

living planet symposium | BONN 23–27 May 2022

TAKING THE PULSE
OF OUR PLANET FROM SPACE



Improving atmospheric correction for OLCI water reflectance products: the SACSO project

François Steinmetz (1), Constant Mazeran (2), Didier Ramon (1), Robert Frouin (3), Julien Chimot (4), Ewa Kwiatkowska (4), David Dessailly (4)
(1) HYGEOS, (2) Solvo, (3) SIO/UCSD, (4) EUMETSAT

2022-05-24, A8.11.2 Colour and Light in the Ocean from Earth Observation - 2

EUMETSAT SACSO study: 2019-2021



<https://www.eumetsat.int/SACSO>

Objective: Develop a **spectral matching atmospheric correction** for Sentinel 3 Ocean Colour (=SACSO), alternative to EUMETSAT operational processor (standard AC in the NIR), applicable to other sensors

→ Build on **Polymer** assets:

- Spectral matching approach allows for a **robust and generic atmospheric correction**: aerosols (in particular absorbing), sun glint, adjacency effects, thin clouds
- Analytical atmospheric reflectance model ; flexibility with respect to the spectral bands
- Account for per-pixel wavelength during the whole atmospheric correction (OLCI and MERIS “smile effect”)
- Sensor-independent Rayleigh look-up table (tabulated in Rayleigh optical thickness)
- Spectral (band shifting) and directional normalization

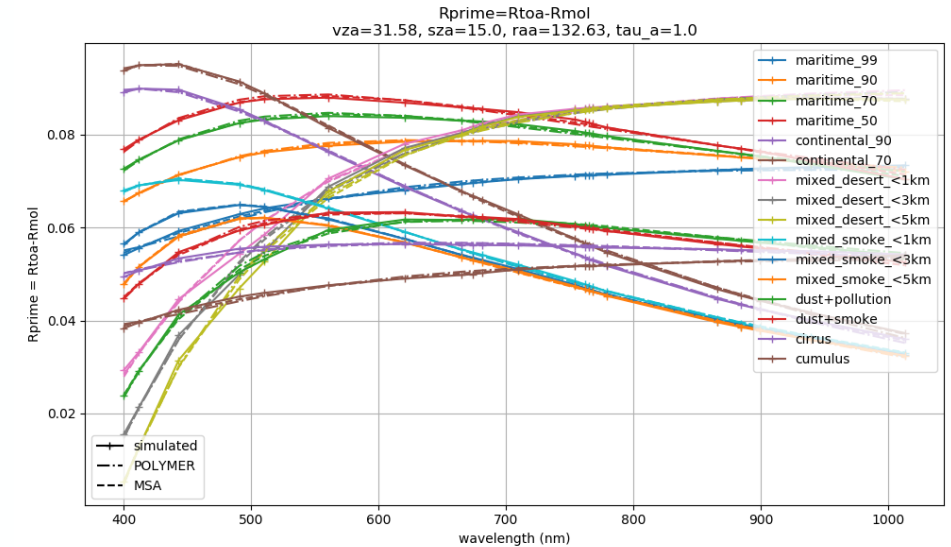
- Polymer atmospheric (+sun glint) reflectance model:

$$\rho_{ag}(\lambda) = T_0(\lambda)c_0 + c_1\lambda^{-1} + c_2\rho_{mol}(\lambda)$$

- The MSA model (« **Multiple scattering approximation** ») is a physically based model extended from the single scattering approximation.

$$\rho_{ag}(\lambda) = \rho_{ago} \left(\frac{\lambda}{\lambda_0} \right)^{-\alpha} \left(\frac{1 + k \left(\frac{\lambda}{\lambda_0} \right)^{-\alpha}}{1 + k} \right)$$

- 3 parameters:
 - ρ_{ago}
 - α : Ångström coefficient
 - $k \in [-1, 0]$, adjusting for the aerosol absorption and multiple scattering, thus decay towards blue bands



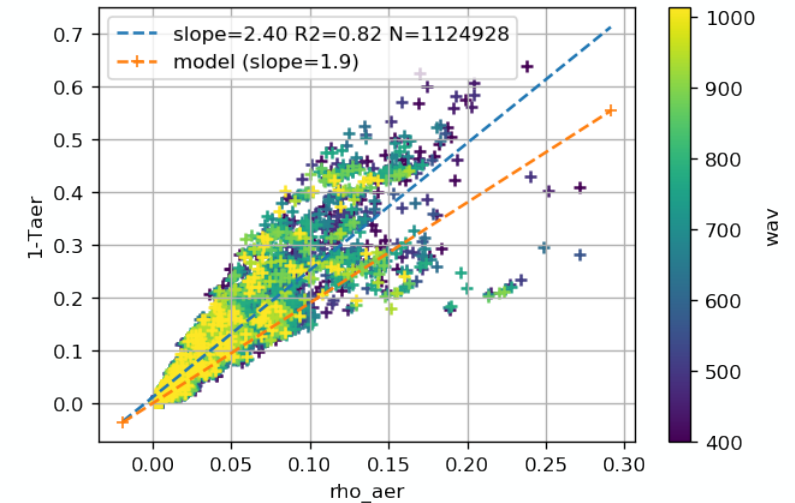
Fit of pure aerosol radiative transfer simulations with $\tau_{aer} = 1.0$ with Polymer and MSA models

- Decomposition of the Rayleigh-corrected signal

$$\rho_{rc}(\lambda) = \rho_a(\lambda) + t(\lambda)\rho_w(\lambda)$$

- In Polymer, assumption that $t(\lambda) = t_R(\lambda)$; thus $t_a(\lambda) = 1$
- In SACSO, based on radiative transfer simulations, the aerosol transmission is estimated as follows:

$$t_a(\lambda) = 1 - 1.9 \cdot \rho_a(\lambda)$$



*Valid for non-absorbing aerosols, outside of sun glint.
For absorbing aerosols, $t_a(\lambda)$ is still underestimated, but this is not seen as the dominant issue with absorbing aerosols.*

