



**eSurge** **esa** 

### Lessons learned assimilating EO data in Venice

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Institute of Marine Sciences (Venice) – CNR

  **eSurge**  
venice

This slide features a background image of a flooded square in Venice with green chairs. It includes logos for eSurge, ESA, and the Institute of Marine Sciences (Venice) - CNR. The title is 'Lessons learned assimilating EO data in Venice' and the presenter is Dr. Marco Bajo.

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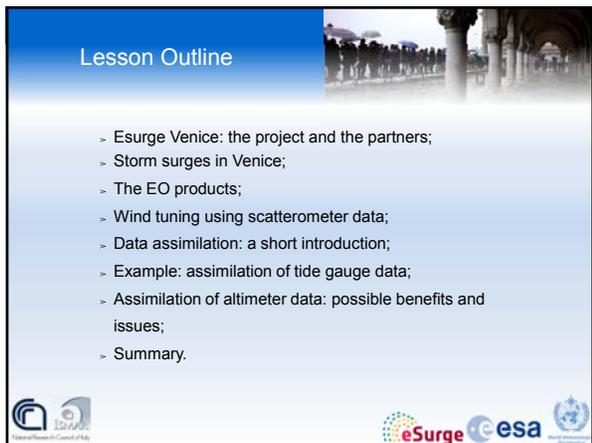
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### Lesson Outline

- > Esurge Venice: the project and the partners;
- > Storm surges in Venice;
- > The EO products;
- > Wind tuning using scatterometer data;
- > Data assimilation: a short introduction;
- > Example: assimilation of tide gauge data;
- > Assimilation of altimeter data: possible benefits and issues;
- > Summary.

  **eSurge** **esa** 

This slide has a blue header with the title 'Lesson Outline'. It contains a bulleted list of seven topics related to the eSurge Venice project. Logos for CNR, ISMAR, eSurge, and ESA are at the bottom.

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### eSurge Venice

eSurge-Venice (ESA Storm Surge for Venice) is a project funded by the European Space Agency, part of its Data User Element (DUE) programme. It aims to increase the usage of Earth Observation (EO) satellite data, within the storm surge community.

<http://www.esurge-venice.eu>

  **eSurge** **esa** 

This slide features a screenshot of the eSurge Venice project website. The text describes the project's funding and goals. A URL is provided. Logos for CNR, ISMAR, eSurge, and ESA are at the bottom.

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### Esurge Venice Partners

- › Institute of Atmospheric Sciences and Climate (ISAC-CNR);
- › Institute of Marine Sciences (ISMAR-CNR);
- › Biophysics Institute (IBF-CNR);
- › Institution Centro Previsioni e Segnalazioni Maree (ICPSM).





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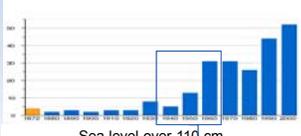
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### Storm surges in Venice

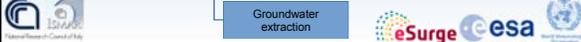


Total sea level	Flooded part
190 cm	88%
140 cm	59%
130 cm	46%
120 cm	28%
110 cm	12%
100 cm	5%




Sea level over 110 cm

Groundwater extraction




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### Operational storm surge models in Venice

#### Statistical models

Based on linear regression, they are still the most used to forecast the sea level in Venice.

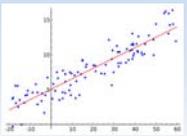
**Predictors** can be tide gauge data, msl pressure data or their gradients, measured wind, model forecast data (wind and pressure), etc.

$$y_i = a_1 X_{i1} + a_2 X_{i2} + \dots + a_n X_{in} + \epsilon_i$$

Residual error

Predictors

Calibration coeff





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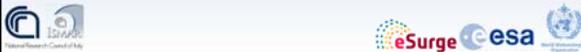
Operational storm surge models in Venice  
Statistical models

**The good:**

- > Very accurate in the short term forecast (6-12 hours)

**The bad:**

- > Difficult to find the best predictors;
- > Limited number of predictors;
- > Possible calibration problems (overfitting);
- > Extreme events' database is short.
- > ...



