

Bottom Albedo Images to Improve Classification of Benthic Habitat Maps

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Outline

- Introduction
 - Study Area
 - Challenges
 - Sensors
(Active/Passive)
- Bottom Albedo Images
- Benthic Habitat Map of La Parguera Reserve
- Conclusions



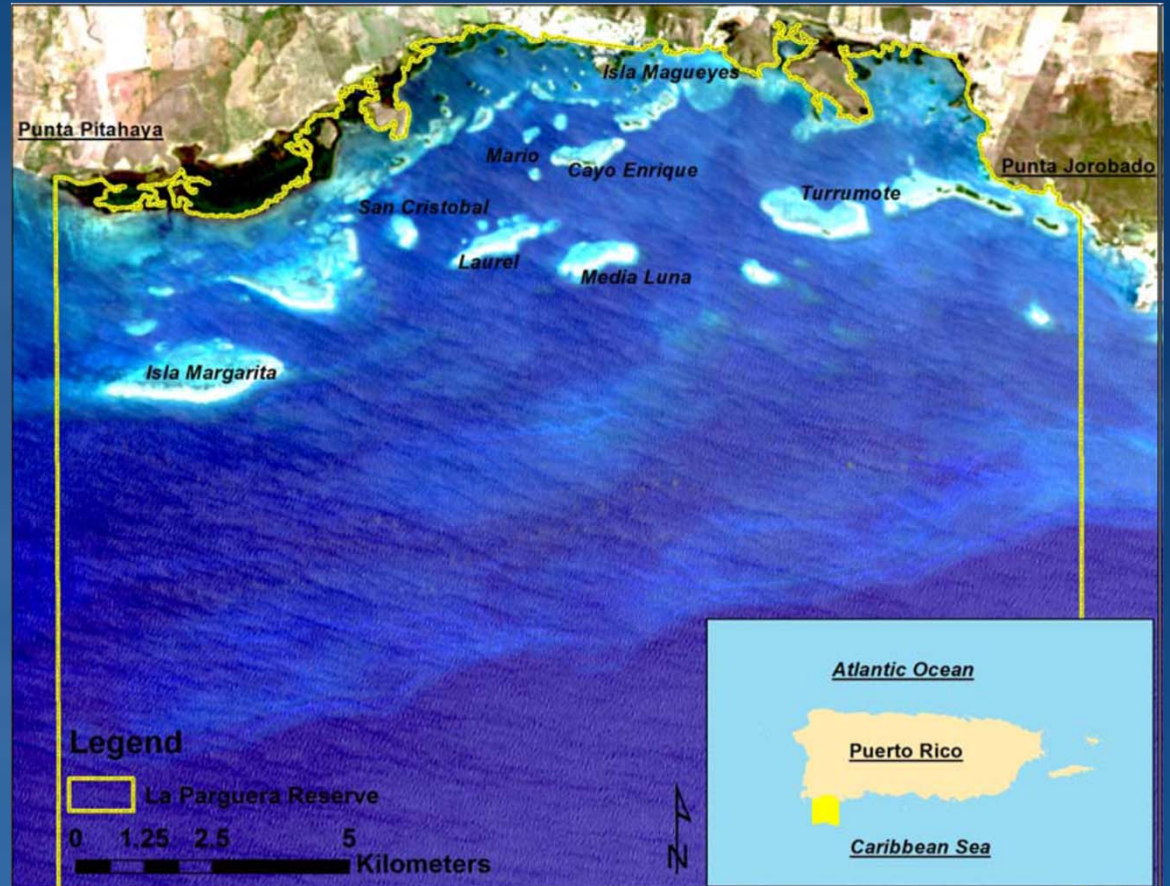
Introduction

Coastal areas

- Important resources
- Ecosystems affected by human-based and natural factors.

