POS2IDON: A TOOL FOR MONITORING MARINE DEBRIS **BASED ON SENTINEL-2 SATELLITE IMAGERY AND** MACHINE LEARNING

Emanuel Castanho^a, André Valente^a, Andrea Giusti^a, João Pinelo^a, Pedro Silva^a

^a AIR Centre – Atlantic International Research Centre, Terceira Island, Azores, Portugal





SCAN THE QR code to access the source-code.

POS2IDON is an open-source data pipeline designed for long-term analyses and monitoring of suspected marine debris accumulations (>10 m) and other ocean features, such as vessels and floating Sargassum. Applications for a plastic-affected region of the Honduras Gulf are here demonstrated, such as post-disaster management and a longterm seasonal analysis to inform stakeholders on patterns and trends of plastic pollution.

Scope and context

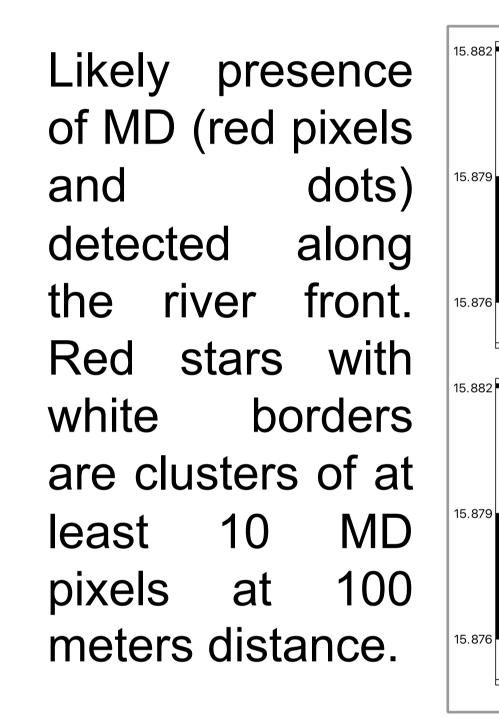
- > Previous studies showed that Copernicus Sentinel-2 can detect artificial plastic targets and large accumulations of marine debris.
- > However, automatic data pipelines are still rare in literature, which hinders scientific advances and monitoring efforts.
- > We developed an open-source data pipeline, POS2IDON, to foster the development of a tool for monitoring plastic in the ocean.
- \succ POS2IDON uses Sentinel-2 and Machine Learning (ML), and provides users several customizable options to detect suspected marine debris and perform different monitoring analyses.

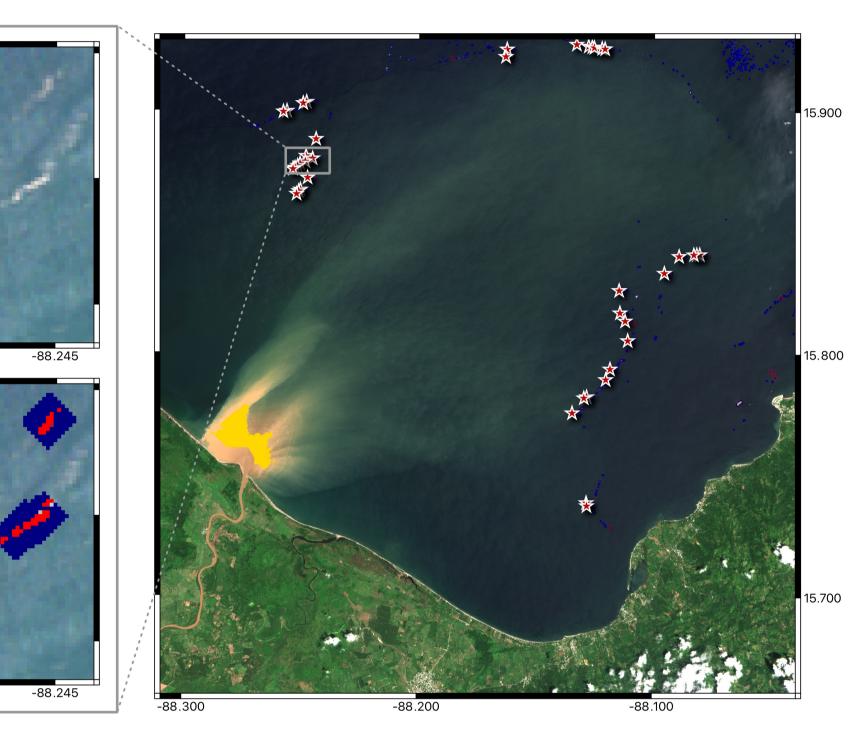
POS2IDON architecture

> POS2IDON downloads (Sentinel-2 L1C through three different data providers), pre-processes (atmospheric correction, and masking for land, cloud, NDWI and/or Band 8), and classifies (using a chosen ML model) all available imagery for the region of interest and period specified by the user.