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Operational storm surge models in Venice  
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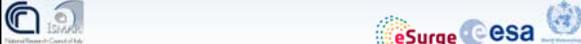
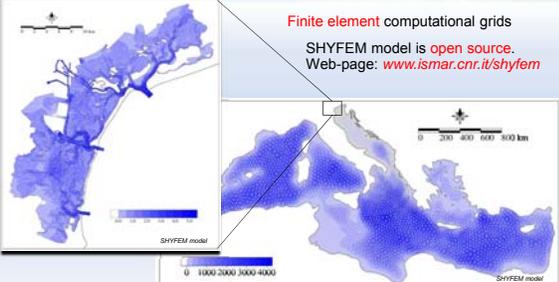
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Operational storm surge models in Venice  
Deterministic model

Finite element computational grids

SHYFEM model is open source.  
Web-page: [www.ismar.cnr.it/shyfem](http://www.ismar.cnr.it/shyfem)



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Operational storm surge models in Venice  
Deterministic model



Shallow water equations

$$\frac{\partial U}{\partial t} - fV = -H \left[ g \frac{\partial \zeta}{\partial x} + \frac{1}{\rho_0} \frac{\partial p_a}{\partial x} \right] + A_H \Delta U + \frac{1}{\rho_0} (\tau_{wx} - \tau_{bx})$$

$$\frac{\partial V}{\partial t} + fU = -H \left[ g \frac{\partial \zeta}{\partial y} + \frac{1}{\rho_0} \frac{\partial p_a}{\partial y} \right] + A_H \Delta V + \frac{1}{\rho_0} (\tau_{wy} - \tau_{by})$$

$$\frac{\partial \zeta}{\partial t} + \frac{\partial U}{\partial x} + \frac{\partial V}{\partial y} = 0$$



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Operational storm surge models in Venice  
Deterministic model



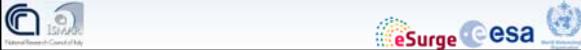
Shallow water equations

$$\frac{\partial U}{\partial t} - fV = -H \left[ g \frac{\partial \zeta}{\partial x} + \frac{1}{\rho_0} \frac{\partial p_a}{\partial x} \right] + A_H \Delta U + \frac{1}{\rho_0} (\tau_{wx} - \tau_{bx})$$

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$$\frac{\partial \zeta}{\partial t} + \frac{\partial U}{\partial x} + \frac{\partial V}{\partial y} = 0$$

Inverse Barometric Effect (IBE)      Wind stress




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Operational storm surge models in Venice  
Deterministic model



Total sea level  $\approx$  surge + tide + bias

Astronomical Tide

If tide-surge non-linear interactions are small the two contributions can be considered linearly independent

→ Tide can be computed for a specific location from the sea level observations, by means of harmonic analysis




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Operational storm surge models in Venice  
Deterministic model



Total sea level  $\approx$  surge + tide + bias

**Bias**

- Due to circulation induced by both average wind-pressure forcing and baroclinic forcing (low frequency oscillations  $\sim$ 2-3 days);
- Only partially reproduced by the model;
- Can be corrected using the mean sea level recorded in the latest days.



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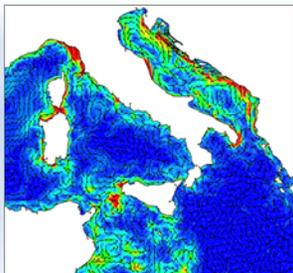
Operational storm surge models in Venice  
Deterministic model

**The good:**

- Good accuracy in medium-long range forecasts;
- no need of a database of parameters correlated to storm surge;
- Automatically improves when atmospheric forecast improves;

**The bad:**

- Less accurate for short range forecasts (<6h);
- Difficult to correct if the atmospheric forecast is wrong;
- Higher computational times.



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**EO data**



The spatial extension of EO data makes them very suitable to be used with deterministic models. Their use in statistic models is more difficult.

**SCATTEROMETER DATA** to improve the **wind forecast** used to force the model.

**ALTIMETER DATA** to improve the **initial state** of the model.



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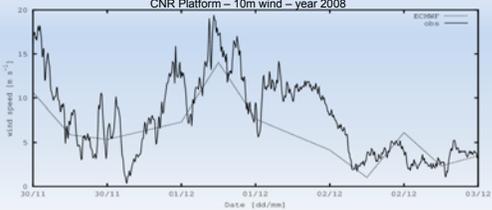
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**EO data scatterometer**



Winds in the Adriatic Sea are underestimated by atmospheric models, due mainly to the complicate orography of the geographical area.



