

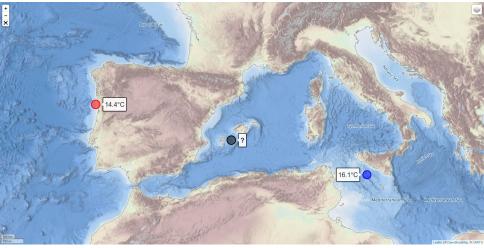
DIVA software and the



@CharlesTroupin, A. Barth, S. Watelet & J.-M. Beckers University of Liège, GeoHydrodynamics and Environment Research

EUDAT Conference, Porto (Portugal), 22-25 January 2018

Can you guess the temperature at the "?"

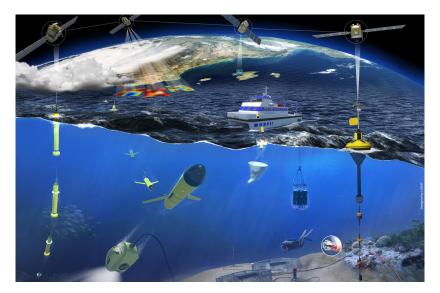




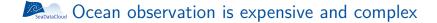
Spatial interpolation: Why is it needed?



Ocean observation is expensive and complex



Credit: www.socib.es

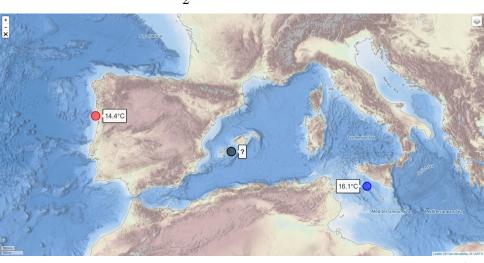


"A measurement not made is a measurement lost forever"

"Collect once, use many times"

Can you guess the temperature at the "?"

$$\frac{14.4 + 16.1}{2} = 15.25^{\circ} \text{C} \qquad ??$$





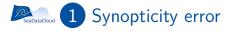
Can you guess the temperature at the "?"

$$\frac{14.4 + 16.1}{2} = 15.25^{\circ} \text{C} \qquad ??$$

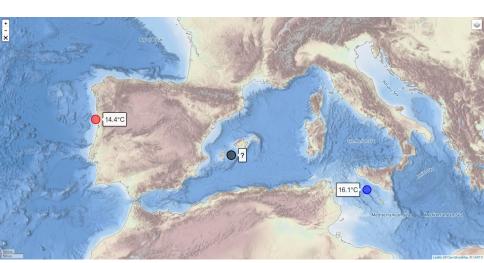




6 reasons why spatial interpolation



Measurements not collected at the same time



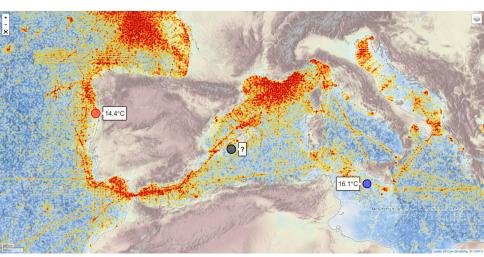


What we measure is not always what we intend to analyse

Example: I want the mean annual temperature off Porto but ships are only at sea when the weather is good

