On the joint use of high resolution tracer images and altimetric data for the control of ocean circulation.

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General objective

Benefit from the complementarity and the richness of several remote observations (altimetric and high resolution tracers) to improve the assessment of oceanic circulation.

Data image assimilation strategy

Explore the feasibility of tracer image inversion for the control of surface dynamics.

Motivation

Ingredients



Ingredients of the talk $\bullet \circ \circ \circ \circ$

Motivation

Submesoscale

- Intermediate scale between Mesoscale and dissipative scales.
- Filaments length: 1-10 km.

Sub-mesoscales generated by mesoscale dynamics To what extend can sub-mesoscales control mesoscales?

Importance of sub-mesoscales

- Impact on larger scale circulation
- Energetic importance

(Capet & al, 2008, Thomas & al, 2008, Klein & al, 2008, Ferrari & al, 2008)



Baroclinic instability in an idealized model (Chlorophyll)

Ingredients of the talk	Motivation	Objective and Strategy
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Altimetry		

- Measure of sea surface height along track
- Geostrophic velocity derived from ssh gradients
- Data: e.g. AVISO (interpolated maps, velocity, ssh)

Use of gridded SSH and velocity: interpolation between tracks



3 satellites are necessary to capture mesoscale signal (Morrow and Le Traon, 2011).
Sub-mesoscales cannot currently be seen by altimetric satellites (Fu and Ferrari, 2008).

Jason and Envisat tracks (15 days)

Ingredients of the talk	Motivation
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Tracer image

Interested in tracers visible from space

- Sea Surface Temperature (near IR, visible)
- Ocean Color: Chlorophyll (visible)







Chlorophyll from MODIS sensor (25-27 April 2011)

Sub-mesoscales filaments revealed by tracer images



Lagrangian tool that gives a measure of ocean stirring Tracer observation and FSLE derived from altimetric velocity have shown similar patterns (Lehahn & al 2008, d'Ovidio & al 2004).



Chlorophyll, South Atlantic, d'Ovidio & al, 2004

Chlorophyll, Pomme area, Lehahn & al, 2008

Motivation

Data assimilation

Data Assimilation aims at finding an optimal compromise between information of different natures, space and time sampling. The sources are generally some observations (satellite, in-situ) and a numerical model.



Conceptual representation of filtering with sequential assimilation, Brasseur, 2006



Path





Path



ngredients of the talk

Motivati<u>on</u>

Objective and Strategy

Complementarity of remote sensing



Jason and Envisat tracks, 15 days before and after December 22, 2004, Tasmania

Sub-mesoscales are not resolved by altimetry.

Ingredients of the talk 00000 Motivation

Objective and Strategy

Complementarity of remote sensing



Jason and Envisat tracks, 15 days before and after December 22, 2004, Tasmania

Chlorophyll, December 22, 2004, Tasmania

Sub-mesoscales are not resolved by altimetry.

Sub-mesoscales are observed using satellite tracer sensors.

Joint use of altimetry and high resolution tracer observation to improve the dynamics.

Future observation

Project SWOT

- High resolution SSH image
- Detect small scale structure

Geostationary color

- Frequent and high resolution chlorophyll images
- Study of structures and biogeochemics

Complex observations \Rightarrow Interest in structure of data

Motivation

Context of the inversion

Feasibility of data image assimilation

Inversion: fixed time assimilation



Ingredients of the talk 00000

Motivation

Objective and Strategy

Context of the inversion

Feasibility of data image assimilation

Inversion: fixed time assimilation



Use of a Data Assimilation approach

The inversion of sub-mesoscale tracer information to correct mesoscale velocity has never been done before

Comparing velocity and tracer images





Need of a proxy



Find the correction of this background the most compatible with tracer information

- The direct measure of the distance between \vec{u} and Tracer is not possible
- Need to find a go-between variable
- Use of Finite-Size Lyapunov Exponents as a proxy (FSLE)

See Gaultier & al, 2012 for details

Comparing velocity and tracer images $\circ \circ \circ$

Method of Inversion

Physical meaning of Lyapunov Exponents



FSLE

$$\begin{split} \lambda &= \frac{1}{T} \times \log(\frac{\delta_{\text{final}}}{\delta_{\text{initial}}}) \\ \text{Inverse of the time T for particles to} \\ \text{be separated from } \delta_{\text{initial}} \text{ to } \delta_{\text{final}} \end{split}$$

