

## PREDICTING INDIVIDUAL CORAL COLONY GROWTH RECRUITMENT & MORTALITY USING NASA NEMO-NET



DREW CHRISTENSEN, VED CHIRAYATH, IMAD TIBERMACINE, SOUFYANE BOUCHELAGHEM, JESSIE YANG, NOAH PIKIELNY, ISAIAH WANG, JAKOB BROWN, NAYONIKA CHOUDHURY SAM PURKIS (PI), UNIVERSITY OF MIAMI NOAA - TOM OLIVER (CO-PI), COURTNEY COUCH (CO-PI)

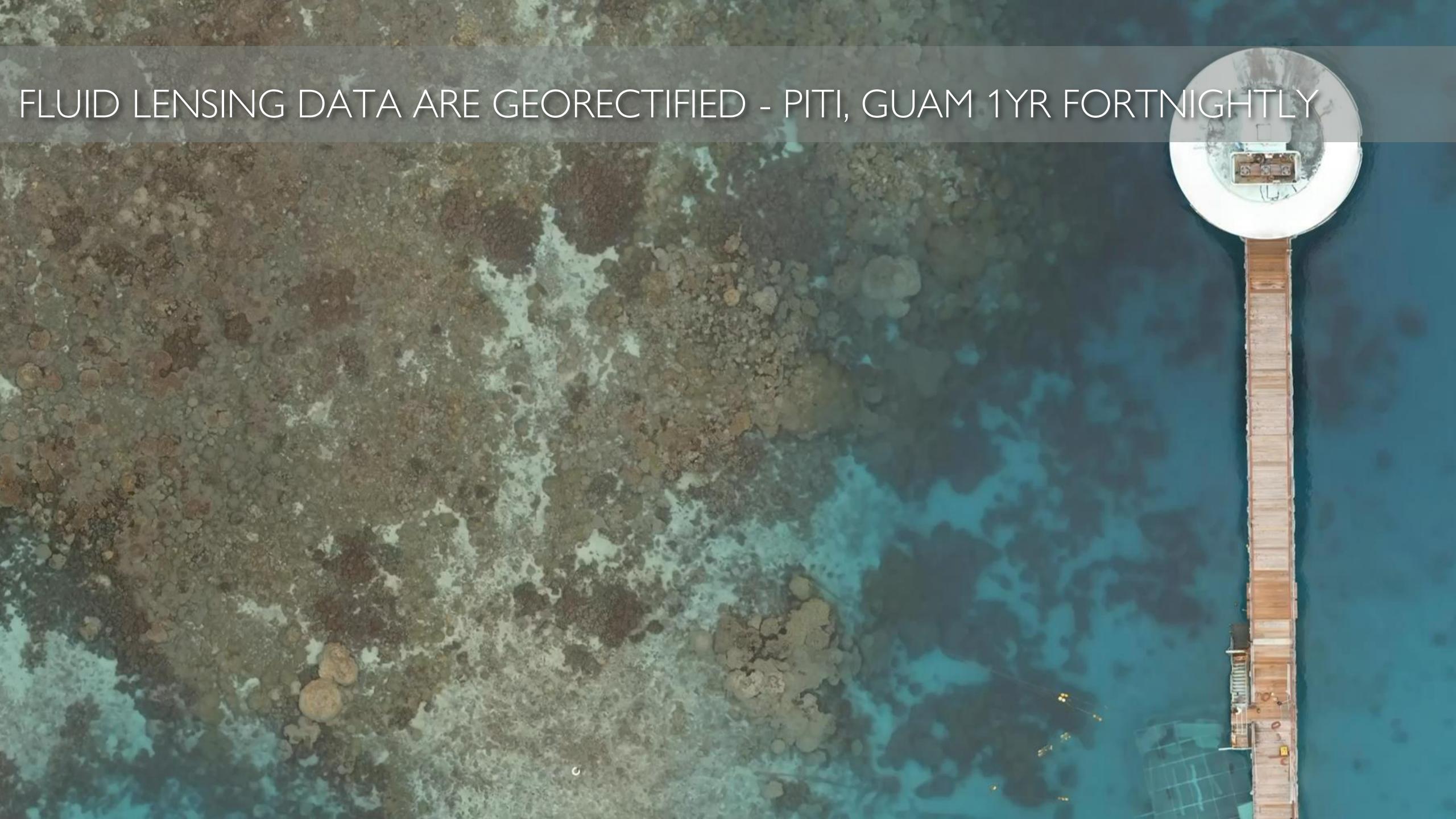










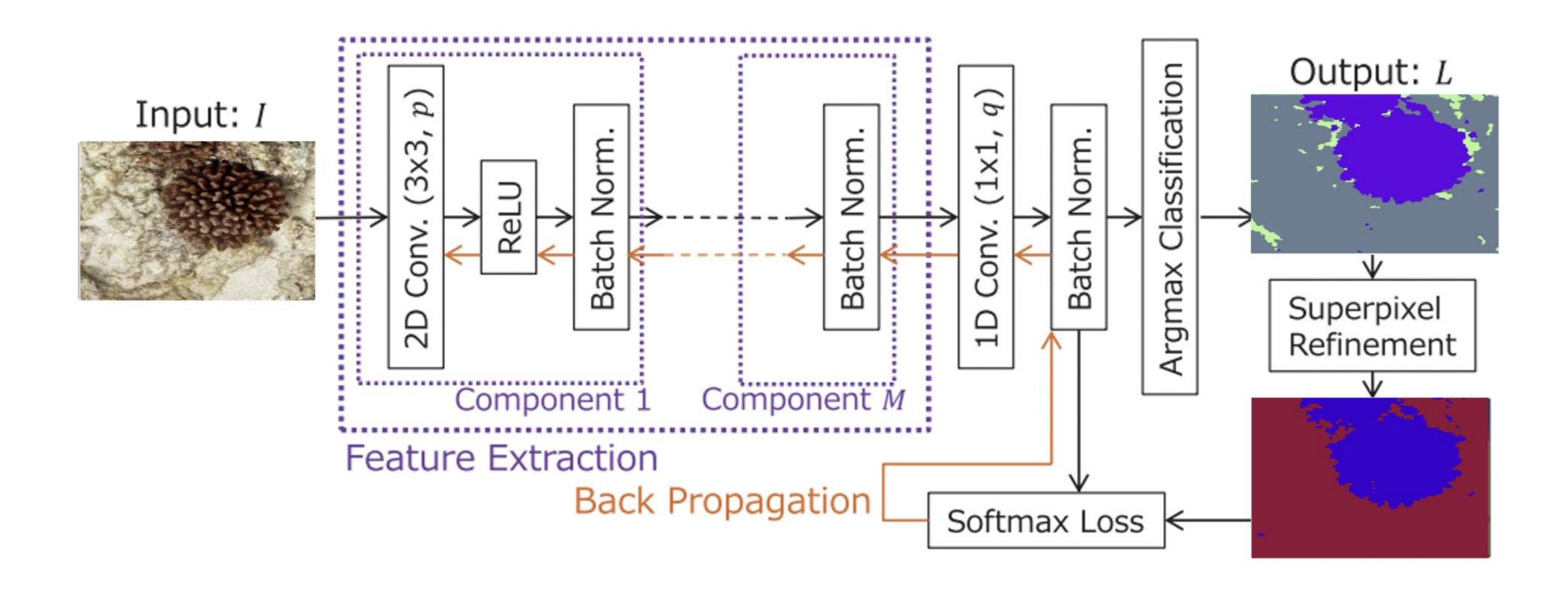


#### SAMPLE TIME SERIES PROVIDED BY NOAA - MISALIGNED AND HETEROGENEOUS



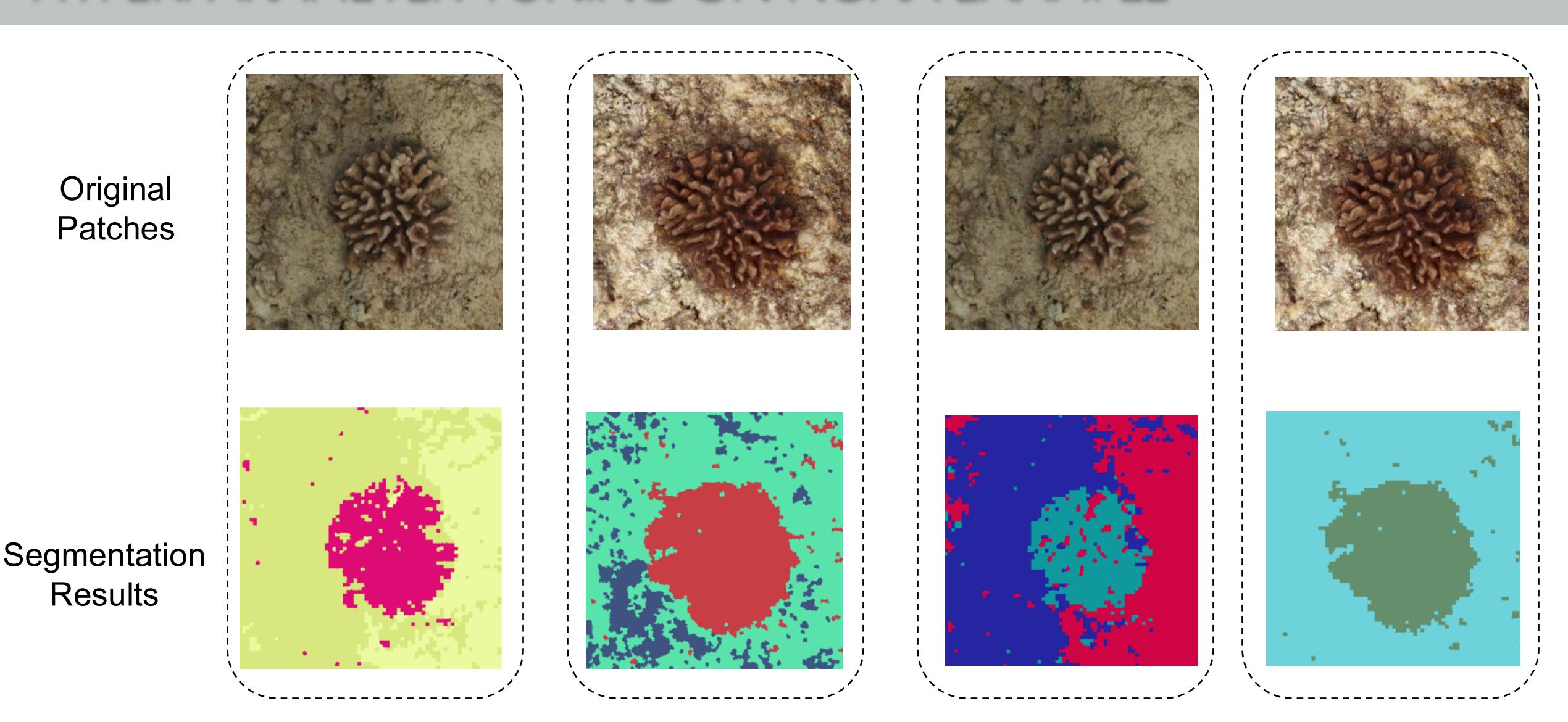


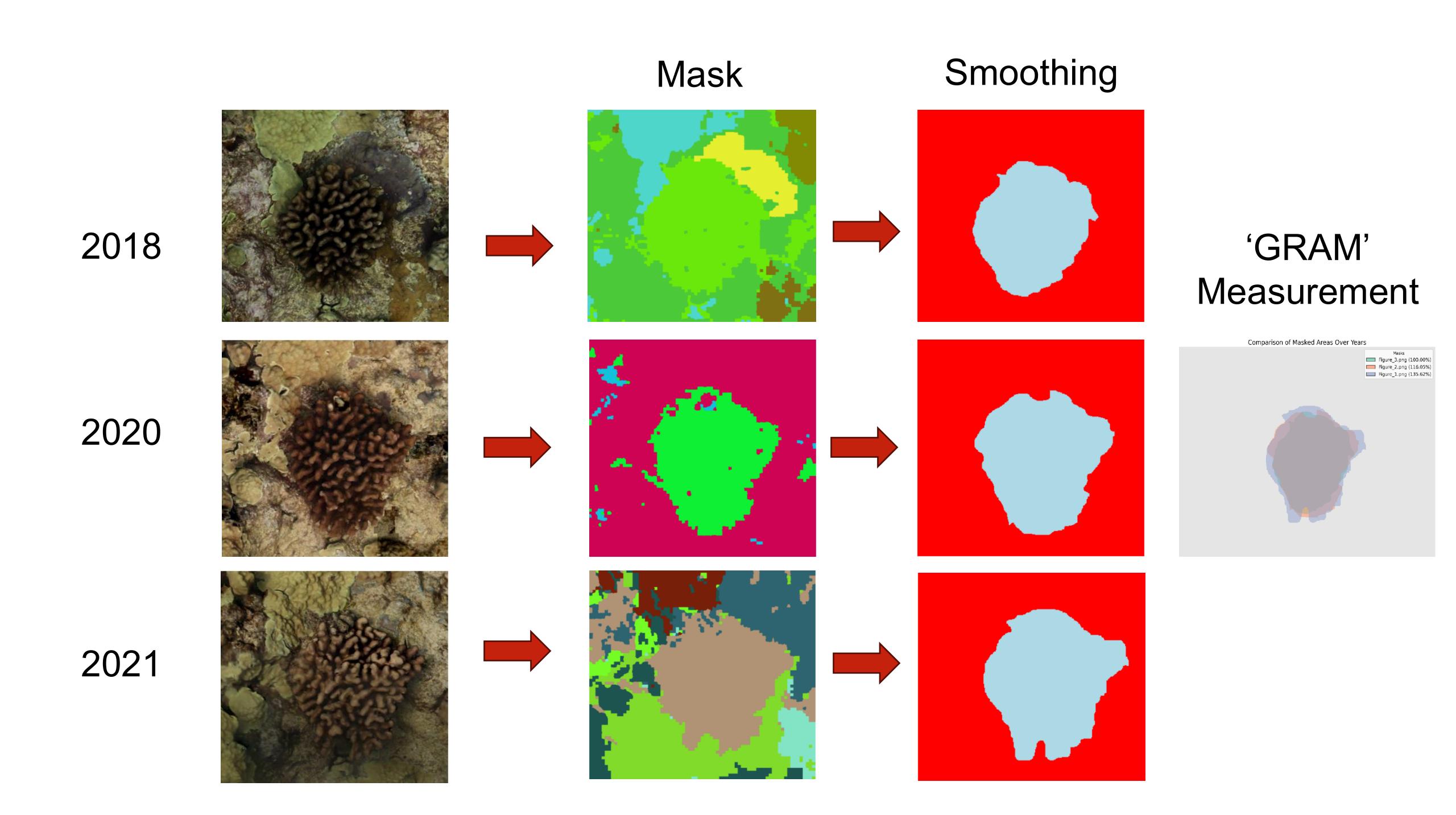
2018



New NeMO-Net Model developed for PICOGRAM

