# Evaluation and intercomparison of GHRSST formatted products at a global scale

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Image: Second secon

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**GHRSST** sciences talks



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## **Barcelona Expert Center on Remote Sensing**





Latest Posts

- Expert Support Laboratory for L0 and L2 of ESA Soil Moisture and Ocean Salinity (SMOS) mission
- We are currently producing and distributing:
  - Sea Surface Salinity (SSS)
    - L3 and L4 SSS products at a global scale (from L0 own production chain)
    - L3 and L4 SSS products in enclosed seas (Mediterranean, Black Sea, Baltic)
    - L3 products in cold waters (Arctic and Antarctic (on-going))
  - Soil Moisture (SM)
    - L3 products at a global scale (from ESA L2)
    - L4 SM products in the Mediterranean region

## **Motivation**



- SST plays a key role in the generation of SMOS SSS products:
  - L2/L3: as an ancillary data for the inversion of radiative models
  - L4: as a template to increase temporal and spatial resolution using multifractal fusion techniques.
- Currently, we are using SST provided by ECMWF for the SSS retrieval (L2/L3), and OSTIA SST as template for the L4 product.



High sensitivity of SSS to SST at the low SSS & low SST regions (critical for Baltic and polar regions)

## Objective



- Assessment on the best SST product, based on the following indicators:
  - Comparison with in situ data: ARGO (fiducial reference measurements)
  - Spectral analysis to assess the effective spatial resolution (feature fidelity)
  - Singularity analysis to assess the dynamical properties
  - Correlated Triple Collocation to estimate the uncertainties of the error for each product [González-Gambau et al. 2020].

#### Datasets



Dataset	Spatial resolution	Source data	Sensor	
AMSR2-REMSS v8.2	0.25°x 0.25°	MW	AMSR2	
CMC v3.0	0,1°x 0,1°	MW+IR+insitu	VIIRS,AVHRR_GAC, AMSR2	
OSTIA v2.0	0.05° x 0.05°	MW+IR	AVHRR, VIIRS, AMSR2, GOES_IMAGER, SEVIRI, SSMIS, SSM/I	
CCI v2.1	0.05° x 0.05°	IR	ATSR, AATSR, AVHRR_GAC	
MUR v4.1	0,01°x 0,01°	MW+IR+insitu	AMSR-E, AVHRR, MODIS, SSM/I, VIIRS, in-situ	
Period	daily maps for 2016 5			5 / ??

#### Comparison with *in situ* data







Collocation criteria:

in situ centered (for every profile we select the closest SST pixel Δ*T*<sub>s</sub> [K] observation)

3 m < depth < 4 m

### Comparison with *in situ* data





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Dataset	Global				
	mode	mean	median	std	N
AMSRE	0.05	0.03	0.02	0.46	47645
СМС	-0.03	-0.04	-0.05	0.35	50928
OSTIA	-0.12	-0.09	-0.08	0.43	50927
CCI	-0.05	-0.07	-0.05	0.44	50927
MUR	-0.11	-0.05	-0.06	0.44	50771

## Spectral analysis

- Reynolds and Chelton (2010) showed that the grid spacing of an SST analysis is not an indicator of its feature resolution (F2T2 GHRSST Meeting 2018)



• Just few L4 SST intercomparison works have explored spectral analysis [Reynolds et al. 2013 and Yang et al. 2021]



#### **Spectral analysis**





- CCI & MUR have the higher effective spatial resolution
- In most regions, CMC and OSTIA do not increase the effective spatial resolution of microwave observations. In SOW region MW AMSR2 drops off at larger λ.

