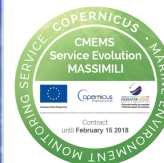
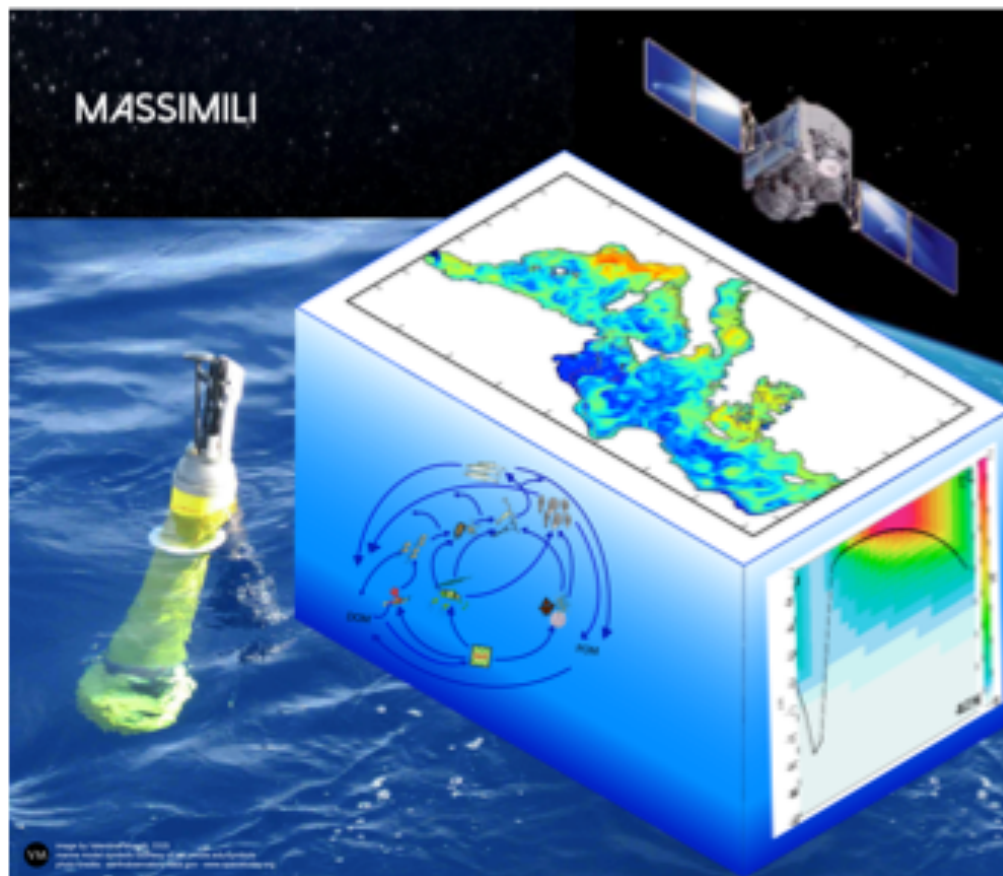


# A multi-data assimilation scheme of Biogeochemical-Argo data for the CMEMS Mediterranean Biogeochemical model

Mariotti L.<sup>1</sup>, Cossarini G.<sup>1</sup>, Salon S.<sup>1</sup>, D'Ortenzio F.<sup>2</sup>, A. Mignot<sup>2</sup>

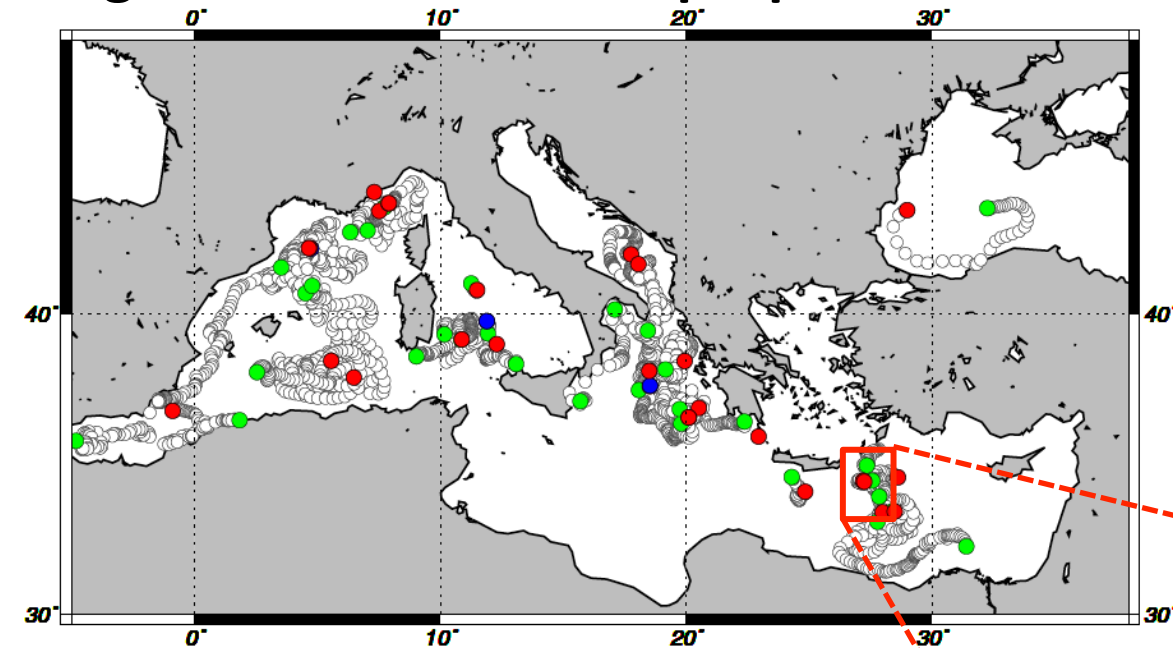
<sup>1</sup>Istituto Nazionale di Oceanografia e di Geofisica Sperimentale (Italy)

<sup>2</sup>Laboratoire d'Océanographie de Villefranche-sur-Mer (France)

