



Using Diva on large datasets: applications and tips

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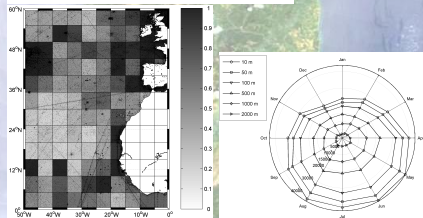
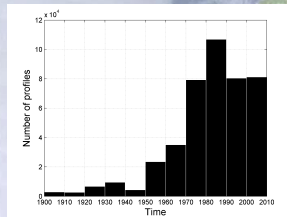
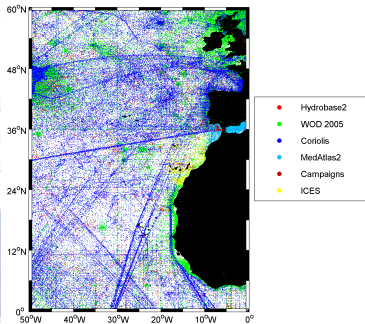
Outline

- 1 some applications of the tools presented yesterday
- 2 some tips from my tries to get a climatology in the Atlantic
- 3 application of Diva + satellite imagery + numerical model
- 4 lots of pictures



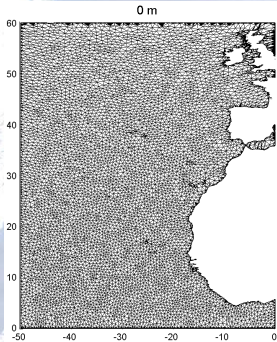
Data

- 0 – 50° W, 0 – 60° N
- WOA + Coriolis + Hydrobase + MedAtlas + campaigns + misc
- from 1890 to August 2008



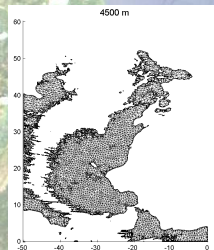
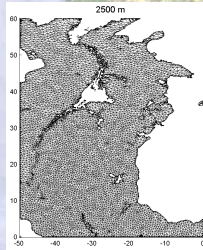
Contour and mesh

- contours from DBDBV (GEBCO too fine)
- $L = 3.0$
- same L for each level



Tips:

- generate the mesh only once for each level
→ implementation in 3Dinfo ?
- choose L not too small (memory issue)
- choose L not too large (resolution issue)



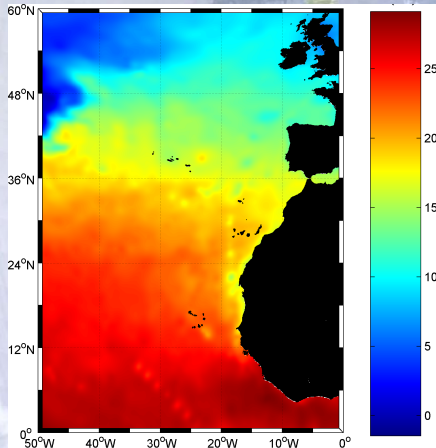
Analysis

Configuration:

- $L = 2.73$ from divafit
- $\lambda = 4.0$ from divafit
- nothing magic

Conclusions:

- divafit too slow with lots of data
→ use divafit nsamples
- visible outliers...



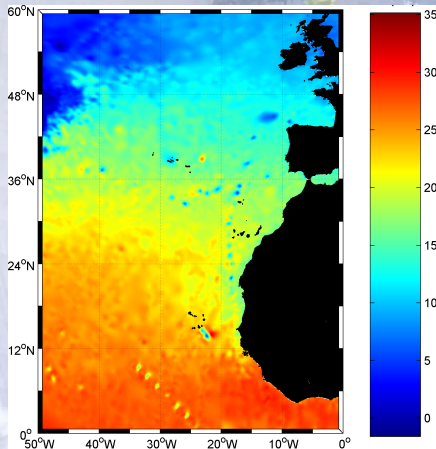
Analysis

Configuration:

- L from divafit
- $\lambda = 300.0$ from divagcv

Conclusions:

- L too small, λ too large
- outliers more visible



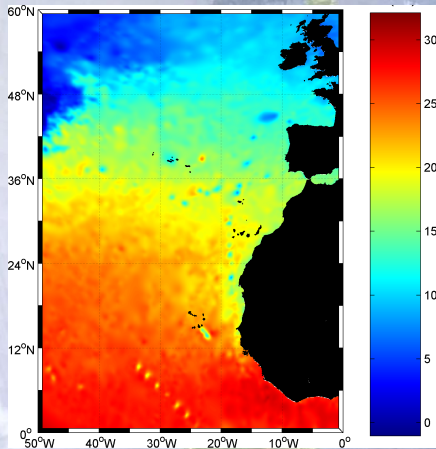
Analysis

Configuration:

- $L = 5$ (experience, physics, ...)
- $\lambda = 300.0$ from `divagcv`

Conclusions:

- λ still too large
- outliers did not disappear (no magic)



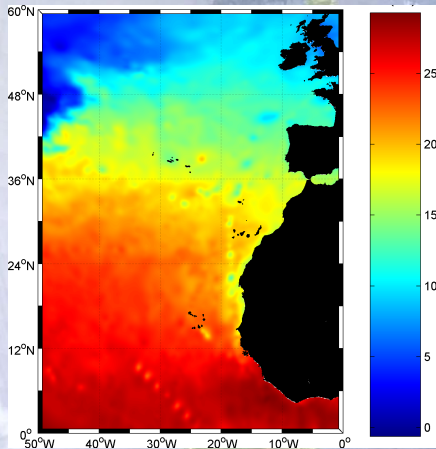
Analysis

Configuration:

- L from divafit
- $\lambda = 14.4$ from
divacvrand 2000 2 -r

Conclusions:

- more realistic estimate of λ
using divacvrand



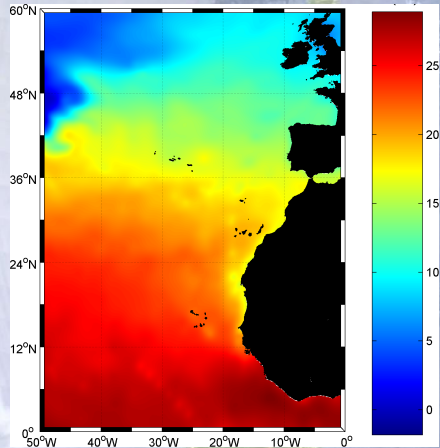
Analysis

Configuration:

- L from divafit
- $\lambda = 1$

Conclusions:

- very smooth field
- outlier still visible



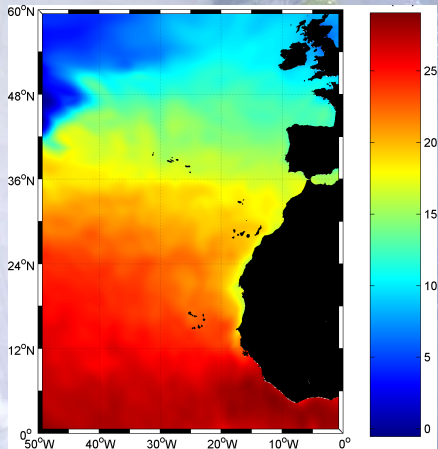
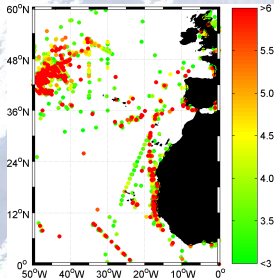
Analysis

Configuration:

- L from divafit
- λ from divacvrand
- outlier removal with divaqcbis

Conclusions:

- 696 outliers detected
- improved results



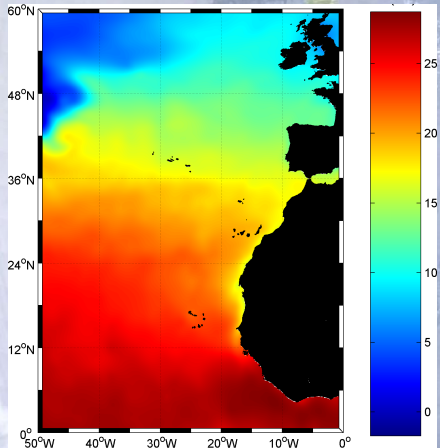
Analysis

Configuration:

- outlier removal
- L from diva_{fit}
- $\lambda = 1$ (with diva_{cv} 63.5)
- semi-normed analysis

Conclusions:

- lots of steps before satisfying field
- need good parameters before using diva_{qc}
- semi-normed analysis helps where few data are available



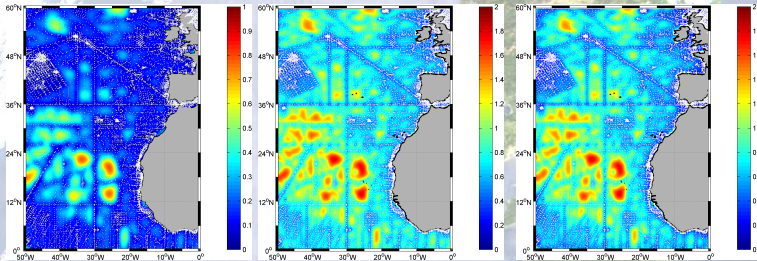
Error fields

Methods:

- 1 *poor man's error estimate* (quick but underestimated)
- 2 hybrid approach, analogy with O.I
- 3 real covariance function → don't use your laptop!

Conclusions:

- 1 highly depends on data coverage
- 2 same distribution of errors
- 3 different orders of magnitude



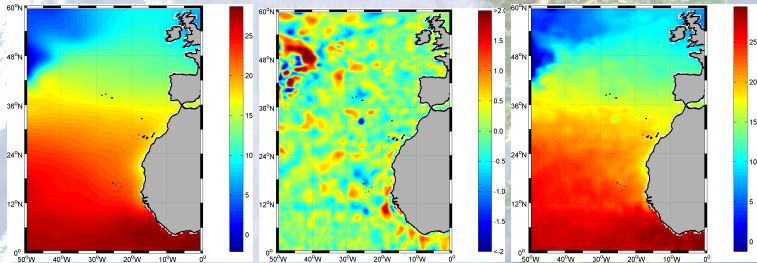
Semi-normed analysis

Methods:

- 1 create a reference field with large L and small λ
- 2 compute anomalies with respect to reference
- 3 analyse the anomalies
- 4 sum up the 2 fields

Utility:

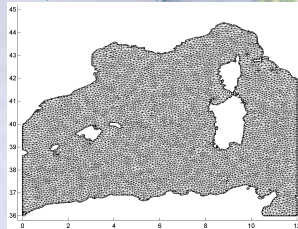
- 1 better field where no data are available



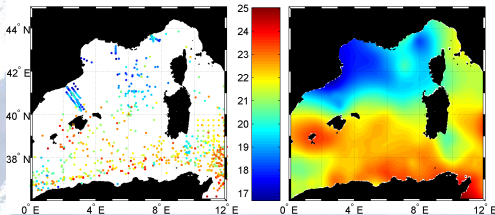
Diva

Data-Interpolating Variational Analysis
[Brasseur et al. (1996), Brankart and Brasseur (1996, 1998)]

- = advanced data-gridding method
- + finite-element resolution
- + coastlines + advection influence
- + error maps
- + many more



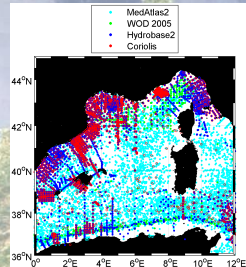
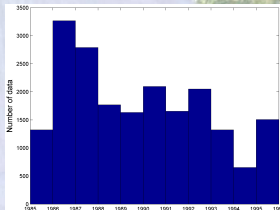
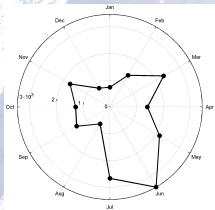
October: 1278 data



Data used

Data set = aggregation between:
WOD05 + MedAtlas2 + Hydrobase2 + Coriolis

- 24293 unique profiles
- observations with depth $< 5\text{ m}$ (82.6% of the profiles)
- 1986: 3267 measurements, 1994: 653
- $> 40\%$ of data between May and July
- mean: 18.4769°C , standard deviation: 3.8773°C
- 99.3% of data have $12 < T < 27^\circ\text{C}$



Introduction

Large data set issues

Other tools

Diva + dineof + GHER model

Monthly temperature fields



Thanks for your attention!