# Sequential assimilation of multi-mission dynamical topography into a global finite-element ocean model



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### Introduction

This study focuses on an estimation of ocean circulation via assimilation of satellite measurements of dynamical ocean topography (DOT) into the global finite-element ocean model (FEOM). The DOT data are derived from a complex analysis of multimission altimetry data combined with a referenced earth geoid. The goal of this work is exploring the feasibility of assimilation of the global altimetric signal based on sequential assimilation technique. Two different sequential assimilation techniques were The difference is significant reaching  $\pm 0.5$  m in some areas. Many of them develop as the result of model adaptation to the bottom topography and present a systematic model bias.

The strong systematic difference between the observations and the model can be reduced by the adiabatic pressure correction suggested by [7, 2]. It works through modifying the pressure field leaving consistent tracer fields. It is introduced by replacing the model density  $\rho_m$  with a combination

### $\rho^* = \alpha \rho_m + (1 - \alpha) \rho_c$



### implemented.

First technique uses the method of adiabatic correction [7, 2]to reduce systematic difference between mean state of the model and the mean DOT. Then, a local SEIK filter (as implemented within PDAF [6]) is used for the assimilation of the time varying DOT and temperature and salinity are updated following the vertical structure of the first baroclinic mode. Second sequential technique uses the local SEIK filter without adiabatic corrections for all model fields while assimilating the same data.

### **Observations**

The DOT data were provided by R. Savcenko and W. Bosch, DFG, Munich, Germany. They were obtained by combining the ENVISAT, GFO, Jason and TOPEX/Poseidon missions data with a referenced geoid provided by the Geo-Forschung Zentrum (GFZ), Potsdam, Germany. The data cover the period between January 2004 and January 2005. They are interpolated onto the model grid so that the observations are available at every point of the model grid every ten days.

The presence of ice makes the altimetry data unusable in the polar areas. The Indonesian region is characterized by complex bottom topography where neither geoid measurements nor model results appear to be accurate enough. This is also true for the Mediterranean Sea. The observational data in these areas were substituted by the values of the RIO05 mean dynamical topography (MDT). These areas are shown in the right panel of figure below as the deep-blue rectangular areas.

with  $\alpha = 0.5$  and  $\rho_c$  the climatological density. The **configuration** of the model with this correction is further referred as  $V_2$ .

The right panel of figure below shows mean difference calculated using  $V_2$  configuration. The difference in the mean fields is reduced compared to the  $V_1$  version in all regions, especially in North Atlantic.



Left: The mean difference between the DOT and model run  $V_1$ . Right: The mean difference between DOT and model run  $V_2$ .

### Assimilation scheme I 5

It is based on a model configuration  $V_2$ . At each time the observations are available, the analysis of the SSH field is carried out applying the local SEIK filter [5] so that the analysis for each water column of the model depends only on observations within a specified influence region. In this study, the influence region is a circle with a radius of 200 km. Using this information, the vertical profiles of temperature and salinity are updated according to the vertical structure of the first baroclinic mode [3] with the amplitude computed from the elevation update, i.e.:

#### 180 210 240

Evolution of RMS error of SSH for the world ocean (except zones corresponding to RIO05 MDT location in the data. The green and yellow solid lines show the errors corresponding to the  $V_1$  and  $V_2$  free simulations (without assimilation), respectively. The blue lines with bullets represent the 10-day model forecasts, while the dotted red lines correspond to the analysis.

### Assimilation scheme II

This assimilation experiment is based on a model configuration without the adiabatical pressure correction. At each time the observations are available, the analysis of the full ocean field is carried out applying the local SEIK filter [5] so that the analysis for each water column of the model depends only on observations within a specified influence region. In this experiment, the influence region is a circle with a radius of 900 km. The observational error covariance matrix is modified according to the distance of the observations to the water column using the 5th order polynomial weighting.







Left: The mean DOT for the period from January 2004 till January 2005.

Right: Standard deviation for the same time period. The deepblue rectangular areas correspond to the locations where the RIO05 MDT was substituted in the data (no variability).

#### Ocean model 3

The study was performed by the Finite-Element Ocean circulation Model (FEOM) [4, 1] configured on a global almost regular triangular mesh with the spatial resolution of 1.5°. There are 24 unevenly spaced levels in the vertical direction. The model is forced at the surface with momentum fluxes derived from the ERS scatterometer wind stresses complemented by TAO derived stresses and relaxed to monthly mean climatology at the surface. It is initialized by mean climatological temperature and salinity. This **configuration** is further referred to as  $V_1$ .

$$T^{a}(x,y,z) = T^{f}(x,y,z) + \delta\eta(x,y) \frac{g\rho_{0}\hat{h}(z)}{\hat{p}(0)} \frac{\partial\bar{T}}{\partial z}(x,y),$$
  
$$S^{a}(x,y,z) = S^{f}(x,y,z) + \delta\eta(x,y) \frac{g\rho_{0}\hat{h}(z)}{\hat{p}(0)} \frac{\partial\bar{S}}{\partial z}(x,y).$$

Here, overbars denote the reference state calculated as a mean from the one year free model run of the  $V_2$  model,  $\hat{p}$  and h are locally defined vertical structures of the first baroclinic modes of velocity and displacement calculated using the local vertical profiles of the Brunt-Väisälä frequency and density from the  $V_2$ model. The function  $\delta \eta(x, y)$  is the analysis increment, i.e. the difference between the analysis of SSH,  $\eta_k^a$  and its forecast  $\eta_k^J$ , and g is the acceleration due to gravity. The velocity field is left unchanged so that it is simply the result of the model evolution.



Experiment with 5th order polynomial weighting and  $L = 9 \times 10^5$ . Left: The mean difference between DOT and analysis. Right: The mean difference between DOT and forecasted fields.



Evolution of RMS error of SSH for the world ocean (except zones) corresponding to RIO05 MDT location in the data) for assimilation scheme II.

### Conclusion

• First assimilation technique leads to a partially successful assimilation approach reducing the rms difference between the model and data from 16 cm to 2 cm. However, it remains suboptimal, showing a tendency in the forecast phase of returning

**Comparison of model results with** observations

After a 10-year spin up from the state of rest, versions  $V_1$  is run for one additional year and the output is stored every 10 days. The left panel of the figure below depicts the difference between the mean DOT and the mean calculated from  $V_1$  model run.

Left: The mean difference between the dynamical topography obtained from the observations and from analysis.

**Right**: The mean difference between the dynamical topography obtained from the observations and from forecasted fields.

toward a free run without data assimilation.

- Second assimilation technique leads to reduction of the rms difference between the model and data from 16 cm to 3 cm. Further, the tendency to return toward the free run is reduced for this technique.
- Both the mean difference and standard deviation of the difference between the forecast and observation data are reduced as the result of assimilation with sequential assimilation methods.

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