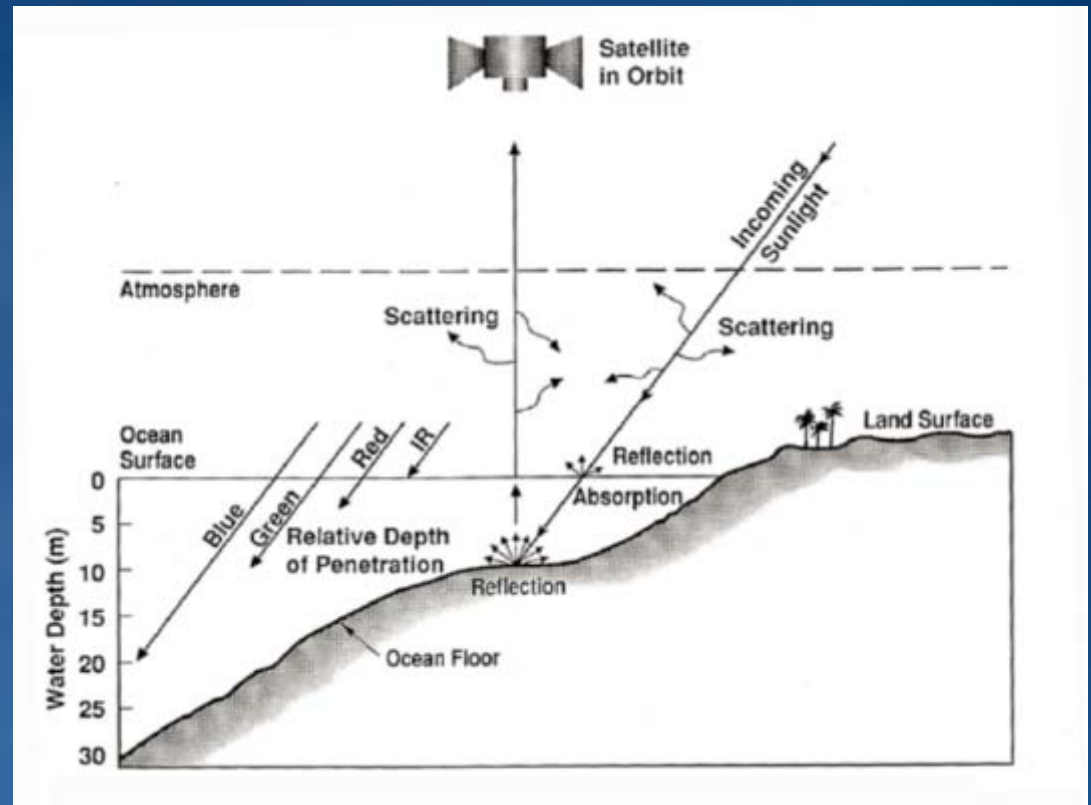
Study Area

- La Parguera
 - DNR Natural Reserve
 - Aprox. 12,500 acres
 - Depth = ~18 m
 - Unique habitats



Challenges

- **Use of Remote Sensing Techniques**
- **Depth**
 - Variable
 - Detection limits
- **Variable substrate**
 - Consolidated
 - Unconsolidated
- **Turbidity**
 - attenuation



(Purkis 2005)

