

# Validating satellite-based seafloor maps for small islands

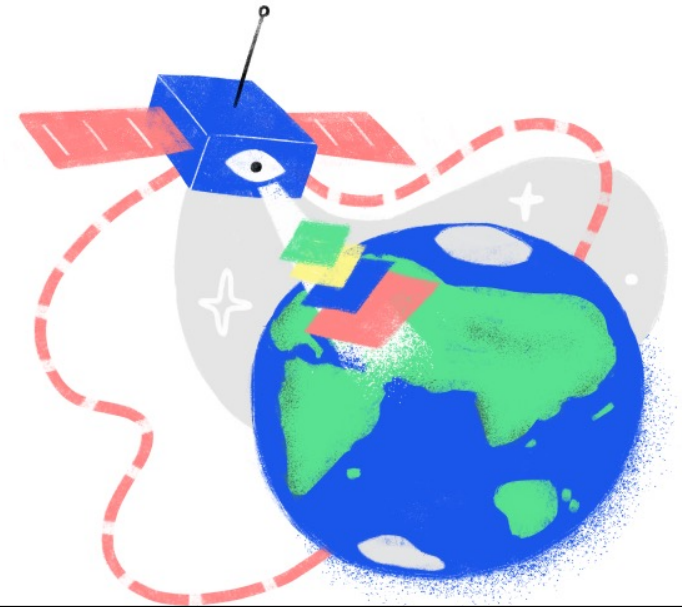
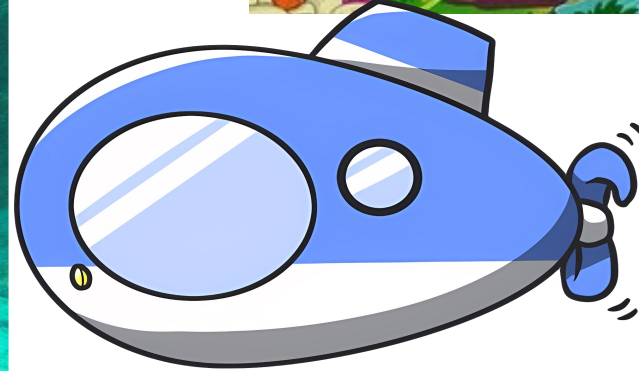
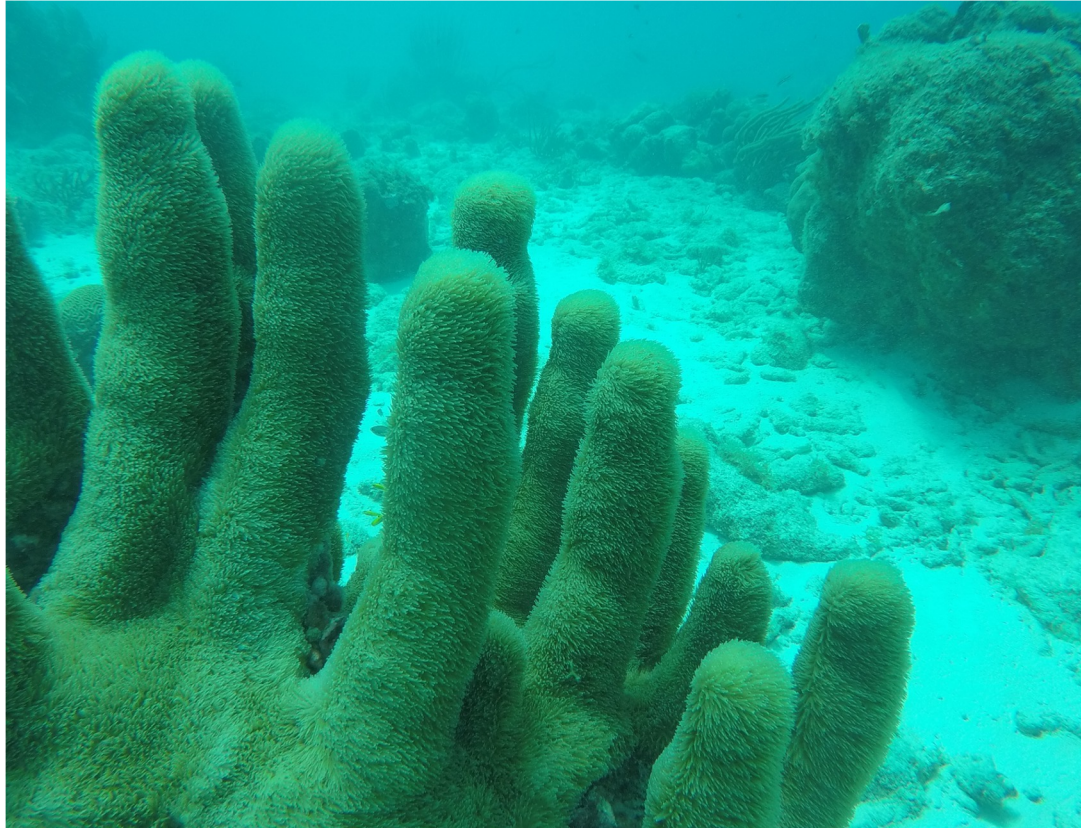


Tatiana Becker, MSc. ing.