#### HYPERPARAMETER TUNING ON NOAA EXAMPLE

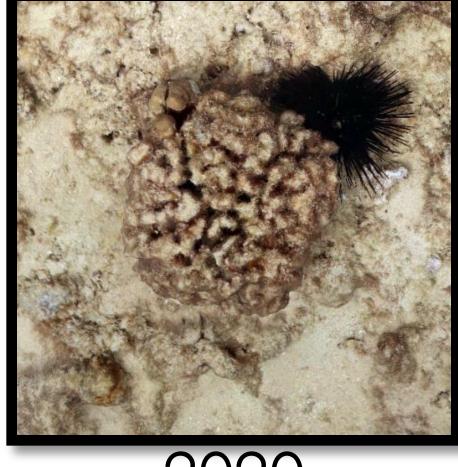




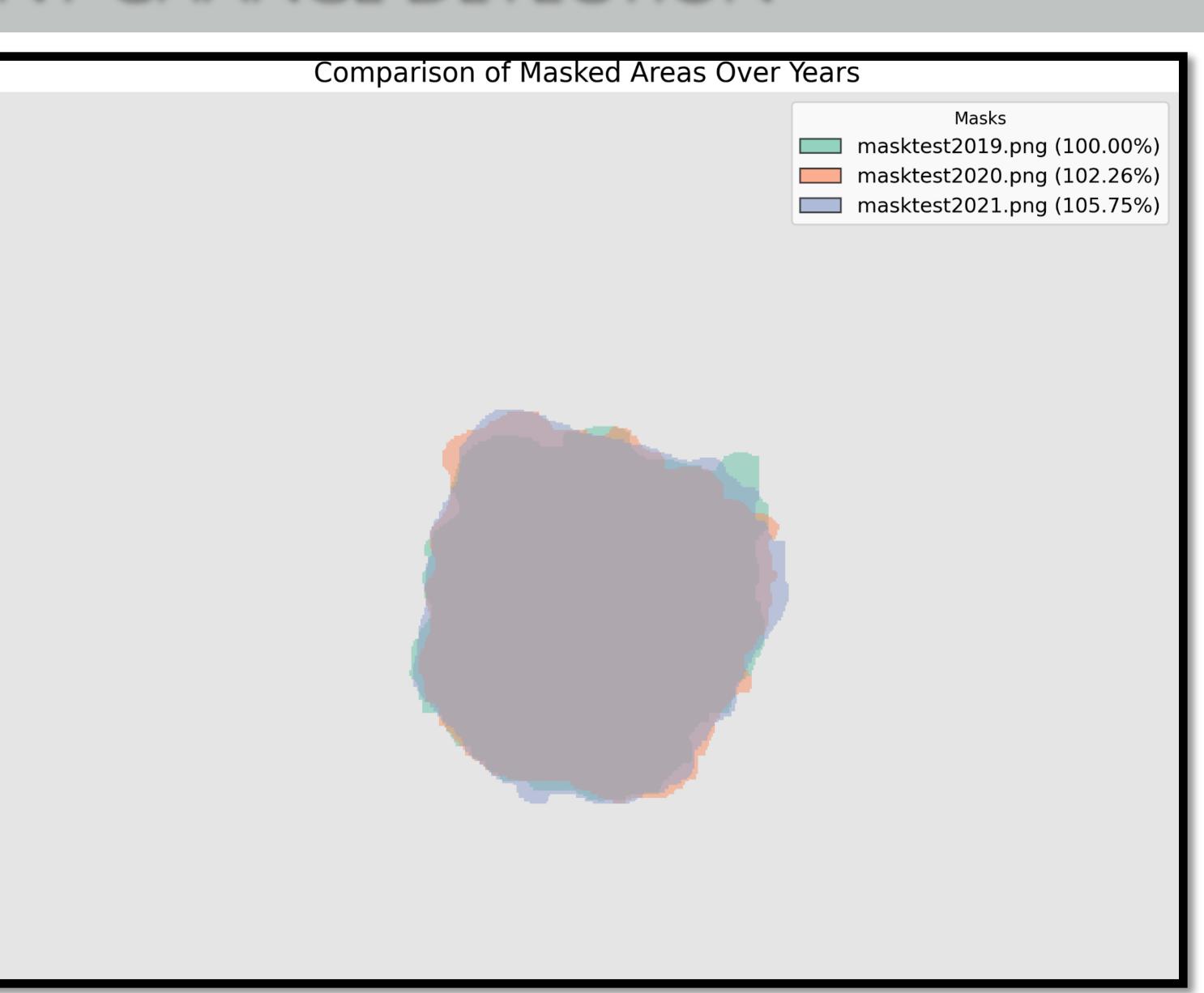
#### INDIVIDUAL CORAL COLONY CHANGE DETECTION



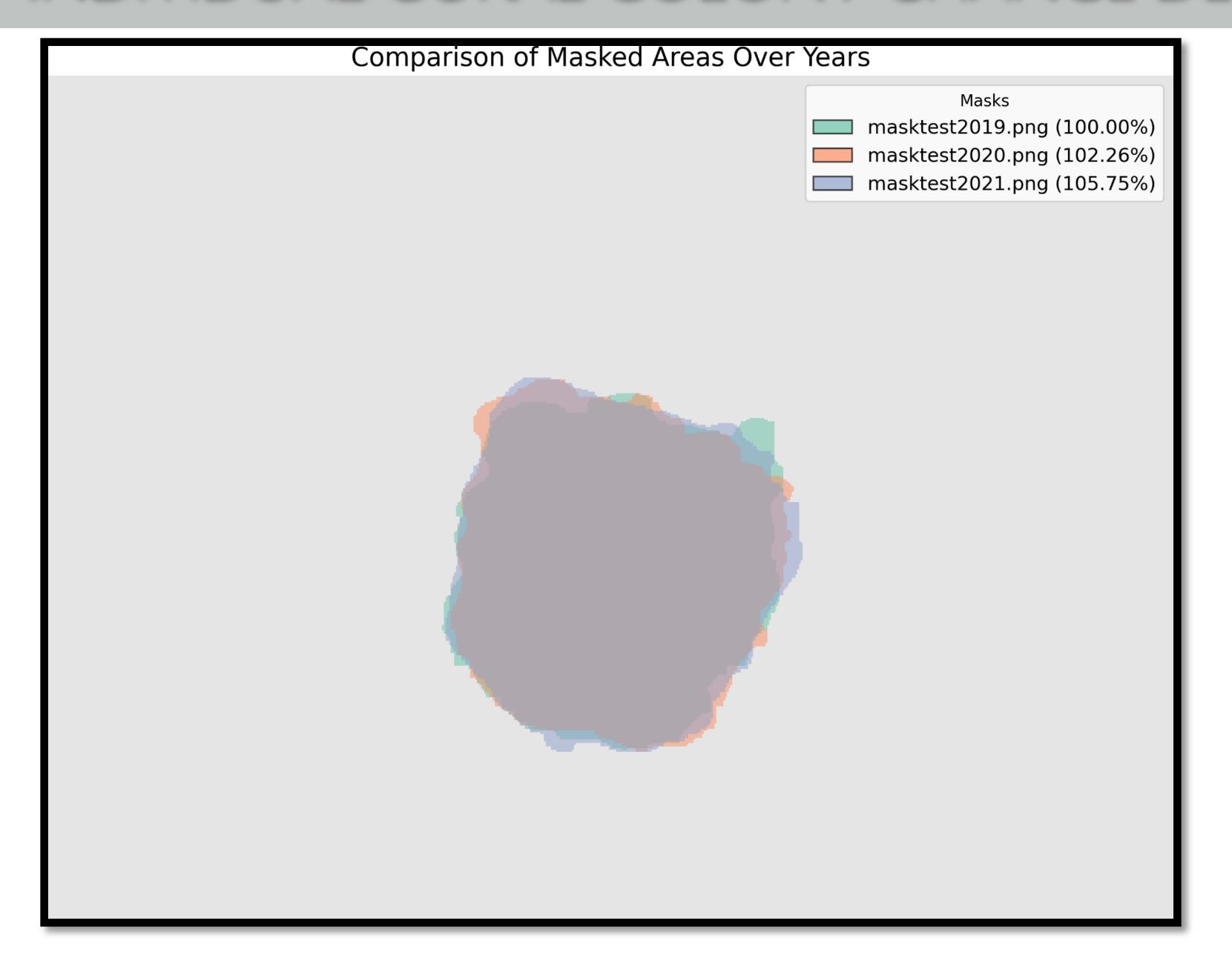








#### INDIVIDUAL CORAL COLONY CHANGE DETECTION



~2.5% increase in planform area per year from 2019 to 2021



#### Homepage

#### HTTP://NEMONET.INFO/PICOGRAM

ABOUT CHALLENGE

**PICOGRAM Data Viewer** 

PUBLICATIONS TEAM

#### **Our Collaborators**

Collaborator 1

Collaborator 2

Collaborator 3

**PREDICTION** 

TIME SERIES

Run a simple prediction

Track the corals time series (set to 4, but it could be adjusted)