- We propose a new metric, the singularity spectrum, to assess the structural and dynamical quality of SST products based on multifractal theory of turbulence. [Turiel et al. 2007, Isern-Fontanet et al. 2007, ]
- Scale surrogate gradients, in the limit  $\ell/\ell_0 \rightarrow 0$  behave as

$$\overline{|\nabla T|}_{\ell}(\vec{x}) \equiv \int_{\mathbb{R}^d} \ell^{-d} G(\ell^{-1}\vec{x}) |\nabla T| (\vec{x} + \vec{x}') d\vec{x}' \sim \left(\frac{\ell}{\ell_0}\right)^{h(\vec{x})}$$

•  $h(\vec{x})$  is the singularity exponent. If  $h(\vec{x}) \in (n, n+1)$  with n being a positive integer,  $\overline{|\nabla T|}_{\ell}(\vec{x})$  is derivable n times but not n+1



• The statistical/geometric properties of the singularity exponents are given by the singularity spectrum

$$D(h') \equiv d_F(\{\vec{x} | h(\vec{x} = h')\})$$

where  $d_{F}(A)$  is the fractal dimension of a set A

- It can be computed from the probability density function of the singularity exponents [Pont et al. 2013]
- D(h) is a convex function and it is scale invariant.
- It is a representation of the turbulent cascade
- It is the step between the canonical (statistical, structure functions) to the microcanonical formalism (geometrical, singularity exponents)























• Singularity spectrum are computed daily and at the native spatial resolution



Higher spatial resolution does not mean better dynamical representation



• There is a range of intermediate gradients that are being excessively smoothen. The turbulent cascade is poorly represented.



We compute the ratio of days in 2016 that present a singularity exponent within the band (0.3<h<0.6)



• We compute the ratio of days in 2016 that present a singularity exponent within the band (0.3<h<0.6) in order to found out the regions where the SST gradients are being smoothen





• To assess if the smoothing of intermediate gradients is caused by the L4 fusion/interpolation scheme, we interpolated OSTIA SST to a 0.25° grid, and repeat the same analysis.





We compute the ratio of days in 2016 that have 0.3<h<0.6



Validation tool of consistency of the dynamical content of SST

- The triple collocation (TC) technique has been widely used to assess the quality of many geophysical variables acquired with different instruments and at different scales.
- We focus in the error characterization of remote sensing maps where having three collocated, completely independent datasets is unlikely.
- We have developed a **new formulation, the CTC (Correlated Triple Collocation)** for the case of **three datasets** that **resolve similar spatial scales**, with **two** of them being **error-correlated datasets** [González-Gambau et al. 2020].
- The derived error maps from CTC analysis allow us to characterize the quality of the different datasets at each location all over the globe.

## **Correlated triple colocation**

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- Rationale and methodology
  - 1. We have three series of spatially and temporally collocated measurements  $x_i$  of the same variable  $\theta$ :

$$x_i = a_i\theta + b_i + \delta_i \quad i = 1, 2, 3$$

- 2. We want to estimate the standard deviation of error  $\delta_i$ , that is,  $\sigma_{\delta_i}$
- 3. Assumptions:
  - a. Errors 1 and 2 are correlated with covariance  $\phi_{12}$  but uncorrelated with error 3.
  - b. The three datasets resolve the same spatial scales, so there is no representative errors
  - c. Measurements are assumed to have intercalibration factor equal to 1
- 4. CTC is less sensitive to statistical fluctuations, allowing a reliable estimation of errors with scarce sampling sizes.

## **Correlated Triple Colocation: selecting triplets**



Dataset	Sensor		
AMSR2-REMSS v8.2	AMSR2		
CMC v3.0	VIIRS,AVHRR_GAC, AMSR2		
OSTIA v2.0	AVHRR, VIIRS, AMSR2, GOES_IMAGER, SEVIRI, SSMIS, <u>SSM/I</u>		
CCI v2.1	ATSR, AATSR, AVHRR_GAC		
MUR v4.1	AMSR-E, AVHRR, MODIS, SSM/I, VIIRS, in-situ		
Period	daily maps for 2016		