[tatiana.becker@ua.aw](mailto:tatiana.becker@ua.aw)

[tatiana@metaboli.ca](mailto:tatiana@metaboli.ca)

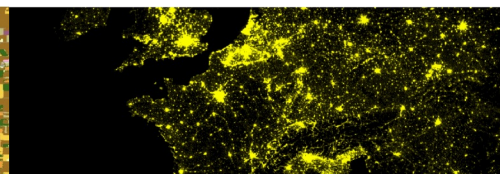
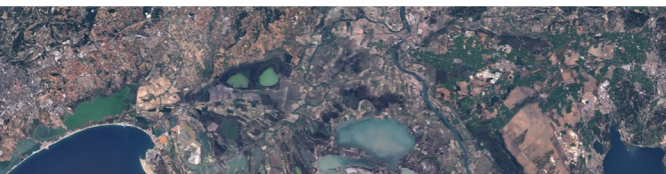
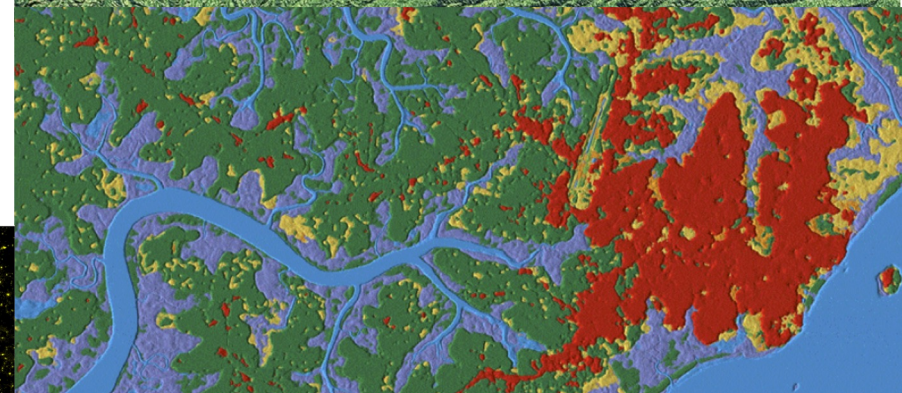
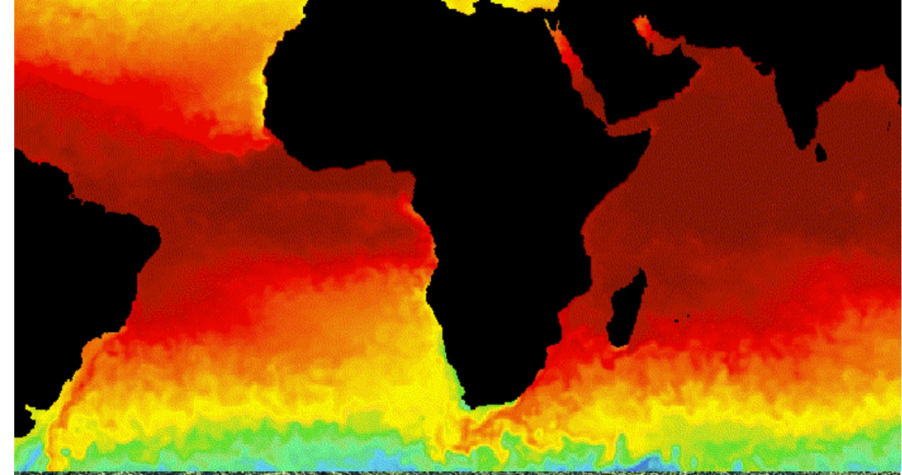
# Benthic Habitats



<b>Intro</b>	<b>Mapping</b>	<b>Validation</b>	<b>Findings</b>	<b>Discussion</b>	<b>Conclusion</b>
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# GOOGLE EARTH ENGINE