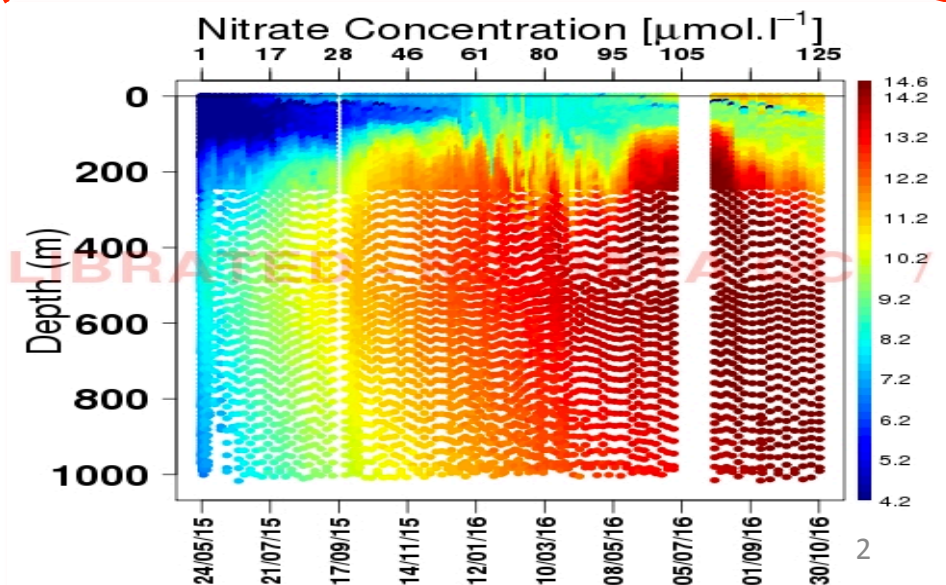


**MASSIMILI: a CMEMS  
service evolution project**

# BGC-Argo floats: operational observations of the biogeochemical vertical properties

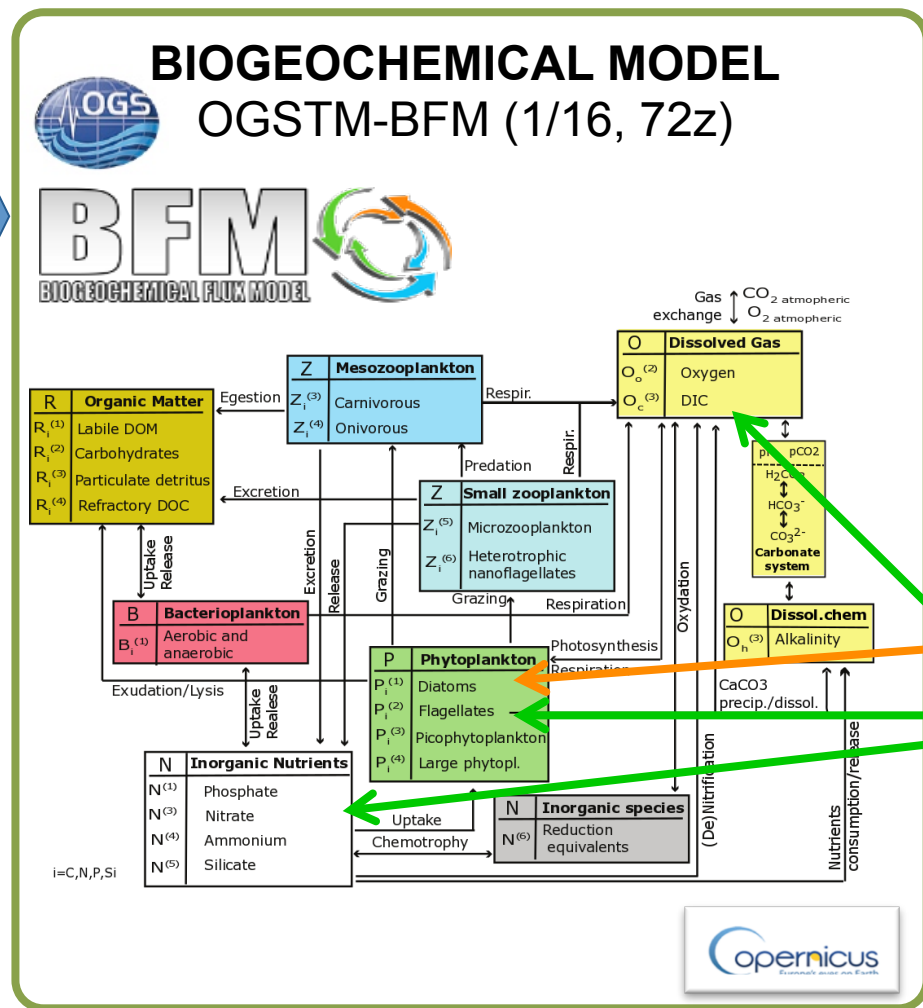


**# BGC-Argo floats:** 12-16  
**BGC vars:** Chla (14), Oxy (12), Nit (8)  
**Mission cycle:** ~ 5 days  
**Parking depth:** 1000m (slow moving)  
**Life-time:** > 2years



**MED-CURRENT**  
Physical forcing

INGV  
terremoti  
vulcani  
ambiente  
ISTITUTO NAZIONALE  
DI GEOPISICA E VULCANOLOGIA



**SATELLITE OBSERVATIONS**  
 Surface chlorophyll concentrations

**ASSIMILATION**  
 3D-VAR scheme

**BGC-ARGO DATA**  
 CHL, [NO<sub>3</sub>, OXY]

**MASSIMILI project: upgrading the 3DVAR-BIO to integrate BGC-Argo float data**

# 3DVAR-BIO code: variational assimilation scheme



The analysis state is given by the minimization of the cost function

$$J(\delta\mathbf{x}_k) = \frac{1}{2} \delta\mathbf{x}_k^T \mathbf{B}_k^{-1} \delta\mathbf{x}_k + \frac{1}{2} (\mathbf{d}_k - \mathbf{H} \delta\mathbf{x}_k)^T \mathbf{R}_k^{-1} (\mathbf{d}_k - \mathbf{H} \delta\mathbf{x}_k)$$

$\delta\mathbf{x} = \mathbf{x} - \mathbf{x}_b$  : perturbation in model space around the background state  $\mathbf{x}_b$

$\mathbf{B}$  and  $\mathbf{R}$  : background and observation error covariance matrices

$\mathbf{H}$  : linearized form of the observational operator, which transforms the state vector  $\mathbf{x}$  in the observation space

$\mathbf{d} = \mathbf{y} - \mathbf{H}\mathbf{x}_b$  : misfit between observations  $\mathbf{y}$  and background estimates

The solution of the cost function can be computed by means of a new control variable  $\mathbf{v}$ :

$$\delta\mathbf{x} = \mathbf{V}\mathbf{v}$$

$\mathbf{V}$  : mapping operator representing the square root of the background-error covariance matrix  $\mathbf{B} = \mathbf{V}\mathbf{V}^T$ , defined as a sequence of linear operators

$$\mathbf{V} = \mathbf{V}_b \mathbf{V}_H \mathbf{V}_v \quad (\text{Dobricic and Pinardi, 2008; Teruzzi et al., 2014})$$

$\mathbf{V}_v$  vertical covariance

→ EOFs vertical profiles

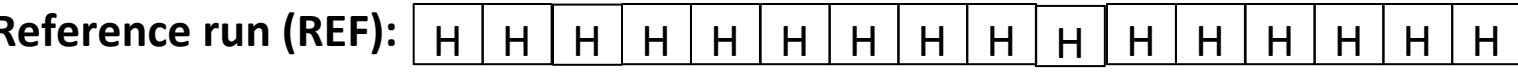
$\mathbf{V}_H$  horizontal covariance

→ Gaussian horizontal filter

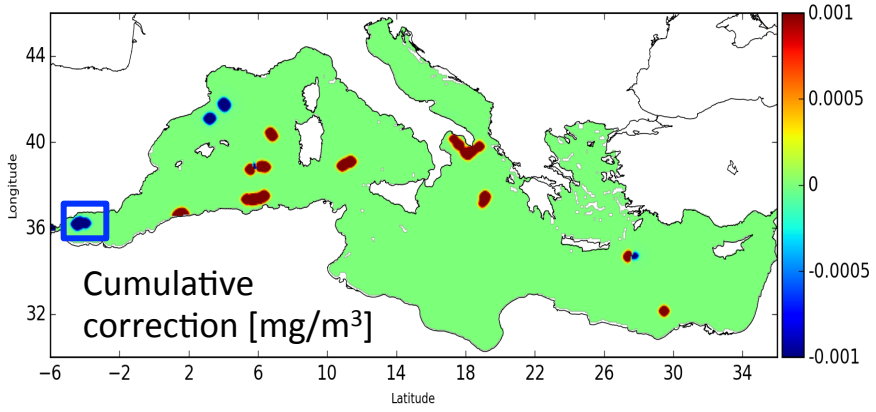
$\mathbf{V}_b$  biological covariance

→ covariance matrix among chlorophyll and other phytopl. model variables (C,N,P,Si in phytopl. groups)

