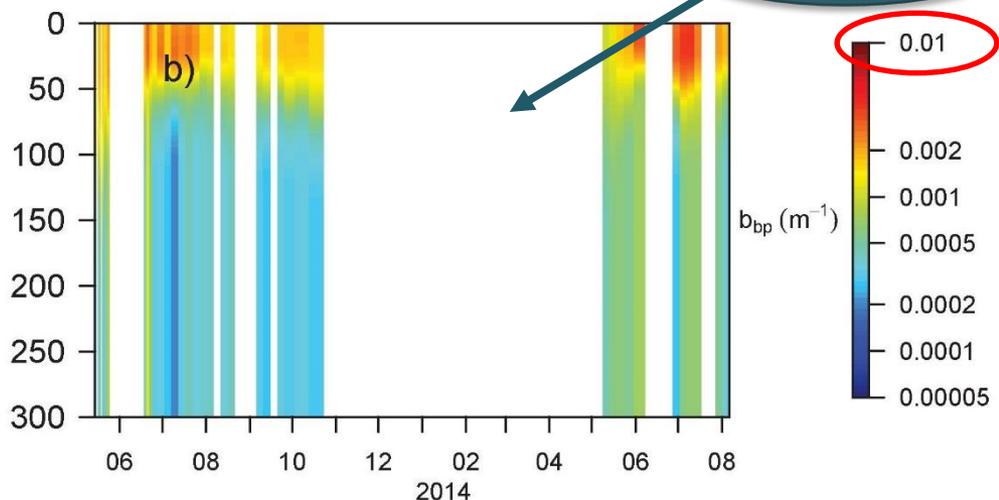
 $b_{bp}$  from Bio-Argo floats $b_{bp}$  from SOCA:

Satellite-Argo (only T/S) matchup

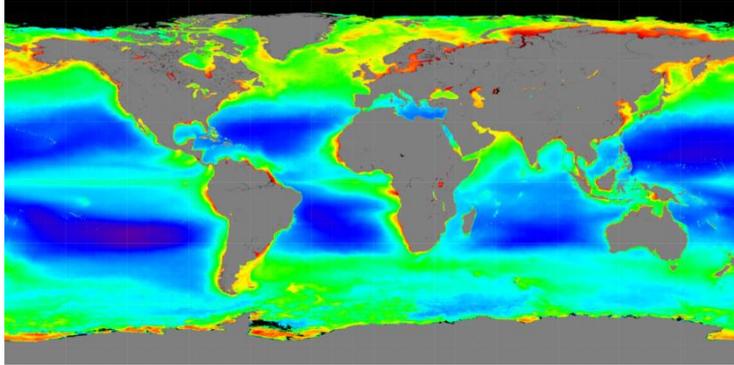


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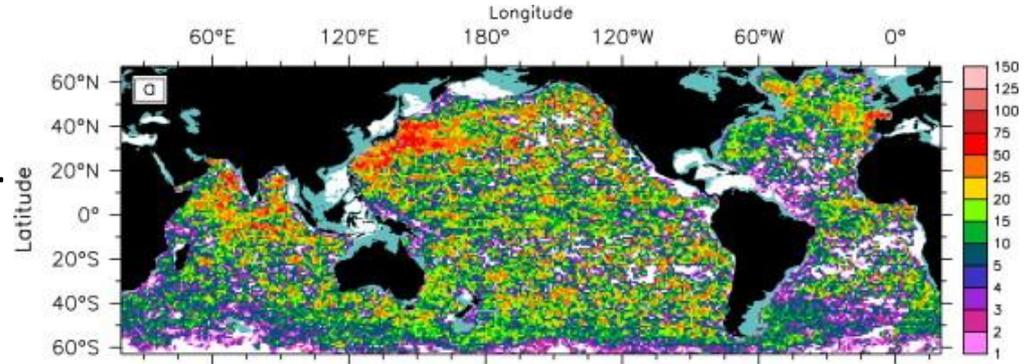
 $b_{bp}$  from Bio-Argo floats $b_{bp}$  from SOCA:  
Satellite-Argo (only T/S) matchupNo matchup =  
no data!