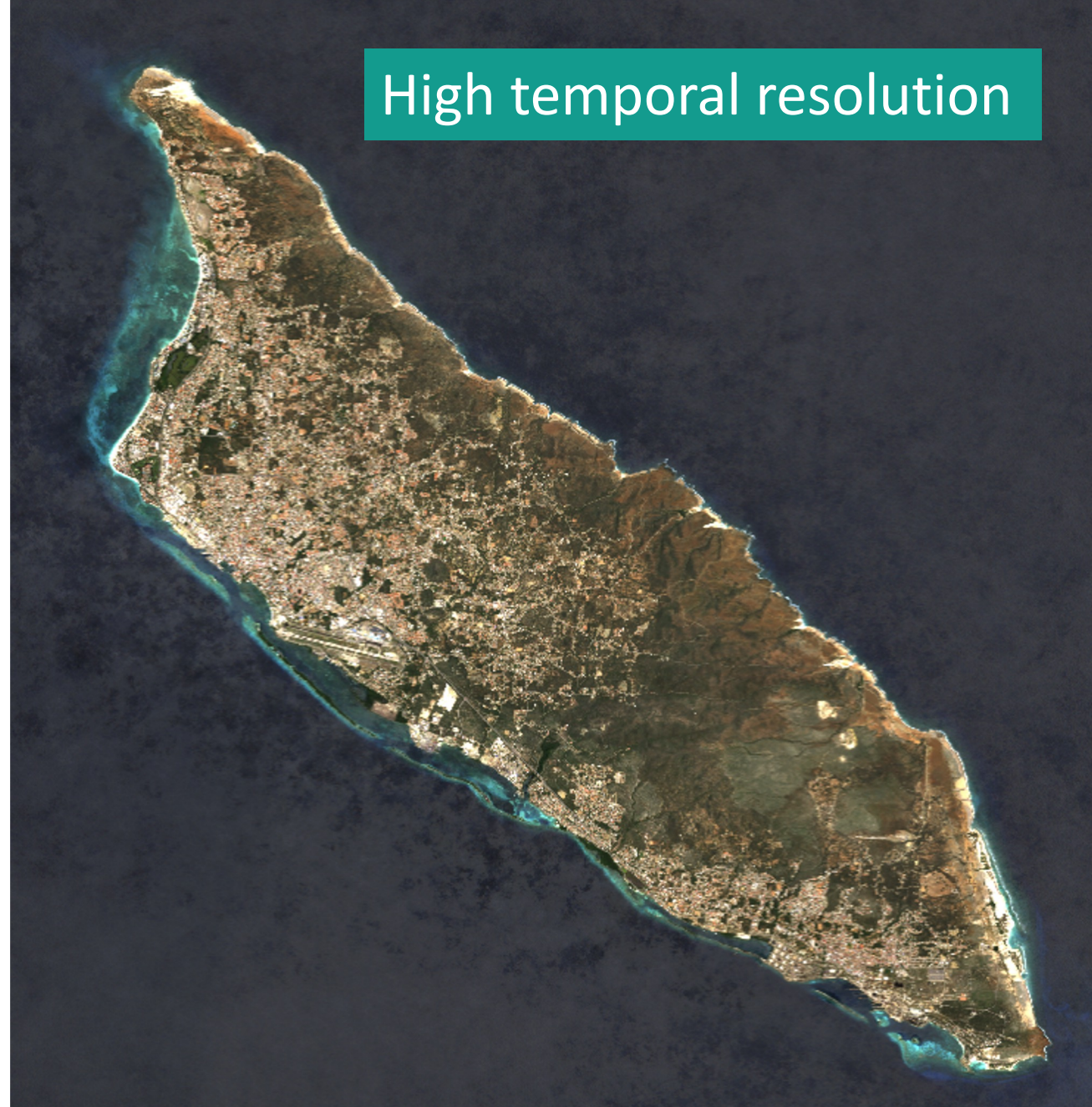
- Simple GIS mapping using Python code
- Connects to available datasets
  - Satellite data and imagery
  - Derived products (data and analysis)
- Web-based
  - No software needed
- Efficient online cloud processing
- Define your own parameters
- Export resulting images and data
  - QGIS, ArcGIS
- Easily share code and results



# Harmonized Sentinel-2 A MSI

B1	443.9nm (S2A) / 442.3nm	Aerosols
B2	496.6nm (S2A) / 492.1nm	Blue
B3	560nm (S2A) / 559nm	Green
B4	664.5nm (S2A) / 665nm	Red
B5	703.9nm (S2A) / 703.8nm	Red Edge 1
B6	740.2nm (S2A) / 739.1nm	Red Edge 2
B7	782.5nm (S2A) / 779.7nm	Red Edge 3
B8	835.1nm (S2A) / 833nm	NIR
B8A	864.8nm (S2A) / 864nm	Red Edge 4
B9	945nm (S2A) / 943.2nm	Water vapor
B11	1613.7nm (S2A) / 1610.4nm	SWIR 1
B12	2202.4nm (S2A) / 2185.7nm	SWIR 2
AOT	Aerosol Optical Thickness	
WVP	Water Vapor Pressure	
SCL	Scene Classification Map	
TCI_R	True Color Image, Red channel	
TCI_G	True Color Image, Green channel	
TCI_B	True Color Image, Blue channel	
MSK_CLDPRB	Cloud Probability Map	
MSK_SNOWPRB	Snow Probability Map	
QA10	Always empty	
QA20	Always empty	
QA60	Cloud mask	