#### Prediction Page

#### **Prediction Page**

Please upload your corals.

First, click here to upload transect or coral colony ortho

Then, click here to run the prediction. This operation can take between 1 to 3 mins based on image size.

#### Prediction Page output

#### **Prediction Page**

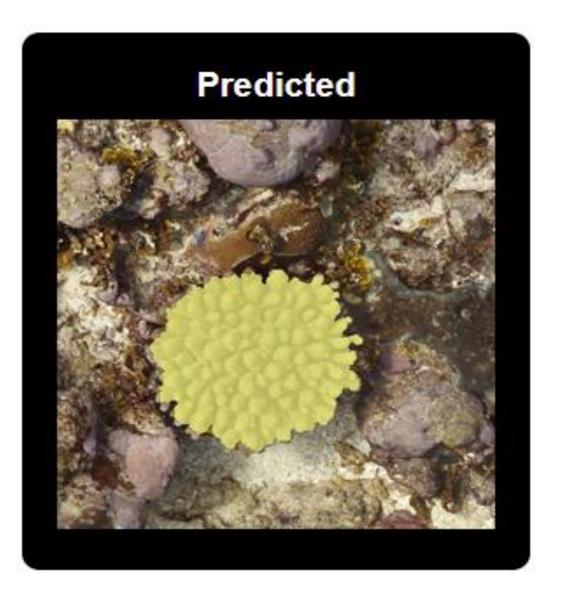
Please upload your corals.

After the model finishes the segmentation, the results will be displayed automatically.

Browse... AM10.png

UPLOAD



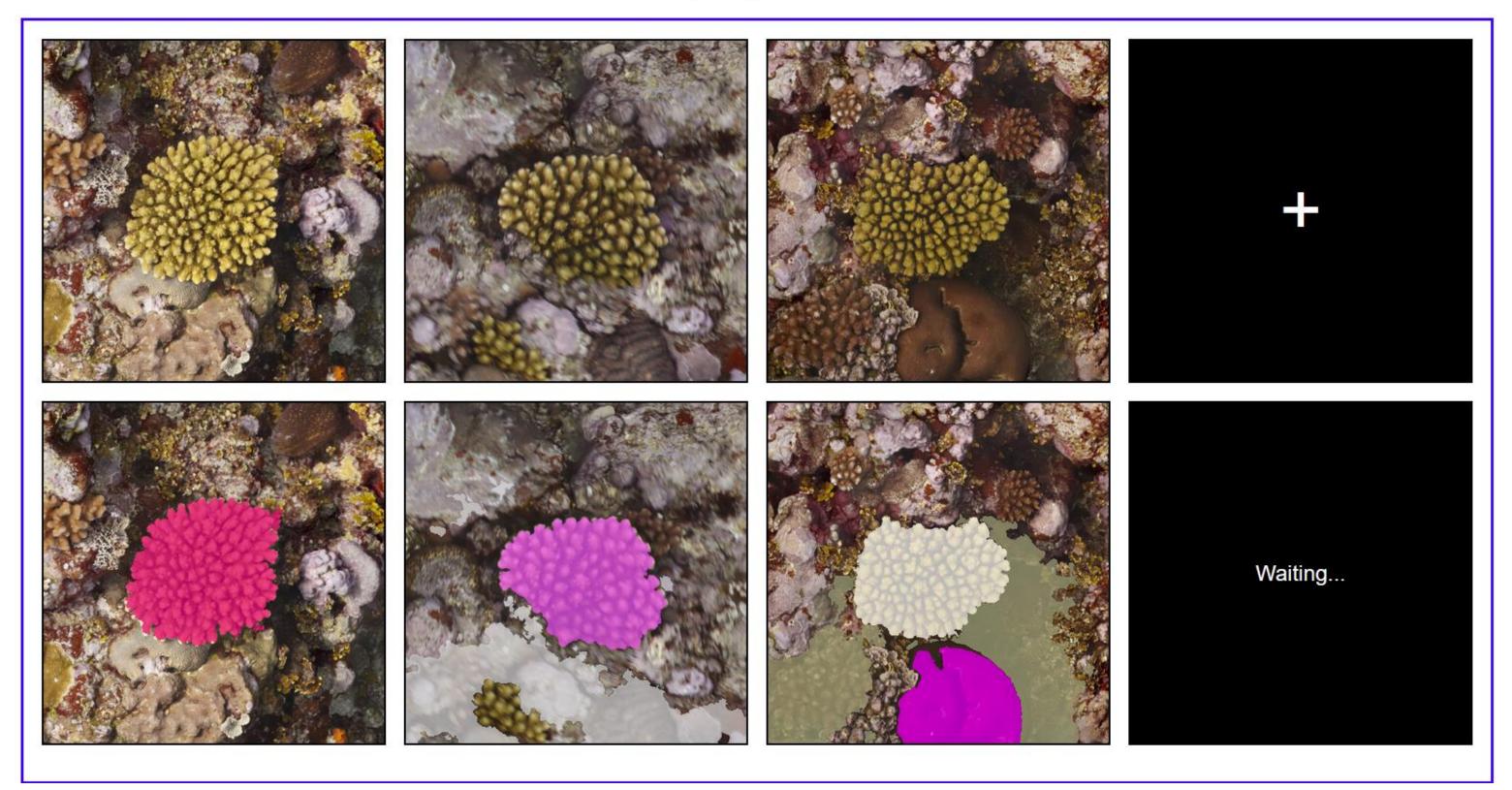


#### Time Series Page Output

**Time Series Page** 

Please upload your time series.