Sensors

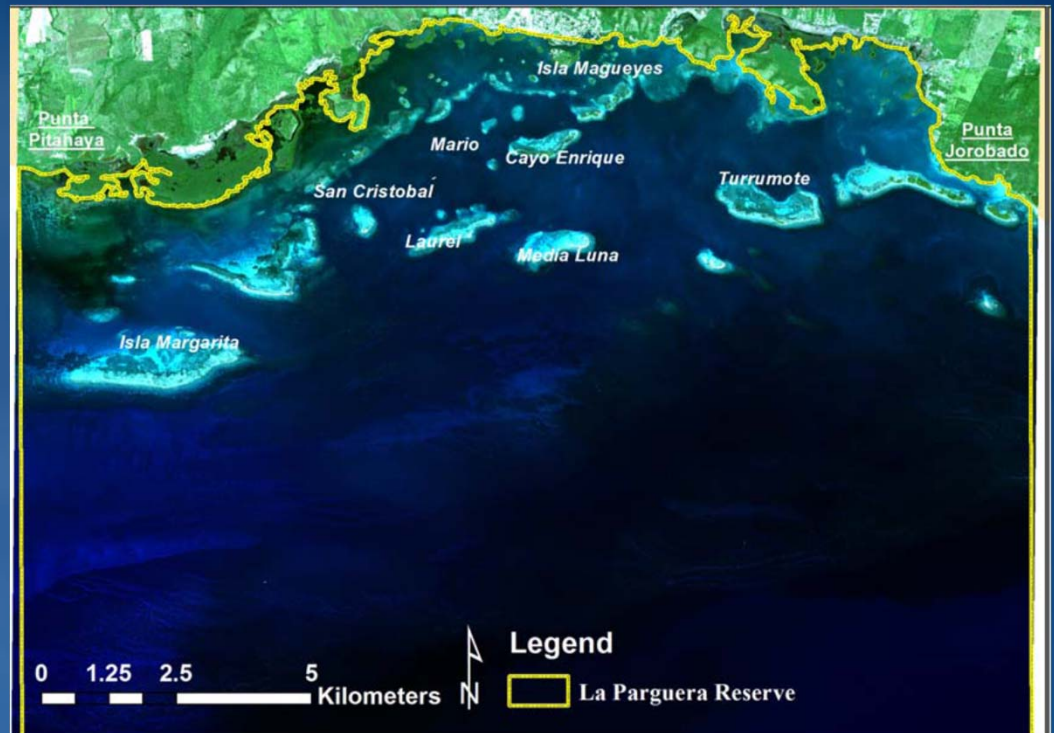
- AVIRIS (Airborne Visible Infrared Imaging Spectrometer)
- December 2005
- 224 Bands (370-2500 nm)
- Hyperspectral
- Visible range: 400-700 nm (32 bands)
- 10 nm bandwidth
- High signal to noise ratio (~1,000:1)