High temporal resolution



Intro

Mapping

Validation

Findings

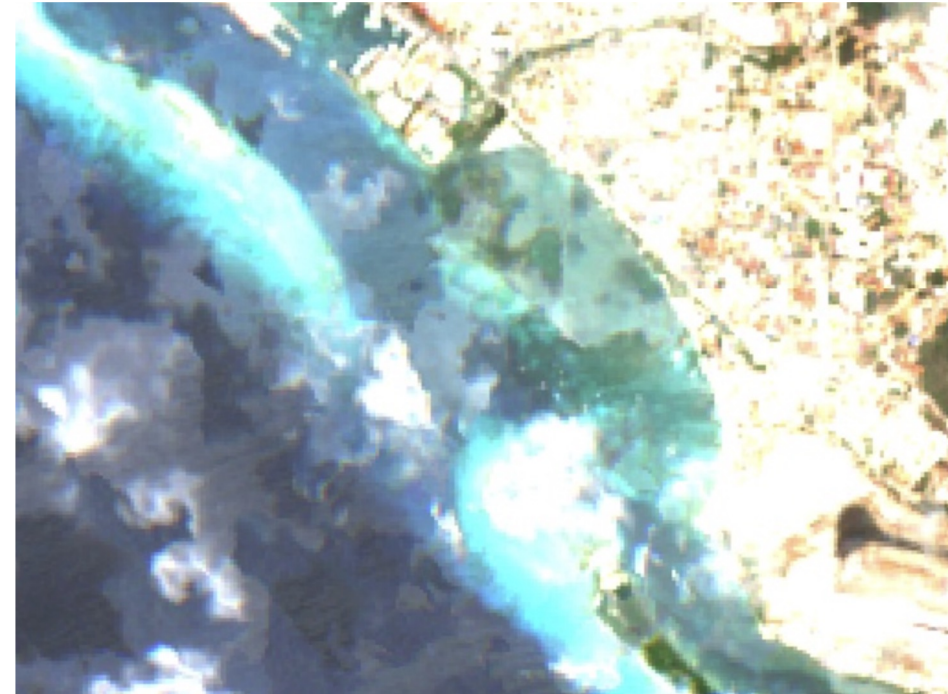
Discussion

Conclusion

# SEAFLOOR MAPPING: Data Processing

## Filtering and Cleaning:

- Time
- Location
- Water
- Bands
- Turbidity
- Sun glint
- Clouds
- Boats
- Image selection
- Image quality
  - Bands

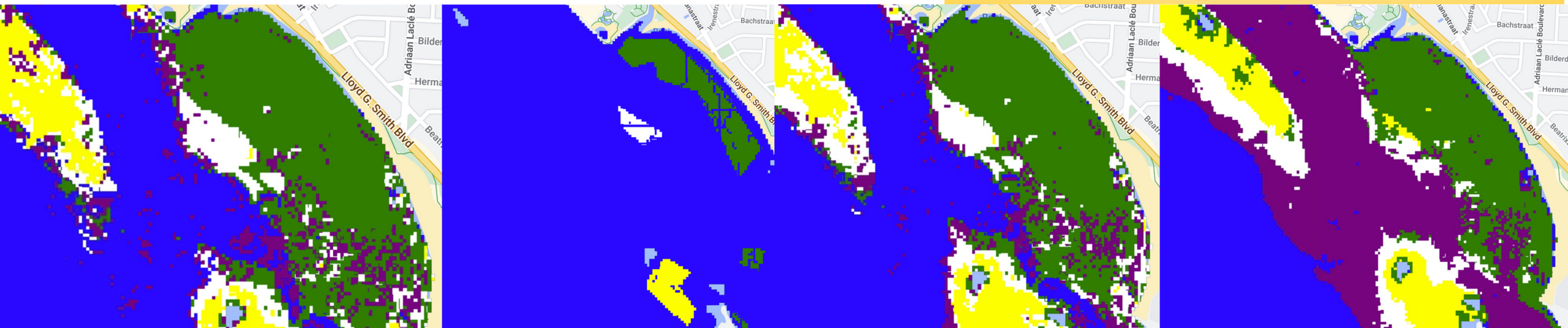


# SUPERVISED CLASSIFICATION

1. Image pre-processing (Li et al., 2021)
2. Training data
3. Classify map testing different models  
(CART, SVM, RF, NB)

## Main Mapping Steps:

1. Draw box around beach of interest
2. Sentinel 2 data for region and dates of interest
3. Mask out bad data
4. Median pixel values
5. Create classifier model using relevant bands and training data
6. Classify map
7. Export map data



