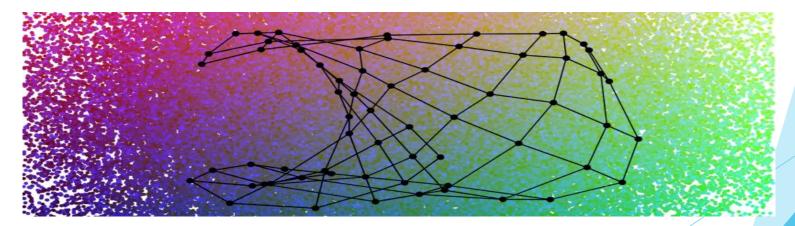


OCEAN CARBON FROM SPACE WORKSHOP 14-18 February 2022

Development of a Machine Learning approach for the estimation of the marine CO<sub>2</sub> Partial Pressure over the global coastal ocean in the frame of the CO<sub>2</sub>COAST project



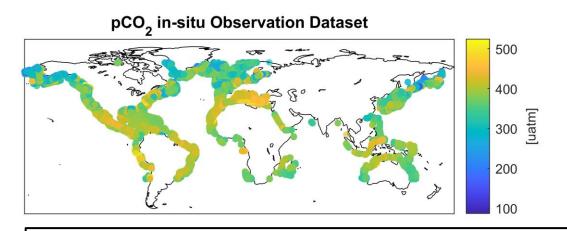
Roy El Hourany, Hubert Loisel, Daniel S.F. Jorge, Vincent Vantrepotte, Cédric Jamet, Sylvie Thiria, Ndeye Niang Keita, Julien Demaria, Marine Bretagnon, Antoine Mangin

While many improvements have been done for about 2 decades by the ocean color community in terms of algorithms development for the monitoring of coastal waters, the application of these algorithms is still limited to local or regional scales.

# CO<sub>2</sub>Coast (2021-2025)

- Estimate the surface-ocean CO<sub>2</sub> partial pressure, pCO<sub>2w</sub>, CO<sub>2</sub> flux, and associated uncertainties from satellite remote sensing over the global coastal waters at 1 and 4 km<sup>2</sup>
- Assessment of the seasonal, inter-annual, and last 25 years trend evolutions of pCO<sub>2w</sub> and CO<sub>2</sub> flux over global coastal waters.
- Analysis of the respective contribution of estuaries and coastal shelf waters, as well as the contribution of the different continental shelf types.

## Global algorithm development; Data



<u>Global pCO2 dataset (SOCAT: https://www.socat.info/):</u> ~500 000 obs with ~90 000 presenting non-missing satellite matchups

<u>Satellite Parameters</u> (daily 4km resolution): **Rrs (6 wavelengths), SST, SSS, acdom, Chla** (from 09/1997: SeaWiFS, MERIS, and present (up to 12/2023) OCR (MODIS, VIIRS, OLCI) sensors + ACRI-ST products)

Challenges: High dimension data,

The relationship between variables might be regionallydependent

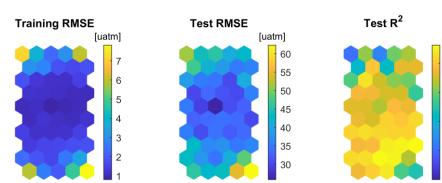
**Clusterwise Regression** 

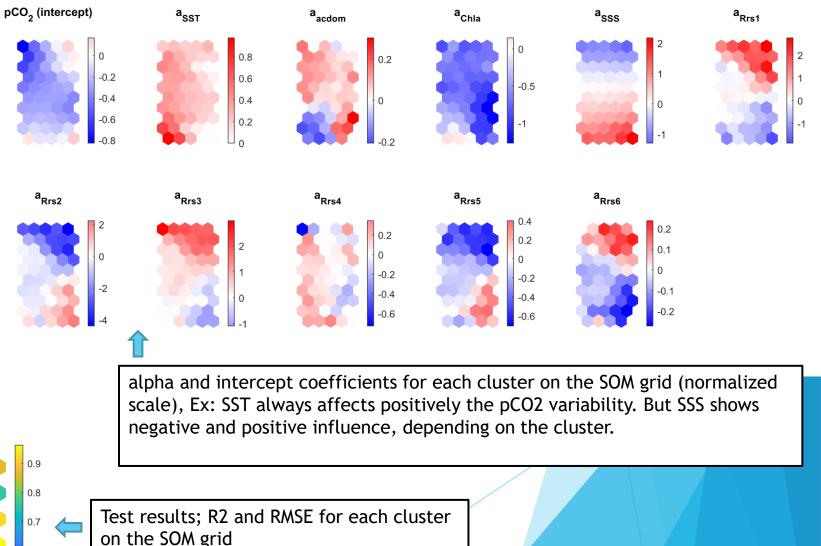
Clusterwise regression is used to classify observations into groups, and that the relationship between pCO2 and satellite parameters is different depending on the group. This process can identify and characterize regional relationships in a global dataset.

