



# Multi-Resolution Variational Method for Ocean Current Estimation from SST and Altimetry Observations

Silèye BA(+), Ronan Fablet(+)

Emmanuelle Autret\*, Bertrand Chapron\*

Dpt Signaux et Communications(+), Telecom Bretagne, Brest

IFREMER\*, Brest

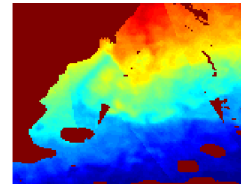
08/03/2012



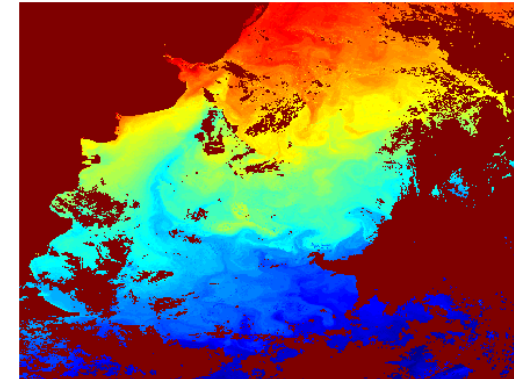
# Geophysical variables

## ■ Sea surface temperature (SST)

- REMSS: low spatial resolution, low missing data rate
- AVHRR-METOP: high spatial resolution, high missing data rate



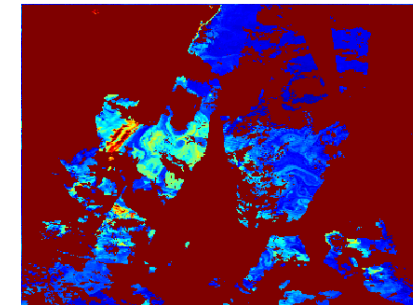
SST-AMSRE  
25km



SST-METOP  
5km

## ■ Chlorophyll concentration (CHL)

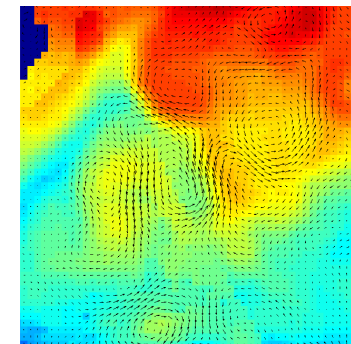
- MERIS: spatial resolution, high missing data rate



CHL-MERIS  
15km

## ■ Altimetry

- Low spatio-temporal resolution (1 observation/week)



ALTIMETRY  
30km



# Problem: missing data interpolation

## ■ Formal problem:

Hidden state (complete geophysical map)

$$Y_t = P_t X_t + \eta_t$$

Projection operator

Gaussian noise  
 $\sim N(0, R)$

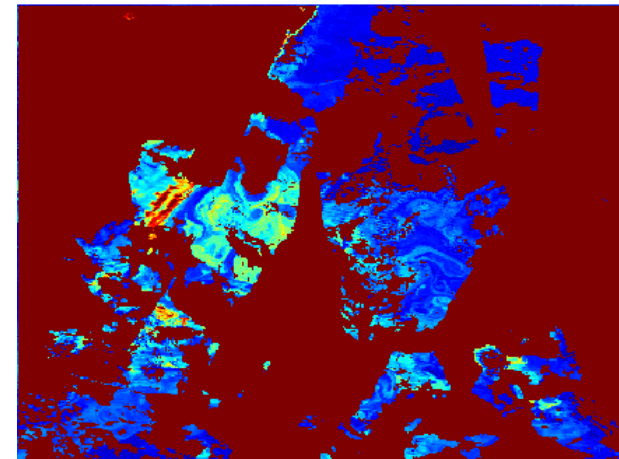
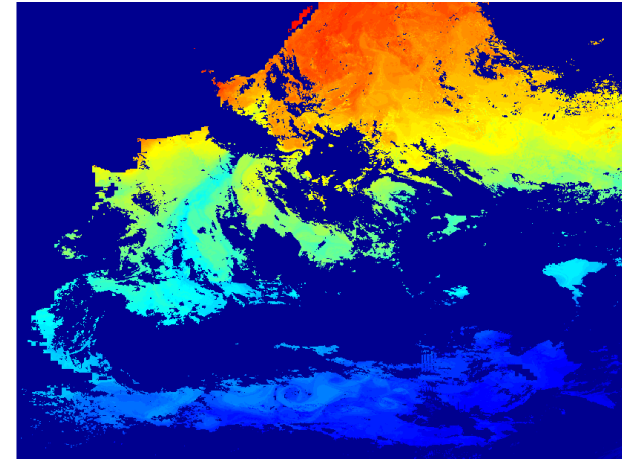
observation with missing data

