

Image-Based Forecast of the Surface Circulation of Black Sea

Isabelle Herlin, Etienne Huot

► To cite this version:

Isabelle Herlin, Etienne Huot. Image-Based Forecast of the Surface Circulation of Black Sea. Black Sea from Space Workshop, Sep 2016, Constanta, Romania. . hal-01349391

HAL Id: hal-01349391 https://inria.hal.science/hal-01349391v1

Submitted on 7 Oct 2016

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers. L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

PROJECT TEAM CLIME **INRIA** Paris 2 Rue Simone IFF 75012 Paris Isabelle.Herlin@inria.fr Etienne.Huot@inria.fr

Image-Based Forecast of the Surface Circulation of Black Sea

Isabelle Herlin - Etienne Huot

FORECAST

Satellite data allow vizualising the surface circulation and the meso-scale structures at the spatial resolution and time frequency of the sensor. These images provide continuous information at fine scale and permit analysing eddies, jets, filaments, mushroom-shaped vortices. Many applications, such as oil spill monitoring, also require the forecast of surface circulation at short temporal horizon. This nowcasting includes two main components: one in charge of estimating the surface dynamics and one of the forecast of future images at a given temporal horizon.

ABSTRACT

OBJECTIVES

A sliding-window technique processes the last 4 images:



The forecast synthetizes future images based on the estimated dynamics and on the last acquisition with the numerical model \mathbb{M}_f :

$$\frac{\partial \mathbf{w}}{\partial t} + \mathbf{w} \cdot \nabla \mathbf{w} = 0$$
$$\frac{\partial I}{\partial t} + \mathbf{w} \cdot \nabla I = 0$$

The numerical accuracy of the synthetized images is crucial for the effective exploitation of the forecasts. Two spatial discretization schemes are combined during the temporal integration:

- A semi-Lagrangian scheme, denoted $\frac{1}{2}L$, which is unconditionally stable and allows a large time step.
- A Godunov scheme, denoted G, which solves the Riemann problem at each inter-pixels boundary. It is adapted for an accurate advection of functions with high spatial gradient values (but CFL conditions).



The state vector \mathbf{X} is composed of motion \mathbf{w} and synthetic image I: $\mathbf{X}(\mathbf{x},t) = \left(\mathbf{w}(\mathbf{x},t) \ I(\mathbf{x},t)\right)^{T}$

SURFACE DYNAMICS ESTIMATION

The surface dynamics is estimated with a data assimilation approach from four image acquisitions and the evolution laws of w and I, summarized in the numerical model \mathbb{M}_e :

$$\begin{aligned} \frac{\partial \mathbf{w}}{\partial t} &= 0\\ \frac{\partial I}{\partial t} + \mathbf{w} \cdot \nabla I &= 0 \end{aligned}$$

• Estimation is computed from the comparison between the satellite images $\mathbf{Y}(t)$ and the synthetic images I(t).

Images I(t) are obtained from the integration of the evolution model \mathbb{M}_e . The discrepancy between $\mathbf{Y}(t)$ and I(t) is equal to $\varepsilon_I(\mathbf{x}, t)$:

$$\varepsilon_I(\mathbf{x},t) = \mathbf{Y}(\mathbf{x},t) - I(\mathbf{x},t)$$



Forecast results:

w from $(\mathbf{Y}_0, \mathbf{Y}_1, \mathbf{Y}_2, \mathbf{Y}_3)$ \rightarrow Last Obs \mathbf{Y}_3



 F_5

VALIDATION

Forecast F_4

Comparison of the acquisitions \mathbf{Y}_i to the corresponding forecasts (at the same time) F_i , F'_i , F''_i :



• The minimization of the discrepancy ε_I is obtained through the minimization of the energy:

 $J(\mathbf{X}(0)) = \iint \frac{\varepsilon_I(\mathbf{x}, t)^2}{R_I(\mathbf{x}, t)} d\mathbf{x} dt + \text{additional terms}$

and a steepest descent method (L-BFGS algorithm). R_I is taken as the variance of the acquisition error.

• The issue of missing data, due to the cloud cover for instance, is solved with $R_I(\mathbf{x},t) = 10^8$, which excludes these pixels from the energy J.



Forecast F'_5

 F_6'



Forecast F_6''

Image courtesy: Gennady Korotaev, Marine Hydrophysical Institute, Sevastopol.

(nría_