- Parameters to the problem x :
 - 2 for the water model: chlorophyll concentration and backscattering ratio, both in log scale :

$$x_w = (\log_{10}(chl), \log_{10}(f_b))$$
 - 3 for the MSA model $x_a = (\rho_{a0}, \alpha, k)$

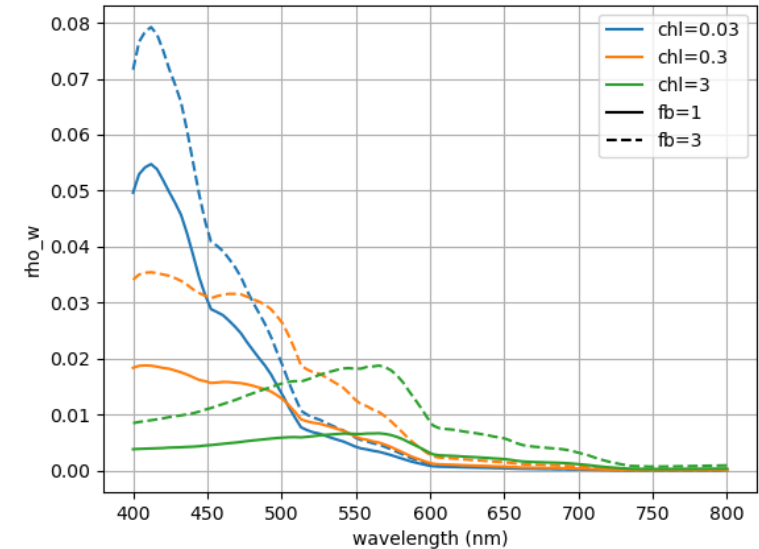
The parameters k and ρ_{a0} can be solved exactly in the minimization problems at each iteration:

$$\frac{\partial \chi^2}{\partial k} = 0 \Leftrightarrow k = f(\rho_{RC}, \rho_w^{mod}, \alpha)$$

$$\frac{\partial \chi^2}{\partial \rho_{a0}} = 0 \Leftrightarrow \rho_{a0} = g(\rho_{RC}, \rho_w^{mod}, \alpha, k)$$

→ In practice, α is the only atmospheric degree of freedom ; ρ_{a0} and k are intermediate variables.

→ The iterative optimization scheme reduces from 5 to 3 parameters



Example for water reflectance spectra generated with the model (Park and Ruddick, 2005)

- Standard mathematical formalism for non-linear least square minimization (NLLSQ):

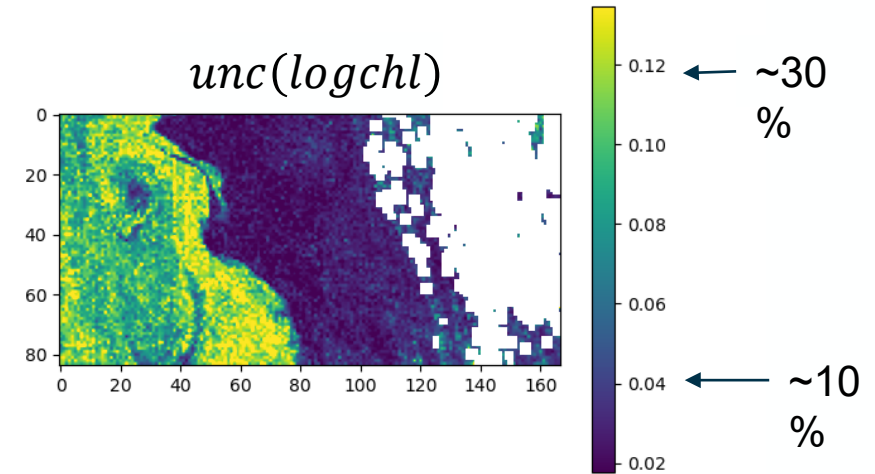
$$\chi^2(\mathbf{x}) = \frac{1}{n_{obs} - n_x} (\boldsymbol{\rho}_{RC}^{mod}(\mathbf{x}) - \boldsymbol{\rho}_{RC})^T \mathbf{W} (\boldsymbol{\rho}_{RC}^{mod}(\mathbf{x}) - \boldsymbol{\rho}_{RC}), \quad \text{where } \mathbf{x} \text{ is the vector of parameters (water+aerosols)}$$

- Ideally, $\mathbf{W} = (\mathbf{C}_{\rho_{RC}^{mod}} + \mathbf{C}_{\rho_{RC}})^{-1}$. If no covariance ($\mathbf{W} = \text{diag}$):

$$\chi^2(\mathbf{x}) = \sum_{i=1}^{n_{obs}} \frac{(\rho_{RC}^{mod}(\lambda_i, \mathbf{x}) - \rho_{RC}(\lambda_i))^2}{\sigma_{mod}^2(\lambda_i) + \sigma_{obs}^2(\lambda_i)}$$

- Variance-covariance of retrieved parameters (\mathbf{C}_x) given by theory of uncertainty propagation:

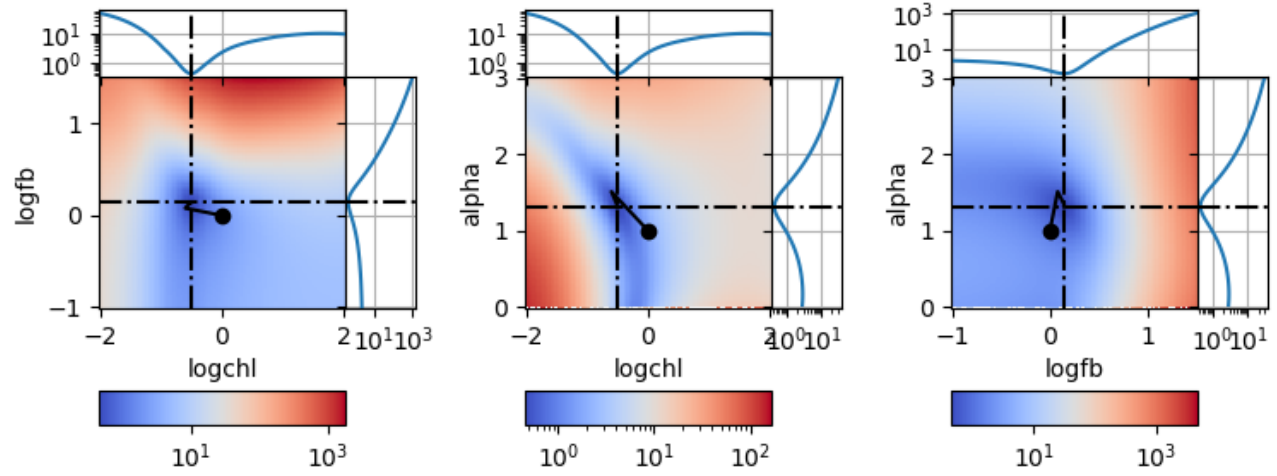
- $\mathbf{C}_x = (\mathbf{J}^T \mathbf{W} \mathbf{J})^{-1}$, at convergence, with $\mathbf{J} = \frac{\partial \boldsymbol{\rho}_{RC}^{mod}}{\partial \mathbf{x}}$
- Uncertainties can then be propagated from \mathbf{x} to $\rho_w(\lambda)$



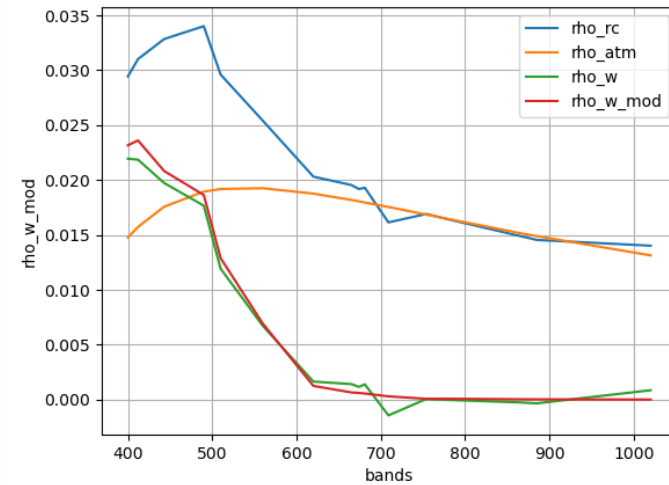
Example of a SACSO uncertainty map

- Key point is the assessment of input uncertainties $\mathbf{C}_{\rho_{RC}^{mod}}$ and $\mathbf{C}_{\rho_{RC}}$

- Iterative minimization of χ^2 in 3 dimensions simultaneously
 - 2 water parameters ($\log_{10} chl, \log_{10} f_b$)
 - 1 aerosol parameters (α)
- Interface with the GSL library: access to many minimization methods (unlike Polymer)
- Final method selected: Levenberg-Marquardt nonlinear least square fitting
- Criteria for selection:
 - **Stability!** Obtaining a stable numerical method was one of the main challenges in this project. Stability of the inversion was already one of the main issues with Polymer.
 - Weak sensitivity to the initialization point
 - Avoid sharp pixel-to-pixel transitions in the images
 - Speed of convergence (to reach an acceptable processing time)

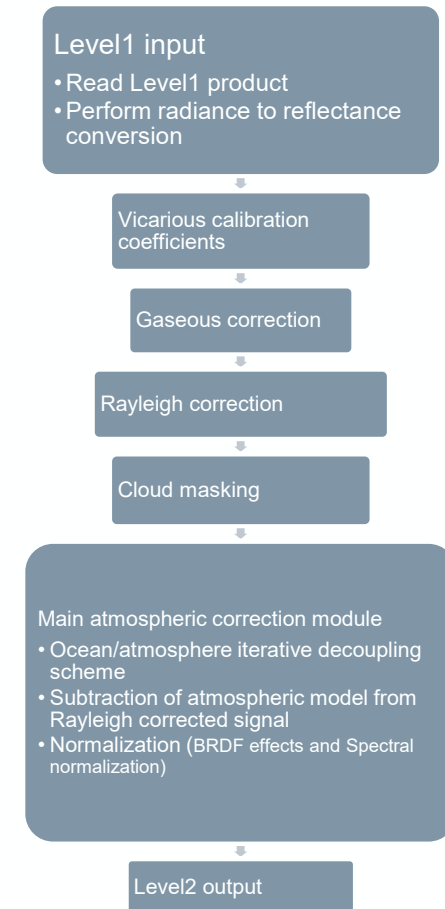


Visualization of the optimization in the 3-dimensional cost function for a sample OLCI spectrum rho_rc



Visualization of the result of minimization of a sample OLCI spectrum rho_rc

- The SACSO processor originates from Polymer
- Implemented in Python, with modules written in cython language for efficiency (converted to C language and compiled)
- Updated for improved modularity
 - Rely on the xarray python module
 - Allowed to directly process EUMETSAT matchup database
 - Processing of EUMETSAT minifiles (NetCDF format)



SACSO implementation scheme

Which products are being compared:

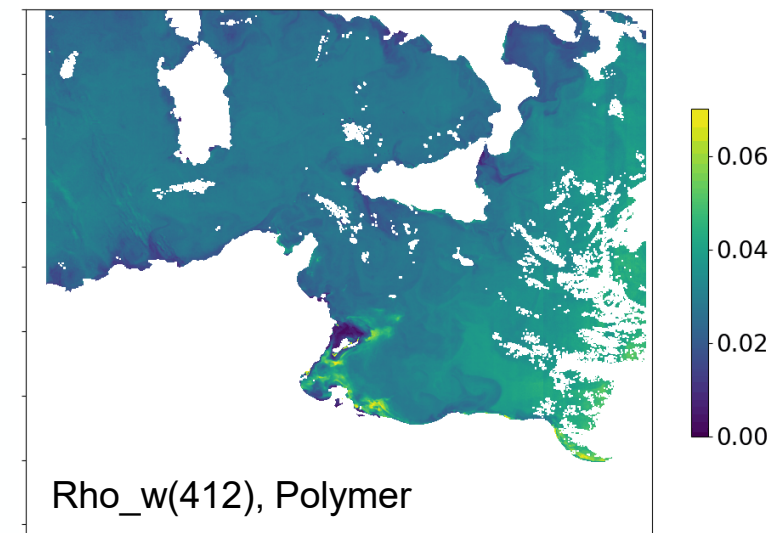
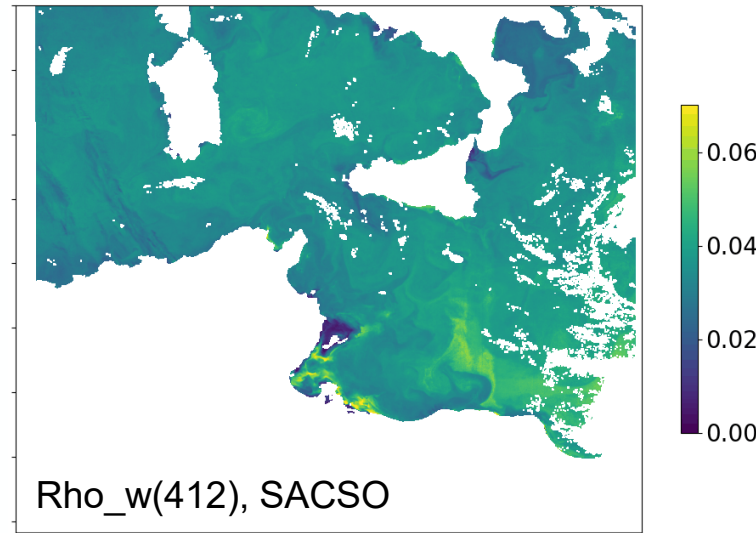
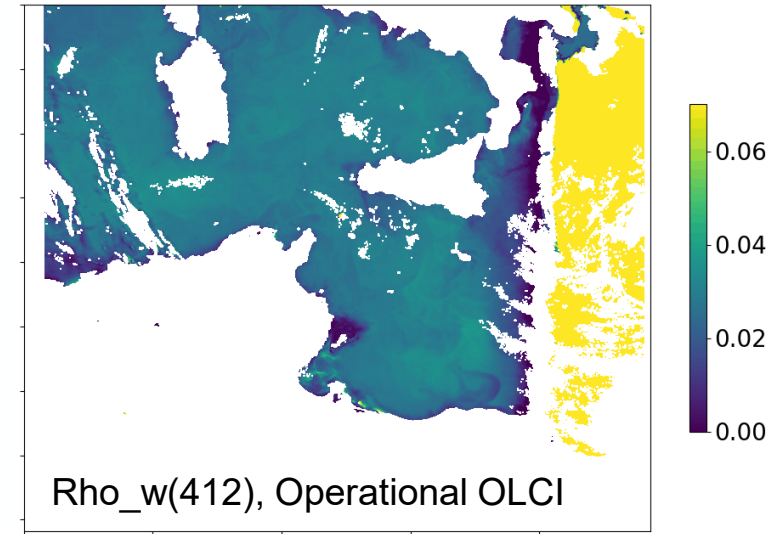
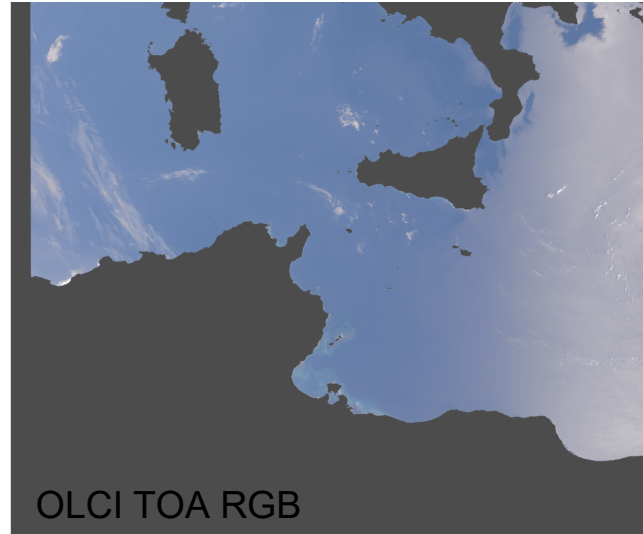
1. SACSO
2. Operational OLCI Level2: IPF collection 3 - OL_L2M.003.01 (PB v2.73)
3. Polymer v4.13

Methods:

1. Visual inspection of 37 selected scenes with a wide range of situations: different types of waters from oligotrophic to extremely turbid, desert dust, polluted dust, volcanic eruptions, ashes, etc (described in the PVP)
2. Validation using in-situ data
3. Global 4-days composites
4. Timeseries over selected points (in particular, South Pacific Gyre)