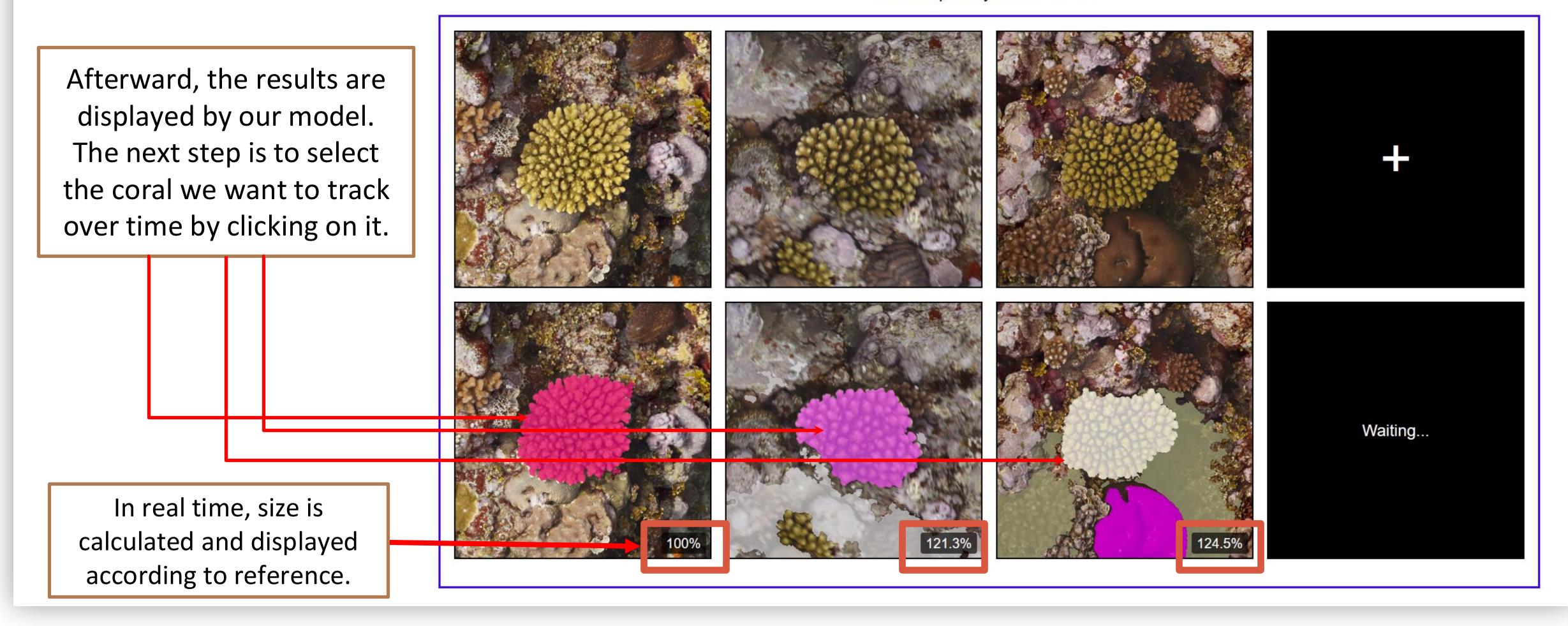
After the model finishes the segmentation, the results will be displayed automatically.



#### Time Series Page Output

**Time Series Page** 

Please upload your time series.