Lyapunov Exponent constitute Lagrangian transport barriers between different regions

FSLE measures the separation rate of particles Tracer cannot cross lines of maximum of FSLE

Comparing velocity and tracer images ○○● Method of Inversion

Proxy FSLE consistent



December 27, 2006



Tracer (SST), South Atlantic region, December 27, 2006

Lyapunov measures stirring in a fluid \rightarrow Link between sub-mesoscale dynamics and biologic stirring. (Lehahn & al, 2008, d'Ovidio & al, 2004)

- Binarisation of FSLE
- Binarisation of the norm of the gradient of tracer filtered image, image processing to reduce the noise (developed by LIK)

Comparing velocity and tracer images ○○● Method of Inversion

Proxy FSLE consistent



December 27, 2006



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- Binarisation of FSLE
- Binarisation of the norm of the gradient of tracer filtered image, image processing to reduce the noise (developed by LJK)

Overview of the method



Method of Inversion

• Cost function:

$$J(\vec{u}) = \|\mathcal{I}_{FSLE}(\vec{u}) - \mathcal{I}_{tracer}\| + \mu \|\vec{u} - \vec{u}_{alt}\|$$

The cost function is strongly non linear, non differentiable, with many local minima.



 $FSLE(\vec{u})$

 $FSLE(\vec{\delta}_u)$

Cost function

• Cost function:

$$J(\vec{u}) = \|\mathcal{I}_{FSLE}(\vec{u}) - \mathcal{I}_{tracer}\| + \mu \|\vec{u} - \vec{u}_{alt}\|$$

The cost function is strongly non linear, non differentiable, with many local minima.



 $FSLE(\vec{u}) \qquad FSLE(\vec{\delta}_u) \qquad FSLE(\vec{u} + \vec{\delta}_u)$

Minimize cost function using known and new methods

Build a subspace of error

Inspired by SEEK filter

• Explore sub-space of error to find the velocity that minimizes the cost function.

Velocity panel using Principal Component Analysis (EOF analysis) with all velocity fields available:

$$\mathbf{u}_{k} = \bar{\mathbf{u}} + \sum_{i=0}^{n} \underbrace{a_{k}^{i}}_{Eigenvalue} \underbrace{\mathbf{u}^{i}}_{EOF}$$

The number of degrees of freedom is reduced, using only 100 or less EOFs.

Need to use new methods : OSMIUM tool

Simulated annealing

Excite a particle to get out of local minimum

Gibbs' Sampler

Sample potential solution around a minimum



Cost function as a function of the number of iterations



Path



Comparing velocity and tracer images

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Strategy of the inversion study



Realistic model data

Test case : small area in the South Atlantic ocean



- Time Range: from 1998 to June 2009, 595 velocity maps
- Velocity field: AVISO, Altimetric data
- **Resolution**: 1/3°, grid points : 18*16
- FSLE Resolution: 1/50°, grid points : 130*120
- Tracer field: SST or Chlorophyll data (MODIS sensor, L2 product)



Realistic model data

FSLE invertible



 $J(\vec{u}) = \|\mathcal{I}_{FSLE}(\vec{u}) - \mathcal{I}_{synthetic}\| + \mu \|\vec{u} - \vec{u}_{alt}\|$

Real data ○○●○

Process model data

Realistic model data

Corrected field



 $J(\vec{u}) = \alpha \|\mathcal{I}_{FSLE}(\vec{u}) - \mathcal{I}_{SST}\| + (1 - \alpha)\|\mathcal{I}_{FSLE}(\vec{u}) - \mathcal{I}_{CHL}\| + \mu \|\vec{u} - \vec{u}_{alt}\|$

Real data ○○●○

Process model data

Realistic model data

Corrected field



 $J(\vec{u}) = \alpha \|\mathcal{I}_{FSLE}(\vec{u}) - \mathcal{I}_{SST}\| + (1 - \alpha)\|\mathcal{I}_{FSLE}(\vec{u}) - \mathcal{I}_{CHL}\| + \mu \|\vec{u} - \vec{u}_{alt}\|$



Lagrangian trajectories from the altimetric Lagrangian trajectories from the velocity field velocity field corrected using tracers

- The trajectory of six particles are represented over the SST
- These trajectories are similar to the filaments observed in SST