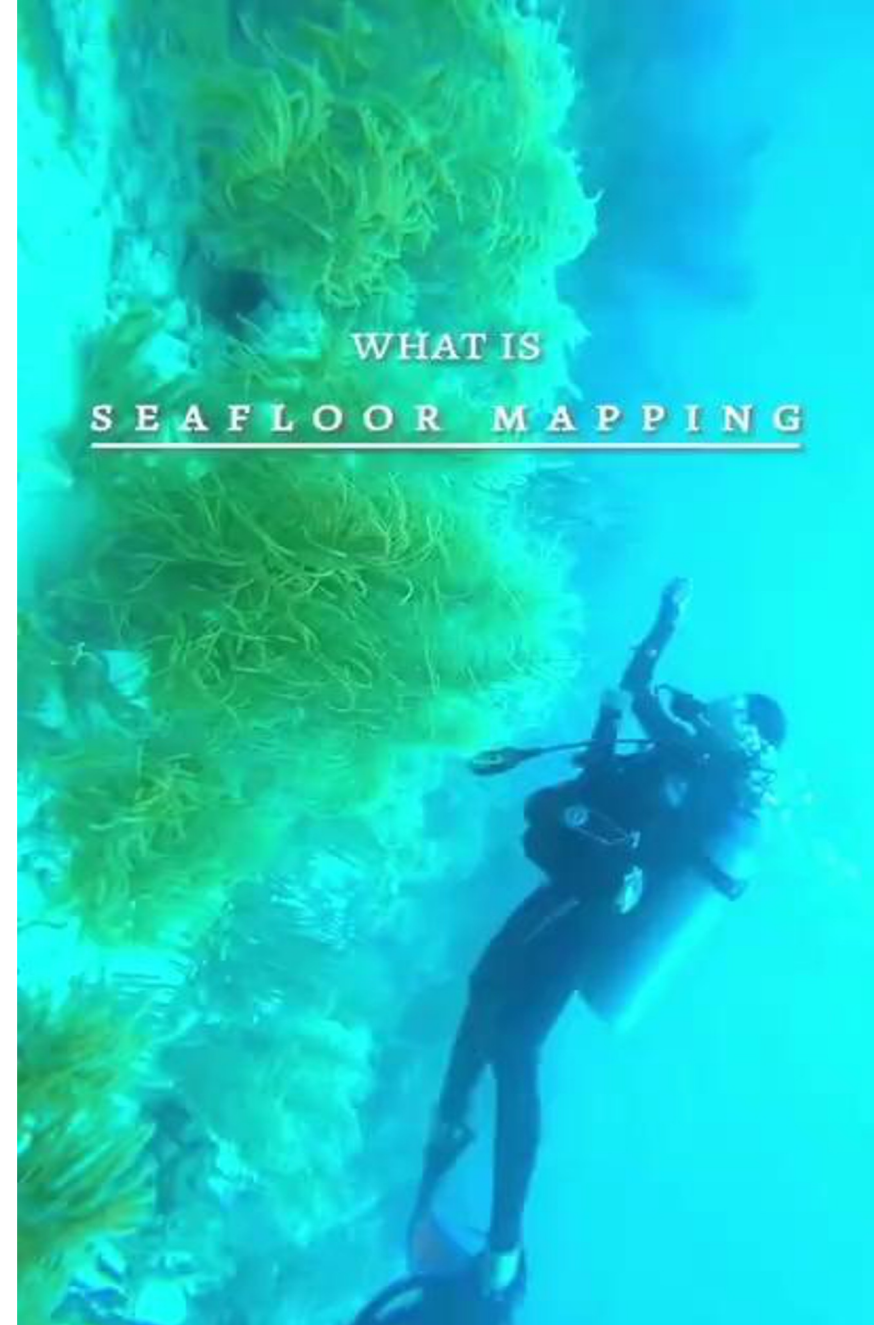
# VALIDATION

## WHY?

- To test the accuracy against real-world conditions
- To quantify the uncertainty for end-users
- To determine the needs for improvements or refinements

## METHOD:

- Sea-truthing according to Roelfsema et al (2019)'s protocol and Stehman & Czaplewski (1998)'s accuracy assessment guidelines
- Transects parallel and perpendicular to shore captures variability along the shore and different depths (max 15 m)
- Assisted by Scubble Bubbles diverss



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# VALIDATION: WORKFLOW



<b>Intro</b>	<b>Mapping</b>	<b>Validation</b>	<b>Findings</b>	<b>Discussion</b>	<b>Conclusion</b>
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# FINDINGS: SENTINEL (SSS) vs Sea-truthing