## SOCA-BBP

Application: climatology of  $b_{bp}$  for June at a global scale

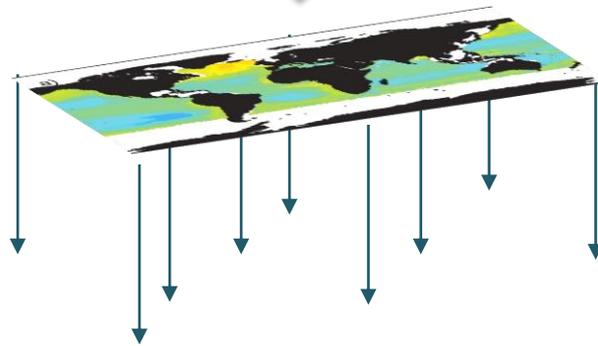
Surface ocean color climatologies

+



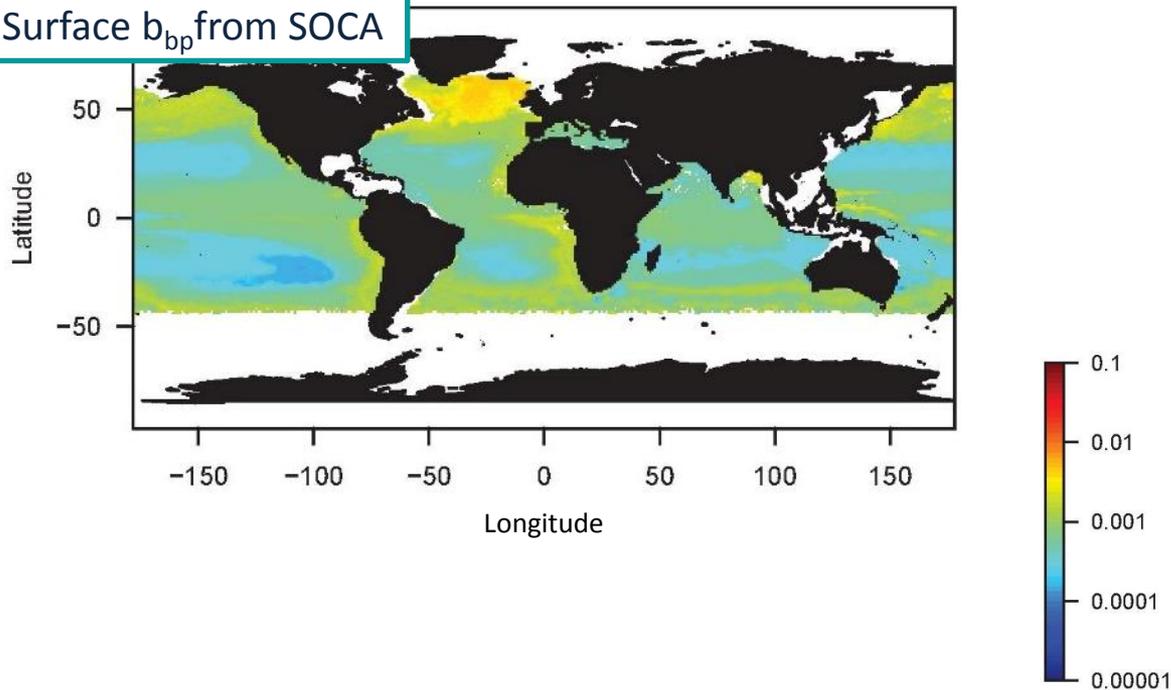
Argo T/S climatology (Roemmich and Gilson, 2009)

SOCA

3D global climatology of  $b_{bp}$

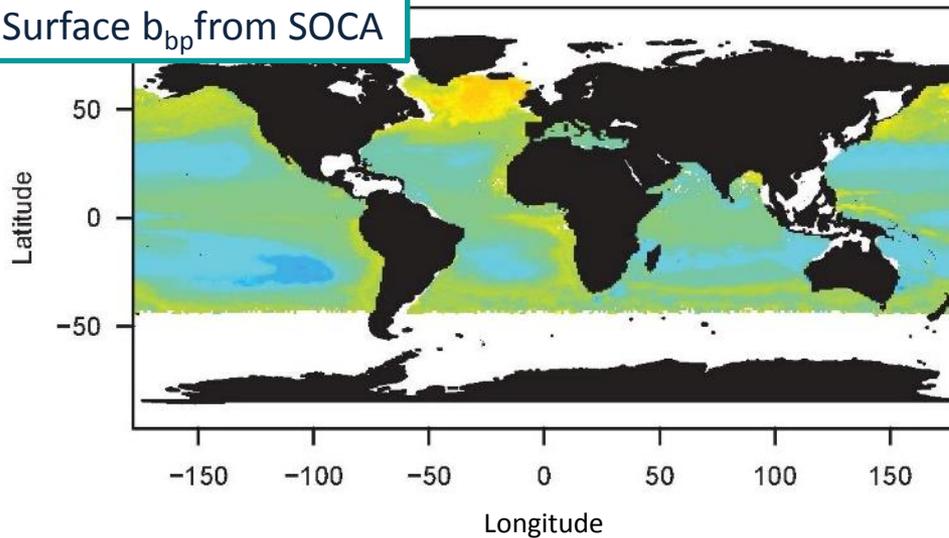
# SOCA-BBP Application: climatology of $b_{bp}$ for June at a global scale

Surface  $b_{bp}$  from SOCA

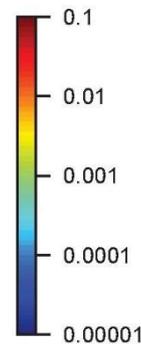
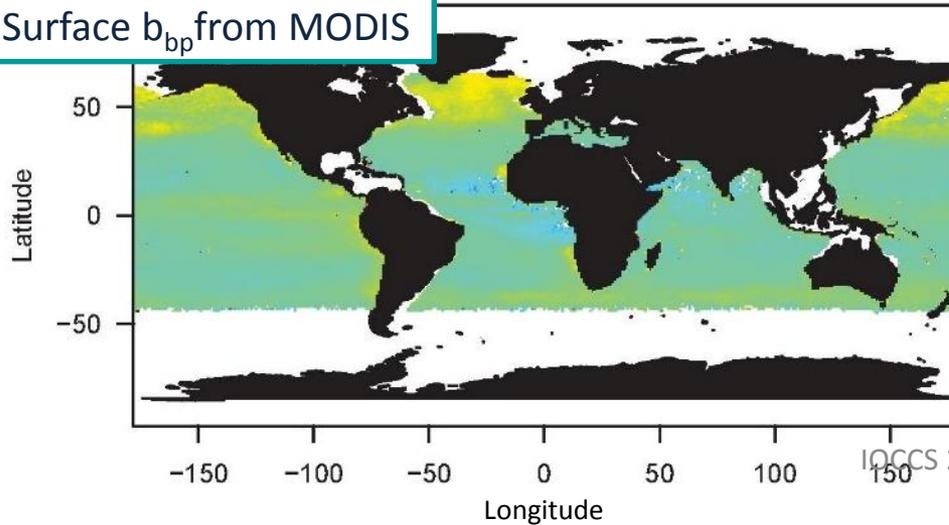