PIFSC/ESD/ARP

# PICOGRAM Tests & Model Comparisons

20250228
Thomas Oliver - NOAA Federal
<Thomas.Oliver@noaa.gov>

Michael Akridge - NOAA Affiliate <michael.akridge@noaa.gov>

#### **Ground Truth**



#### PICOGRAM 1.0

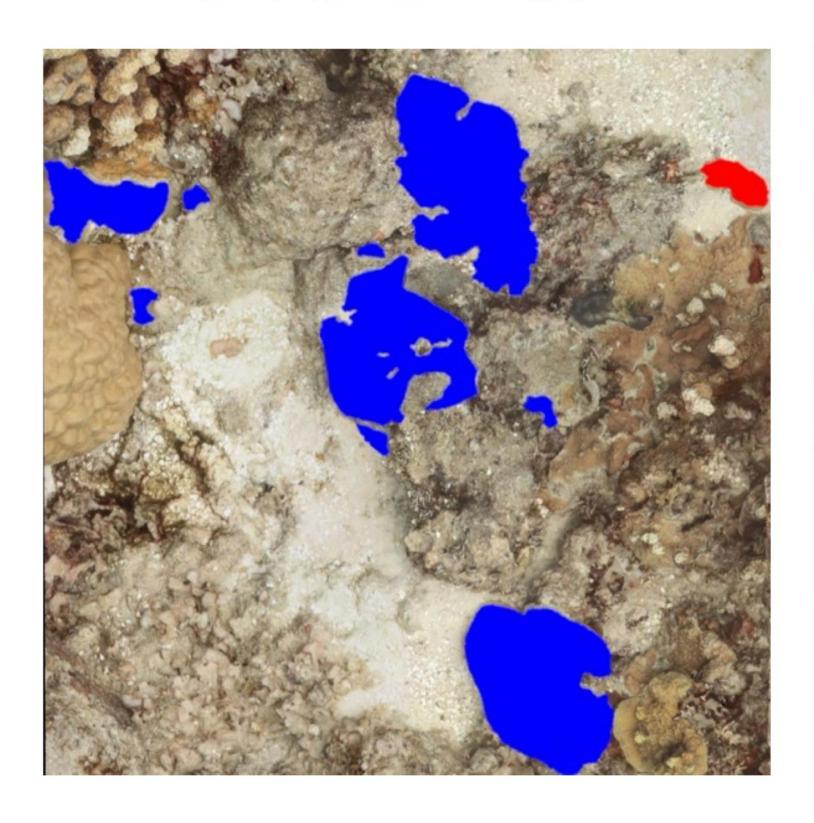


#### PICOGRAM 2.0





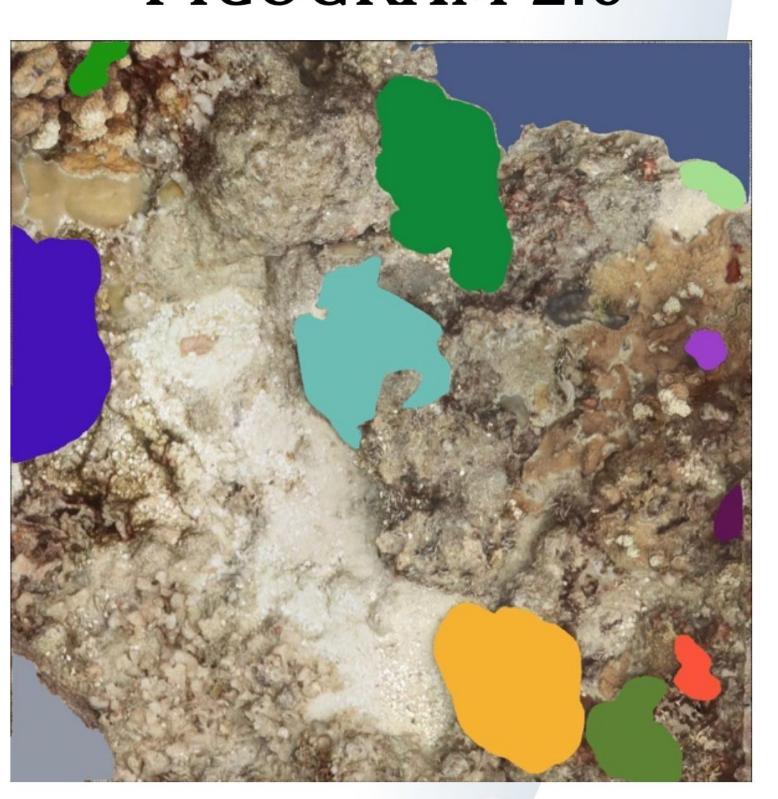
#### **Ground Truth**



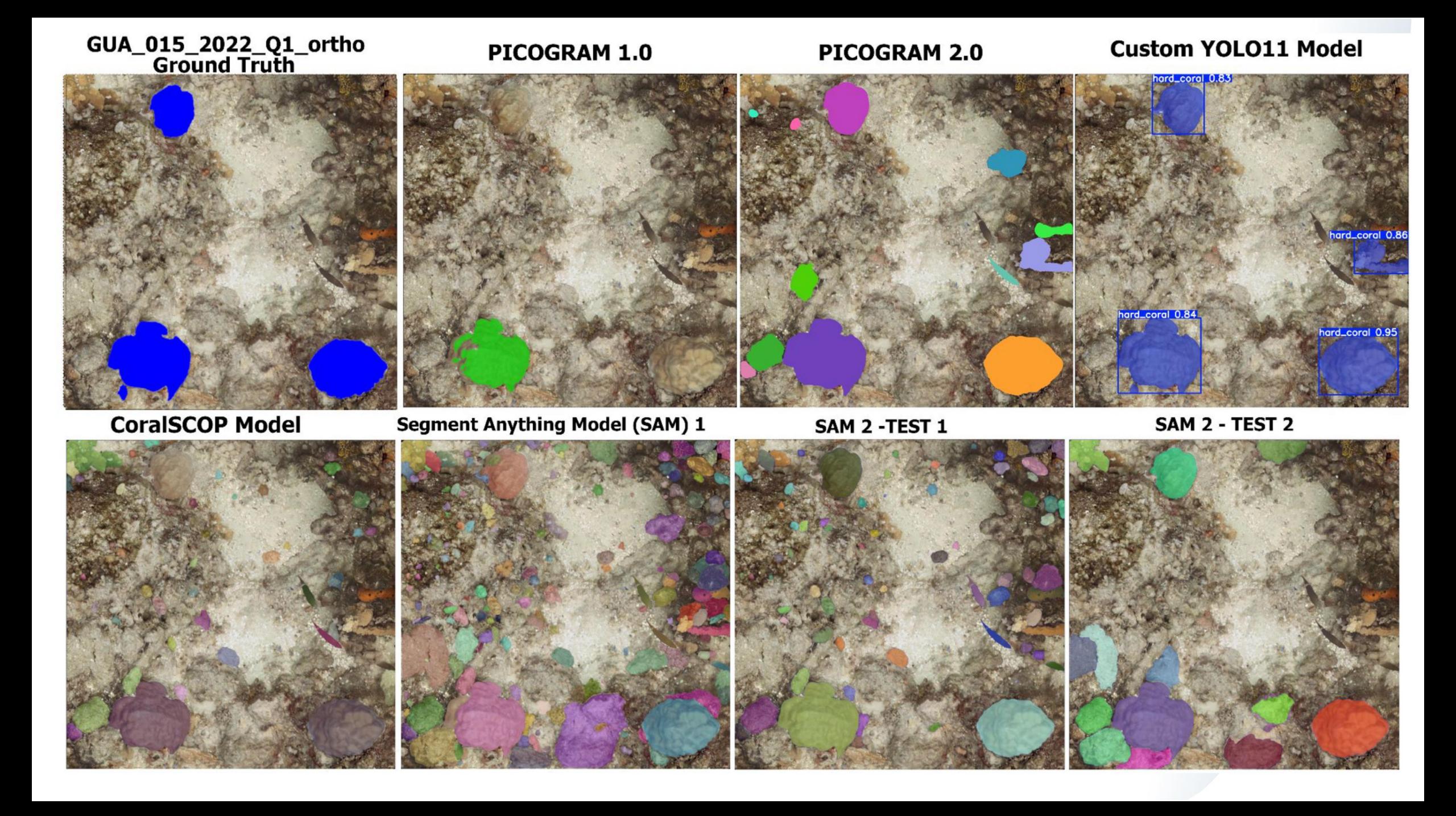
PICOGRAM 1.0



#### PICOGRAM 2.0







#### PRESS RELEASE

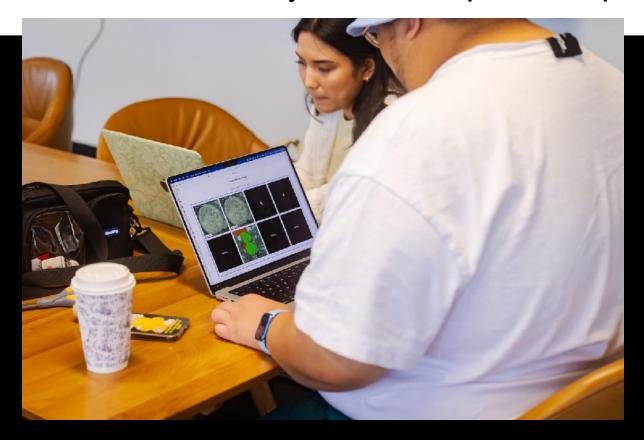


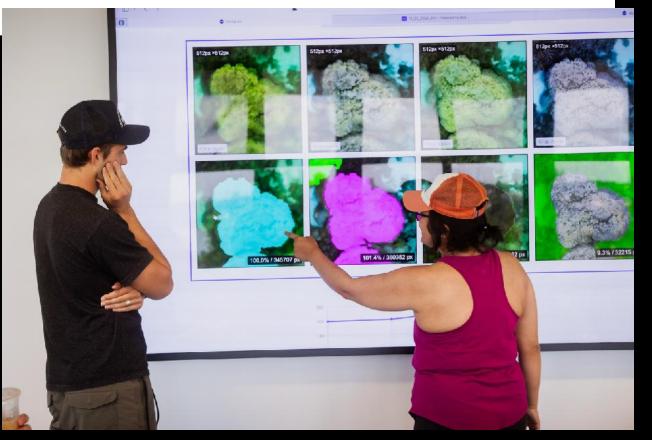
FOR IMMEDIATE RELEASE

#### University of Guam and University of Miami Partnership Use Guam Datasets to Monitor Coral Reef Changes using new NeMO-Net & PICOGRAM Tools

Beneath the surface of the Earth's oceans, coral reefs are undergoing significant changes. Transformations like coral bleaching are visible to the naked eye, while others like coral growth and mortality occur at the centimeter scale and would be much more difficult to measure over time. As these biodiverse and economically vital marine ecosystems face ongoing threats such as rising sea temperatures and human interference, researchers from the University of Guam (UOG) and the University of Miami's Rosenstiel School are working together to explore various solutions that could advance coral reef monitoring and mapping through machine learning tools.

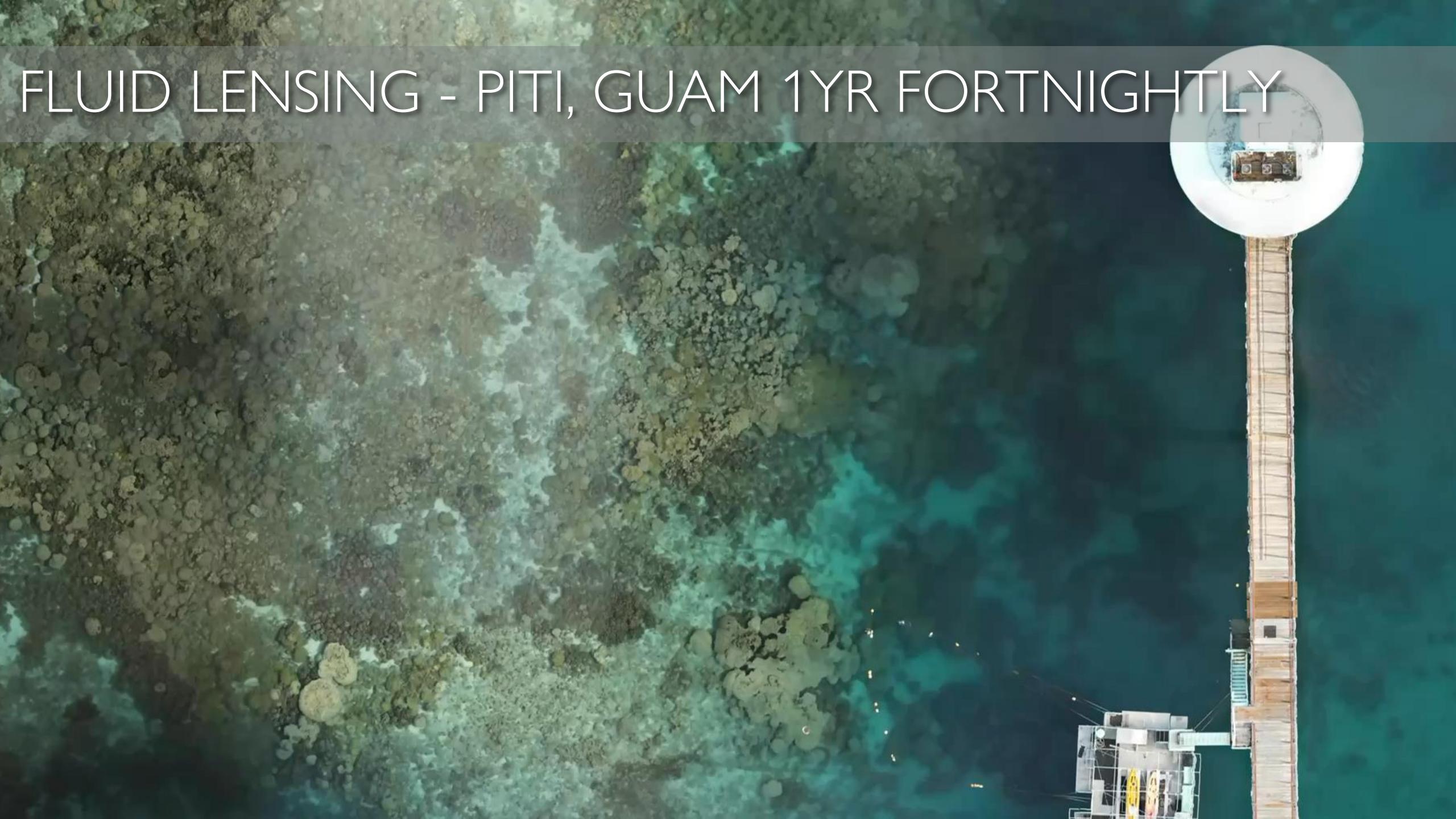
One collaboration is the *Prediction of Individual Coral Growth, Recruitment, and Mortality* (PICOGRAM) program, a computer-based tool in development by NASA & NOAA that uses machine learning to track coral colony size over extended time periods, supported by NASA's Ecological Forecasting Program and led by PI Prof. Sam Purkis at the University of Miami in partnership with NOAA.



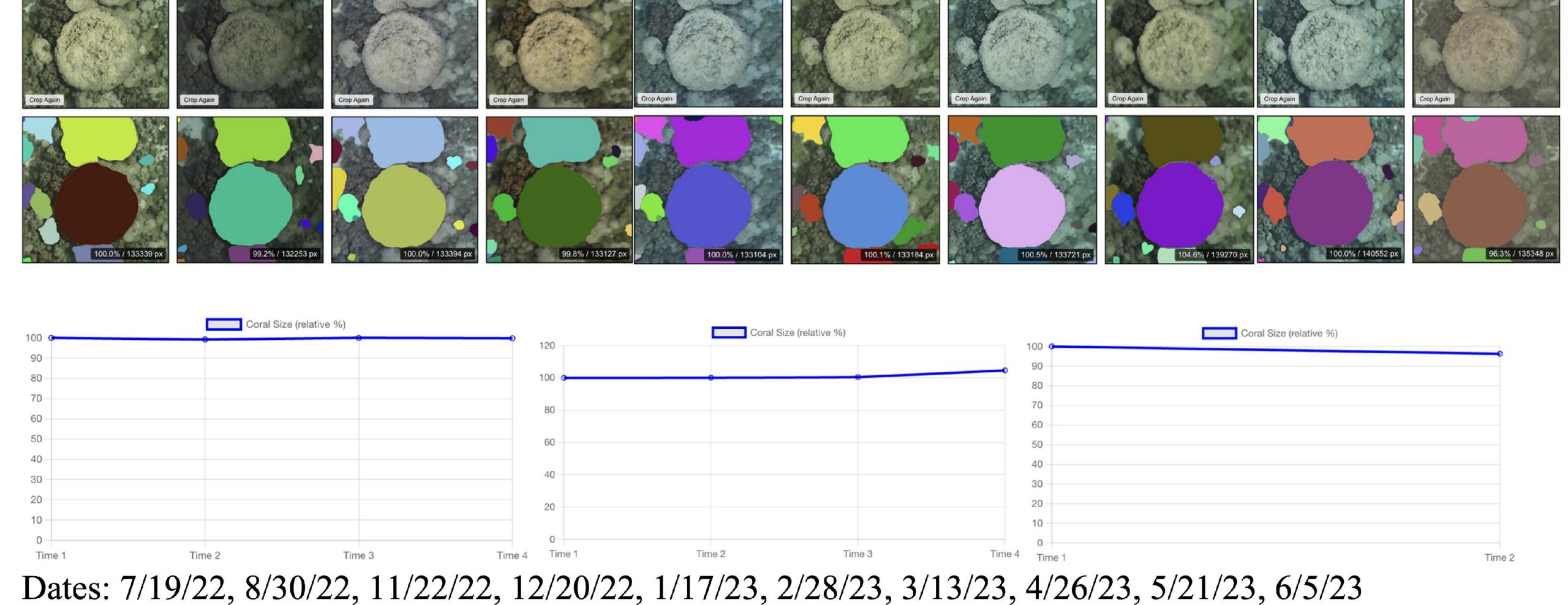








#### PICOGRAM COLONY CHANGES OVER 1 YR - AUTOMATICALLY DETECTED AND MEASURED





#### End-User Survey: NASA NeMO-Net, Fluid Lensing, NeMO-Net Data Viewer, & PICOGRAM

As part of our NASA-funded projects, we are collecting feedback from end users to better understand how project tools and data products are utilized in practice. Your insights will assist us in evaluating impact, enhancing future iterations, and ensuring long-term sustainability of these projects.