correctness  
■ incorrect  
■ correct  
Google Hybrid

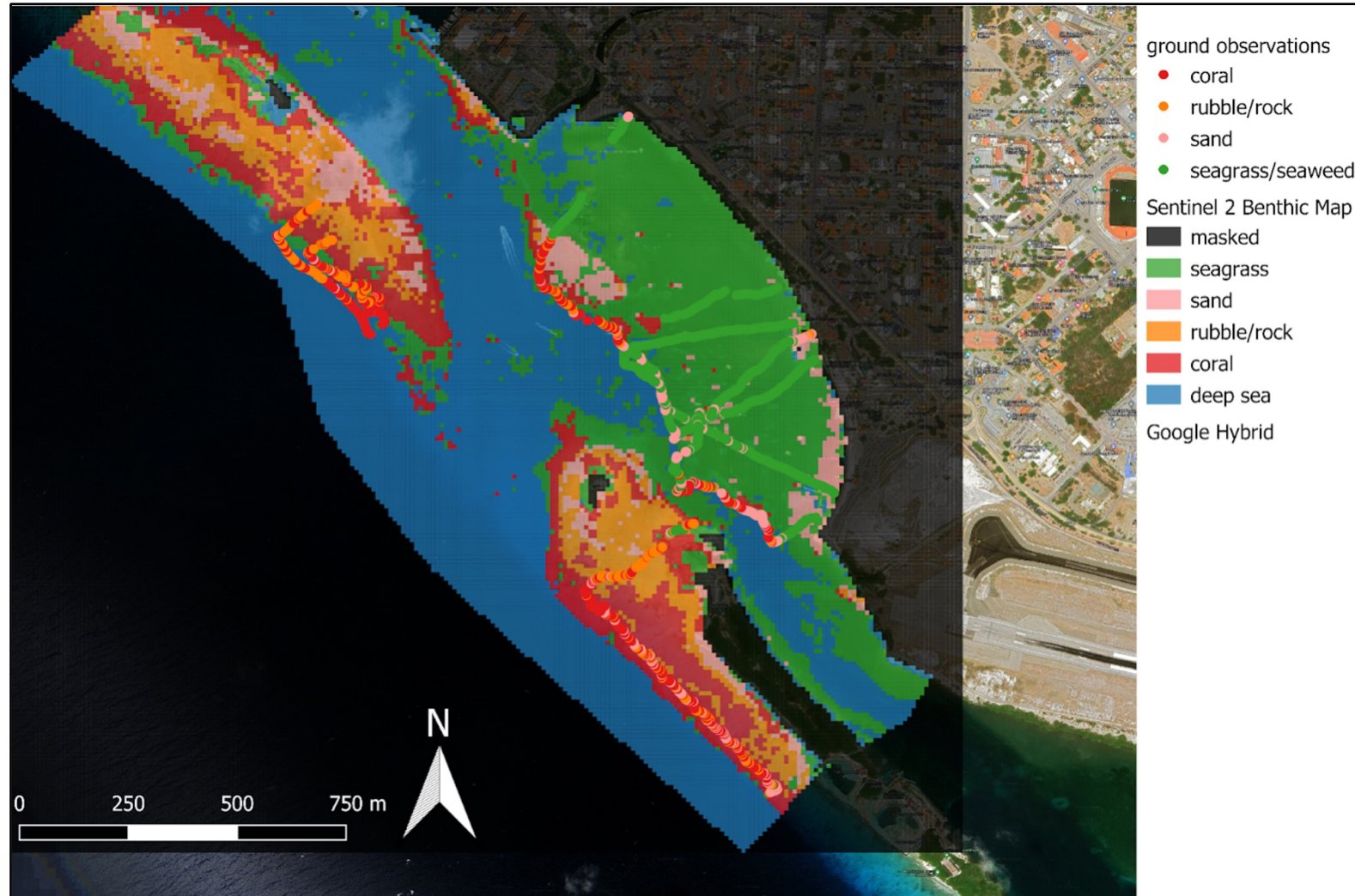
Testing Accuracy:

46.3%

Validation Accuracy:

60.7%

# FINDINGS: SENTINEL (SSS) benthic habitat map



# FINDINGS: vs Allen Coral Atlas

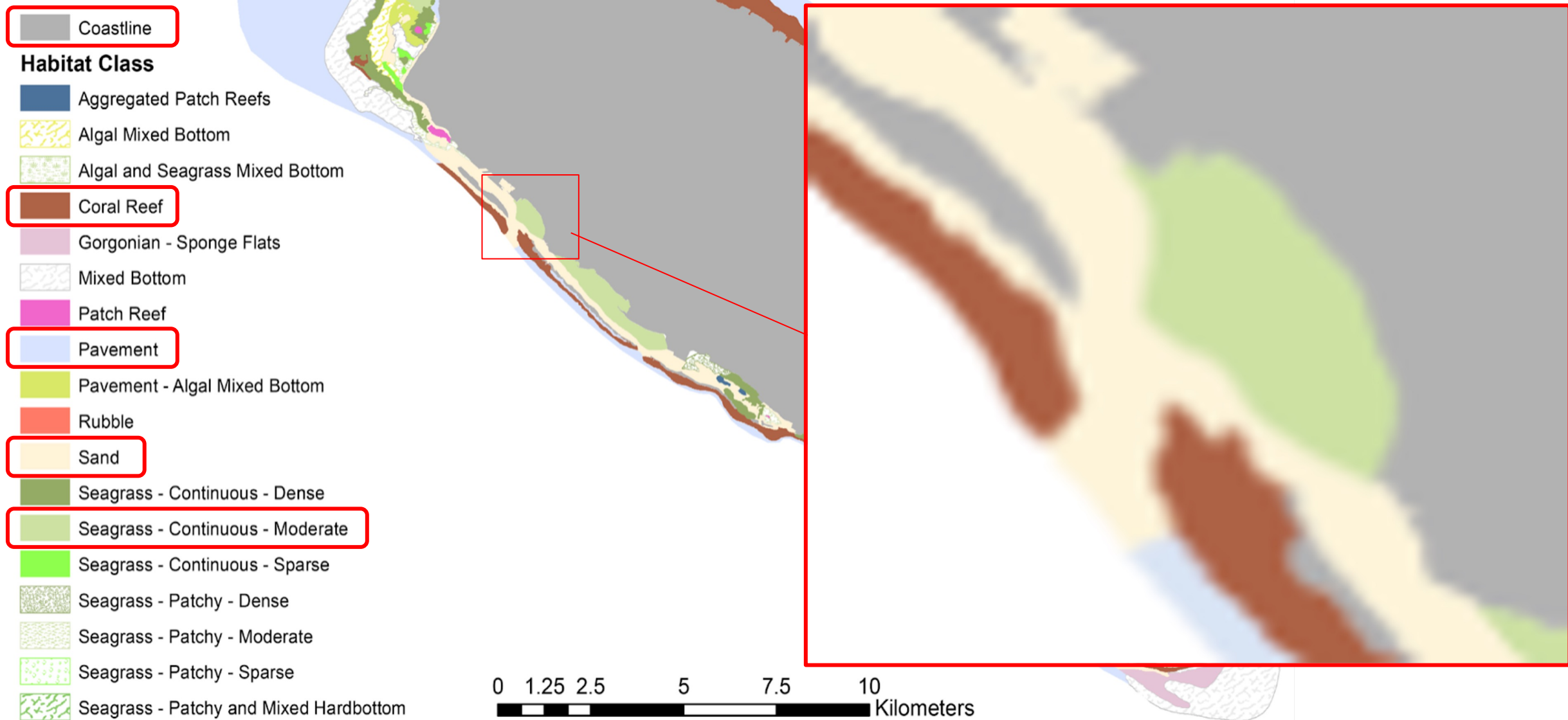


ground observations

- Coral
- Not applicable
- Rock
- Rubble
- Sand
- Seagrass
- Seaweed
- Unclear
- Water

Allen coral atlas map Image  
Google Hybrid

# FINDINGS: vs Carmabi Benthic Habitat Map



# Discussion

- Balance sea-truthing effort:
  - Emphasis on habitat vs cover
  - increasing sampling in macroalgal habitat
- Limited accuracy of used GPS devices
- Timeframe correspondence of Field and Sentinel imagery
- Sea-truthing classification labels:  
Redefining?

# Conclusions

- Moderate accuracy
  - Similar or even better compared to other literature (e.g. Fauzan et al., 2017, Wicaksono et al., 2021 )
  - Better accuracy indicated for local-scale mapping, compared to existing seafloor maps for Aruba
  - Depth limit of 10 m
- More accurate in shallow depths
  - Fast changes (bleaching, disease, rapid invasions)
  - Slow notable changes, considering the availability of long-term data

# Acknowledgements



# References

Allen Coral Atlas (2022). Imagery, maps and monitoring of the world's tropical coral reefs. doi.org/10.5281/zenodo.3833242

Fauzan, M. A., Kumara, I. S. W., Yogyantoro, R., Suwardana, S., Fadhilah, N., Nurmalasari, I., Apriyani, S., & Wicaksono, P. (2017). Assessing the capability of sentinel-2A data for mapping seagrass percent cover in Jerowaru, East Lombok. *Indonesian Journal of Geography*, 49(2), 195–203. <https://doi.org/10.22146/ijg.28407>

Li, J.; Knapp, D.E.; Lyons, M.; Roelfsema, C.; Phinn, S.; Schill, S.R.; Asner, G.P. (2021) Automated Global Shallow Water Bathymetry Mapping Using Google Earth Engine. *Remote Sens.* 2021, 13, 1469. <https://doi.org/10.3390/rs13081469>

Roelfsema C.M. K. Markey, E. Kennedy, E. Kovacs, R. Borrego, H. Fox, B.Bambic, B.Free, K. Rice and S.R. Phinn (2019). Protocol for Georeferenced Benthic Photoquadrat Surveys. Remote Sensing and Research Centre, School of Earth And Environmental Sciences, University of Queensland, Brisbane, Australia.

Stehman, S. V, & Czaplewski, R. L. (1998). Design and Analysis for Thematic Map Accuracy Assessment - an application of satellite imagery. *Remote Sensing of Environment*, 64(January), 331–344.

Vermeij, M., Marhaver, K., Estep, A., Sandin, S. (2020). Coral Reefs Study Aruba 2019. Carmabi Foundation Curaçao.