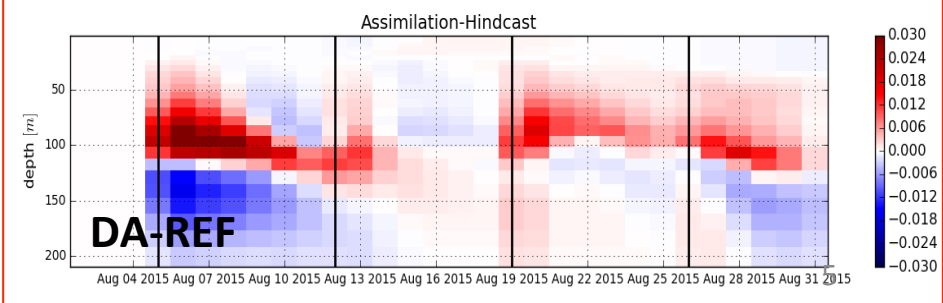
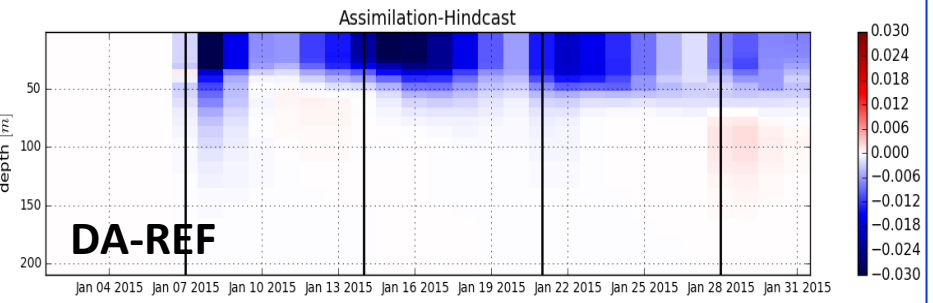
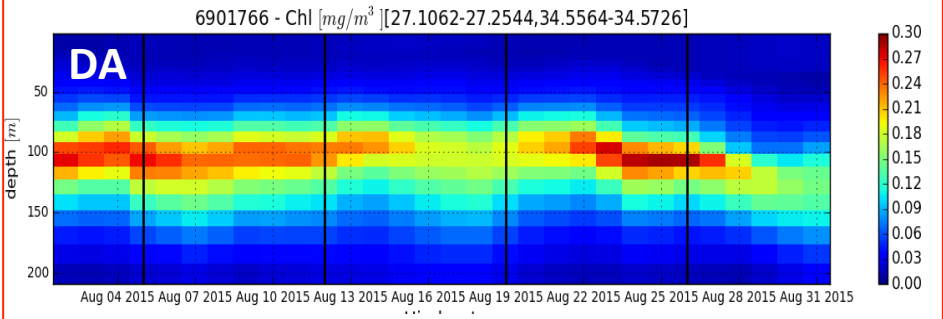
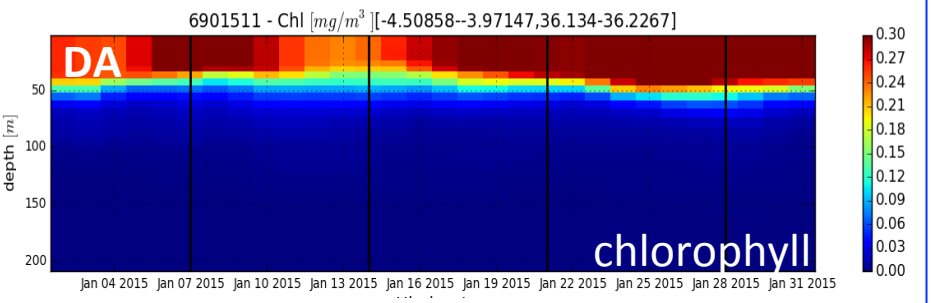
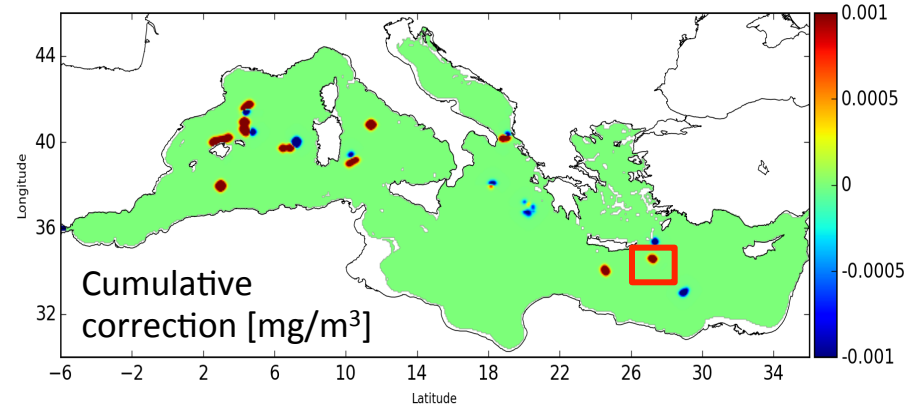
# Assimilation versus reference: winter and summer runs



**Winter: 1 Jan – 31 Jan 2015**



**Summer: 1 Aug – 31 Aug 2015**



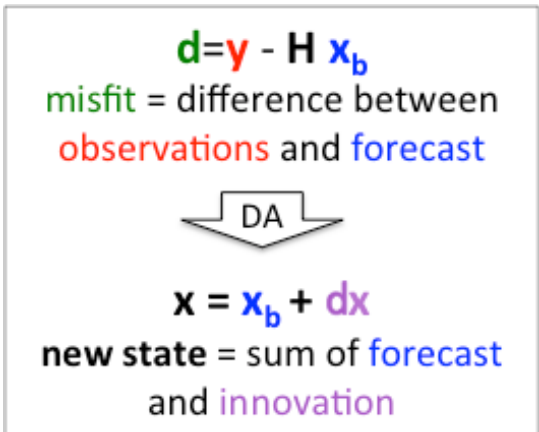
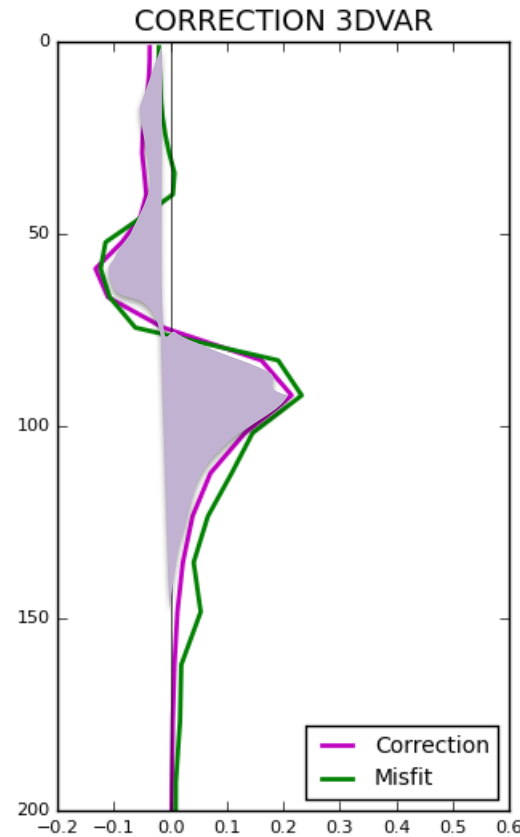
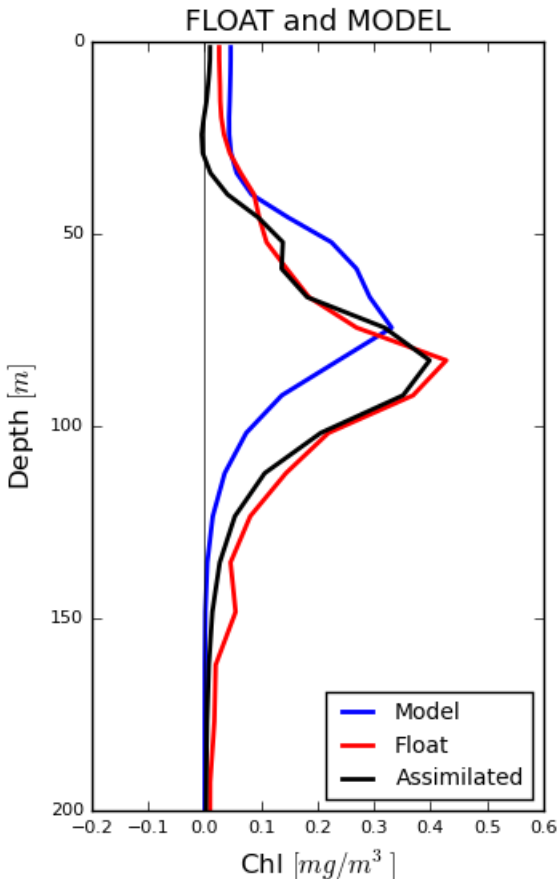
# Estimating the impact of the assimilation