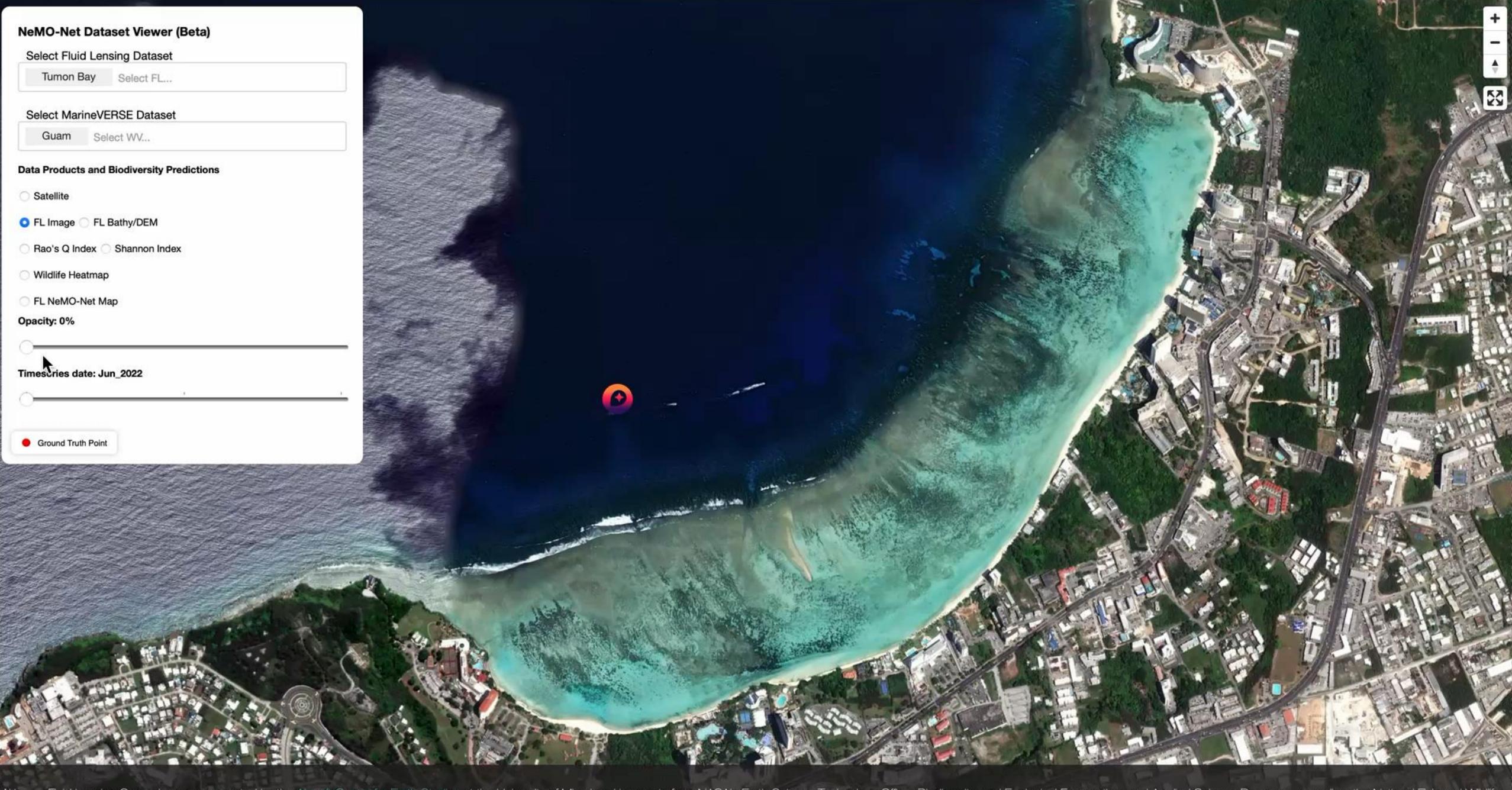
# Respondent Information Name Title Email

12:29 UNIVERSITY OF MIAMI ROSENSTIEL SCHOOL OF MARINE, ATMOSPHERIC, AND EARTH SCIENCE AIRCRAFT CENTER for EARTH STUDIES **End-User Survey: NASA NeMO-Net, Fluid Lensing,** NeMO-Net Data Viewer, & **PICOGRAM** As part of our NASA-funded projects, we are collecting feedback from end users to better understand how project tools and data products are utilized in practice. Your insights will assist us in evaluating impact, enhancing future iterations, and ensuring long-term sustainability of these