## ■ Objective:

- recover the full geophysical maps
  - SST and altimetry

## ■ Our solution:

- Multi-resolution multi-modal variational data assimilation





## Variational estimation : general principle

### ■ Data:

- Observation sequence  $Y_t, t \in [t_0 t_f]$  of geophysical variable  $X_t$
- Dynamical model  $M(X_t)$ : temporal evolution of  $X_t$

### ■ Variational cost function

$$J(X) = C(X, Y) + C(X)$$

### ■ $C(X, Y)$ : observation model

- similarity between reconstructed variables and observations

### ■ $C(X)$ : temporal consistency term

- Consistency of variables temporal evolution
- Build on top of the dynamical model  $M(X_t)$



## Variational model

State model:

$$X = (\theta, \omega_g, u_a)$$

$\theta$  SST

$\omega_g$  geostrophic vorticity

$u_a$  ageostrophic velocity

Dynamic model:

$$\begin{cases} \partial_t \theta + (u_a + u(\omega_g)) \nabla \theta + \kappa \operatorname{div}(\nabla \theta) \\ \partial_t \omega_g + u(\omega_g) \nabla \omega_g = \zeta_t \\ X_{t_0} = X_0 + \vartheta_t \end{cases}$$

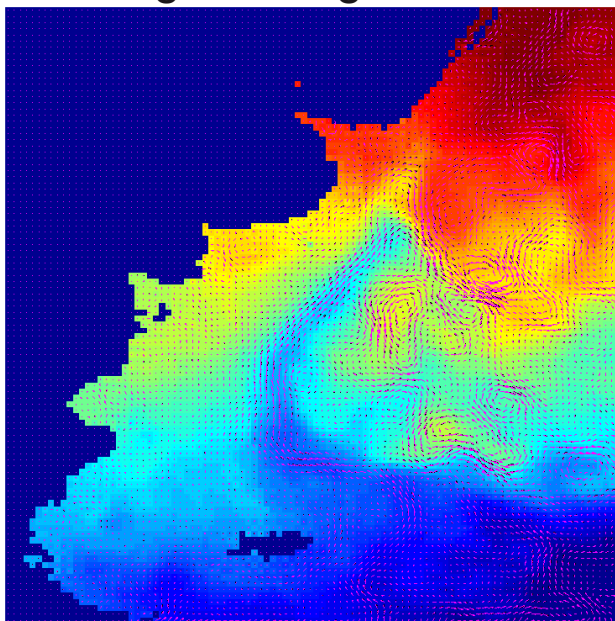
Variational Problem:

$$J(\theta, \omega_g, u_a) = \underbrace{E_{Obs}(\theta, \omega_g)}_{\text{observation term}} + \underbrace{E_{Dynamic}(\theta, \omega_g, u_a) + E_{Dynamic}(\omega_g)}_{\text{temporal consistency}} + \underbrace{E_{MultiResol}(\theta_{HR}, \theta_{LR})}_{\text{multi-resolution}}$$

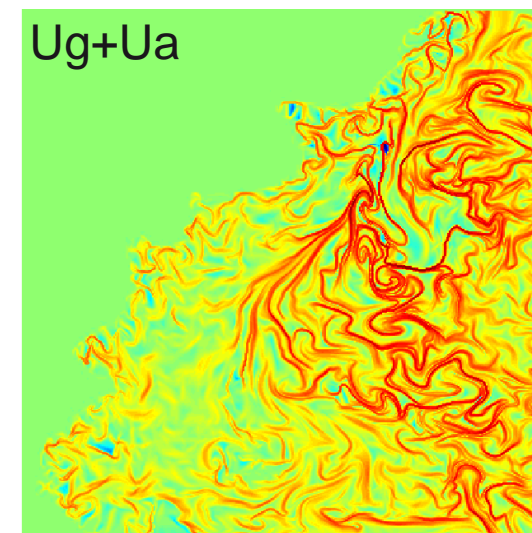
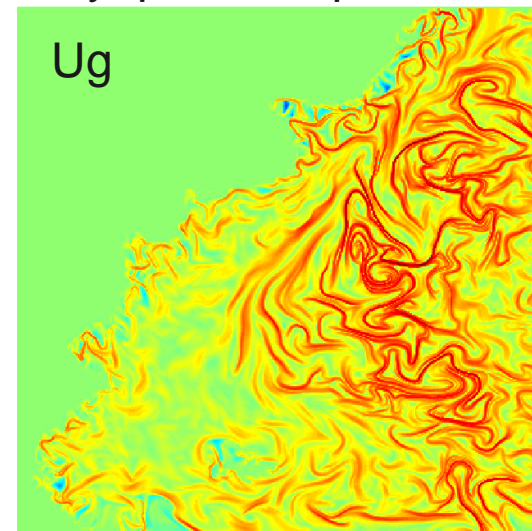


