

Mechanistic Drivers of the Particulate Backscatteringto-Chlorophyll *a* Relationship and Bias-assessment of Phytoplankton Carbon Algorithms

Camila Serra-Pompei<sup>1</sup>, Anna Hickman<sup>2</sup>, Gregory Britten<sup>1</sup>, Stephanie Dutkiewicz<sup>1</sup>

<sup>1</sup>Massachusetts Institute of Technology (MIT, USA). <sup>2</sup>National Oceanography Centre, University of Southampton (UK)

The particulate backscattering coefficient  $(b_{bp})$  has been suggested to be a good proxy for phytoplankton carbon biomass  $(C_{phyto})$  and is used in some NPP global models.

But  $C_{phyto}$  data is scarce in space and time, therefore  $b_{bp}$ -Chl relationships can be used to derive these algorithms

Problems:

- $b_{bp}$  is a proxy of all particles in the ocean, not only phytoplankton
- Scarce C<sub>phyto</sub> field data and biased in space and time

We use a global ocean circulation model (MITgcm) with optics embedded in it and the Bgc-Argo data-set to:

- Understand how well does  $b_{bp}$  estimate  $C_{phyto}$
- $\circ$   $\;$  Understand  $b_{bp}$  -Chl relationships and their potential to obtain  $\mathcal{C}_{phyto}$

# Is $b_{bp}$ a good proxy of Cphyto?

- We use the MITgcm model to investigate  $b_{bp}$ - $C_{phyto}$  relationship
- We get an algorithm by fitting a linear regression to the  $b_{bp}$ - $C_{phyto}$  relationship of the surface pixels of the MITgcm
- We compare the C<sub>phyto</sub> estimated by the algorithm relative to the "real" C<sub>phyto</sub> of the model



#### Results:

- Algorithm tends to deviate by a factor less than 2 in most regions
- Worst fits are in winter of high latitudes, in those regions phytoplankton has a very low contribution to  $b_{bp}$  (where heterotrophic bacteria and detritus dominate the  $b_{bp}$  signal)

#### Understanding the $b_{bp}$ -Chl relationship

- Use the Darwin model to understand the  $b_{bp}$ -to-Chl relationship
- Use BGC-Argo data to compare trends



Results:

- A linear regression in the linear scale fits relatively well the b<sub>bp</sub>-Chl trend (i.e. we do not get a bi-linear trend in the linear scale as observed in other studies)
  - This is also seen in the Argo data-set
- The bi-linear trend in the log-scale is a visual artifact from having a positive intercept. This positive intercept is mainly driven by a background b<sub>bp</sub> of NAPs
- Once the background b<sub>bp</sub> is removed, the trend becomes somewhat linear and a log-log fit seems better. Still, large variability.





# Bgc-Argo data-set

Log<sub>10</sub>b<sub>bp</sub>(700) [m<sup>-1</sup>]

-2

-1

## Using $b_{bp}$ - Chl to get $C_{phyto}$ ?

## Algorithm $b_{bp}$ - $C_{phyto}$

Month=" R<sup>2</sup>=0.90, RMSD=0.16 (-50<Lat.<65) R<sup>2</sup>=0.90, RMSE=0.23 (-90<Lat.<90)



Month=4 R<sup>2</sup>=0.94, RMSD=0.14 (-50<Lat.<65) R<sup>2</sup>=0.86, RMSE=0.21 (-90<Lat.<90)



Month=7 R<sup>2</sup>=0.93, RMSD=0.14 (-50<Lat.<65) R<sup>2</sup>=0.83, RMSE=0.32 (-90<Lat.<90)



Month=10 R<sup>2</sup>=0.93, RMSD=0.14 (-50<Lat.<65) R<sup>2</sup>=0.78, RMSE=0.47 (-90<Lat.<90)



1/10 1/3 3 Factor difference (Cphytomod/Cphyto)

#### Biases in $b_{bp}$ -based algorithms?





## Algorithm using $b_{bp}$ - *Chl*

Month=1 R<sup>2</sup>=0.88, RMSD=0.18 (-50<Lat.<65)

Month=7



R<sup>2</sup>=0.92, RMSD=0.16 (-50<Lat.<65) R<sup>2</sup>=0.85, RMSE=0.21 (-90<Lat.<90)









Month=10 R<sup>2</sup>=0.92. RMSD=0.15 (-50<Lat.<65) R<sup>2</sup>=0.80, RMSE=0.45 (-90<Lat.<90)







• Chl-  $b_{bp}$  relationships can be used to obtain  $C_{phyto}$  with similar performance if we had real  $C_{phyto}$  data

• The problem is how to convert  $b_{hp}$  to C<sub>phyto</sub> ...



b<sub>bp</sub>(470) [m<sup>-1</sup>]

×10<sup>-3</sup>

- Assumptions regarding conversions from  $b_{bp}$  to  $C_{phyto}$  can result in large differences
- Sampling bias in Graff et al 2005 did not have a strong effect in algorithm performance (tested, but not shown here)  $\rightarrow$  differences in the linear regression across regions might not be that high



# **Conclusions:**

- *b<sub>bp</sub>*-based algorithm deviates by a factor of 2 in most regions
  - Some regions are heavily overestimated, specially winter high latitudes (b<sub>bp</sub> signal is dominated by detritus and heterotrophic bacteria)
- No bi-linear trend at linear scale (either in the MITgcm or Argo data):
  - Bi-linear trend in the log scale emerges from having a positive intercept
- Algorithm derived from Argo data is similar to two of the existing algorithms, but assumptions on conversion factors need to be better constrained
- Sampling biases do not have a strong effect on the overall performance of the algorithm.

# Knowledge gaps and next steps:

- Clearly, the limiting factor is the lack of C<sub>phyto</sub> data
- Use the Argo b<sub>bp</sub>-Chl data to see differences across regions/biomes
- Explore what has the largest uncertainty to estimate  $C_{phyto} \rightarrow b_{bp}$  or Chl?
  - Uncertainties related to  $b_{bp}$  seem similar to the ones driven by differences in Chl: $C_{phyto}$  ratios
  - However, b<sub>bp</sub> gives a notion of Chl:C<sub>phyto</sub> ratios