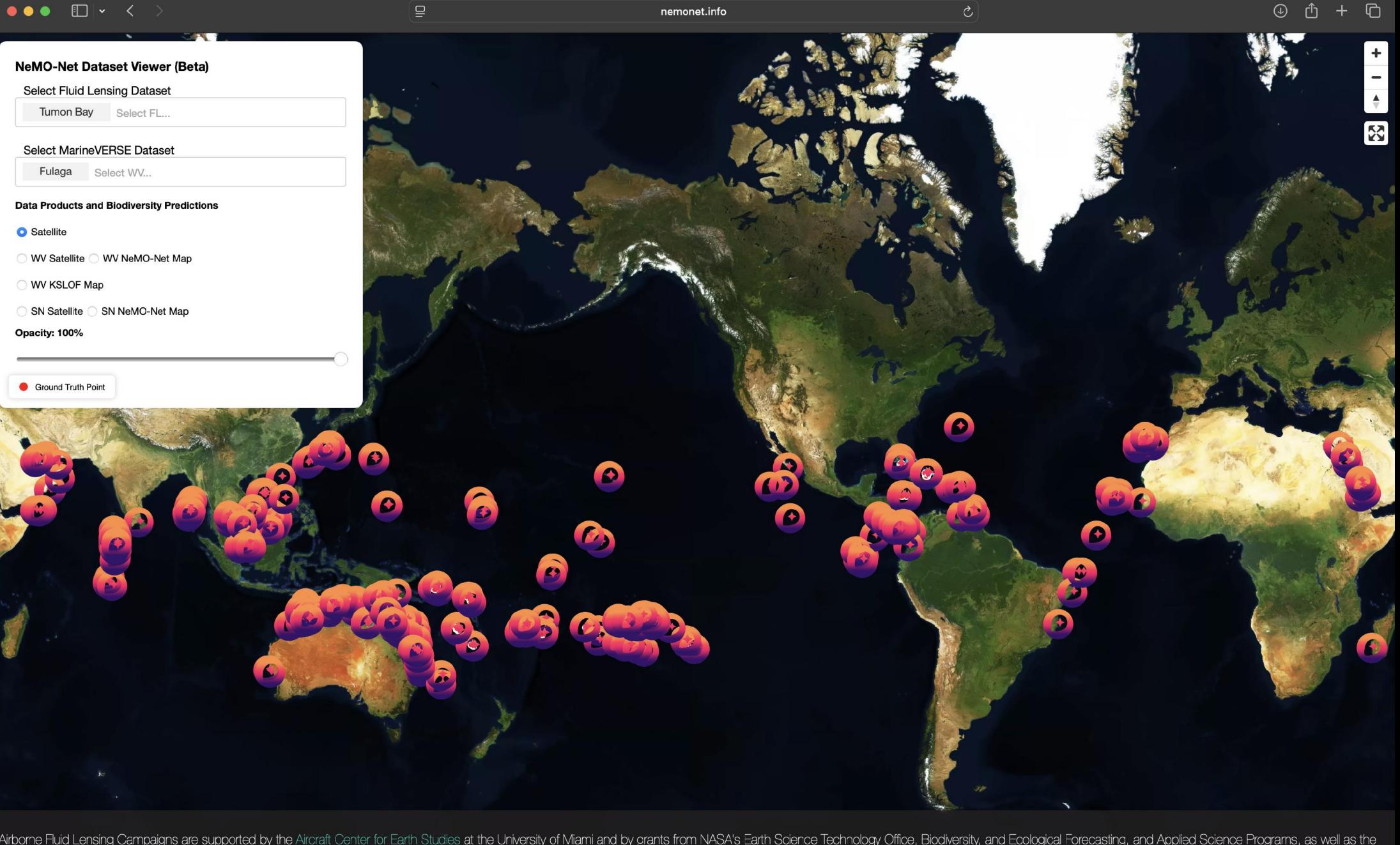
**Respondent Information** 

Name

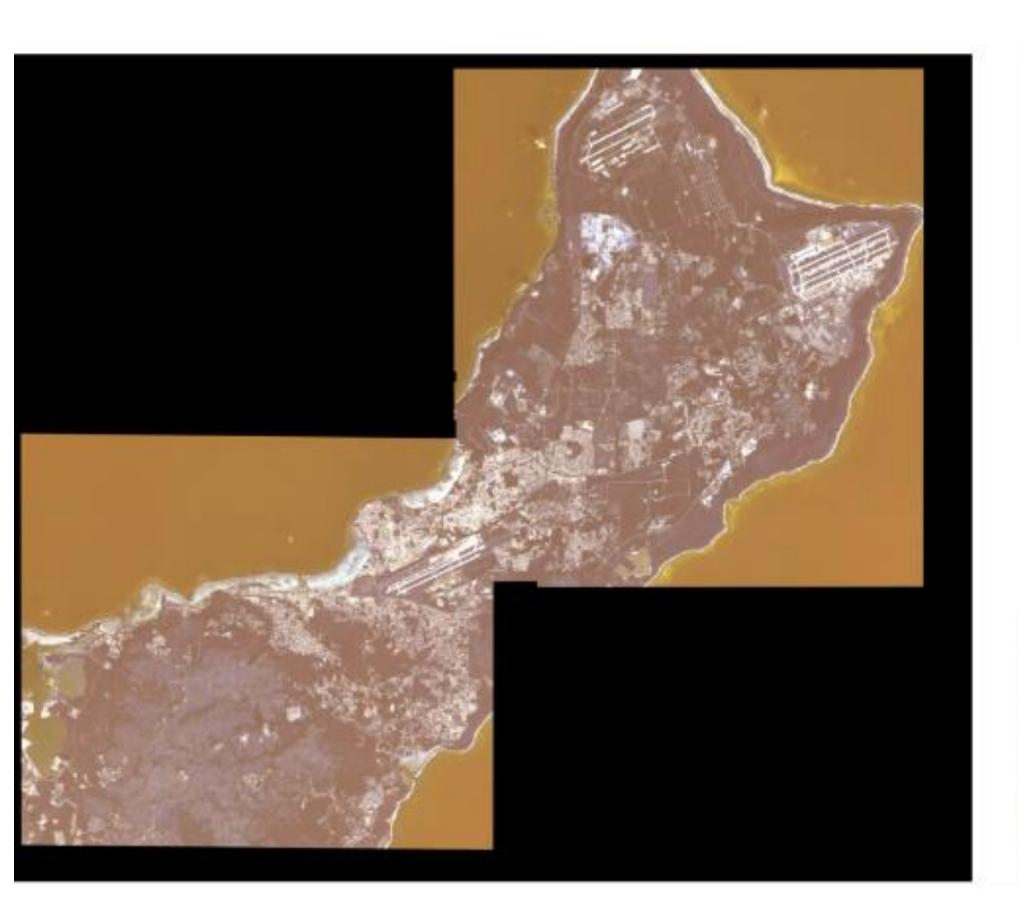
projects.

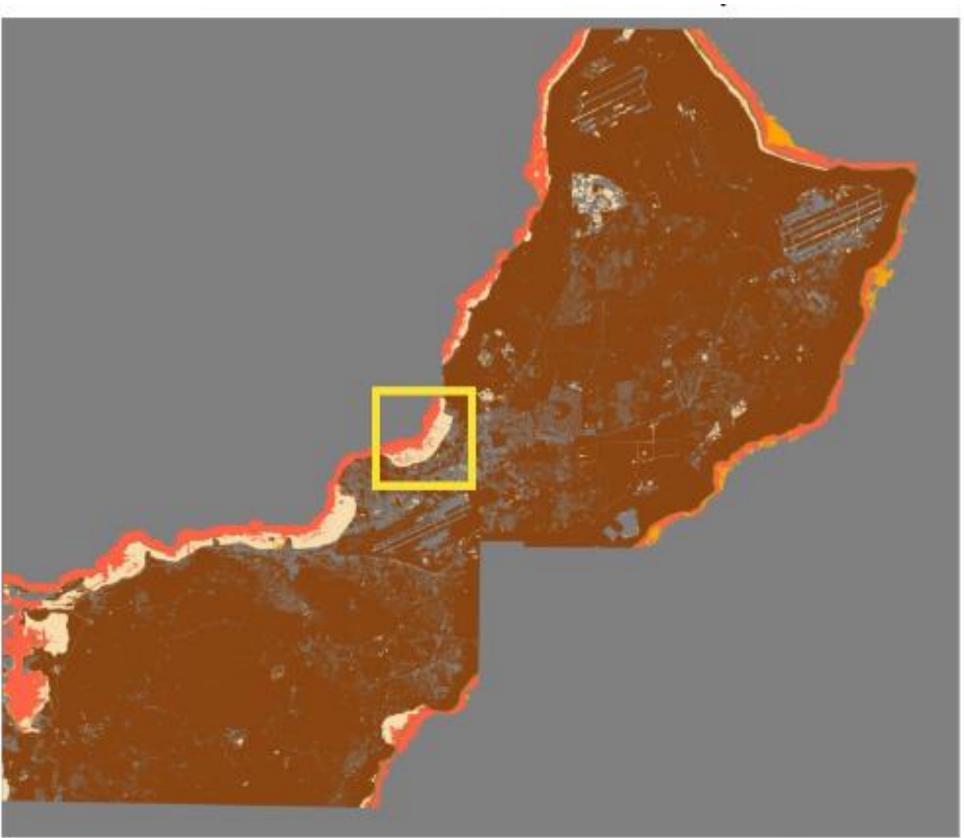


Airborne Fluid Lensing Campaigns are supported by the Aircraft Center for Earth Studies at the University of Miami and by grants from NASA's Earth Science Technology Office, Biodiversity, and Ecological Forecasting, and Applied Science Programs, as well as the National Fish and Wildlife Foundation (NFWF). Requests for datasets may be submitted at the ACES website.