# Clusterwise regression using SOMClusterwise

- Classify individuals into G groups while having the best model within each of the groups.
- Classification and learning carried out simultaneously

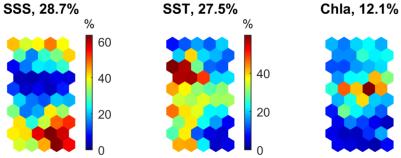
Each cluster of the SOM is characterized by an intercept and a slope for every variable [pCO2= intercept+ $\Sigma a var * var$ ]. The resultant slopes shows how each variable affects the variation of the pCO2 in each cluster

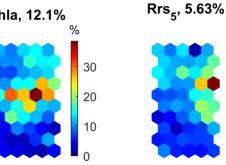


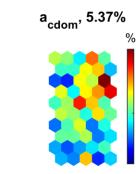


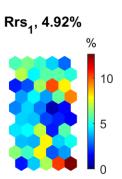
The importance of Satellite-derived variables used to accurately estimate pCO2 values vary from a cluster to another. This result upholds the regional aspect of the relationship between pCO2 and the oceanic parameters.

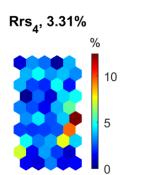
#### Importance of variables; Random Forest analysis applied in each SOMCluster







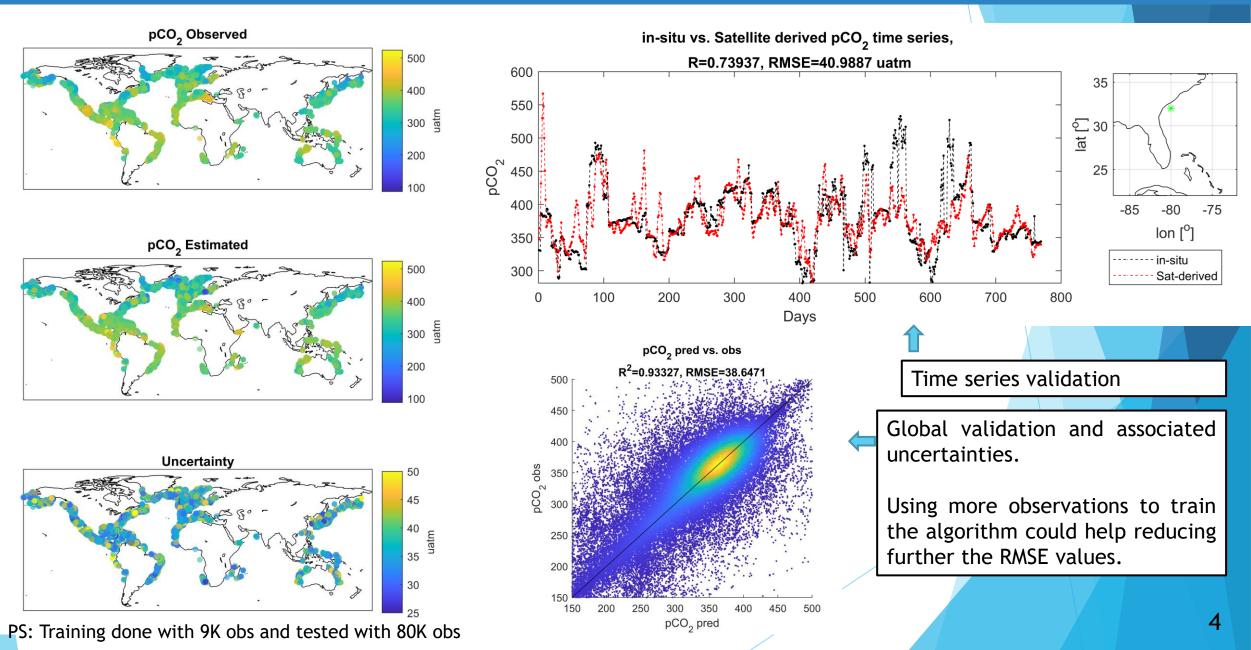




Variables' importance (%) shown on the SOM grid for each cluster.

The variables were ordered on this figure following an increasing overall mean importance shown in each plot's title

# Tests applied at the global and temporal scales highlight a significant accuracy while estimating pCO2 from satellite data based on a clusterwise regression approach.



### Next steps:

- Algorithm adjustments: Incorporating more datapoints in the training process, choice of optimal hyperparameters (number of clusters and the number of training iterations) to get the best significant models with the least error possible for each cluster,
- Development of a quality control mask and Geographic interpretation of the clusters,
- Production of a daily Satellite-derived global dataset

## future aims:

- Assessment of the seasonal, inter-annual, and last 25 years trend evolutions of pCO2w and CO2 flux over global coastal waters.
- Analysis of the respective contribution of estuaries and coastal shelf waters, as well as the contribution of the different continental shelf types.

Thank You