Float 6901649 (41.417;4.3427)

## 1) III: index of innovation intensity

$$III = \int_z |innovat|$$

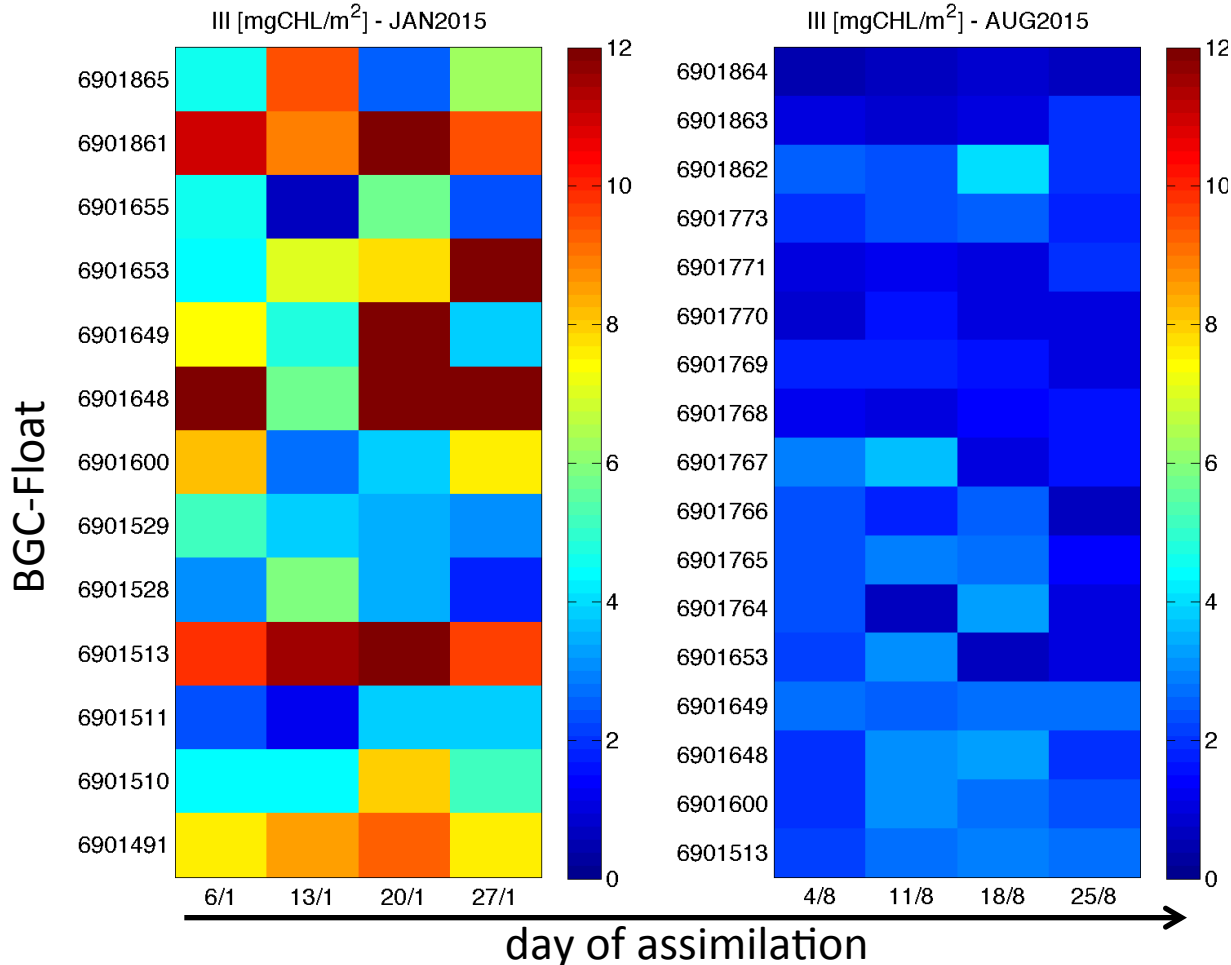


# Estimating the impact of the assimilation



## 1) III: index of innovation intensity

$$III = \int_z |innovat|$$



$\langle III \rangle_{Jan} = 6.9 \text{ mg/m}^2$

$\langle III \rangle_{Aug} = 1.8 \text{ mg/m}^2$

# Estimating the impact of the assimilation



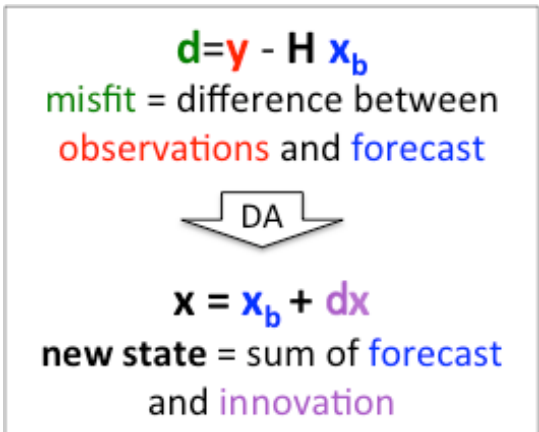
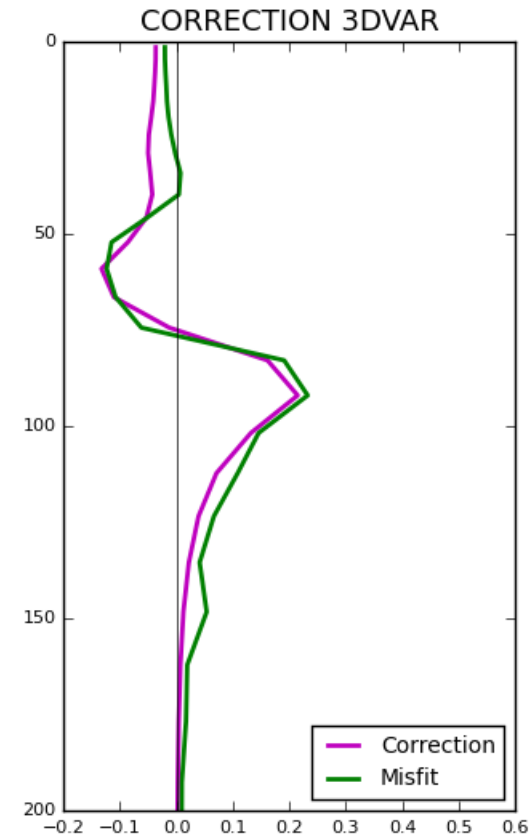
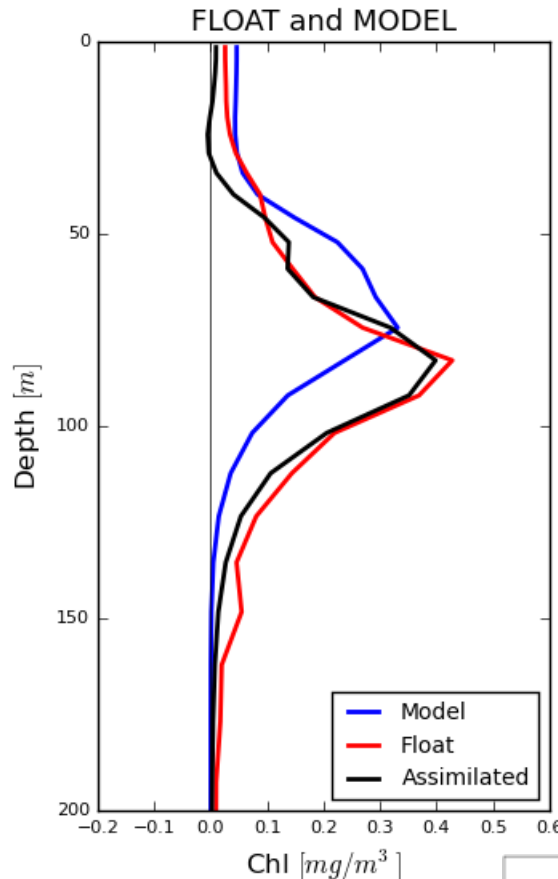
Float 6901649 (41.417;4.3427)