Logos: CNR, eSurge, esa, Copernicus

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**EO data scatterometer**



We considered 12 Storm Surge Events (SEVs) happened in Venice. The highest in the last years.

- SEV 2010-02-28
- SEV 2010-11-10
- SEV 2010-11-19
- SEV 2010-11-21
- SEV 2010-11-26
- SEV 2010-12-23
- SEV 2011-02-16
- SEV 2012-10-27
- SEV 2012-10-31
- SEV 2012-11-02
- SEV 2012-11-11
- SEV 2012-11-28



Logos: CNR, eSurge, esa, Copernicus

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**EO data wind tuning**



- ECMWF wind speed ( $w_e$ ) was tuned using scatterometer wind speed ( $w_s$ );
- These quantities were averaged over a specific temporal window around each SEV;
- A similar procedure was followed for directions,  $\Delta\theta$ .

$$w_c = w_e + (w_s - w_e / w_s) w_e$$

$$\theta_c = \theta_e + \Delta\theta$$

Logos: CNR, eSurge, esa, Copernicus

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EO data  
wind tuning



Wind stress

The wind stress have been computed using a boundary layer model derived from Liu et al. (1979). The necessary bulk quantities input to the model are the SST, the dry and wet air temperatures, and the atmospheric pressure at the sea level. While the SST fields have been obtained from satellite observations (such as GHRSSST data), the other quantities were derived from ECMWF.




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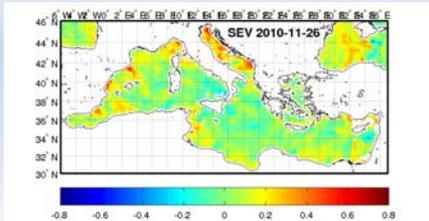
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EO data  
wind tuning



Normalised bias  $\Delta w/w_{scat}$  between ASCAT and ECMWF related to SEV 2010-11-26 for the Mediterranean Sea.





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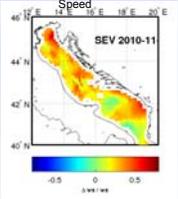
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EO data  
wind tuning

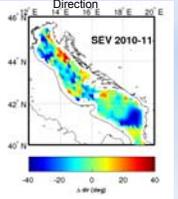


SEV 2010-11-26

Speed



Direction






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EO data  
wind tuning

Statistics of the **tuned** data shows a longer distribution's right tail, meaning higher speeds in the extreme events.

The figure contains four sub-charts. The top row shows two wind roses: 'Wind rose (original data)' and 'Wind rose (corrected data)'. The bottom row shows two histograms: 'Distribution of the speed (original data)' and 'Distribution of the speed (corrected data)'. The histograms show that the corrected data has a longer right tail, indicating higher wind speeds in extreme events. Logos for eSurge, esa, and the University of Cantabria are at the bottom.

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EO data  
wind tuning

Statistics of the **tuned** data shows a longer distribution's right tail, meaning higher speeds in the extreme events.

The figure contains four sub-charts: two wind roses and two histograms, comparing original and corrected wind data. The corrected data histogram shows a longer right tail. Logos for eSurge, esa, and the University of Cantabria are at the bottom.

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EO data  
wind tuning

- TYPE1 = Original wind fields;
- TYPE2 = Tuned wind fields.

- RES1 = Full Resolution (1/8 degree);
- RES2 = Middle resolution (1/4 degree);
- RES3 = Low resolution (1/2 degree).

The bottom of the slide features logos for eSurge, esa, and the University of Cantabria.