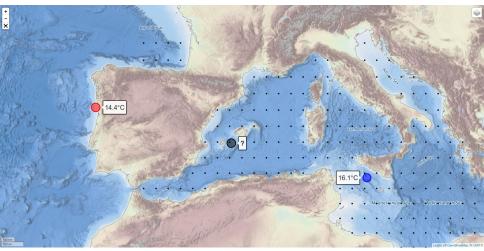


A lot of observations, but not everywhere





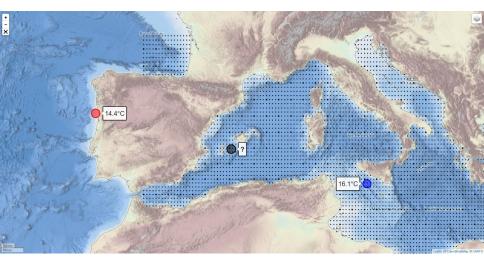
Need to interpolate at many locations







4 Need to interpolate at many locations





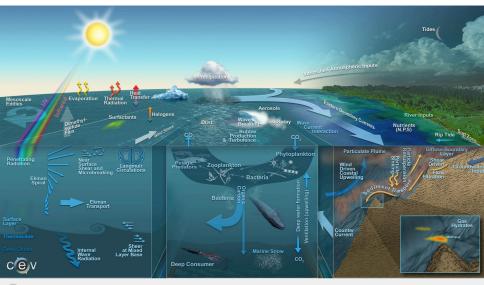
Land acts as a physical barrier







6 A lot of processes taking place...







How do we do it?

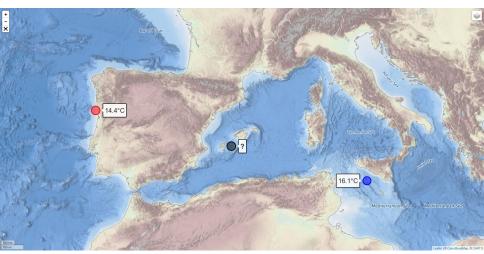
Minimisation of a cost function taking into account:

- Closeness to the observations
- 2 Regularity/smoothness of the solution

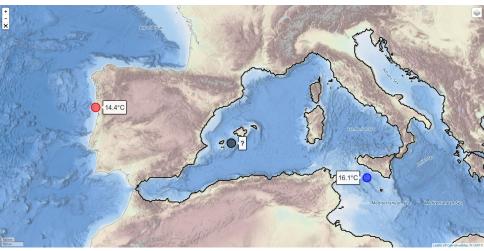
$$J[\varphi] = \sum_{i=1}^{N} \mu_i \left[d_i - \varphi(x_i, y_i) \right]^2 + \int_{D} \left(\nabla \nabla \varphi : \nabla \nabla \varphi + \alpha_1 \nabla \varphi \cdot \nabla \varphi + \alpha_0 \varphi^2 \right) dD,$$

solved by a finite-element technique

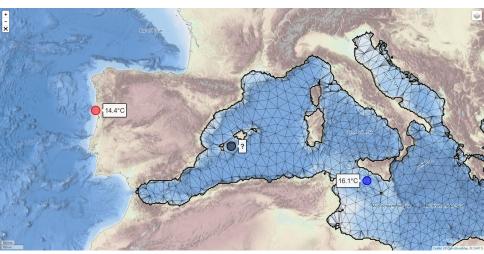
- https://github.com/gher-ulg/DIVA
- **DOI** 10.5281/zenodo.836727



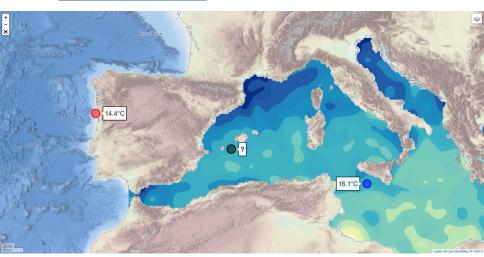
- https://github.com/gher-ulg/DIVA
- **DOI** 10.5281/zenodo.836727



- https://github.com/gher-ulg/DIVA
- doi DOI 10.5281/zenodo.836727



- https://github.com/gher-ulg/DIVA
- doi DOI 10.5281/zenodo.836727





2013: Octave or MATLAB

2016: julia

faster, better, stronger

divand-1.0: n-dimensional variational data analysis for ocean observations

A. Barth^{1,*}, J.-M. Beckers¹, C. Troupin², A. Alvera-Azcárate¹, and L. Vandenbulcke^{3,4}

1GHER, University of Liège, Liège, Belgium

²IMEDEA, Esporles, Illes Balears, Spain

3 seamod.ro/Jailoo srl, Sat Valeni, Com. Salatrucu, Jud. Arges, Romania

4CIIMAR, University of Porto, Porto, Portugal

* Invited contribution by A. Barth, recipient of the EGU Arne Richter Award for Outstanding Young Scientists 2010.