## Results: circulation estimation

Black arrow= $U_a+U_g$   
Magenta= $U_g$



Lyapunov exponents



- Convergence structures are better resolved with the geostrophic and the ageostrophic velocities





**Thanks for your attention**  
**Questions & Commentaires**



# Single (SST) resolution variational method

- Hidden state sequence :
- Temporal evolution driven by the system

$$\begin{cases} \partial_t \theta + M(\theta, \omega_g, u_a) = \xi_t, & M(\theta, \omega_g, u_a) = (u_a + u(\omega_g)) \nabla \theta + \kappa \operatorname{div}(\nabla \theta) \\ \partial_t \omega_g + u(\omega_g) \nabla \omega_g = \zeta_t \\ X_{t_0} = X_0 + \vartheta \end{cases}$$

$$X = (\theta, \omega_g, u_a)$$

$\theta$  SST

$\omega_g$  geostrophic vorticity

$u_a$  ageostrophic velocity

- Variational problem

$$J(\theta, \omega_g, u_a) = \underbrace{\int_{t_0}^{t_f} E(\theta_t, \omega_g) dt}_{\text{observation term}} + \underbrace{\int_{t_0}^{t_f} \left\| \partial_t \theta - M(\theta, \omega_g, u_a) \right\|_{Q_\theta^{-1}}^2 dt + \int_{t_0}^{t_f} \left\| \partial_t \omega_g - u(\omega_g) \nabla \omega_g \right\|_{Q_\omega^{-1}}^2 dt}_{\text{temporal consistency}}$$

$$E(\theta, \omega_g) = \sigma_\theta^{-1} \left\| Y^\theta - P_\theta \theta \right\|_2^2 + \beta \left\| \nabla \theta \right\|_2^2 + \sigma_\omega^{-1} \left\| Y^\omega - P_\omega \omega_g \right\|_2^2$$

SST observation

vorticity observation (altimetry)





## Multi-resolution variational method

### ■ Similar underlying dynamics

- Various geophysical variables follow the same ocean circulation
  - Same main frontal structures
- Examples:
  - SST observation at various resolution (REMSS and METOP SST)

### ■ Solution:

- Multi-resolution variational processing with frontal structure constraints

### ■ Multi-resolution observation model:

Regularity term: impose reconstructed variables to share main frontal structures

$$E_J(\theta_1, \theta_2) = E(\theta_1) + E(\theta_2) + \int_{\Omega} g_a(|\nabla \theta_1|) \rho_{\varepsilon} \left( \left\langle \frac{\nabla \theta_2}{|\nabla \theta_2|}, \frac{\nabla \Lambda \theta_1^{\perp}}{|\nabla \Lambda \theta_1|} \right\rangle \right) dp$$

high resolution

low resolution

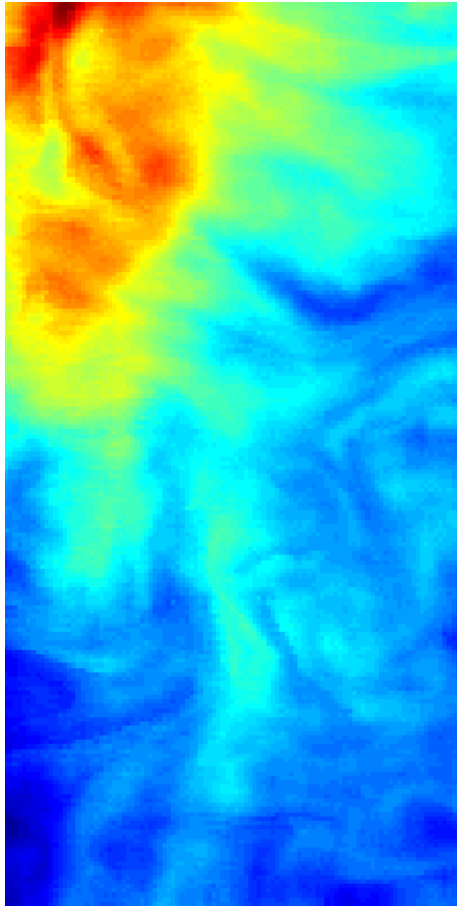
$$E(\theta) = \sigma^{-1} \|Y^{\theta} - P_{\theta} \theta\|_2^2 + \beta \|\nabla \theta\|_2^2 \quad \text{Independant cost}$$

$$g_a(x) = 1 - \frac{1}{1+ax} \quad \text{Frontal structure marker (edge detector)}$$

