

6th Diva user workshop

Theory and 2-D version

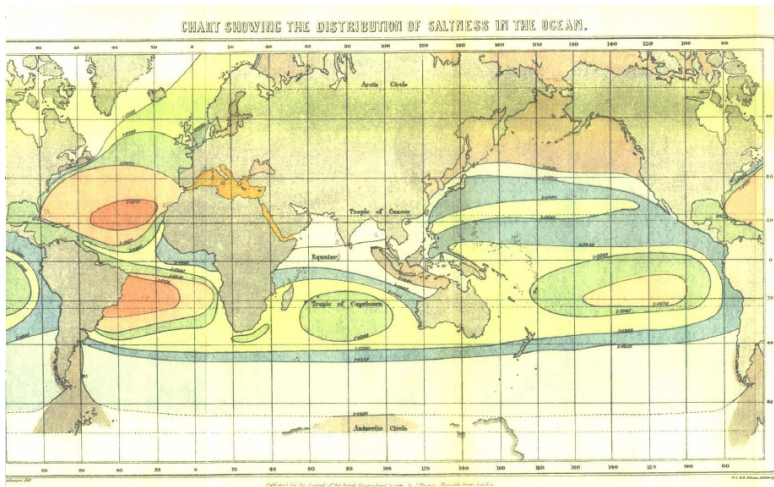
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University of Liège, Belgium
<http://modb.oce.ulg.ac.be/>

Roumaillac, October 8–12, 2012



Diva: interpolation (gridding) of in situ data



What is Diva?

Data
Interpolating
Variational
Analysis

What is Diva?

- a method to produce gridded field
- a set of scripts and Fortran programs

What is not Diva?

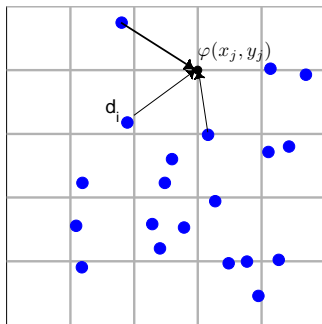
- a plotting tool
- a *black-box*
- a numerical model



gridding and plotting are different tasks

DIVA: Data-Interpolating Variational Analysis

N_d data points d_i • \rightarrow gridded field



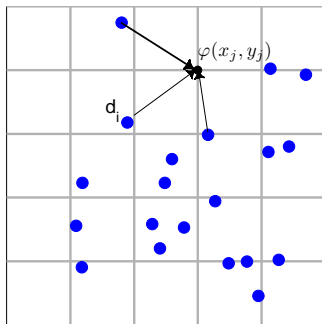
Formulation: minimize cost function $J[\varphi]$

$$\min J[\varphi] = \sum_{i=1}^N \mu_i [d_i - \varphi(x_i, y_i)]^2 \quad \text{data-analysis misfit}$$

$$+ \int_D (\nabla \nabla \varphi : \nabla \nabla \varphi + \alpha_1 \nabla \varphi \cdot \nabla \varphi + \alpha_0 \varphi^2) dD \quad \text{field regularity}$$

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Four ways to use Diva

- 1 Command line (batch processing)
- 2 Ocean Data View
- 3 Diva-on-web
- 4 Matlab package

```
divastripped : bash
////////////////////////////////////
Mesh generation
////////////////////////////////////

Reading parameters

Copying output files for visualisation
in directory ./output/meshvisu/

-----
Mesh is created
-----

////////////////////////////////////
going to analyse a field
////////////////////////////////////

Number of data points: 3

VARBAK: 1.0

Errors will be calculated

Output of results copied in ./output/

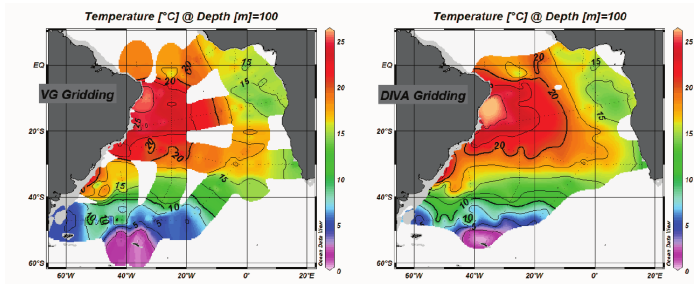
Creating netcdf file for field and associated error
in directory ./output/ghertonetcdf/

-----
Analysis is finished
-----

Check the results in
./ouput/ghertonetcdf/results.nc (netcdf)
[charles@gher13 divastripped]$
```

Four ways to use Diva

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[R. Schlitzer (2009), Ocean Data View User's Guide, Version 4.2]

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<http://gher-diva.phys.ulg.ac.be/web-vis/diva.html>