Applications and potential services

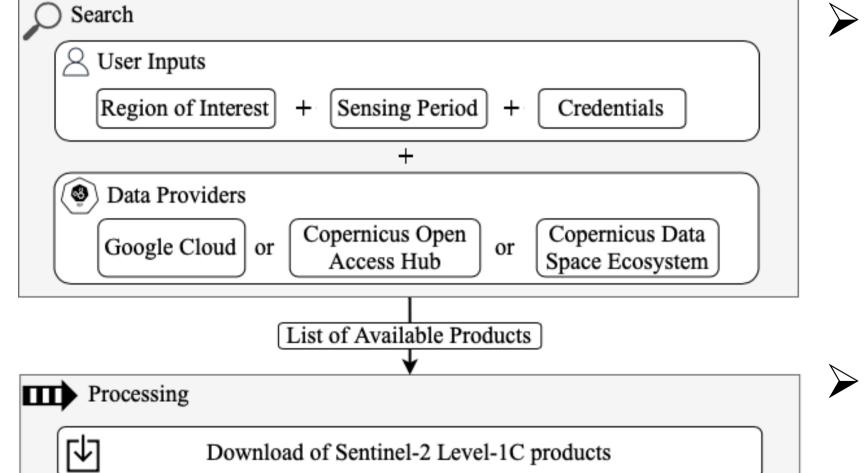
> Post-disaster management after flooding events in heavily polluted regions to direct cleanup activities and assess the input of debris into the ocean. Next figure shows POS2IDON applied to a major plastic debris event in Honduras Gulf in 2020/09/18.





- Supported classification models include Random Forest (RF) and XGBoost. An improved U-NET option is under testing.
- > RF and XGBoost models were trained with an extended version of the MARIDA spectral library, totalizing 14 including Marine Debris, classes, Macroalgae, Ships, Foam, Floating Turbid Water Clear and and Phytoplankton Blooms. Satisfactory performances provide confidence in using these models within POS2IDON.

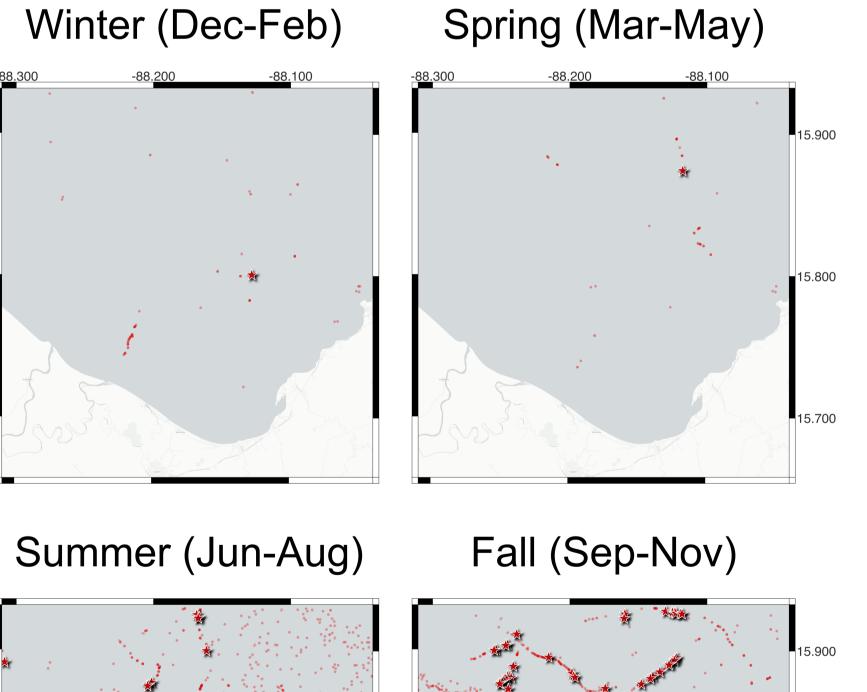
	XGBoost		
	loU	Recall	F1
Marine Debris	0.73	0.84	0.84
Macro Average	0.89	0.93	0.94
	Random Forest		
	loU	Recall	F1
Marine Debris	0.68	0.83	0.81
Macro Average	0.86	0.91	0.92