- Spatial resolution: ~3m



AVIRIS mosaic

Sensors

- Worldview 2 (WV2)
- December 2011
- 8 bands, 5 visible
- Multispectral
- ~2 m spatial resolution
- “Coastal band” (425nm)

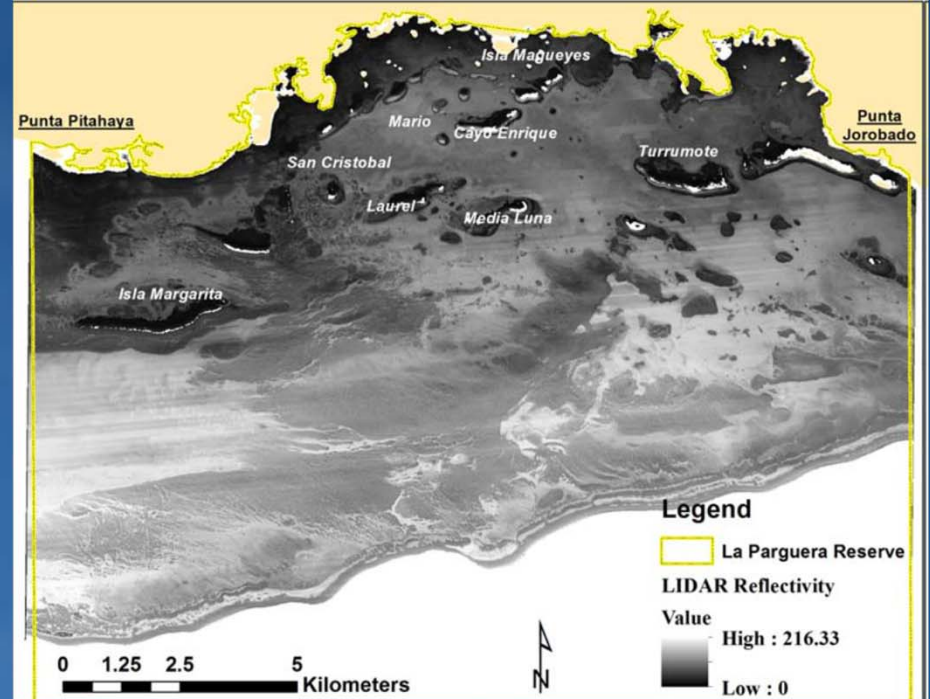
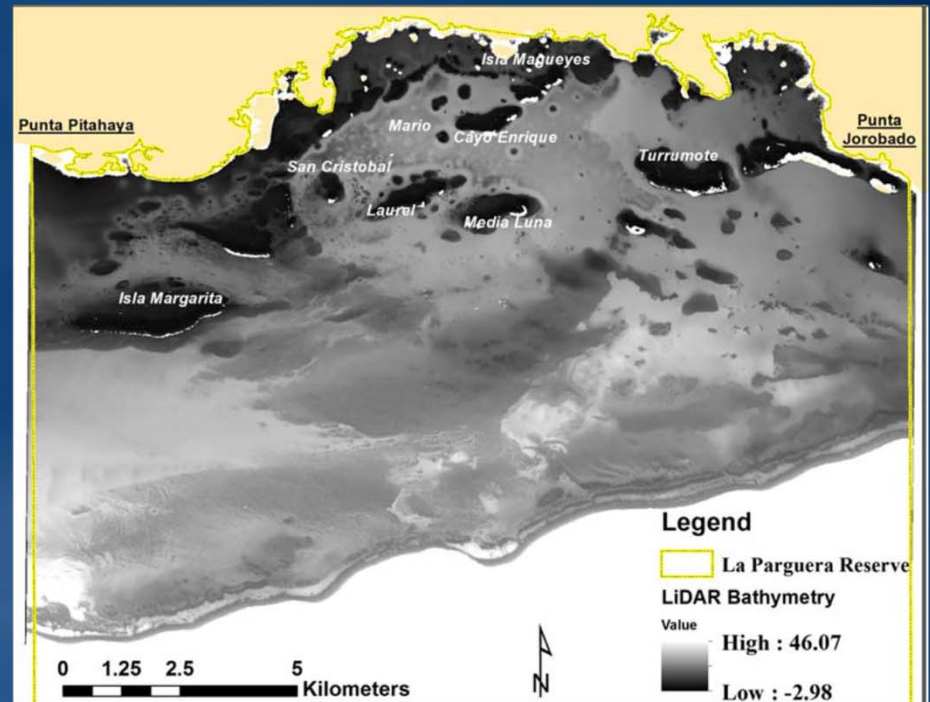