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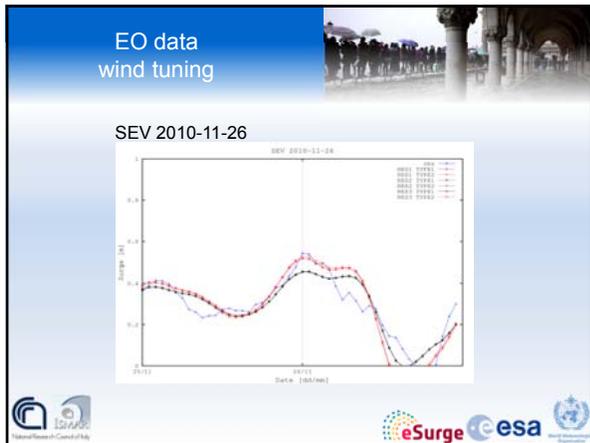
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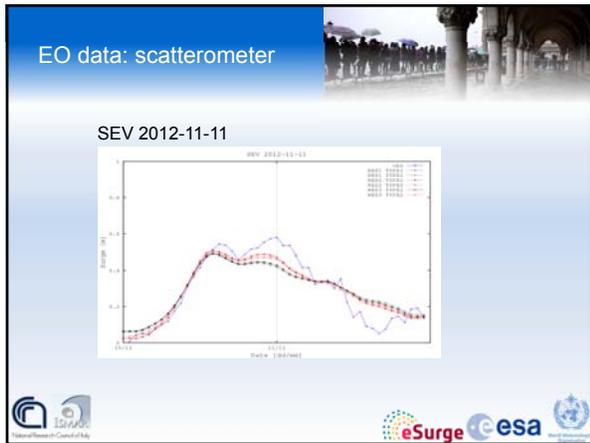
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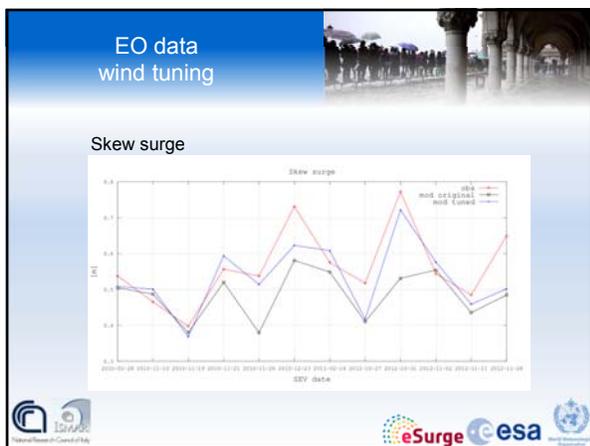
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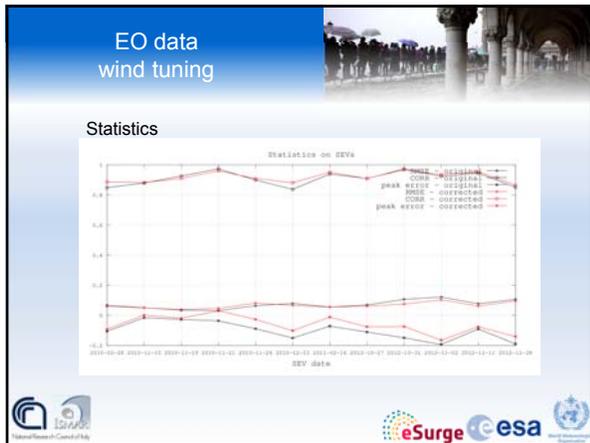
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### EO data wind tuning

Statistics

	RMSE [m]	Correlation	Peak error [m]
<i>Original data</i>			
<b>Full res.</b>	0.072	0.908	-0.102
<b>½ res.</b>	0.072	0.906	-0.102
<b>¼ res.</b>	0.073	0.904	-0.104
<i>Tuned data</i>			
<b>Full res.</b>	0.066	0.918	-0.062
<b>½ res.</b>	0.066	0.917	-0.063
<b>¼ res.</b>	0.067	0.915	-0.067

Logos: eSurge, esa

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### EO data wind tuning

Statistics

	RMSE [m]	Correlation	Peak error [m]
<i>Original data</i>			
<b>Full res.</b>	0.072	0.908	-0.102
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<i>Tuned data</i>			
<b>Full res.</b>	0.066	0.918	-0.062
<b>½ res.</b>	0.066	0.917	-0.063
<b>¼ res.</b>	0.067	0.915	-0.067

Logos: eSurge, esa

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**EO data: scatterometer**  
important remarks

What is the **best temporal window** to collect scatterometer data to tune the forecast wind fields?

- If it is too large it could not well represent that storm event;
- Scatterometer data should cover a wide part of the computational domain.



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**EO data**  
wind tuning

Changes in **spatial resolution** do not affect too much the performances.

That is because storm surges are mesoscale phenomena and the correctness of the **average wind speed and direction** are more important than the resolution of local features.