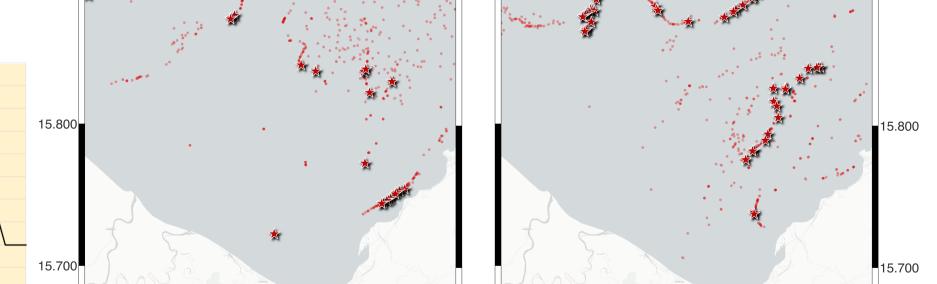


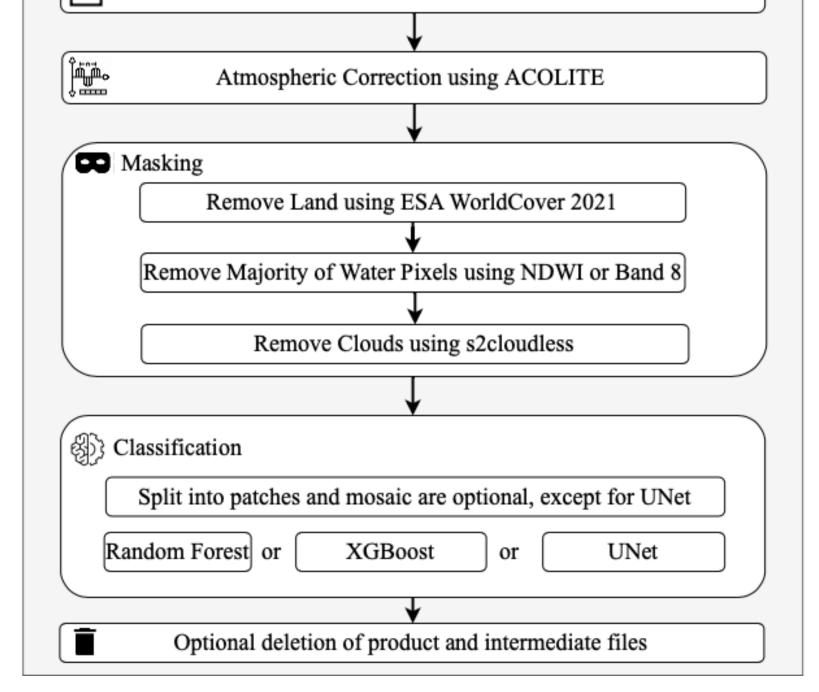
- > POS2IDON is developed mostly in Python, with some options in Julia to speed up processing and advantage take Of dedicated GPU.
- parameters The input used in this study include

Long-term seasonal analysis taking advantage of regular Sentinel-2 imagery (5 days) since 2018, useful to better understand patterns and trends in marine debris. POS2IDON is used to analyse Honduras Gulf for all 2020 (72 images).

Each of the classified images with suspected MD were human verified. Those with false artifacts and positives were removed. Dryer seasons (winter and spring) have less detection. debris in agreement with rivers being the major source of pollution.





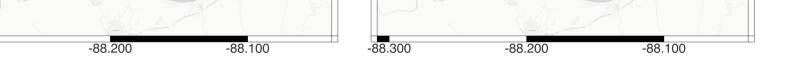


POS2IDON architecture

a land buffer of 50, an NDWI threshold of 0.01, an NDWI dilation of 6, a cloud mask threshold of 0.05, cloud а mask average of 10 and a cloud mask dilation of 50.

processing steps Post and human visualization critical to deal with are false positives.





Monthly precipitation in Honduras

Challenges and future work

- Reach out to stakeholders and test POS2IDON unique tool.
- \succ False detections due to foam, sun glint and thin clouds need more sophisticated processing modules (e.g. leveraging meteooceanographic data), as well as human validation of the results.
- \succ More in-situ data for validation of plastic-debris events using community-driven geo-referenced photographic databases, to enhance spectral libraries and improve classification models.

Acknowledgements: This work was supported by the Azores ECOBLUE project (financed by Iceland, Liechtenstein and Norway, through the EEA Grants) and LabPlas project (funded by the European Union's Horizon 2020 research and innovation programme under grant agreement No 101003954).

0,50

0.45

0,40 0.35

= 0,30 کر 2 0,25

E 0,20 0,15

0,10

0,05