WV2 Imagery

Sensors

- **LiDAR SHOALS**

- 2006
- (LADS) Mk II Airborne System.
- Infrared beam (1064 nm)
- Green beam(532nm)
- 4 x4 meters bathymetry surface
- 5x5 meters intensity surface



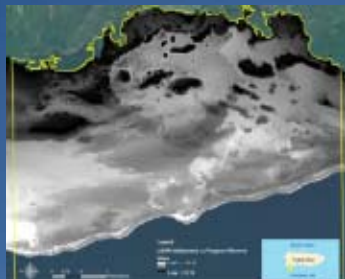
METHODS



AVIRIS image



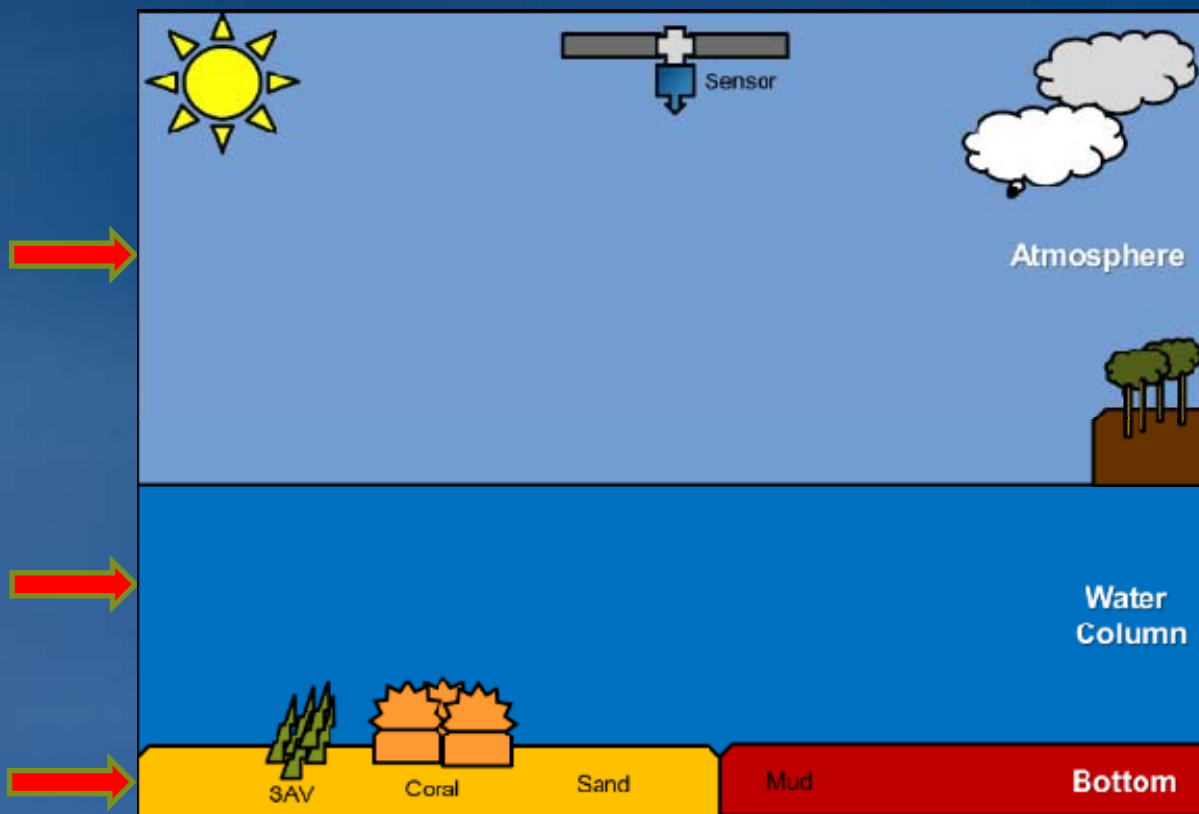
WV2 image



LiDAR SHOALS



Bio-optical sampling

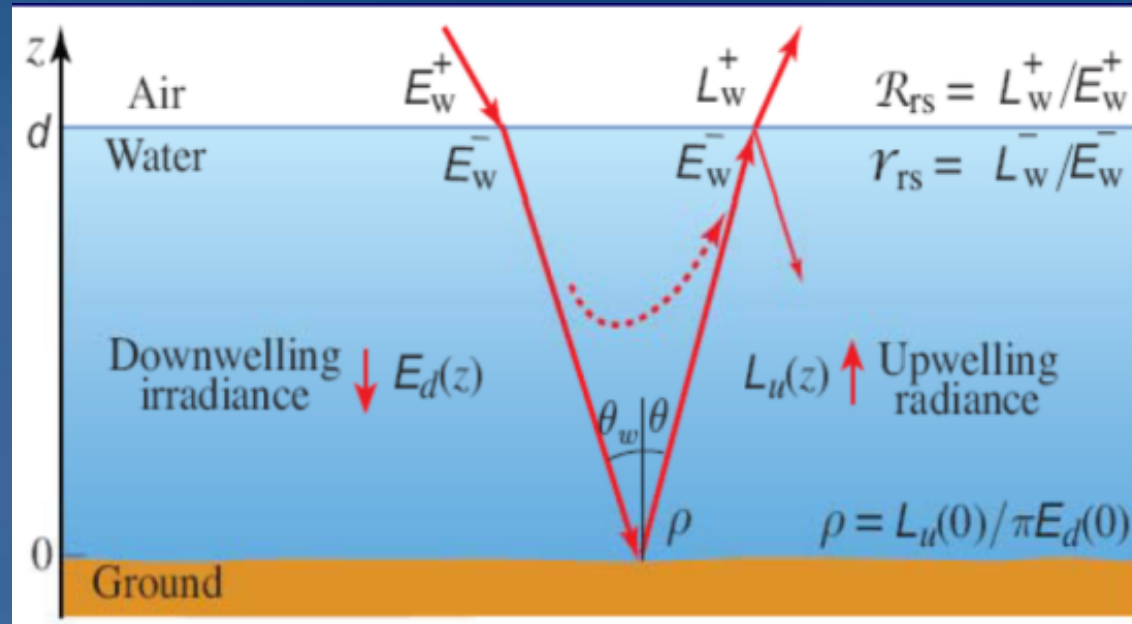


Pre-processing Steps (ArcGIS 10.3 and ENVI 5.2)
co-registration,
landmask,
atmospheric correction