# SOCA-BBP Application: climatology of $b_{bp}$ for June at a global scale

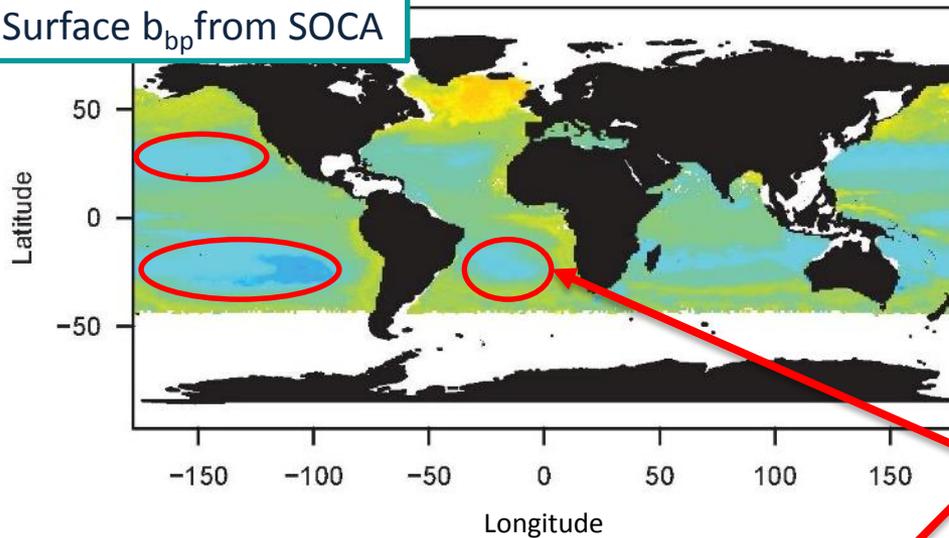
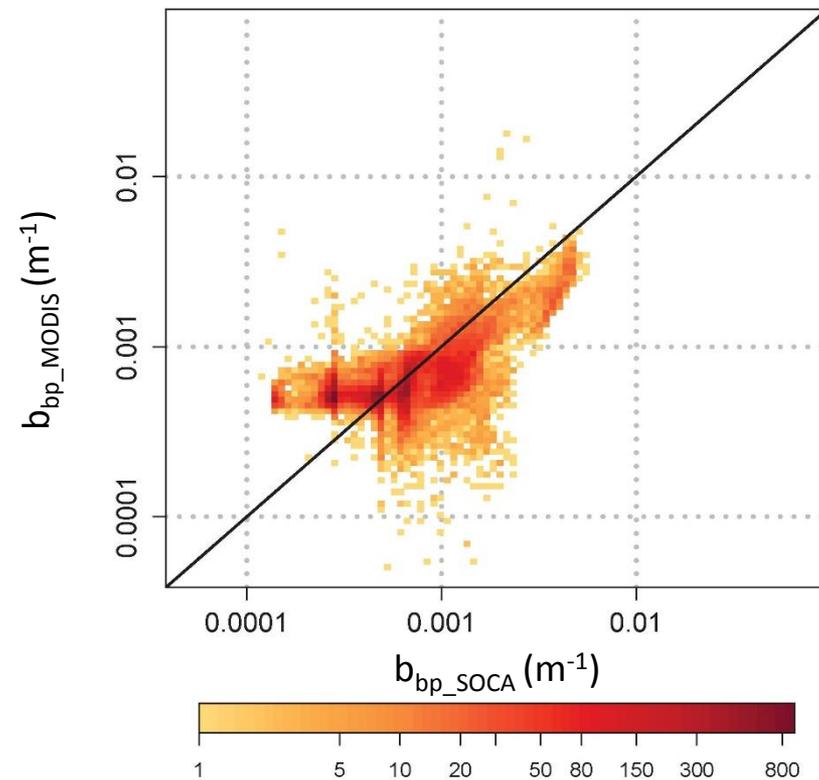
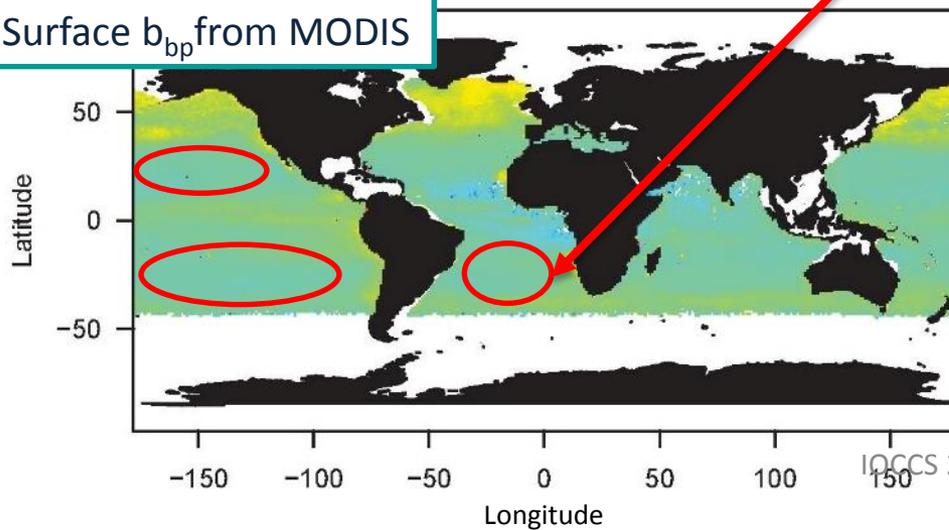
Surface  $b_{bp}$  from SOCA