High Resolution coupled physico-biogeochemical model



Model configuration: Levy, 2002

- NEMO dynamics coupled with LOBSTER biochemical model
- Channel domain: $478 \times 500 \times 4$ km ($240 \times 252 \times 30$ grid points)
- Horizontal resolution: 2 km
- Sub-mesoscale and mesoscale structures result from an unstable baroclinic jet



Inversion parameters

- Time range: 90 states of the model before the study date
- Background velocity: Velocity 5 days after the study date
- Corresponding FSLE: 2 km resolution
- Tracer Image: Chlorophyll and/or Sea Surface Temperature image of the study date at 2 km resolution

Process model data

Realistic model data

$J(ec{u}) = \|\mathcal{I}_{FSLE}(ec{u}) - \mathcal{I}_{CHL}\| + bg$



Background velocity



True velocity



Chlorophyll

Process model data

Realistic model data 0000000

$J(\vec{u}) = \|\mathcal{I}_{FSLE}(\vec{u}) - \mathcal{I}_{CHL}\| + bg$



Chlorophyll

$J(\vec{u}) = \|\mathcal{I}_{FSLE}(\vec{u}) - \mathcal{I}_{SST}\| + bg$





SST

Corrected velocity

True velocity

Process model data

Realistic model data 0000000

$J(\vec{u}) = \|\mathcal{I}_{FSLE}(\vec{u}) - \mathcal{I}_{SST}\| + bg$



SST



SST

Chlorophyll



$J(\vec{u}) = \alpha \|\mathcal{I}_{FSLE}(\vec{u}) - \mathcal{I}_{SST}\| + (1 - \alpha) \|\mathcal{I}_{FSLE}(\vec{u}) - \mathcal{I}_{CHL}\| + bg$



SST

Chlorophyll

Process model data

Realistic model data

$J(ec{u}) = \|\mathcal{I}_{FSLE}(ec{u}) - \mathcal{I}_{struct}\| + bg$



Process model data

Realistic model data

$J(\vec{u}) = \|\mathcal{I}_{FSLE}(\vec{u}) - \mathcal{I}_{struct}\| + bg$



Process model data

Realistic model data

Tracer contribution





Cost function as a function of iterations (semi-log)

Error on the velocity as a function of iterations (semi-log)

Good Results even if the image processing is rough

- SST filaments are easier to detect and to use to correct dynamical fields
- Chlorophyll filaments help the convergence
- Merging SST and Chlorophyll enables us to detect structure from the dynamics only

Improving the image processing may improve the estimation of the velocity

Realistic model data

Realistic model of the Solomon sea



High Resolution realistic model of the Solomon Sea (Nathacha Djath)

- Oynamics : NEMO-OPA code, sub-mesoscale permitting
- Horizontal resolution : $\frac{1}{36}^{\circ}$
- Vertical resolution : 46 levels
- Forcing : ERA-INTERIM
- Time range : 1989-2006

Process model data

Realistic model data

Similarity FSLE-tracer



Angle between FSLE and SST, histogram

Similar histograms for other daysSimilar histogram for SSS and SPICE

Inversion parameters

Sub-space of error

High-resolution model very turbulent: hard to build a consistent error sub-space.

In the following, the sub-space is idealized.

> 50 data from the model preceding the chosen study date.

Inversion parameters

- Sub-space of error: EOF analyses on the variation of the model between 5 days
- Background velocity: Velocity 5 days after the study date
- Tracer Image: Sea Surface Salinity and/or Sea Surface Temperature image

Realistic model data

area A



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Realistic model data

area A



area B



SST

SSS





SST

SSS

Realistic model data

Cost function



Realistic model data

Norm of the remaining error, area A



Information on dynamics reaveled by SST, SSS and SPICE differ
Inversion of spiciness filaments give the best result

Realistic model data

Norm of the remaining error, area A



 \succ Information on dynamics reaveled by SST, SSS and SPICE differ \succ Inversion of spiciness filaments give the best result

Conclusions



Limits



Prospects and Conclusions

Richness of future image observation



Prospects and Conclusions

Data image assimilation

Growing of complexities in model and observation

- Increase of non-linear effects
- Non Gaussian statistics



Data assimilation in a high resolution model

 At large scale, altimetry can control dynamics in the ocean.
At small scales, using tracer images to control complex structures. Thank you for your attention