The screenshot displays the Diva web interface. At the top, there are navigation tabs: 'Upload', 'Grid', 'Analysis' (which is highlighted in green), 'Statistics', 'Download analysis', 'Link or embed', 'Report a problem', and 'Help'. Below the 'Analysis' tab, the section is titled 'Analysis with Diva'. It contains two input fields: 'Correlation length [deg]:' with the value '0.6' and 'Signal to noise ratio:' with the value '2'. Below these fields is a 'diva!t!' button. Underneath, the 'Optional parameters' section includes 'Maximum rel. error (from 0 to 1):' with the value '0.3' and a 'Previous' button. To the right of the 'Maximum rel. error' field is a 'Make analysis/Update' button. The main part of the interface is a map of the North Atlantic region, overlaid with a color-coded data field. A vertical color scale legend is positioned to the right of the map, ranging from 20.7 (dark blue) to 21.9 (dark red). The map includes navigation controls (arrows and a zoom-in button) and a coordinate display at the bottom right showing '23.10000, 41.52090'.

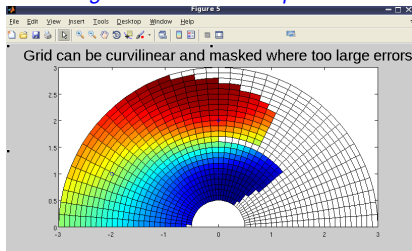


Four ways to use Diva

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Package available at

<http://modb.oce.ulg.ac.be/mediawiki/upload/divaformatlab.zip>



Requires executables of Diva (mesh and analysis)

A little bit of history

Code development (1990-1996):

- Variational Inverse Method (VIM) (Brasseur, 1991, JMS, JGR)
- cross-validation (Brankart and Brasseur, 1996, JAOT)
- error computation (Brankart and Brasseur, 1998, JMS; Rixen et al., 2000, OM)

2D-analysis (2006-2007):

- set of bash scripts (`divamesh`, `divacalc`, ...)
- Fortran executables
- parameters optimization tools
- Matlab/Octave scripts for plotting

3D-analysis (2007-2008):

- superposition of 2D layers
- automated treatment and optimization
- stability constraint (Ouberdous et al.)
- advanced error computation (Troupin et al., 2012, OM)

A little bit of history

4D-analysis (2008-2009):

- start from ODV spreadsheet
- *detrending* (with J. Carstensen, DMU)
- NetCDF 4-D climatology files

Web tools

- On-line analysis (Barth et al., 2010, Adv. Geosci.)
<http://gher-diva.phys.ulg.ac.be/web-vis/diva.html>
- Climatology viewer:
<http://gher-diva.phys.ulg.ac.be/web-vis/clim.html>

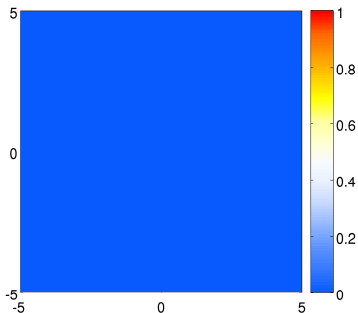
2010 (on-going)

- transition to Fortran95
- multivariate approach
- data transformation tools
- 4-D graphical interface
- implementation of *source/decay* terms
- ...

General: user-driven developments

A simple example with a few points

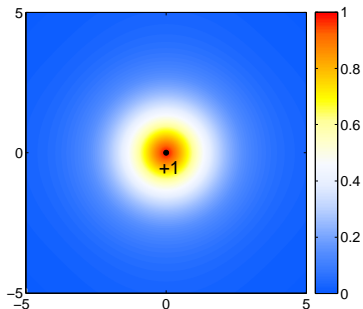
Zero background



- influence around data point
- variable radius of influence
- distance between data points
- separation by obstacles
- confidence in measurements

A simple example with a few points

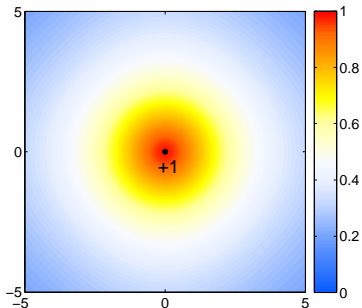
Square domain, data at center



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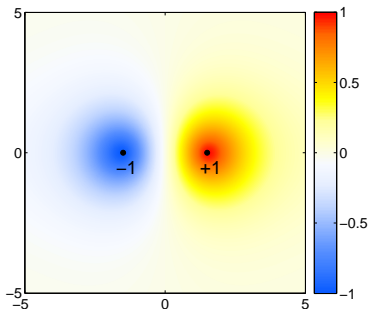
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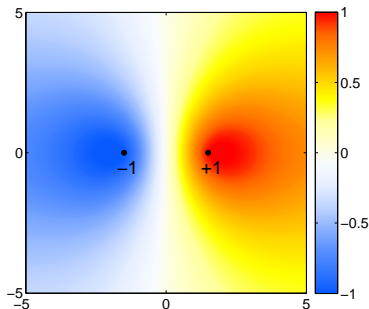
Two symmetric data points