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#### Classes

Coral

Coral Fore Reef

Reef Crest - Coralline Algal Ridge

Algae

Seagrass

Sediment & Rubble

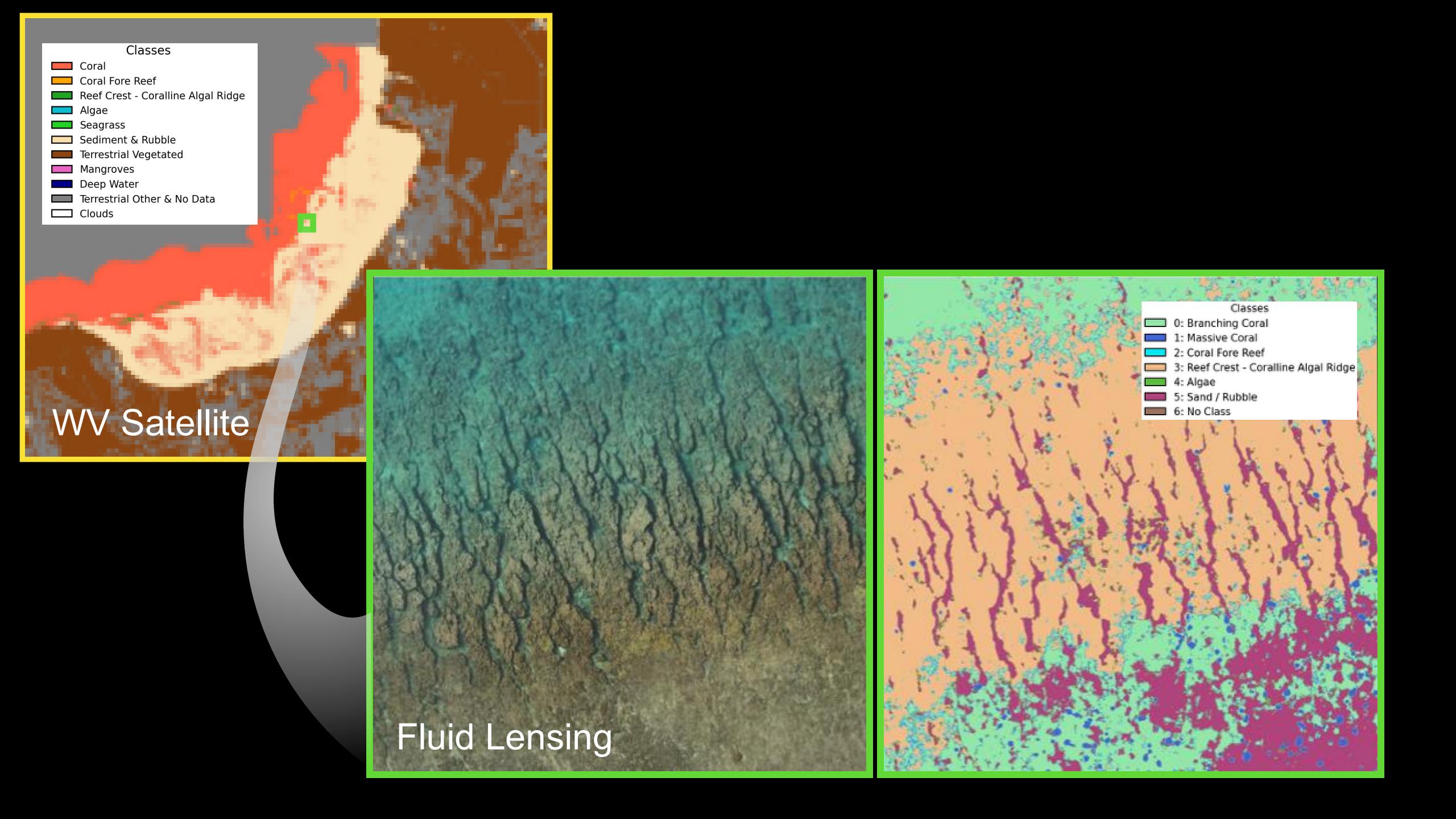
Terrestrial Vegetated

Mangroves

Deep Water

Terrestrial Other & No Data

Clouds



#### THANK YOU!

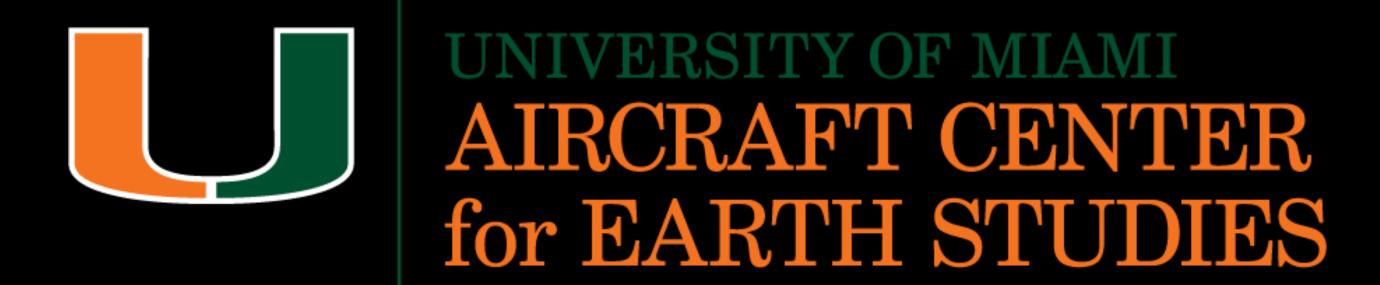


#### ACES.EARTH.MIAMI.EDU



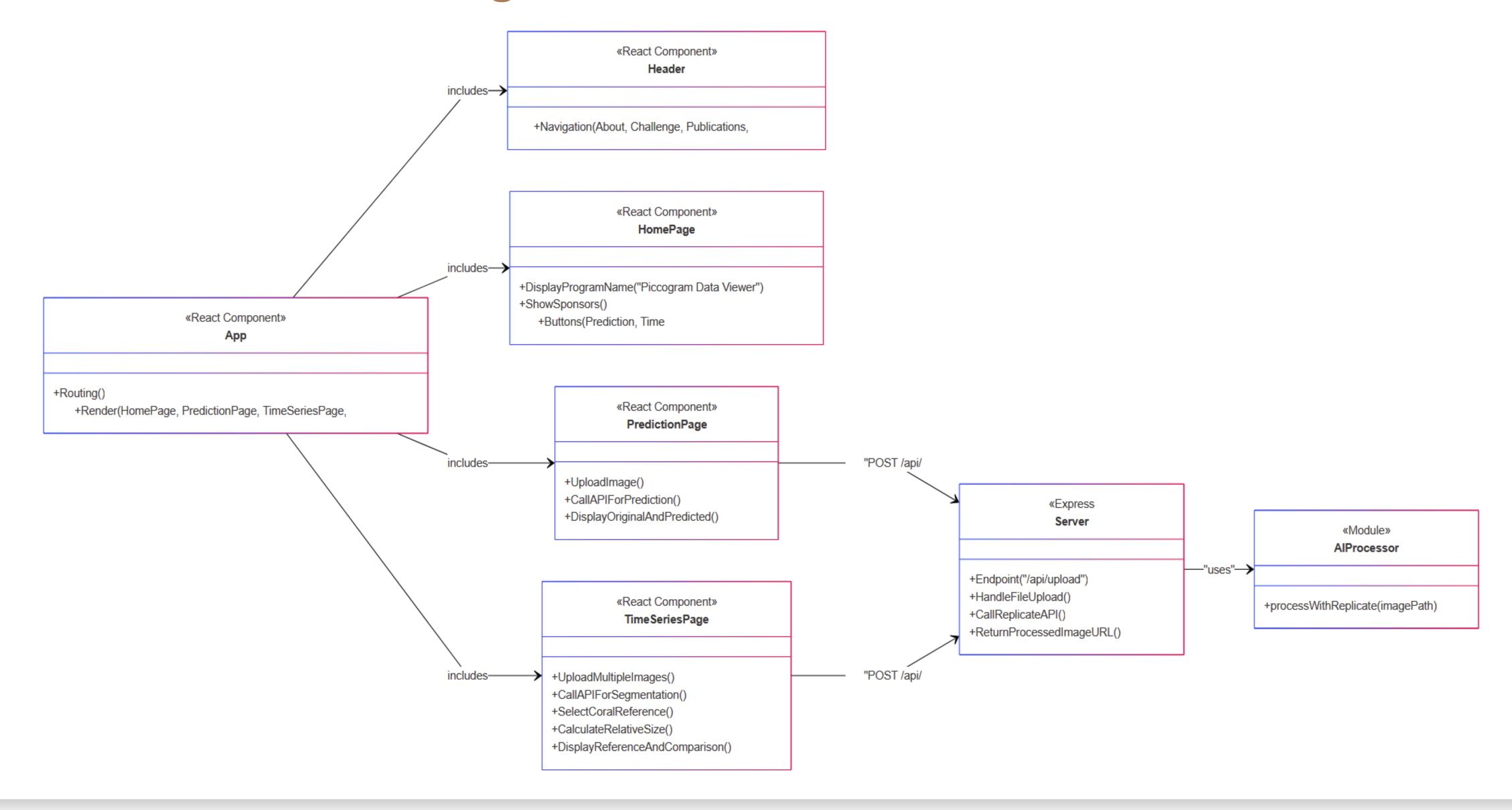


O' @ACESUMIAMI



### BACKUP SLIDES

#### Class Diagram for PICOGRAM Data Viewer





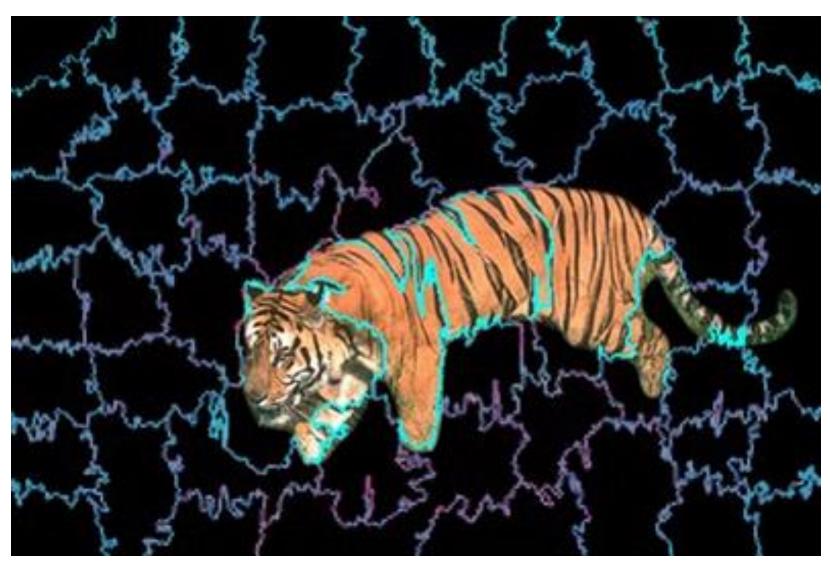
#### UNSUPERVISED IMAGE SEGMENTATION BY BACK-PROPAGATION

#### A new approach with NeMO-Net CNN

After applying SLIC (Simple Linear Iterative Clustering) segmentation and setting compactness, the algorithm uses a CNN to extract feature representations directly from the input image without labeled data.

The CNN then assigns initial labels to pixels by selecting the highest activation across its feature maps, creating an initial segmentation map. This leverages the CNN's ability to learn patterns in an unsupervised manner, providing a foundation for iterative label refinement in subsequent steps.





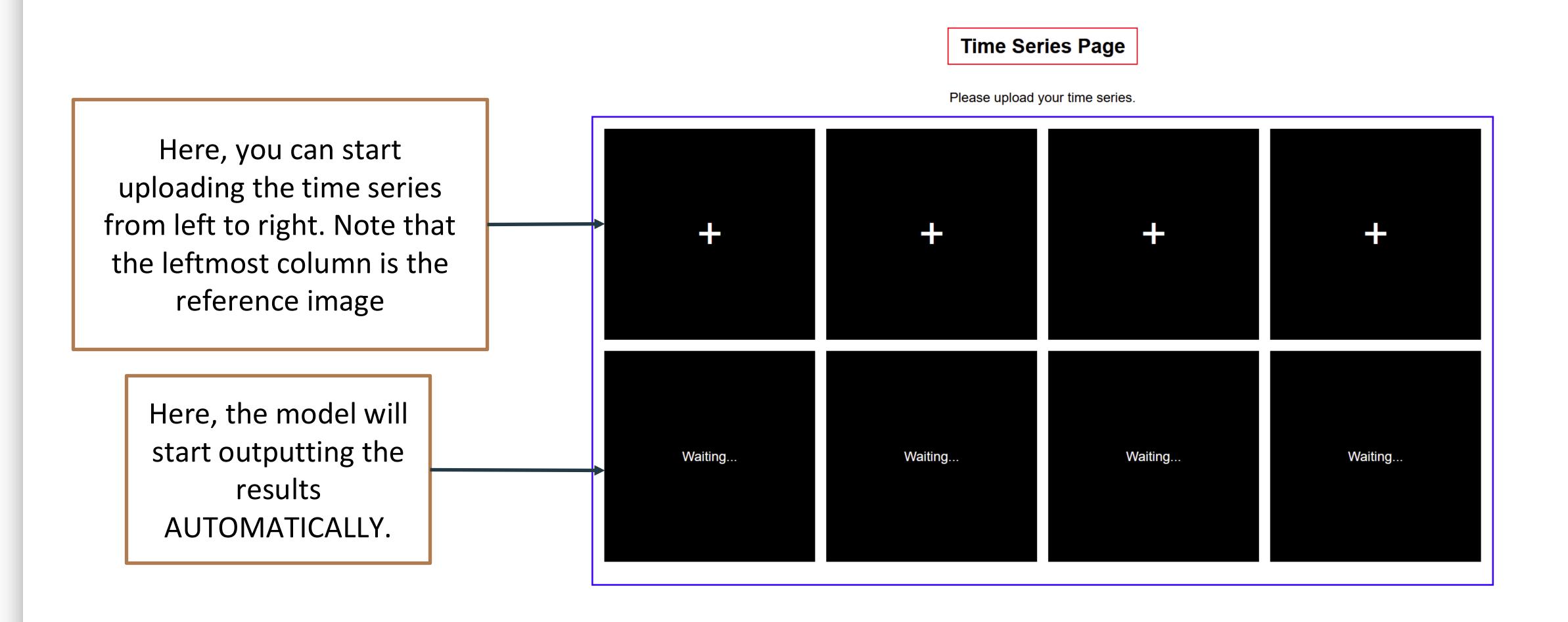
Sample SLIC segmentation

#### TESTING ON TIME SERIES DATA AND SPECIFIC COLONY TYPES

Original Patches 

Segmentation Results

#### Time Series Page



#### PICOGRAM 2.0 vs SAM 2.1 Model Comparisons