Bottom Albedo Map from water column correction (ArcGIS 10.3)
Benthic Habitat Map of La Parguera Reserve (ArcGIS 10.3)

Water Column Correction Bottom Albedo Images

- $R_{rs}(\lambda) = R_{rs}^w(\lambda) + R_{rs}^b(\lambda)$ (Lee et al. 1994, 1999)

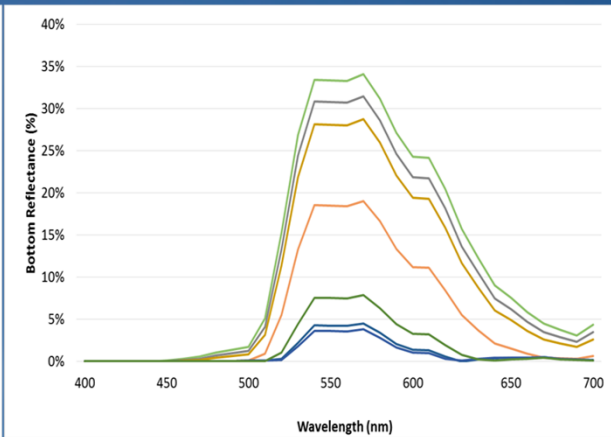
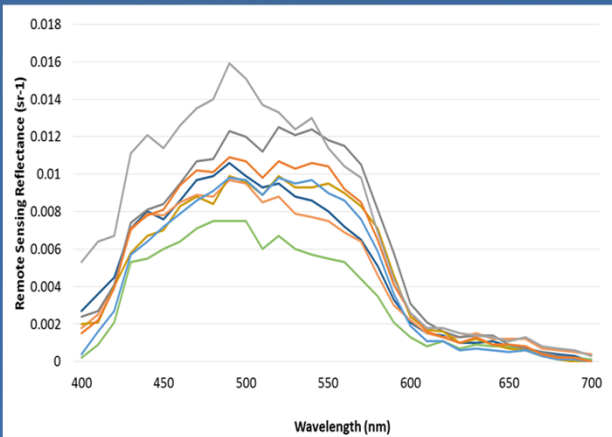
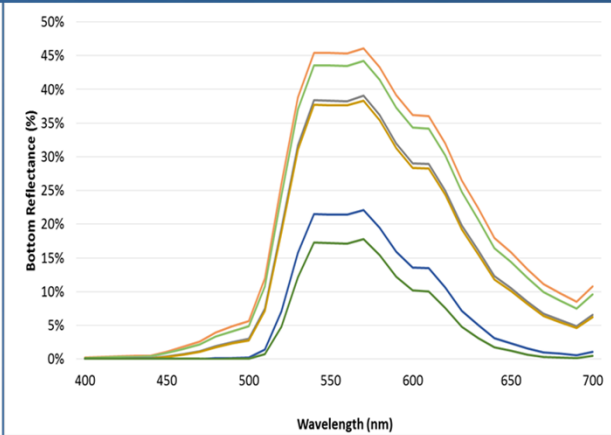
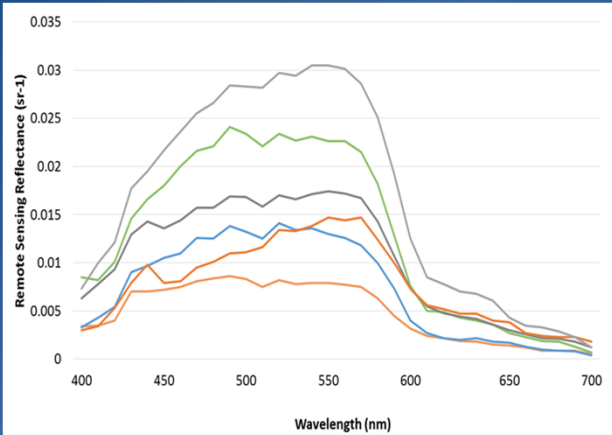
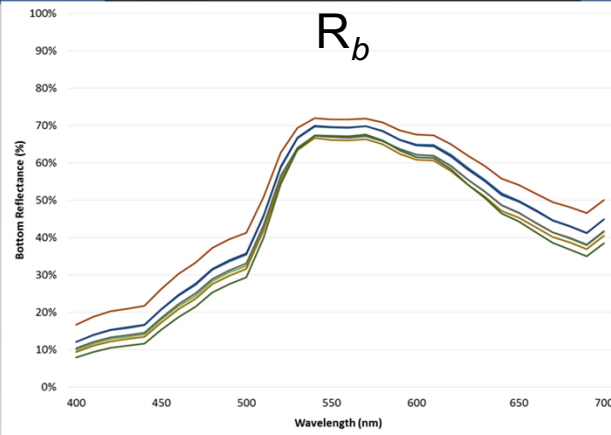
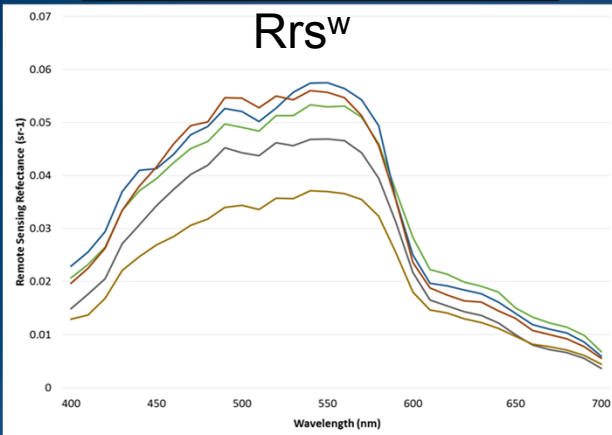