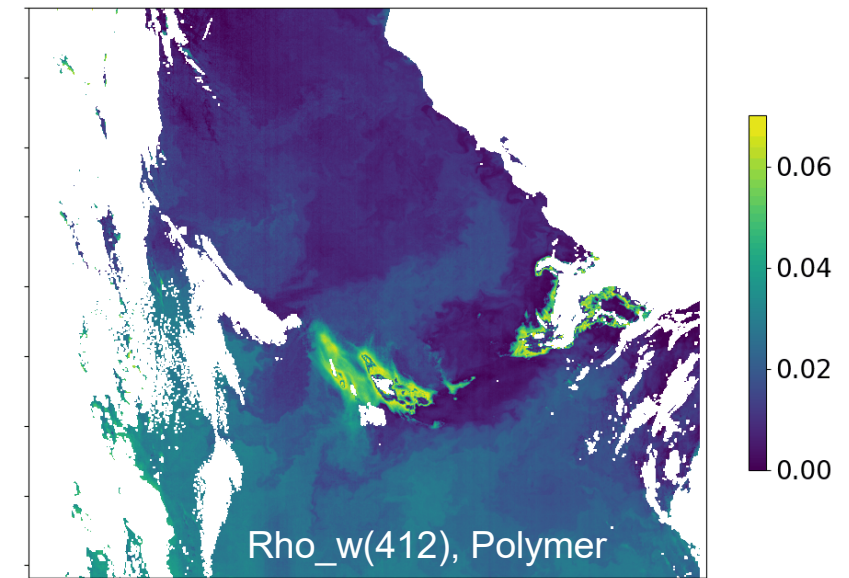
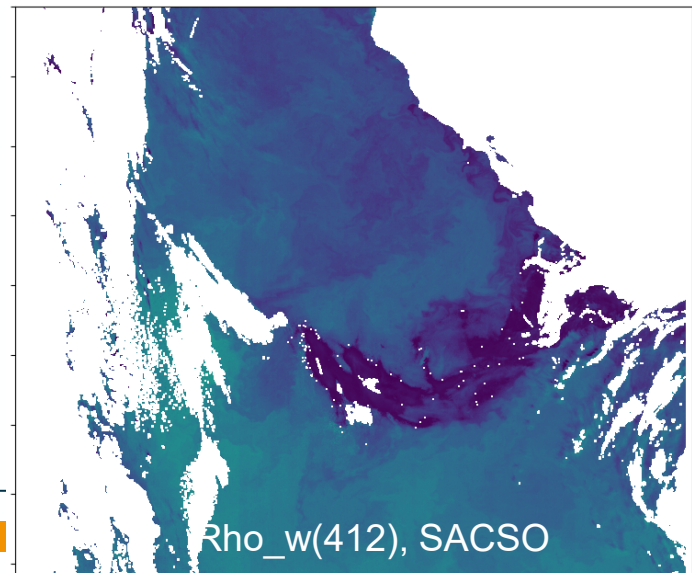
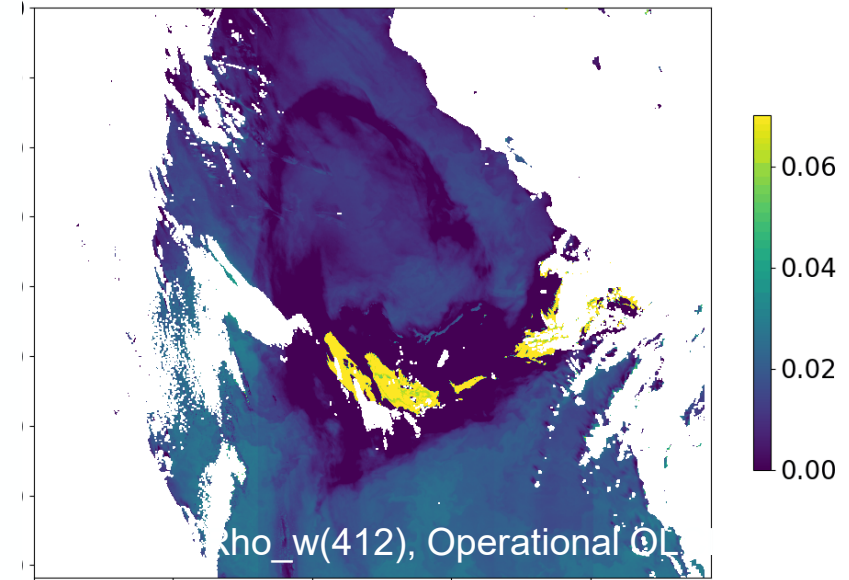
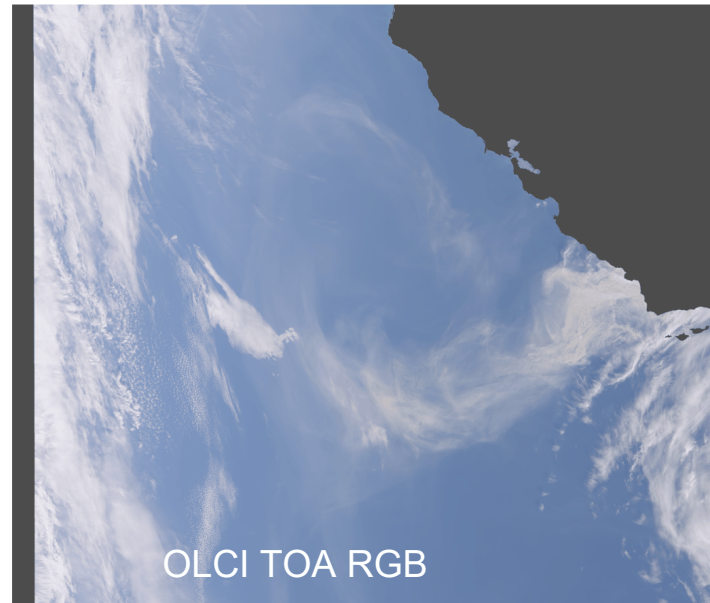
Example 1: Mediterranean Sea

- S3A_OL_1_EFR____20170529T092331
- Quality flags are shown on/off for each processor
- Consistent results between Polymer and SACSO
- Glint correction works well with SACSO