Surface  $b_{bp}$  from MODIS



# SOCA-BBP Application: climatology of $b_{bp}$ for June at a global scale

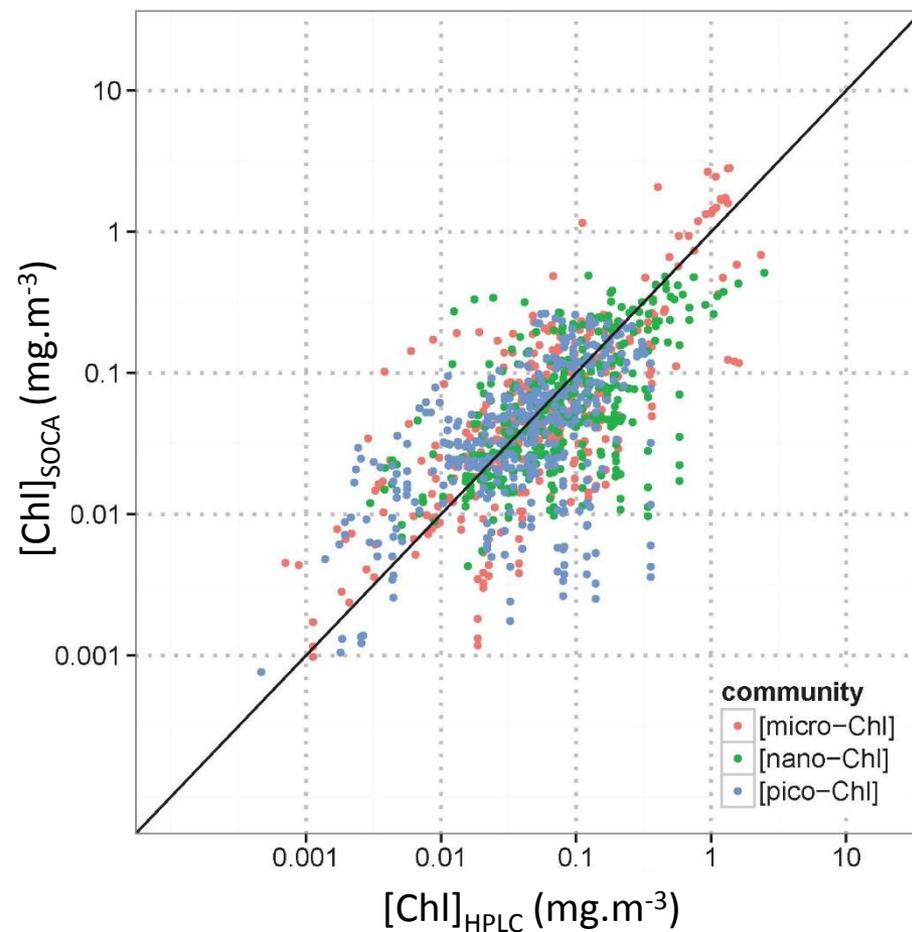
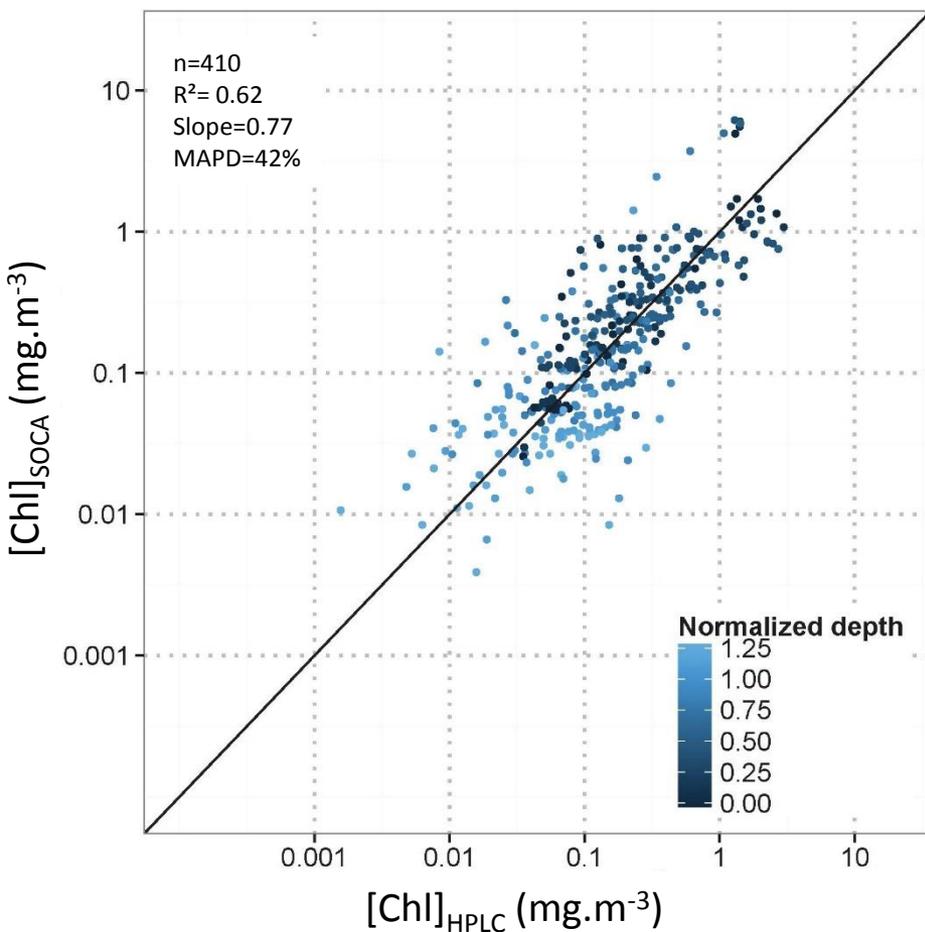
Surface  $b_{bp}$  from SOCASurface  $b_{bp}$  from MODIS