1) III: index of innovation intensity

$$III = \int_z |innovat|$$

2) IIP: index of innovation power

$$IIP = 1 - \int_z \left| \frac{innovat - misfit}{misfit} \right|$$





# Estimating the impact of the assimilation

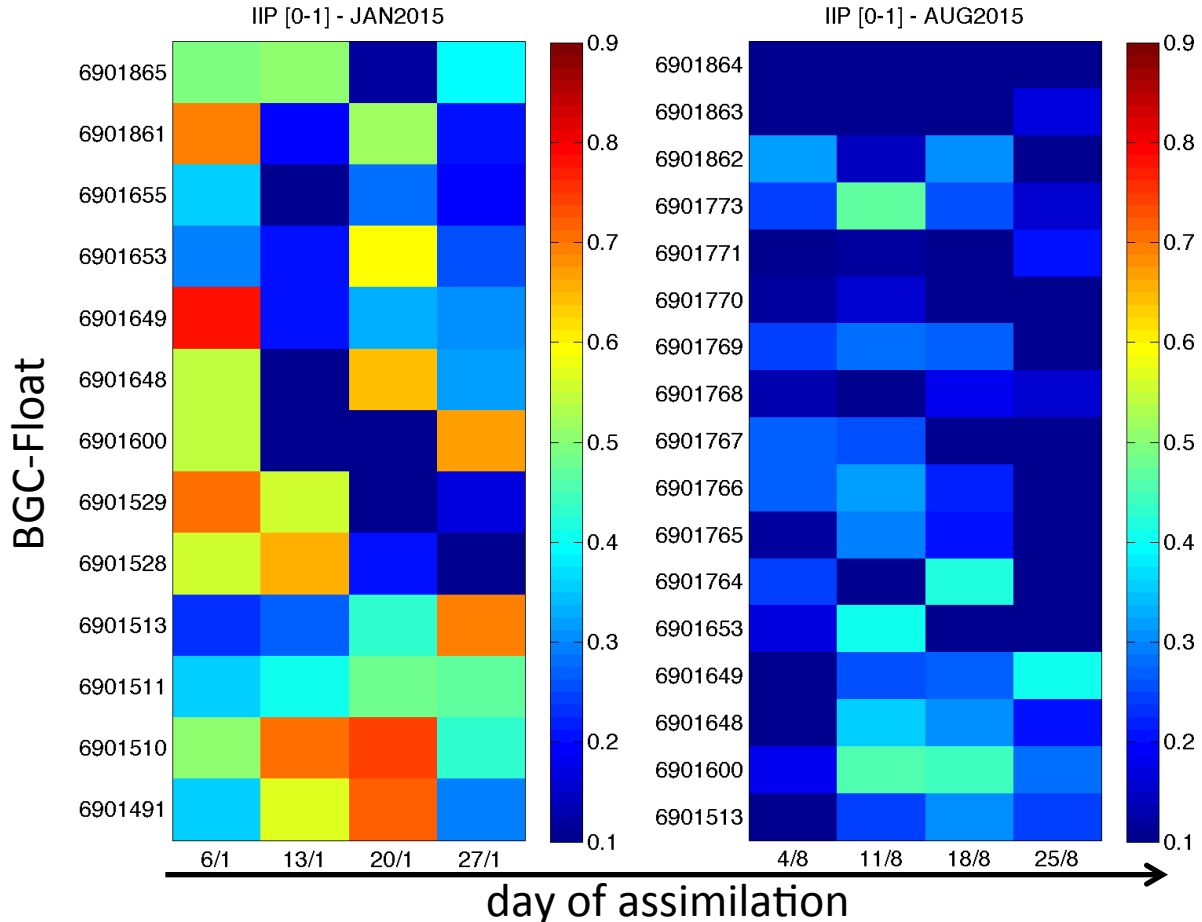


1) III: index of innovation intensity

$$III = \int_z |innovat|$$

2) IIP: index of innovation power

$$IIP = 1 - \int_z \left| \frac{innovat - misfit}{misfit} \right|$$



$\langle IIP \rangle_{Jan} = 39\%$

$\langle IIP \rangle_{Aug} = 20\%$

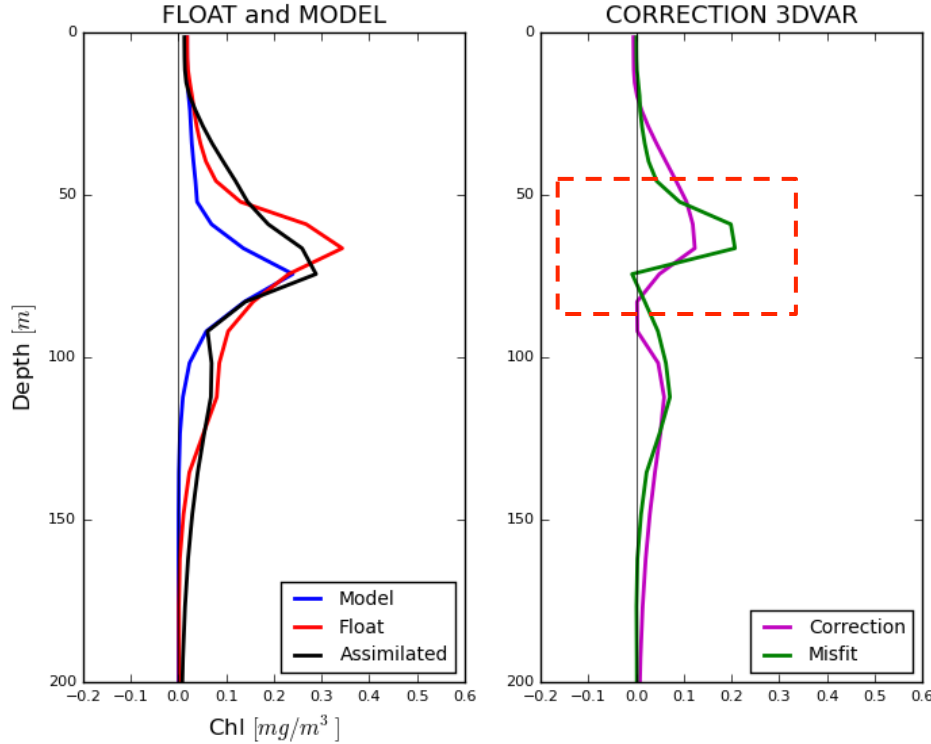
# Sensitivity of DA to the vertical covariance operator ( $V_v$ )