- influence around data point
- variable radius of influence
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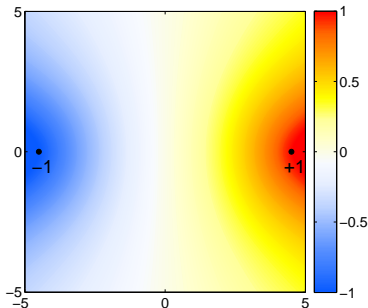
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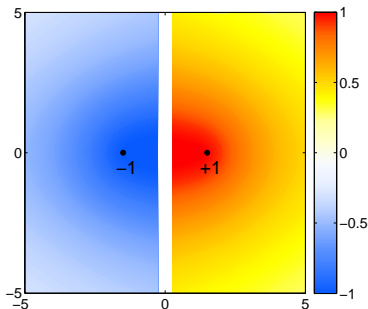
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A simple example with a few points

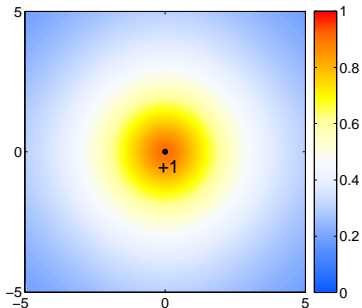
Physical barrier



- influence around data point
- variable radius of influence
- distance between data points
- separation by obstacles
- confidence in measurements

A simple example with a few points

Data at center



- influence around data point
- variable radius of influence
- distance between data points
- separation by obstacles
- confidence in measurements

Analysis parameters are related to data

Non-dimensional version:

$$L = \text{length scale} \quad \rightarrow \quad \tilde{\nabla} = L \nabla \quad (1)$$

$$\rightarrow \quad D = L^2 \tilde{D} \quad (2)$$

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- $\alpha_0 \rightarrow L$ for which data-analysis misfit \simeq regularity term: $\alpha_0 L^4 = 1$
- $\alpha_1 \rightarrow$ influence of gradients: $\alpha_1 L^2 = 2\xi, \quad \xi = 1$
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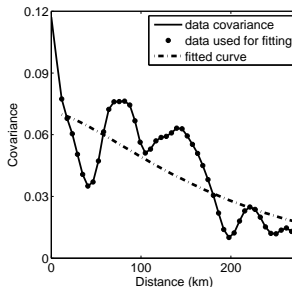
Coefficients α_0 , α_1 and μ_i related to

- 1 Correlation length L
- 2 Signal-to-noise λ
- 3 Observational noise standard deviation ϵ_i^2

Main analysis parameters

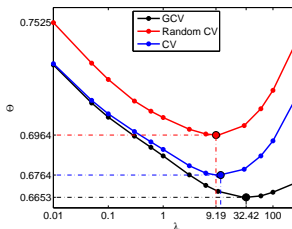
Correlation length L :

- Measure of the *influence* of data points
- Estimated by a least-square fit of the covariance function



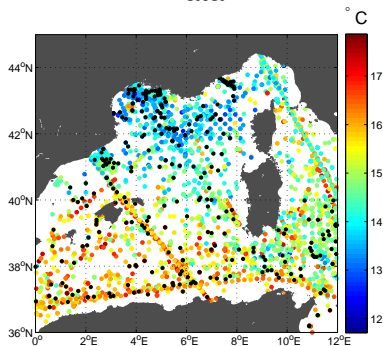
Signal-to-noise ratio λ :

- Measure of the *confidence* in data
- Estimated with Generalized Cross Validation techniques