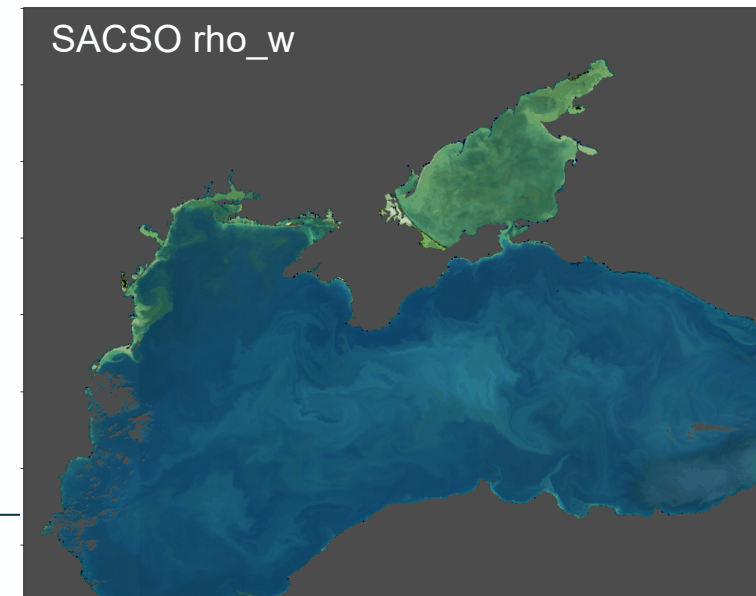
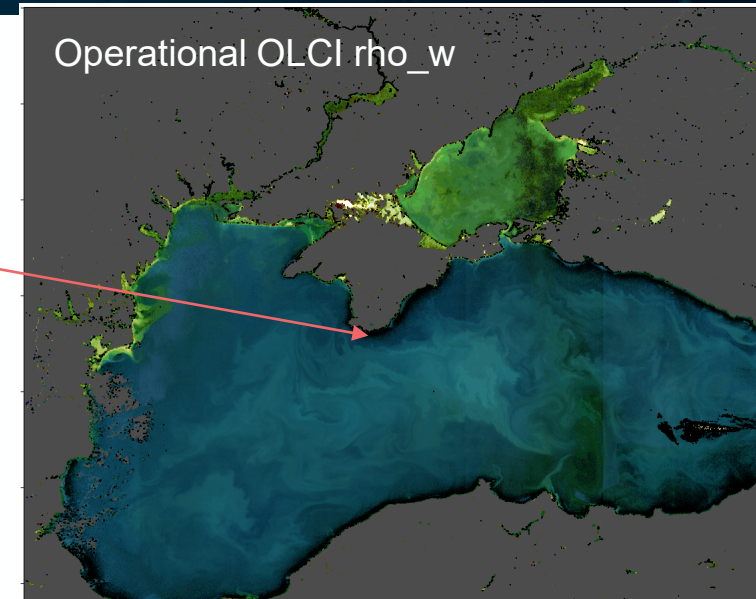
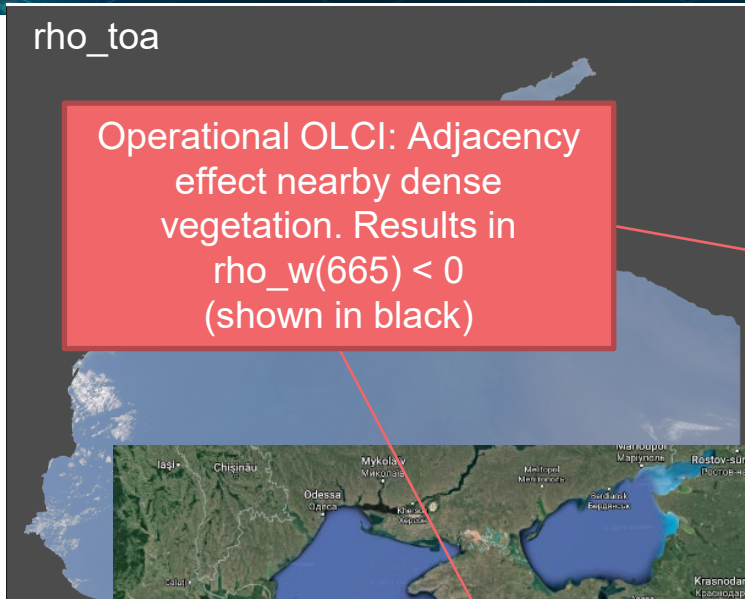
Example 2: California fires

- S3A_OL_1_EFR____20171210T183320
- SACSO (like Polymer) is much less sensitive to the absorbing aerosols than the Operational OLCI.
- SACSO flags are more effective than Polymer to mask the ash plume.