## SOCA-CHL

performance for [Chl] and assemblage of phytoplankton communities retrievals

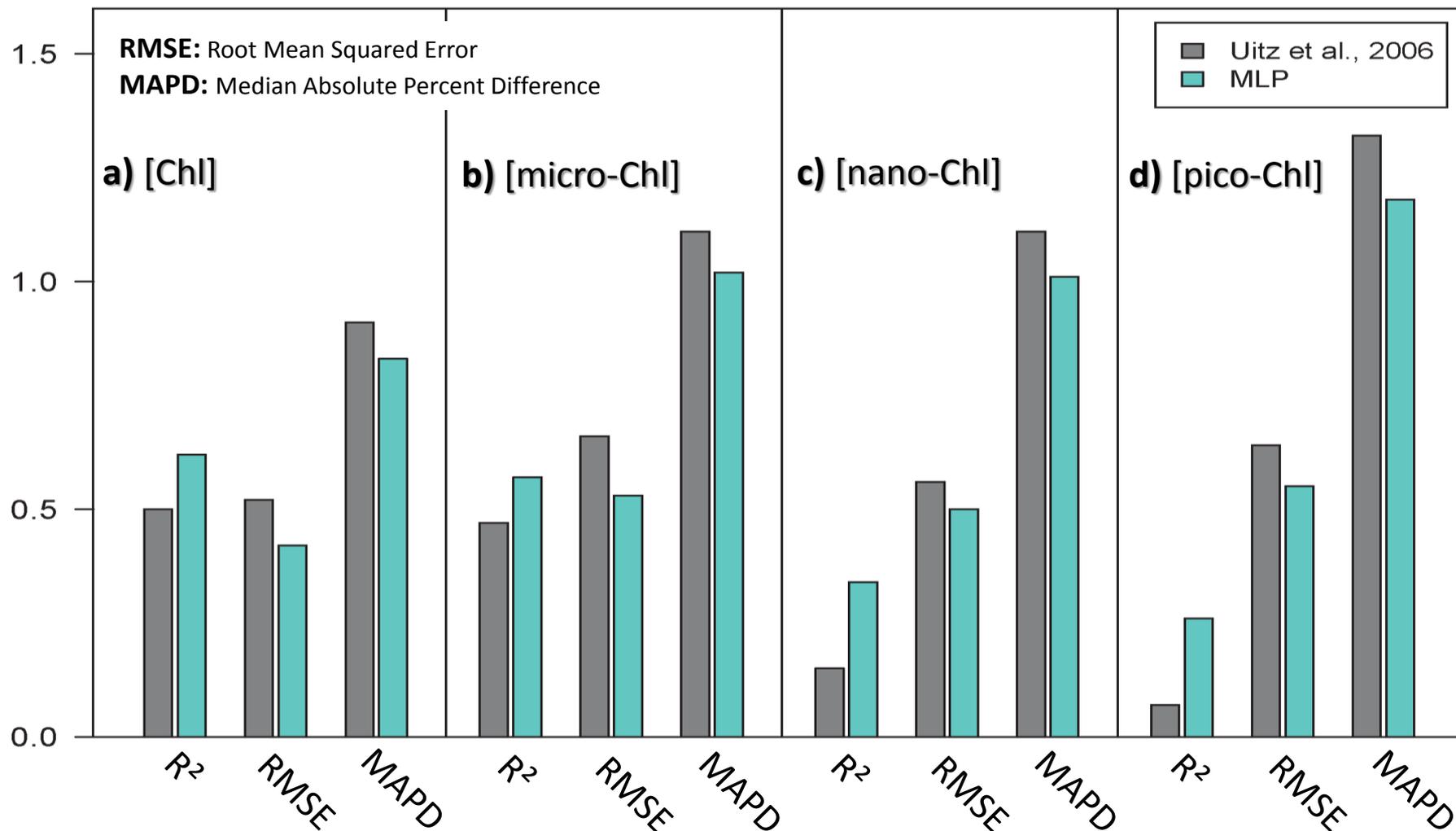
Validation of final results with cruises data (HPLC)