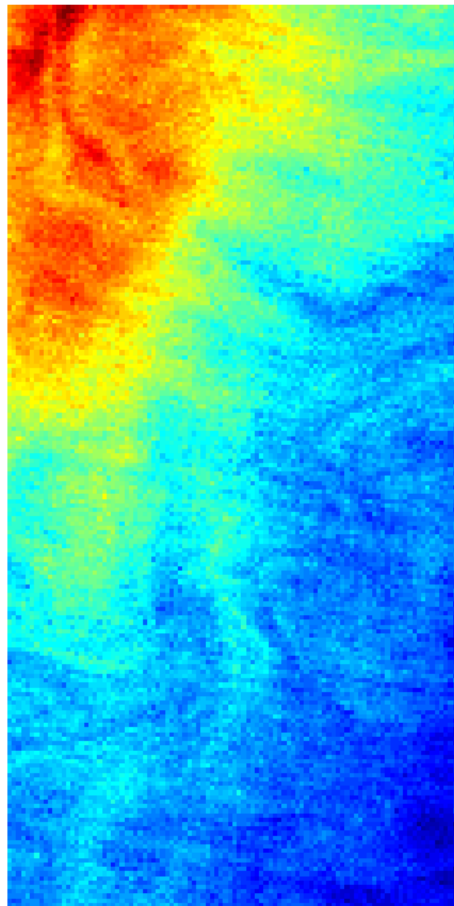
$$\rho_{\varepsilon}(x) = \sqrt{x^2 + \varepsilon} \quad \text{Absolute value difrentiable approximation}$$



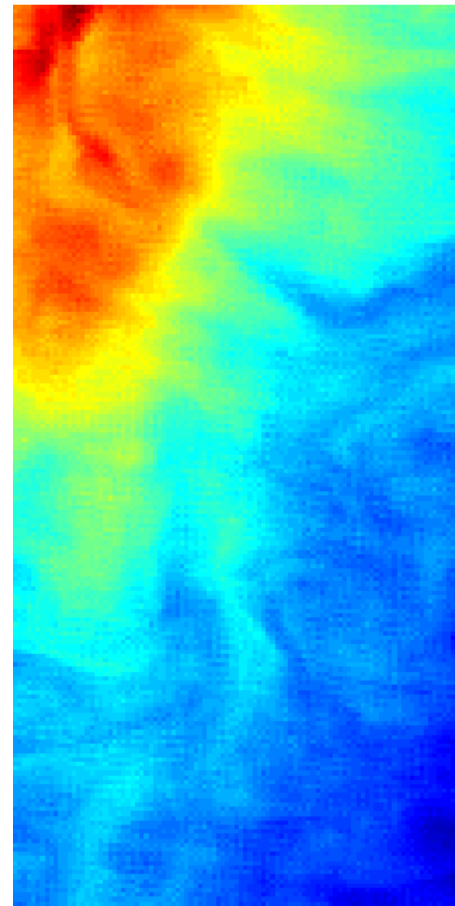
## Results: frontal structure prior effects



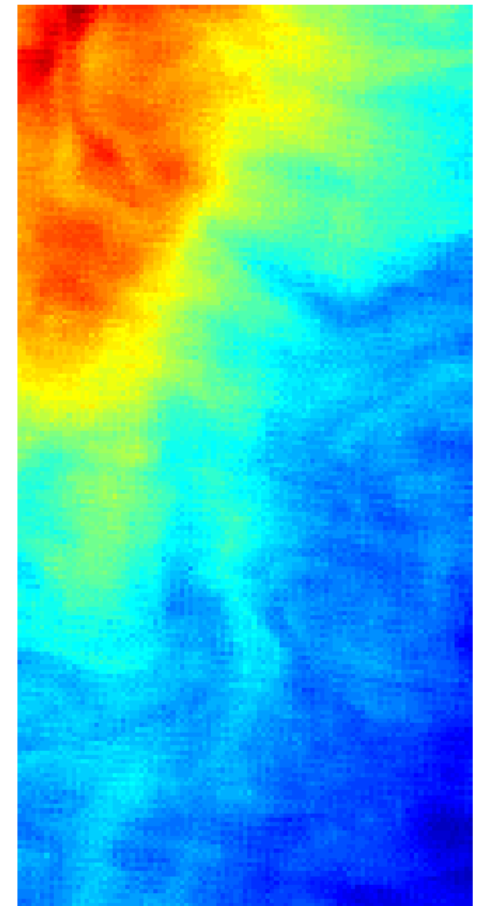
BT



SST



Regularized SST  
BT fronts Prior



Regularized SST  
BT div. Prior