**But**

All the configurations tested come from the **same atmospheric model**. Models with lower resolutions give lower average wind (e.g., previous ECMWF model versions).

Statistics are very similar with tuned wind, but the reproduction of the **maximum peaks** improves.



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**Part II**  
**EO data: altimeter**

Altimeter data provide **spatial information** on the sea level.

These data can be **assimilated** to improve the **analysis** state of the model and, consequently, the forecast.

STEPS:

- Find all the **tracks** in the computational **domain** and in the **assimilation window**;
- **Process them** in order to have quantities that can be assimilated (i.e., residual water level);
- Use them with a data assimilation **technique**.



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**EO data: altimeter find tracks**

Example of altimeter tracks – SLA - SEV October 31, 2004

Logos: CNR, eSurge, esa, ESA

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**EO data: altimeter TWLE**

Total Water Level Envelope is computed from the SSH corrected for ionospheric delay, wet/dry tropospheric delays, sea state bias, loading tides and solid earth tides.

Oceanic tides and the local response to atmospheric forcing are left.

Logos: CNR, eSurge, esa, ESA

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**EO data: altimeter Tide**

The storm surge model computes the response to atmospheric forcing. The astronomical tide must be subtracted from the TWLE.

How to find the tidal correction?

- Use the tidal correction provided with the data (from a model);
- Provide a new correction using a high resolution tidal model;
- Use a storm surge model with the tidal forcing included.

Logos: CNR, eSurge, esa, ESA

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EO data: altimeter  
Tide

- Run a tidal model for about 1 year;
- Find the coefficients from a harmonic analysis of the sea level in each node;
- Find the tide along the altimeter's tracks with an interpolation of the values in the nearest nodes.



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EO data: altimeter

When is useful to assimilate altimeter data?

The initial state can be more or less important, depending on what processes the model must reproduce (complete calm, strong circulation, seiches, ...).

Find the spin-up time of the system to understand if the initial state is important or not.



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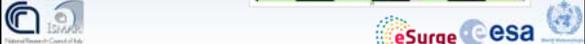
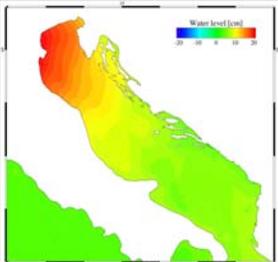
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EO data: altimeter  
Seiche reproduction

Seiches are important in the Adriatic Sea.

Seiches are free oscillations, happening in enclosed or semi-enclosed basins, following an initial perturbation of the sea level (i.e., a storm surge event).



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### EO data: altimeter Seiche reproduction

In the Adriatic Sea the seiches have a main period of ~22h and a second period of ~11h.

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### EO data: altimeter Assimilation schemes

**PURPOSE:** to improve the accuracy of the **initial state**, using observations to **minimise** the errors.

The **analysis** state minimises a **cost function** where different sources of errors are considered.

real time assimilation

non-linear methods

(4D-Var or) 4D-PSAS with model error

↙ EKF ↘

intermittent 4D-Var or 4D-PSAS

↙ ↘

3D-Var or 3D-PSAS

↙ ↘

Optimal Interpolation (OI)

↙ ↘

Cressman Successive Corrections nudging

↙ ↘

Interpolation of observations

From Boutier, F., et al. (1999)

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### EO data assimilation

#### 4D-VAR cost function

Find an initial perturbation  $\delta x_0$ , which minimises:

$$J(\delta x(t_0), \dots, c_i, \dots) = \frac{1}{2} \delta x^T B^{-1} \delta x + \frac{1}{2} \sum_{i=0}^n (H_i \delta x(t_i) - d_i)^T R^{-1} (H_i \delta x(t_i) - d_i) + \frac{1}{2} \sum_{i=1}^n c_i^T Q_i^{-1} c_i$$

Background errors
Observation errors
Model errors

Dimension analysis

$$J(x) = (x - x_0)^T B^{-1} (x - x_0) + (y - Hx)^T R^{-1} (y - Hx)$$

See Courtier, P., (1997)

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### EO data assimilation

4D-PSAS is a "dual formulation" of 4D-VAR. The minimisation is made in the observation space.