→ finally, 41 profiles to validate SOCA-CHL



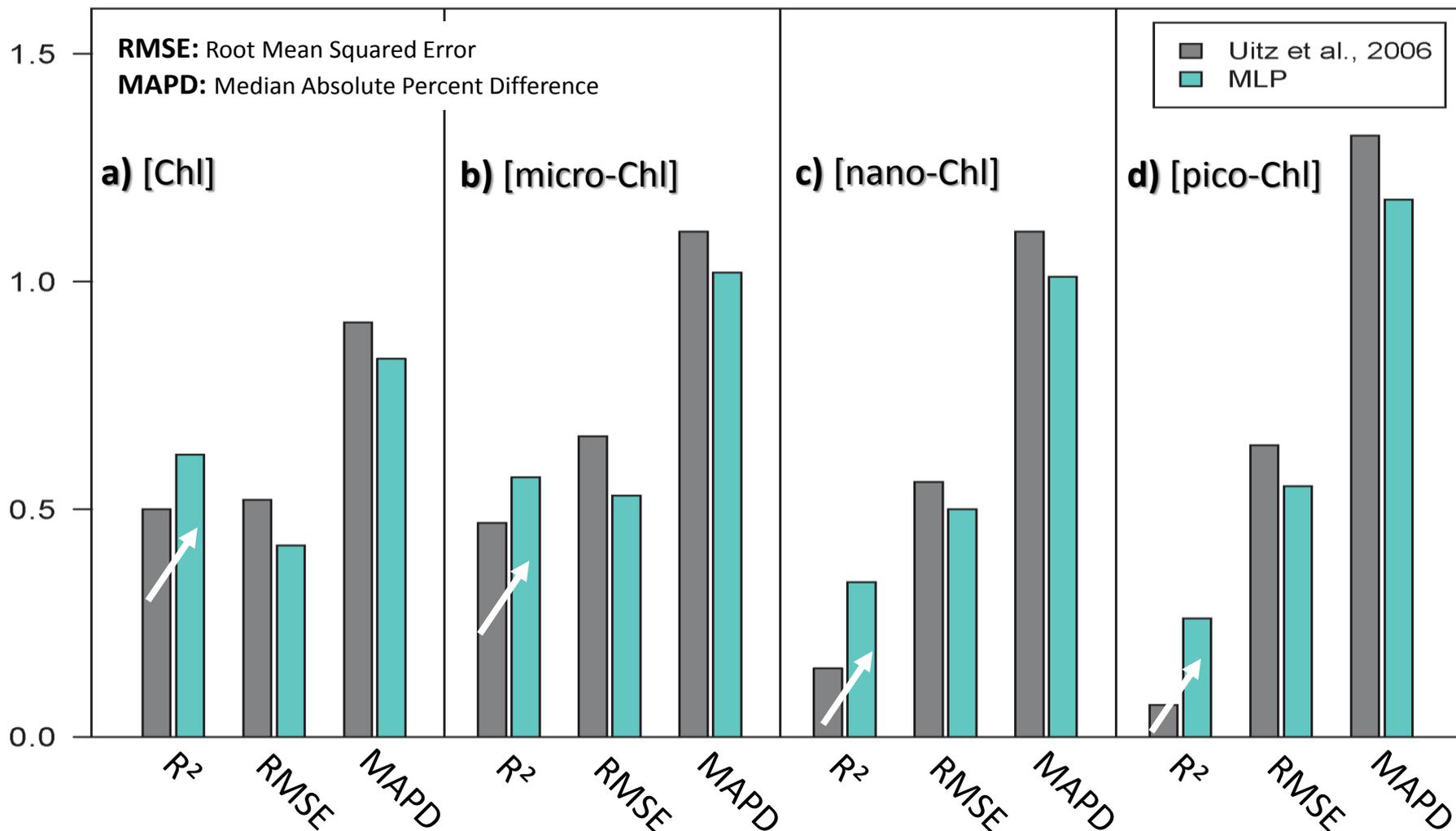
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performance for [Chl] and assemblage of phytoplankton communities

→ Comparison with the parameterization of Uitz *et al.* (2006)