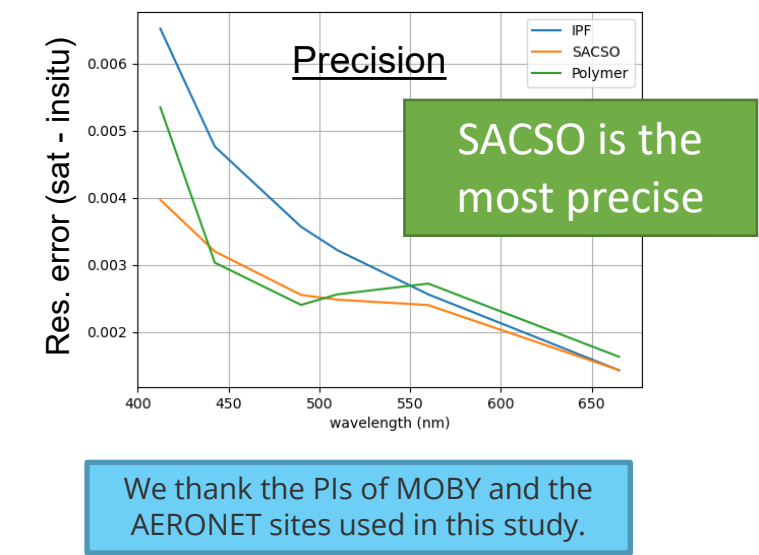
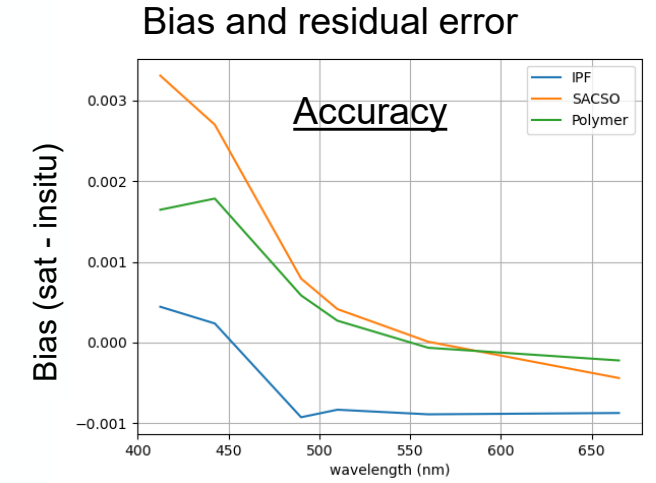
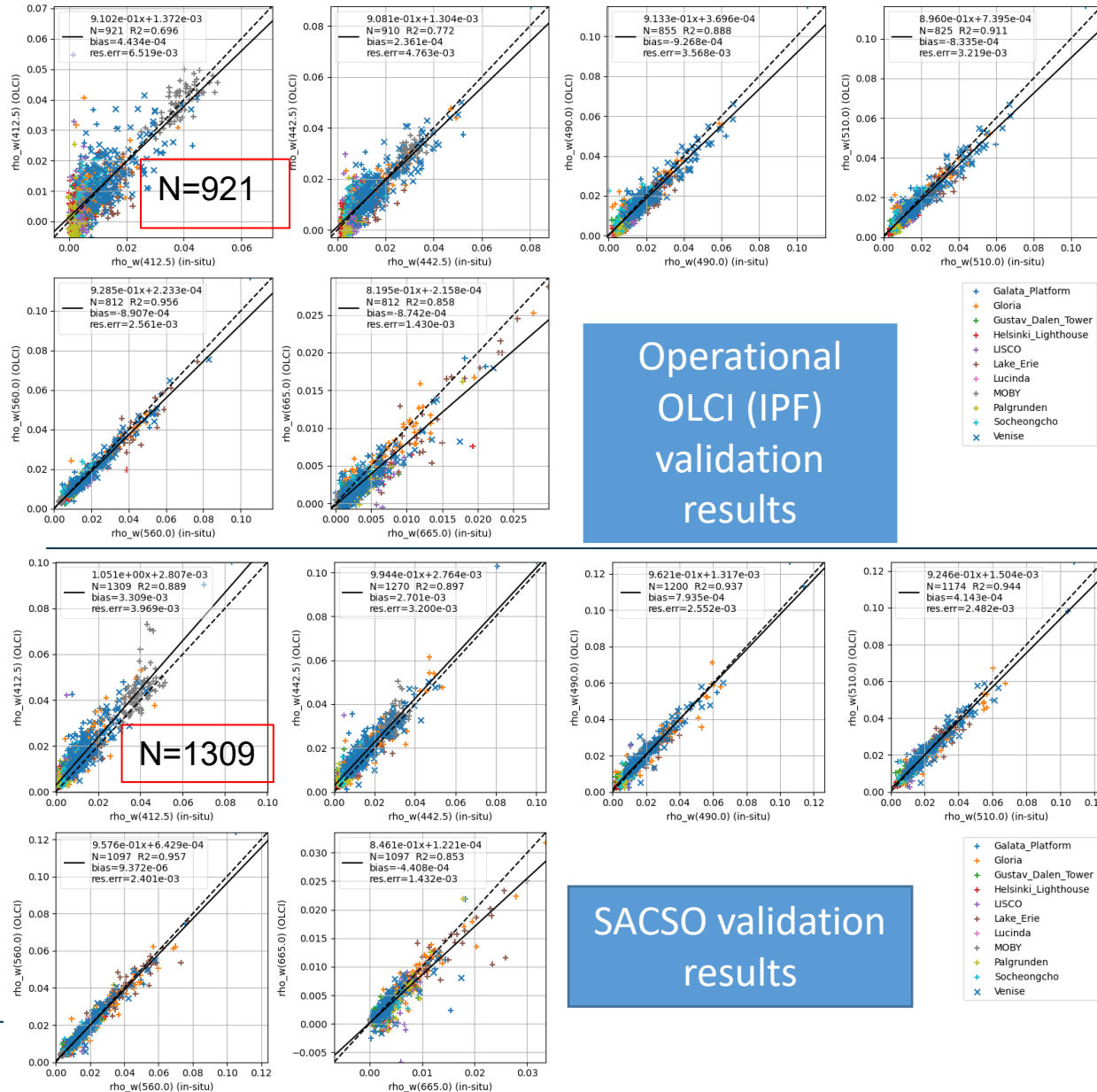
Example 3: Black Sea

- S3A_OL_1_EFR___20170913T080543
- RGB visualization, with R=665, G=560 and B=443
- Same colour scales for rho_w and rho_a (max=0.2 ; log scale to enhance dark areas)
- Allows to visualize the decoupling between the ocean and atmosphere components in “natural” colours