Set of the vertical EOFs:  $V_v = S\Lambda^{1/2}$

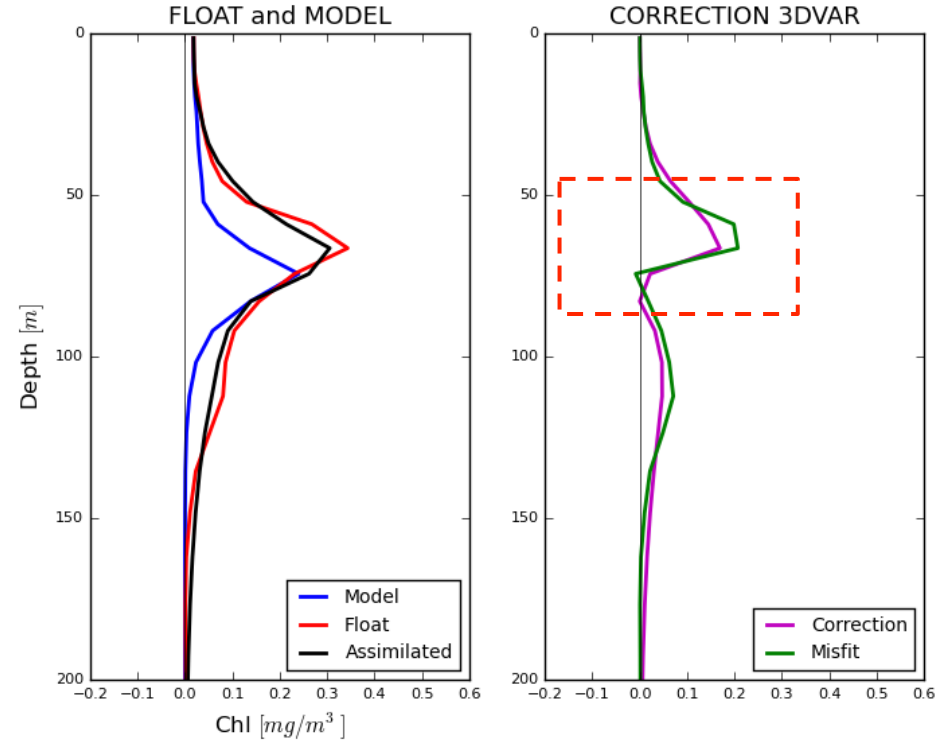
## 5 EOFs

Float 6901600 (38.003;2.9779)



## 25 EOFs

Float 6901600 (38.003;2.9779)



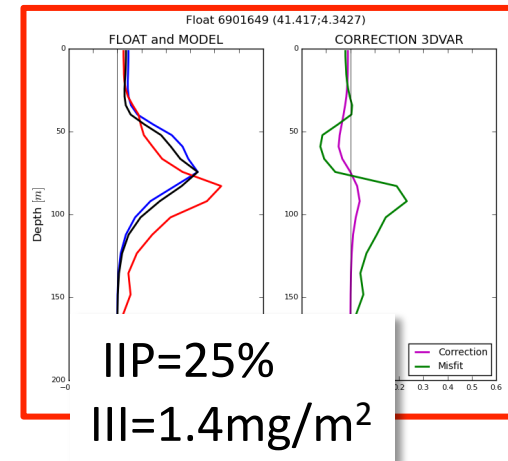
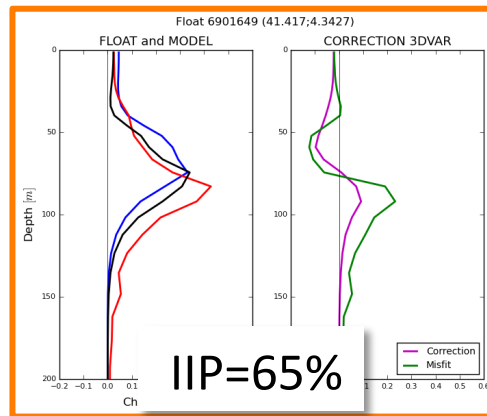
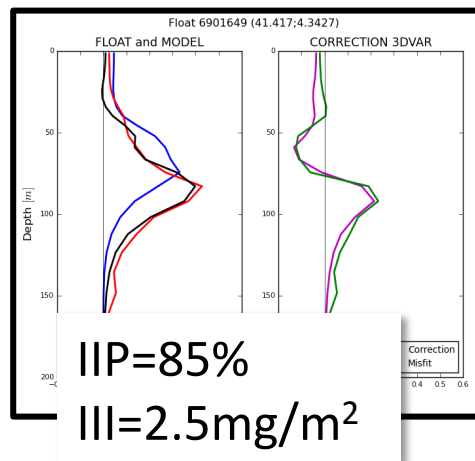
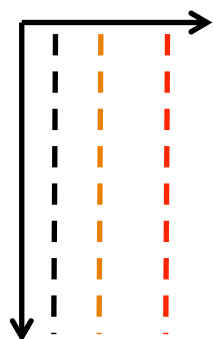
index of innovation intensity III: no significant changes

index of innovation power IIP: significant increase (mostly in summer)

# Sensitivity of DA to the observation error definition (R)

Statistics of obs. error are not perfectly known [Desroziers et al., 2005]

Constant error profile (proportional to instrumental error [0.01, 0.05, 0.1 mg/m<sup>3</sup>])



high error value → the power of assimilation decreases

The aim is to identify criteria useful for tuning the observation error covariance in order to have an effective assimilation while the consistency between model and observation errors is preserved.

Two methods:

- Desroziers et al., 2005, definition of the a-posteriori diagnostic observation and background errors
- “Estimation of BGC-Argo chlorophyll fluorescence and nitrate observational errors”  
Alexander MIGNOT, LOV CNRS/UPMC – France (POSTER)

# Estimating the impact of the assimilation



1) III: index of innovation intensity

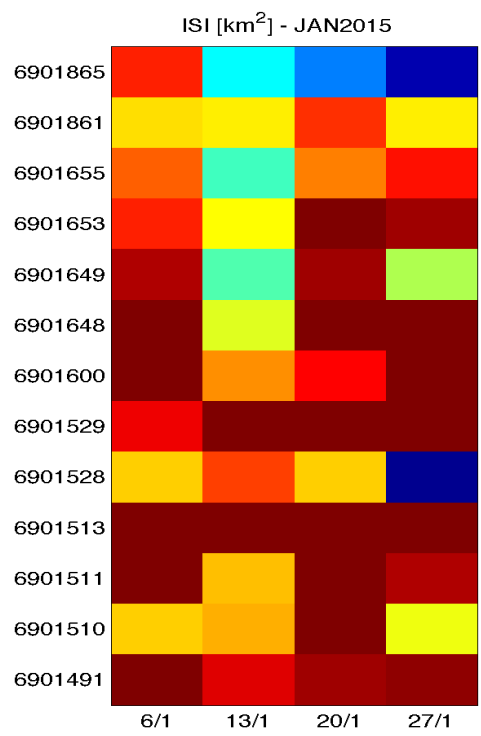
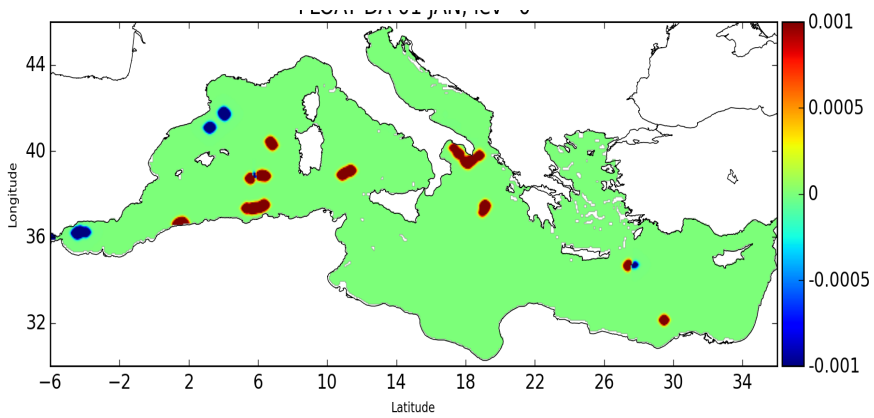
$$III = \int_z innovat$$