- SSM/I sensor is only present in MUR and OSTIA. We'll consider these two datasets as the one uncorrelated and perform the CTC for the triplets:
  - CCI\_CMC\_MUR
  - CCI\_AMSR2\_MUR
  - AMSR2\_CMC\_MUR
  - CCI\_CMC\_OSTIA
  - CCI\_AMSR2\_OSTIA
  - AMSR2\_CMC\_OSTIA
- The errors of the shared sensors will be consider as part of the SST variability. We obtain 3 different estimations (one per triplet) of the standard deviation of SSM/I error for each dataset (OSTIA and MUR).
- All datasets have been reinterpolated to the coarser grid (0.25°)

#### **Correlated Triple Colocation MUR**



BFC

#### CTC MUR: SSM/I





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## CTC OSTIA: SSM/I





## Estimation of SSM/I error uncertainty



- The standard deviation of the SSM/I error is about 0.3-0.4 K, in accordance with in situ assessment
- CTC is a powerful tool for estimating the standard deviation of the errors, and providing its spatial distribution (land sea contamination, RFI, atmospheric corrections....)





- Choosing the best L4 SST analysis is not easy. GHRSST provides the users with some guidelines [Beggs 2020, GHRSST Office 2023]
- The comparison wrt in situ data does not give the "global picture".
- We proposed a new metric in order to assess SST products from a dynamical and structural perspective (the singularity spectrum) [González-Haro et al., in prep.]
- Correlated Triple Colocation is a powerful tool for estimating the standard deviation of the errors, and providing the spatial distribution of the errors (land sea contamination, RFI, atmospheric corrections....)





- Our final election: we'll use the MUR L4 SST product at its effective spatial resolution (0.1°)
  - we are currently retrieving the SSS using MUR SST at 0.1 for the regional product in the Southern Ocean
  - assessment of the impact of new SST on the retrieval (stay tuned for next GHRSST meeting)

#### **Further work**



- Further efforts should be addressed to:
  - characterize the effective spatial resolution for smaller regions (10°x10°) at a global scale (short/mid term)
  - use CTC at L2 to characterize the standard deviation of the errors of the different sensors (depending on available resources)
  - understand which methodological choice of producers have higher impact on not preserving the expected singularity spectrum, i.e not respecting the turbulence cascade (Feature Fidelity Task Team F2T2??)
  - improve interpolation/fusion schemes to respect the turbulence cascade

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# **Extra Slides**

#### SST definitions





#### **Spectral analysis**





#### **Spectral Analysis**





#### **Spectral Analysis**

























## **CTC:** Rationale and methodology



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- We have looked for a linear transformation of  $x_1, x_2$  into variables  $x_1^{+}, x_2^{+}$  with errors  $\delta x_1^{+}, \delta x_2^{-}$ , which are uncorrelated.
- We define the auxiliary scaling parameters u,v as:

$$u = \frac{s_2 - s_{12}}{s_1 + s_2 - 2s_{12}}$$
;  $v = \frac{s_1 - s_{12}}{s_1 + s_2 - 2s_{12}}$ 

s : Variance of the measurements x

- s\_: Covariance between the measurements i, j
- The order-2 moments of the uncorrelated-error variables are defined:
- Finally, the error variances and error covariance of the original  $x_1, x_2$  are given by:  $\int \sigma_1^2 = v^2 s_1' + s_2' - s_{22}'$

$$\begin{cases} \sigma_{\delta_1}^2 &= v^2 s_1' + s_2' - s_{23}' \\ \sigma_{\delta_2}^2 &= u^2 s_1' + s_2' - s_{23}' \\ \sigma_{\delta_3}^2 &= s_3 - s_{23}' \\ \phi_{12} &= -uvs_1' + s_2' - s_{23}' \end{cases}$$

V. González-Gambau et al., 2020: Triple Collocation Analysis for Two Error-Correlated Datasets: Application to L-Band Brightness Temperatures over Land. Remote Sens. 2020, 12, 3381. DOI: 10.3390/rs12203381