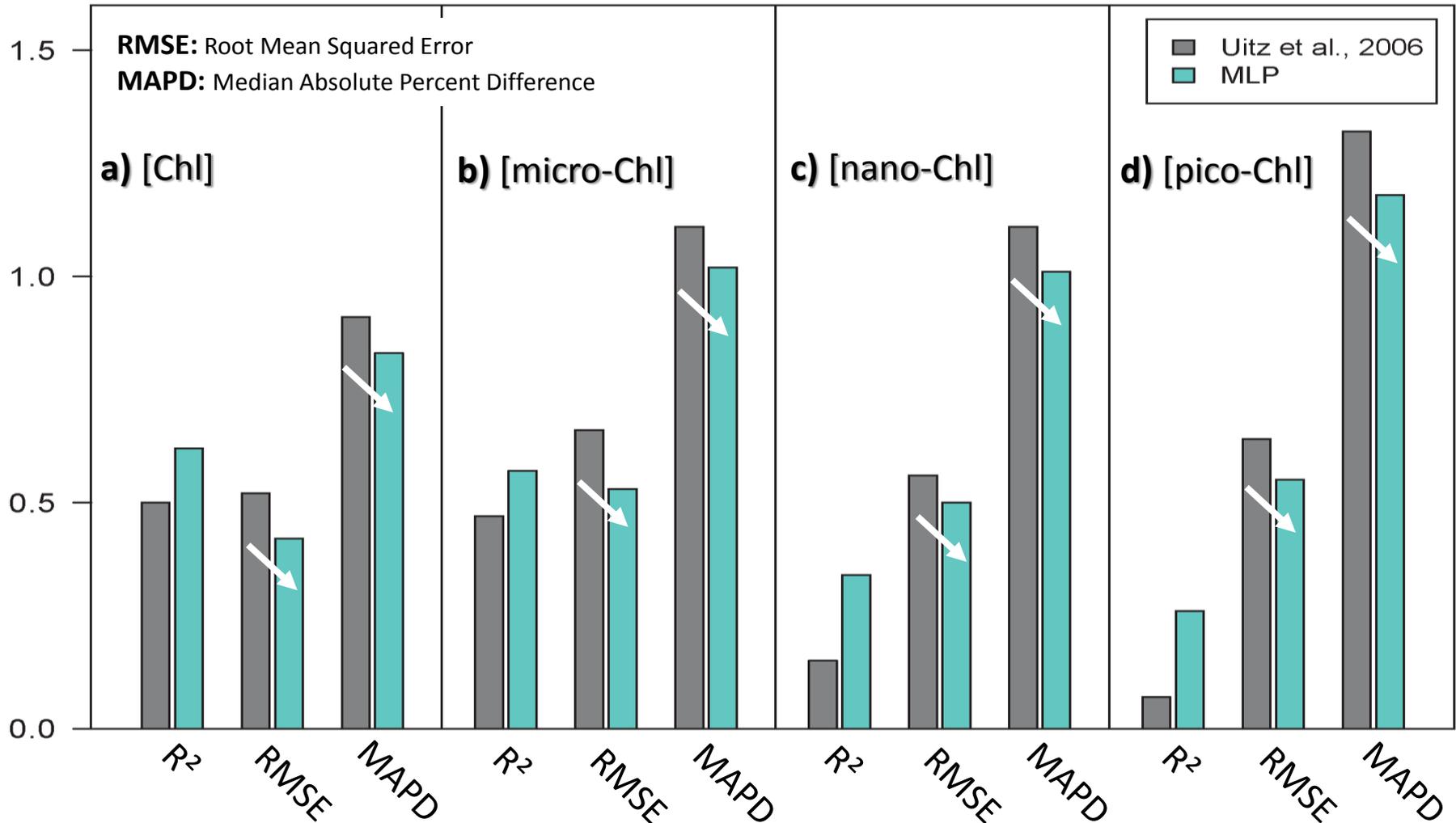
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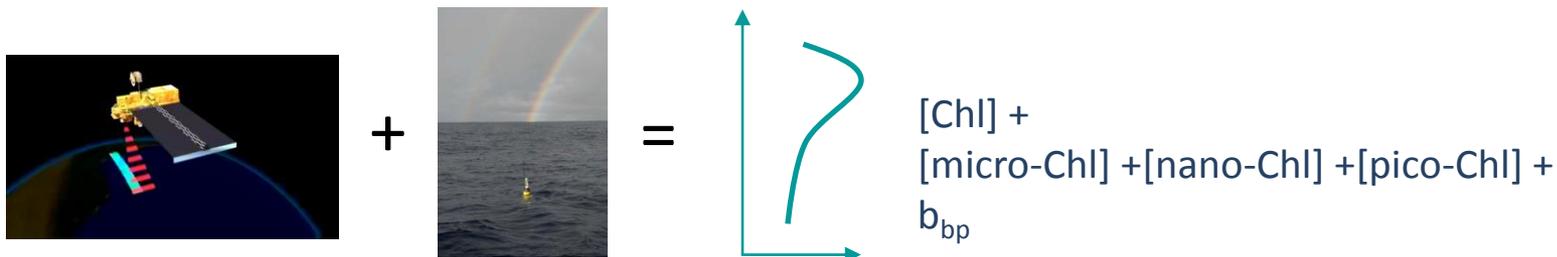
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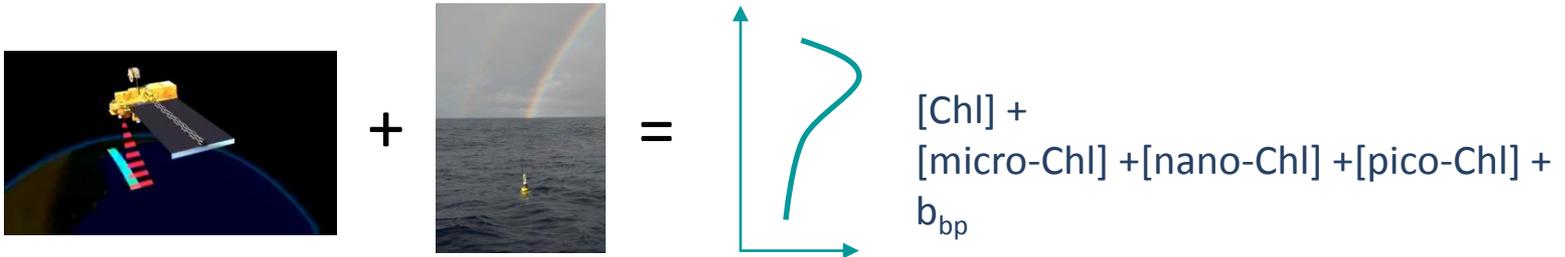
## Conclusion and perspectives :