2) IIP: index of innovation power

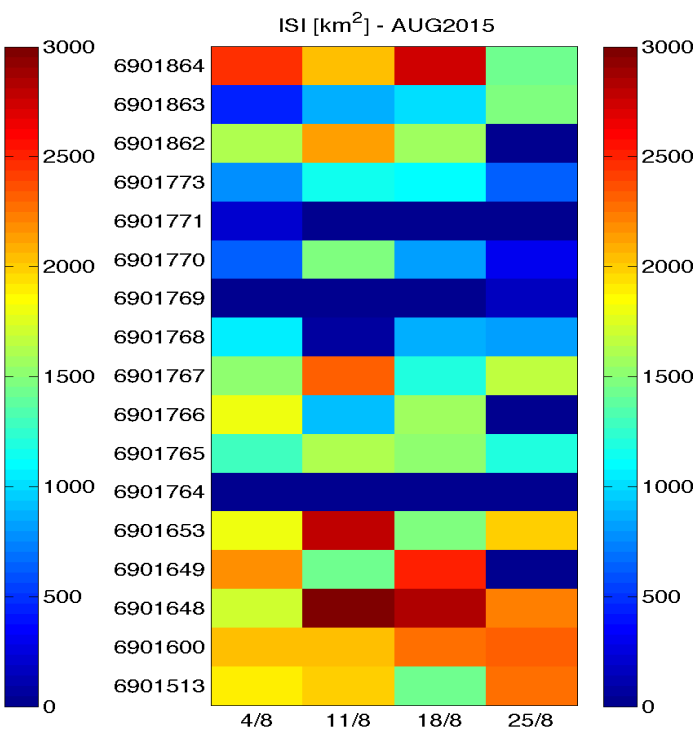
$$IIP = 1 - \int_z \left| \frac{innovat - misfit}{misfit} \right|$$

3) ISI: index of spatial innovation

$$ISI = \int_{x,y} SURF(|innovat| > T_{inn})$$



$\langle ISI \rangle_{Jan} = 2500 \text{ km}^2$   
2.5% of med



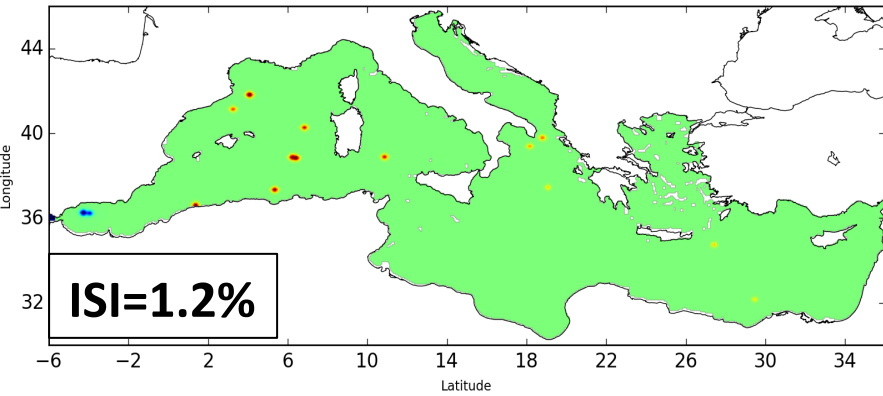
$\langle ISI \rangle_{Aug} = 1500 \text{ km}^2$   
1.5% of med

# Sensitivity of ISI to horizontal covariance operator

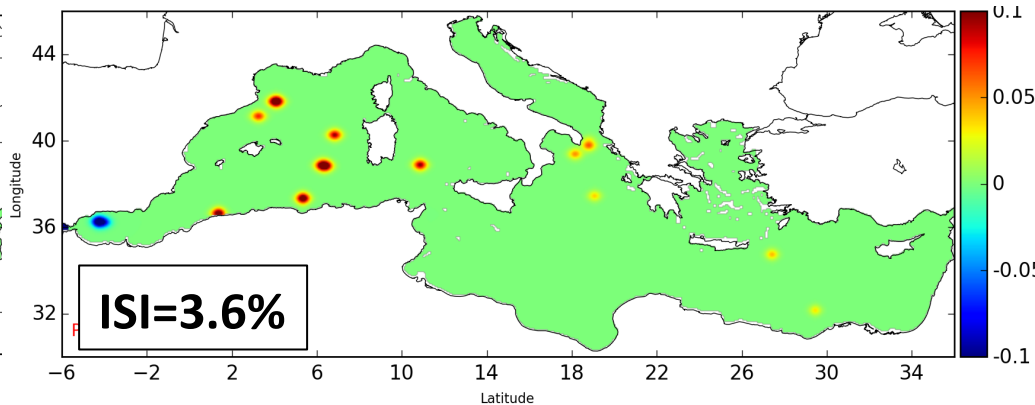
Gaussian recursive filter based on smoothing coefficients (corr. radius length scale)

Corr. radius length scale ( $L_r$ ) is no uniform in x and y and varies with depth.

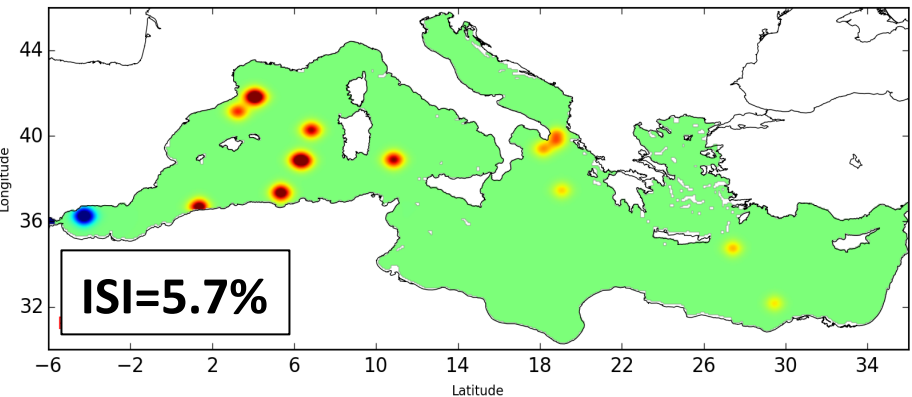
**$\langle L_r \rangle = 5\text{km}$**



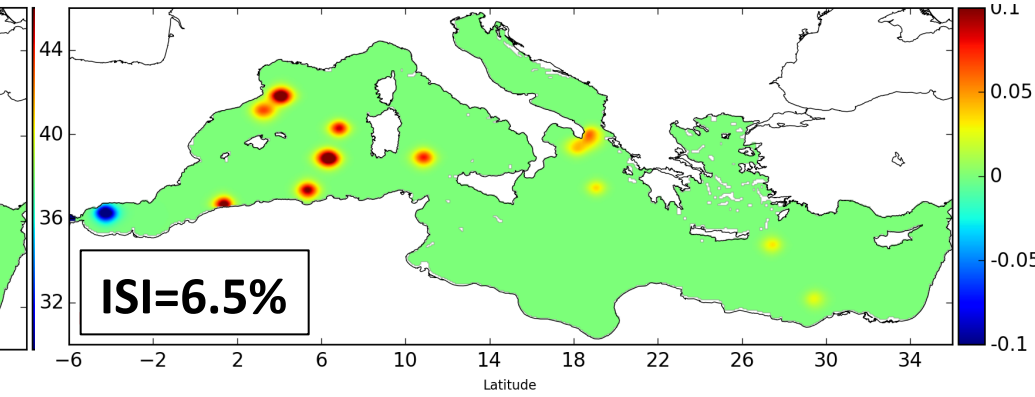
**$\langle L_r \rangle = 10\text{km}$**



**$\langle L_r \rangle = 15\text{km}$**



**$\langle L_r \rangle = \text{not uniform}$**



Maps of the innovation (mgCHL/m<sup>3</sup>)