4D-Psas cost function and gradient:

$$F(u) = \frac{1}{2} u^T \left( I + \sqrt{R^{-1}} G D G^T \sqrt{R^{-1}} \right) u - u^T \sqrt{R^{-1}} d$$

$$\nabla_u F = \left( I + \sqrt{R^{-1}} G D G^T \sqrt{R^{-1}} \right) u - \sqrt{R^{-1}} d$$

with:

$$d = y^o - Hx^b$$

$$\delta x = B H^T \sqrt{R^{-1}} u$$

$$G = \left( \dots, H, M_{(t_i, t_0)} \dots \right)$$

$$D = B \oplus \sum Q$$

See Courtier, P., (1997)




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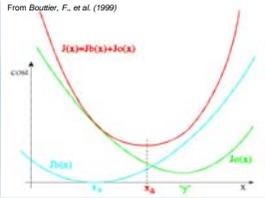
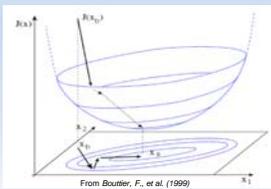
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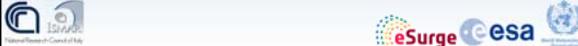
### EO data assimilation

From Bouvier, F., et al. (1999)

Cost function and minimisation algorithm

From Bouvier, F., et al. (1999)




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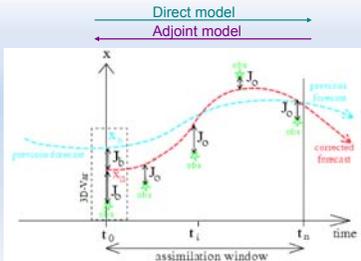
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### EO data assimilation

Direct model  
Adjoint model



From Bouvier, F., et al. (1999)




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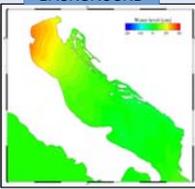
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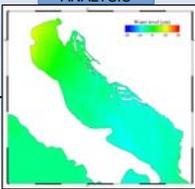
### EO data Assimilation of tide gauge data

Assimilation of residual levels from tide gauges in the Adriatic Sea

**BACKGROUND**



**ANALYSIS**



Logos: eSurge, esa, and other institutional logos.

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### EO data Assimilation of tide gauge data

Tide gauges can provide a good observational description near the coasts.

*But in the open sea and near some coasts they are not present or can be not available*

Altimeter data can supply to this lack



Logos: eSurge, esa, and other institutional logos.

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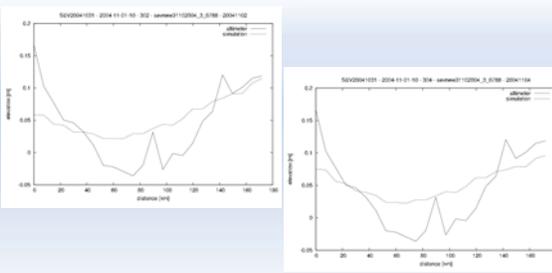
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### EO data altimeter vs model



Logos: eSurge, esa, and other institutional logos.

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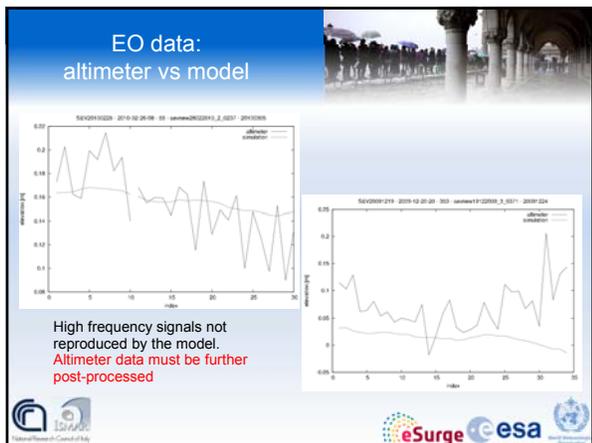
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- ### Concluding remarks
- What are the main sources of the forecast errors in the specific case? Is the wind bad estimated? Is the initial state important?
  - What EO data are available operationally in the study area?
  - Tuning the wind with scatterometer data improves the forecasts in all the storm surge events considered;
  - Initial state can be improved with data assimilation;
  - Altimeter data can provide informations where other observations are not available;
  - Altimeter data must be carefully post-processed before being assimilated.
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