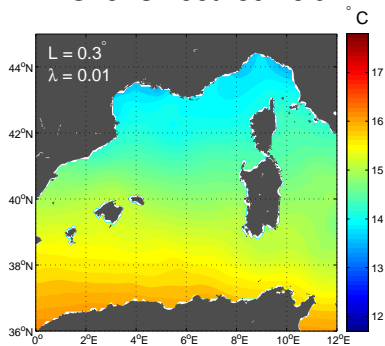


Parameters: smoothness vs. data influence

Data

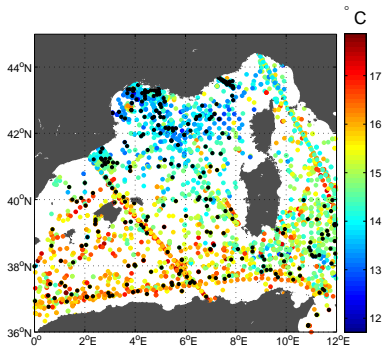


Over-smoothed field

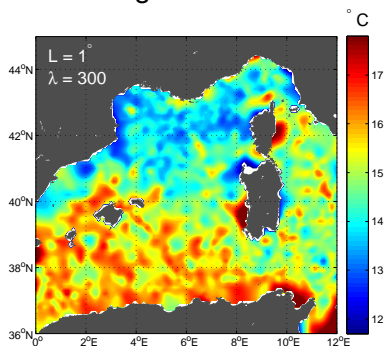


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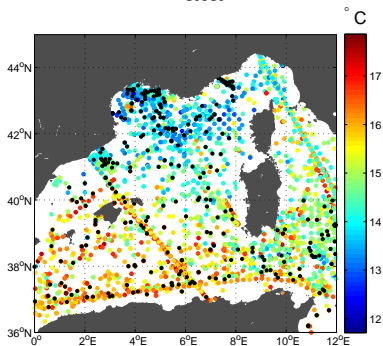


Too strong influence of data

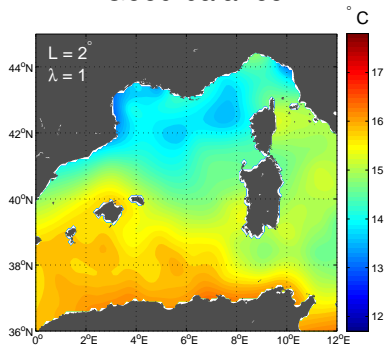


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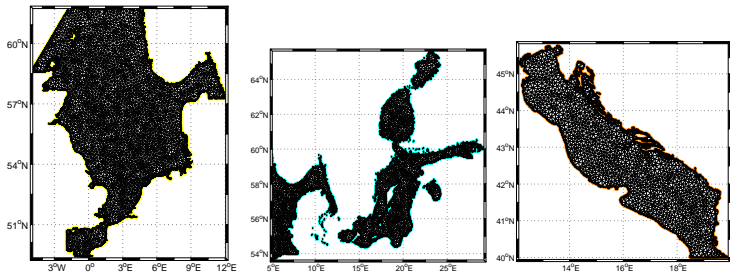


Good balance



Minimization with a finite-element method

Field regularity → plate bending problem → finite-element solver



Advantages:

- boundaries taken into account
- numerical cost (almost independent on data number)
- no *a posteriori* masking

Minimization with a finite-element method

Triangular FE only covers sea:
$$J[\varphi] = \sum_{e=1}^{N_e} J_e(\varphi_e) \quad (3)$$

In each element: $\varphi_e(\mathbf{r}_e) = \mathbf{q}_e^T \mathbf{s}(\mathbf{r}_e)$ with
$$\begin{cases} \mathbf{s} & \rightarrow \text{shape functions} \\ \mathbf{q} & \rightarrow \text{connectors} \\ \mathbf{r}_e & \rightarrow \text{position} \end{cases} \quad (4)$$

(4) in (3) + variational principle

$$J_e(\mathbf{q}_e) = \mathbf{q}_e^T \mathbf{K}_e \mathbf{q}_e - 2\mathbf{q}_e^T \mathbf{g}_e + \sum_{i=1}^{N_{de}} \mu_i d_i \quad (5)$$

where
$$\begin{cases} \mathbf{K}_e & \rightarrow \text{local stiffness matrix} \\ \mathbf{g} & \rightarrow \text{vector depending on local data} \end{cases}$$

Minimization with a finite-element method

On the whole domain:
$$J(\mathbf{q}) = \mathbf{q}^T \mathbf{K} \mathbf{q} - 2\mathbf{q}^T \mathbf{g} + \sum_{i=1}^{N_d} \mu_i d_i \quad (3)$$

Minimum:
$$\mathbf{q} = \mathbf{K}^{-1} \mathbf{g} \quad (4)$$

$$\mathbf{q} = \mathbf{K}^{-1} \mathbf{g} \quad (5)$$

- Stiffness matrix
- Connectors (new unknowns)
- Charge vector

Mapping of data on FEM \rightarrow transfer operator $\mathbf{T}_2 \rightarrow \mathbf{g} = \mathbf{T}_2(\mathbf{r})\mathbf{d}$
Solution at any location \rightarrow transfer operator $\mathbf{T}_1 \rightarrow \varphi(\mathbf{r}) = \mathbf{T}_1(\mathbf{r})\mathbf{q}$

Results obtained at any location $\rightarrow \varphi = \mathbf{T}_1(\mathbf{r})\mathbf{K}^{-1}\mathbf{T}_2(\mathbf{r})\mathbf{d}$