# Estimating the impact of the assimilation



1) III: index of innovation intensity

$$III = \int_z |innovat|$$

2) IIP: index of innovation power

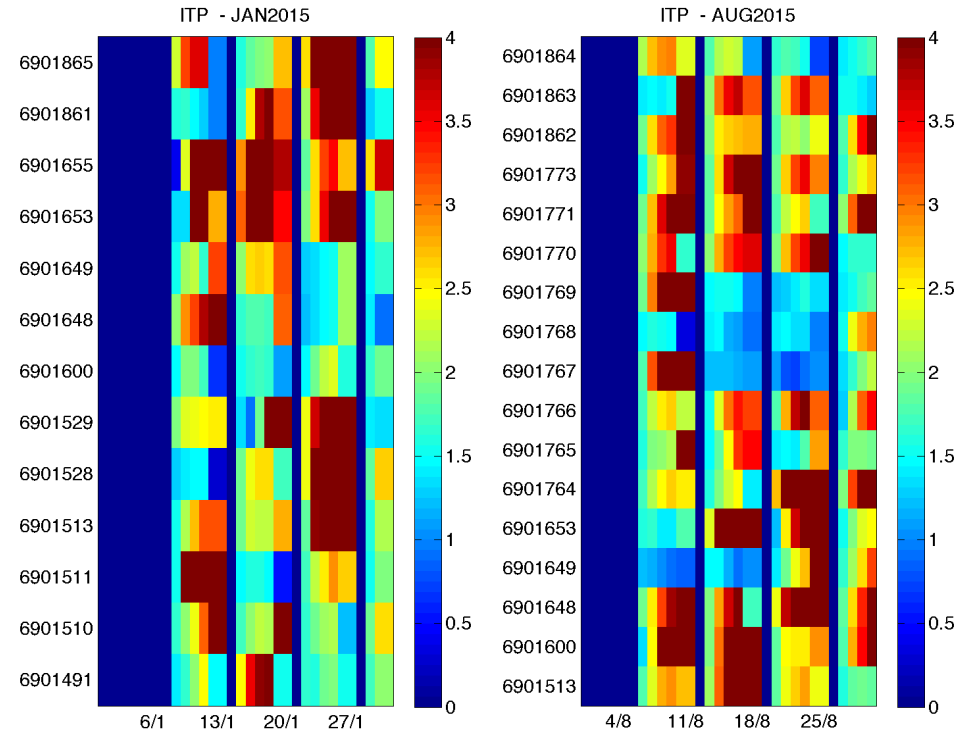
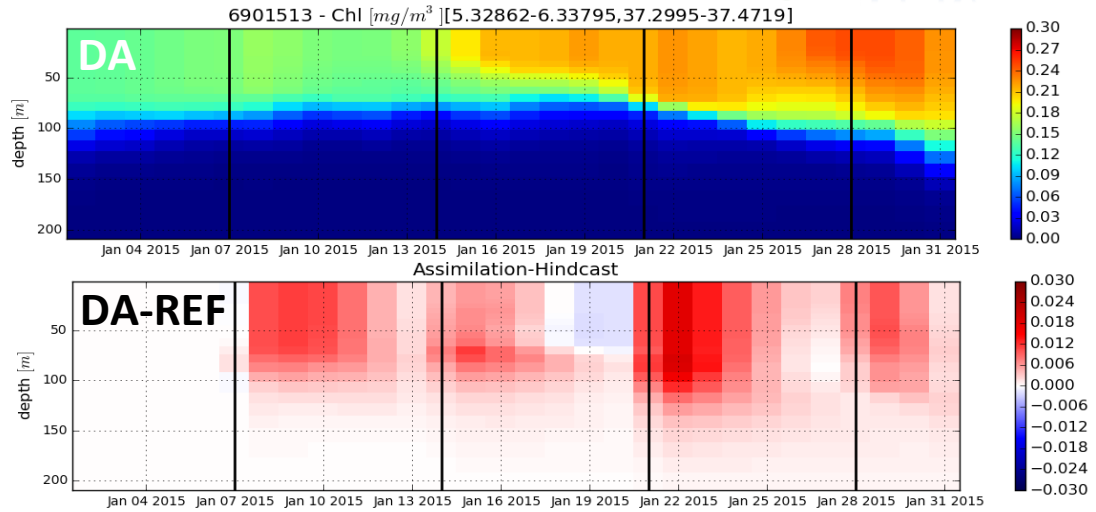
$$IIP = 1 - \int_z \left| \frac{innovat - misfit}{misfit} \right|$$

3) ISI: index of spatial innovation

$$ISI = \int_{x,y} SURF(|innovat| > T_{inn})$$

4) ITP: index of temporal persistence

$$ITP = \int_{x,y,z} |(DA-REF)_0 / (DA-REF)_t|$$

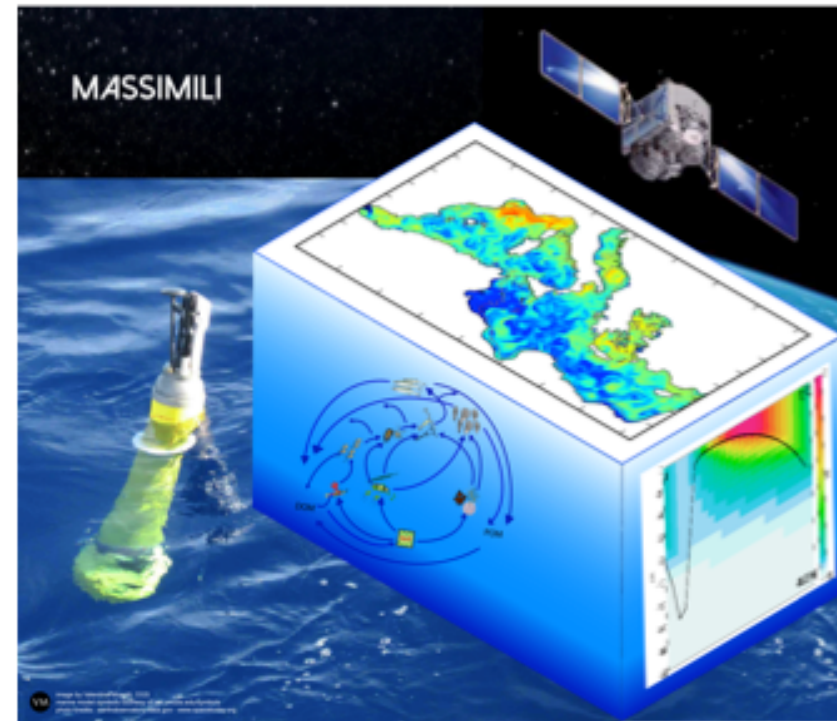


$\langle ITP \rangle_{Jan} > 2$  in 2.9 days

$\langle ITP \rangle_{Aug} > 2$  in 2.4 days<sup>14</sup>

## Conclusions

- BGC-Argo float data crucial to constrain the CMEMS forecast model systems
- Upgraded 3DVAR-BIO assimilation scheme:
  - no uniform correlation radius
  - vertical EOFs
  - observation error
  - frequency of assimilation cycle



## Next challenges of the MASSIMILI project (by 2018):

- Consistency of multivariate assimilation (CHL, NIT, OXY) → Vb operator (biological operator)
- Consistency between SAT and FLOAT chlorophyll data for multiplatform assimilation
- Evaluation of the performance of assimilation

Thanks