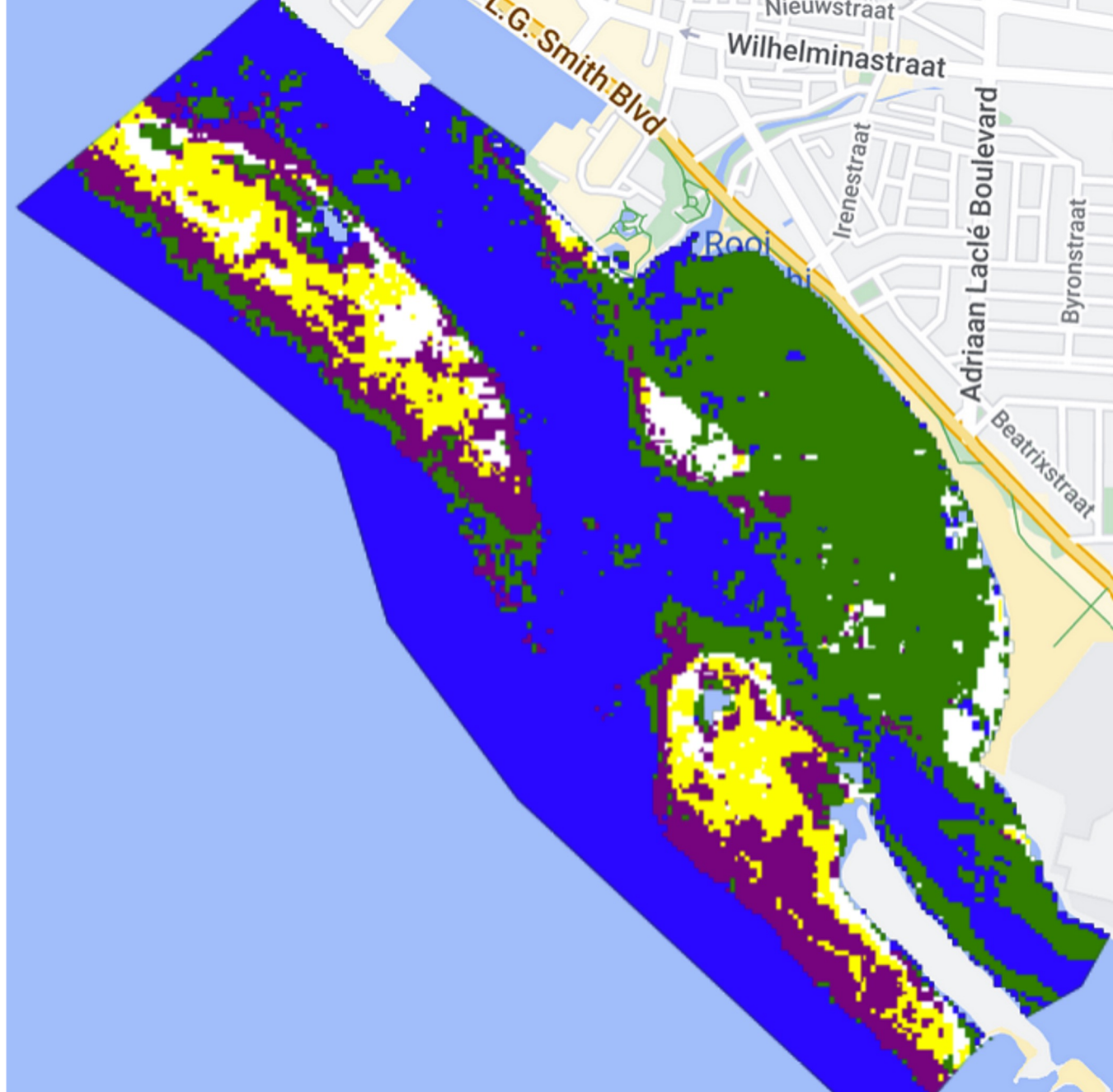
Wicaksono, P., Wulandari, S. A., Lazuardi, W., & Munir, M. (2021). Sentinel-2 images deliver possibilities for accurate and consistent multi-temporal benthic habitat maps in optically shallow water. *Remote Sensing Applications: Society and Environment*, 23(June), 100572. <https://doi.org/10.1016/j.rsase.2021.100572>



# MAPS

## To do

- Review training data
  - Groundtruthing
  - Kayak images, zooniverse, AI
- Mask deeper water
- Add indices for training
- Calculate areas
- Repeat for Aruba
- Test on other islands



Scripts Docs Assets

- in2031
- rifmtdsv
- seafloor
- seafloor-aruba
- seafloor-aruba-connected
- seafloor-thumb
- seafloor\_X
- seafloor\_cart\_s2
- seafloor\_svm+cart
- sentinelSR\_export
- statiamdsv

```

seafloor-aruba *
80
81 // Train a CART classifier with default parameters.
82 var trained = ee.Classifier.smileCart().train(training, label, bands);
83
84 // Classify the image with the same bands used for training.
85 image = image.clip(roi);
86 var classified = image.select(bands).classify(trained);
87 print(classified);
88
89 Map.addLayer(image,
90     {bands: ['B4_median', 'B3_median', 'B2_median'], min: 0, max: 20},
91     'image');
92 Map.addLayer(classified,
    
```

Inspector Console Tasks

Use print(...) to write to this console.

- Image (1 band) JSON
- Image (1 band) JSON
- 77373.18039215682