**Argo T/S profiles + satellite matchup**

**→ Profiles of  $b_{bp}$  + [Chl] + phytoplankton communities**

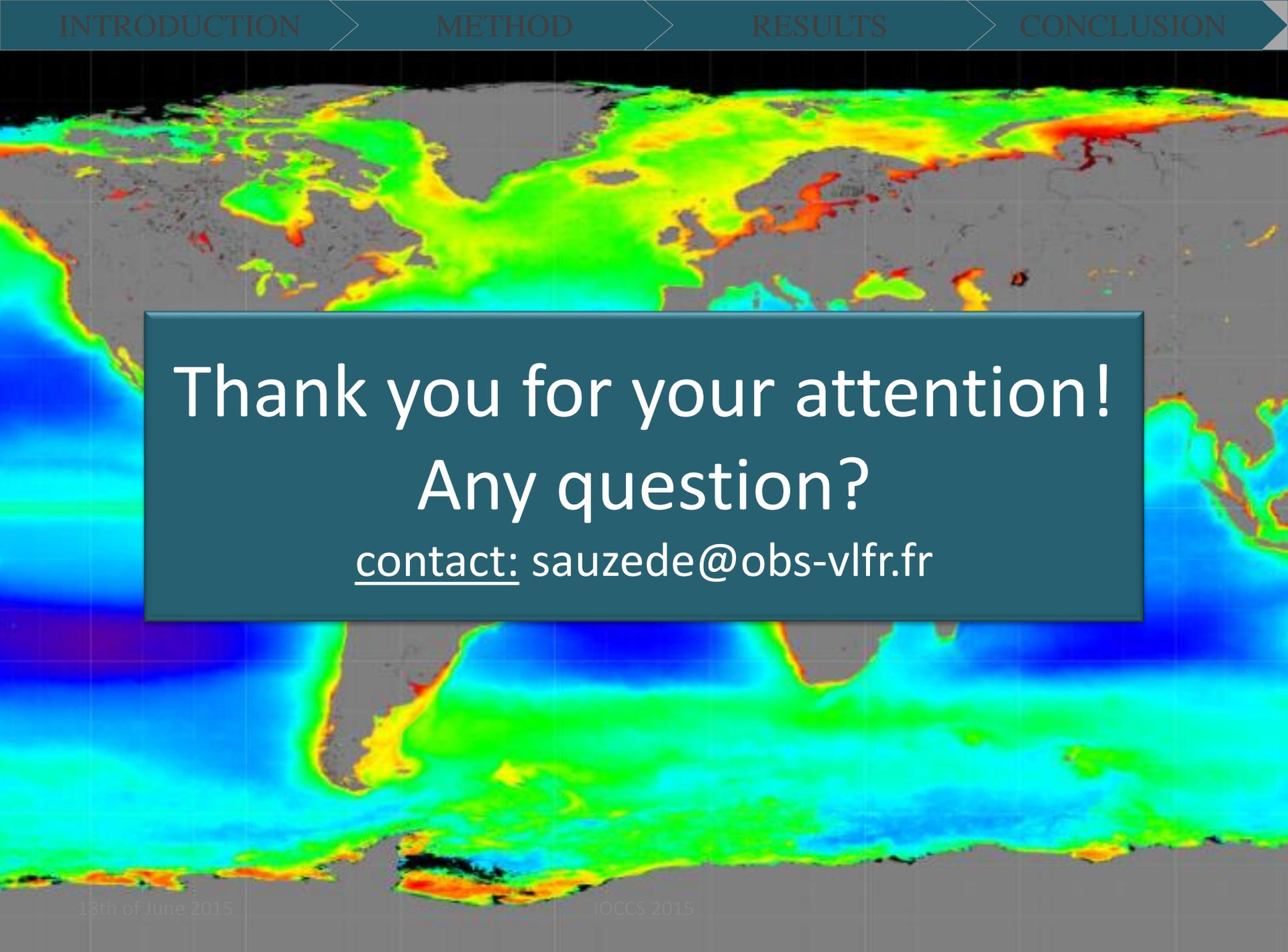
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### Argo T/S profiles + satellite matchup

#### → Profiles of $b_{bp}$ + $[Chl]$ + phytoplankton communities

- SOCA-BBP: first results of the application of SOCA-BBP at the regional and global scales are very promising. SOCA-BBP suggests an overestimation of  $b_{bp}$  by satellite in the subtropical gyres.
- SOCA-CHL: improvement of the retrieval of the  $[Chl]$  associated with the total phytoplankton biomass and with the three size classes compared to the model of Uitz *et al.* (2006).
- SOCA method have multiple potential applications such as the initialisation/validation of biogeochemical models, the retrieval of POC from  $b_{bp}$  ...
- Improvement of SOCA performance using more Bio-Argo profiles (available soon)



Thank you for your attention!

Any question?

contact: [sauzede@obs-vlfr.fr](mailto:sauzede@obs-vlfr.fr)