Correspondence to: A. Barth (a.barth@ulg.ac.be)

Received: 7 June 2013 - Published in Geosci, Model Dev. Discuss.: 23 July 2013

Revised: 18 October 2013 - Accepted: 12 December 2013 - Published: 29 January 2014



| https://www.geosci-model-dev.net/7/225/2014/gmd-7-225-2014.pdf



https://github.com/gher-ulg/divand.jl



DIVAnd: generalised, n-dimensional interpolation

$$\begin{split} &K^{n,m}(r)\\ &= c^{n,m} \frac{(2\pi)^{-\frac{n}{2}}}{2(1-m)} r^{\frac{2-n}{2}} \int_0^\infty \mathbf{J}_{\frac{n-2}{2}}(kr) k^{\frac{n-2}{2}} \frac{d}{dk} \left(\frac{1}{(1+k^2)^{m-1}}\right) dk\\ &= c^{n,m} \frac{(2\pi)^{-\frac{n}{2}}}{2(m-1)} r^{\frac{4-n}{2}} \int_0^\infty \mathbf{J}_{\frac{n-4}{2}}(kr) k^{\frac{n-4}{2}} \frac{k}{(1+k^2)^{m-1}} dk\\ &= \frac{1}{4\pi(m-1)} \frac{c^{n,m}}{c^{n-2,m-1}} K^{n-2,m-1}(r) \end{split}$$

n is the dimension

m is the highest derivative

where

 $K^{n,m}$ is the Kernel

 $J_{\nu}(r)$ is the Bessel function of first kind or order ν



Problem

Solution in DIVA

1 Synopticity error 2 Representativeness error

Regularity constrain in cost function

3 Many observations

Numerical cost (almost) independent on the number of data points

4 Interpolate at many locations

Finite-element solver

5 Anisotropy

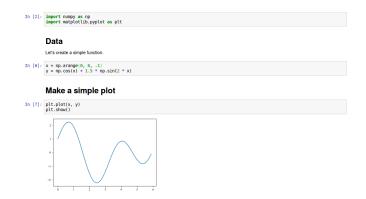
Finite-element solver

6 Currents

Advection included in the cost function

Notebooks: user-interface

- Documentation, including equations and export to pdf
- 2 Code fragments for different steps of the interpolation
- 3 Figures illustrating the data or intermediate results





NATURE | TOOLBOX





Interactive notebooks: Sharing the code

The free IPython notebook makes data analysis easier to record, understand and reproduce.

Helen Shen

05 November 2014

http://www.nature.com/news/ interactive-notebooks-sharing-the-code-1.16261

Notebooks: workflow description

Provide the jupyter-notebooks along with the data product (interpolation)

Make easier the reproducibility and peer-review



Why do we need

- Virtual
- Research
- **E**nvironments?

Computational resources

Storage and inversion of huge matrices

Typical case:

Horizontal grid: 500×500

Vertical levels: 50 depth levels

Time periods: 20



People connect, access the data, and work!



Installed/deployed once, used many times



Installing is sometimes much harder than running the code...





Management of multiple instances of the single-user Jupyter notebook server



• https://github.com/jupyterhub/jupyterhub

Demo available at https://hub-test.oceanbrowser.net/
(deployed at CINECA via Docker)



Authentication

Inputs: CDI data and user data

Results of the interpolation

Outputs: data products, climatologies, gridded fields





✓ Spatial interpolation is a frequent but not trivial operation in ocean sciences



- ✓ Spatial interpolation is a frequent but not trivial operation in ocean sciences
- ✓ **Specific** tools (DIVA, DIVAnd) have been designed for data interpolation



- ✓ Spatial interpolation is a frequent but not trivial operation in ocean sciences
- ✓ **Specific** tools (DIVA, DIVAnd) have been designed for data interpolation
- ✓ With a VRE, more users can access more easily SeaDataCloud resources (metadata, data & tools)



Tools

Leaflet DIVA

DIVAnd

Map layers

EMODnet Bathymetry Earth At Night 2012

MedSea observations

Temperature and salinity observation collection V1.1

The temperature at the "?"