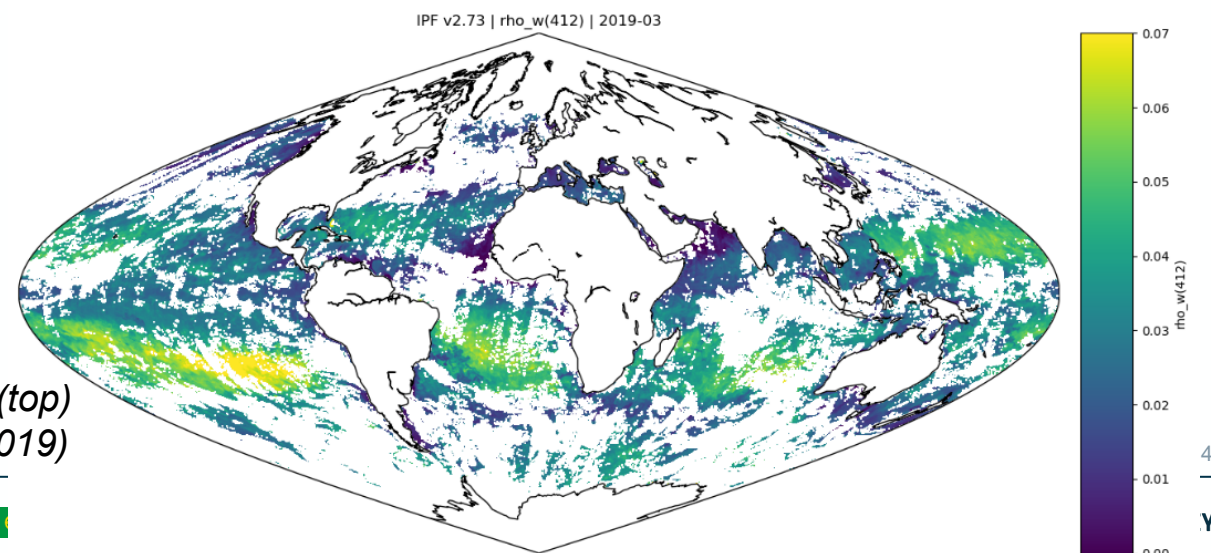
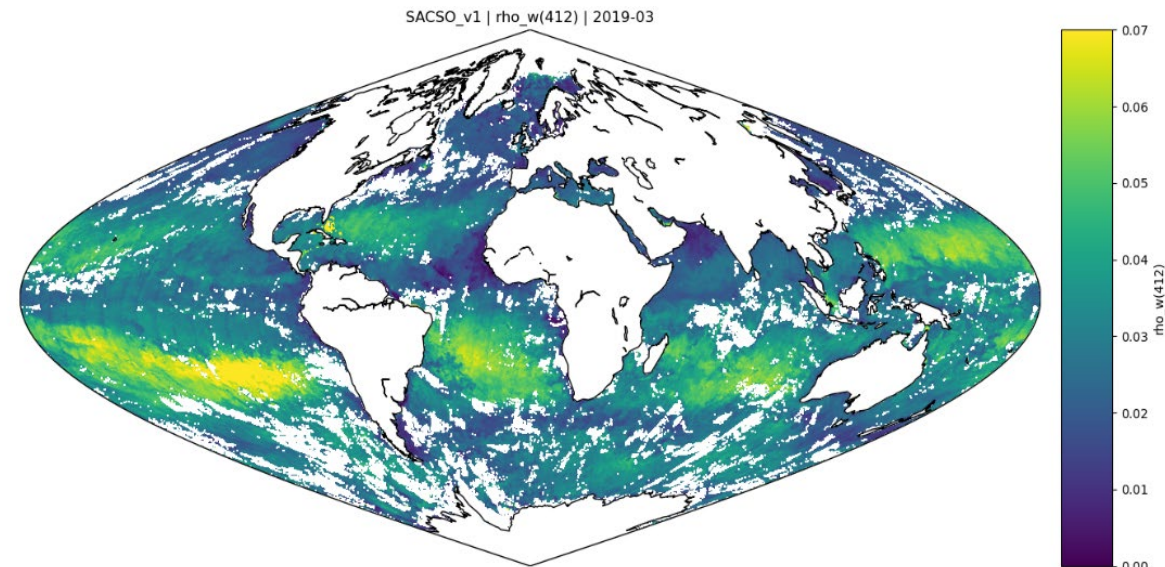
Validation results

- Validation using Aeronet-OC and MOBY data
- EUMETSAT matchup database was processed directly with SACSO
- Results include early dedicated system vicarious coefficients (SVC)
- Operational OLCI is the most accurate in the blue, but with a lower R²
- SACSO, like Polymer, retrieves significantly more valid (useful) pixels



Polymer: issues of stability and bias in the very clear waters of the South Pacific Gyre (SPG).

- Global 4-days composites and timeseries analysis have shown a better consistency between SACSO and Operational OLCI than Polymer and Operational OLCI in the SPG
- But: impact of system vicarious calibration (SVC) in the SPG to be further studied



*Global 4-day composite of SACSO (top)
and Operational OLCI (1-4 March 2019)*

- The spectral matching approach allows to improve the robustness of the aerosol correction, compared with the standard atmospheric correction
 - Retrieved marine reflectances are less sensitive to the aerosols, in particular absorbing ones
 - Robust to the sun glint (even the brightest part of the sun glint) and adjacency effects
 - Simple analytical aerosol reflectance modelling
 - The analytic atmospheric model and stable inversion scheme anomalous pixel to pixels transitions
 - Significantly increased spatial coverage
- Weakness: more dependency on the water reflectance model
 - SACSO: water reflectance model for the whole VIS+NIR range
 - Operational OLCI AC: water reflectance model only in the NIR (Bright Pixel Correction)
- SACSO includes an uncertainty propagation scheme

Many improvements implemented in SACSO compared to the baseline Polymer:

- A more constraining aerosol reflectance model, allowing for **more stable results over oligotrophic (clear) waters**. However an over-correction is still observed **over extremely turbid waters**.
- A more rigorous and generic definition of the cost function
- A **more stable numeric inversion scheme**
- SACSO takes into account the aerosol transmittance
- An uncertainty propagation scheme and associated flags

Drawback: the computational efficiency is currently about 3 times lower than Polymer because of the more complex inversion scheme ; to be optimized

- Some points to be further consolidated (uncertainties definition, flagging, processing time, water reflectance model...)
- To be tested on other sensors

Thanks for your attention.