- $R_{rs}^w(\lambda) \approx 0.05 [b_b(\lambda) / a(\lambda) + b_b(\lambda)] [1 - e^{-3.2[a(\lambda) + b_b(\lambda)]Z}]$
- $R_{rs}^b(\lambda) \approx 0.173 \rho(\lambda) e^{[-2.7a(\lambda) + b_b(\lambda)]Z}$

AVIRIS

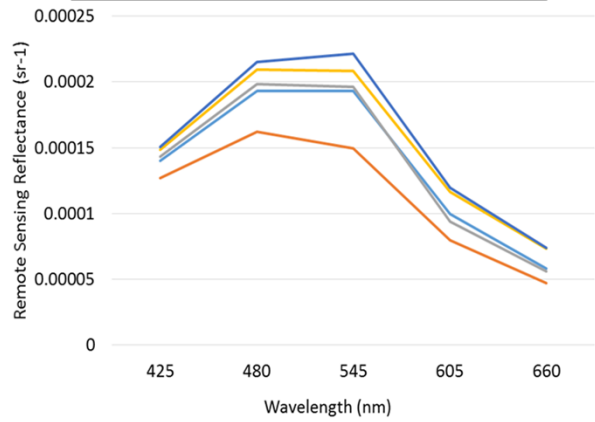
Before Water Column Correction

After Water Column Correction

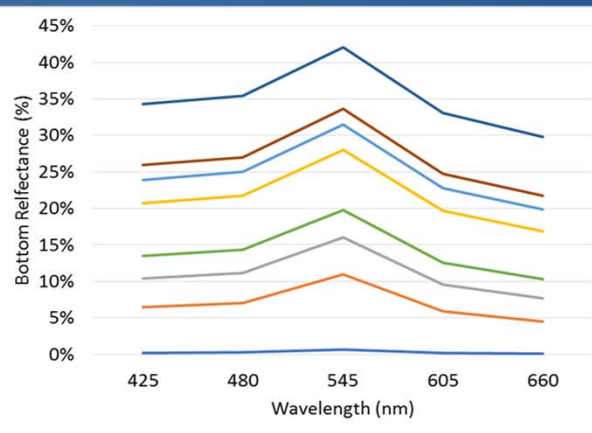
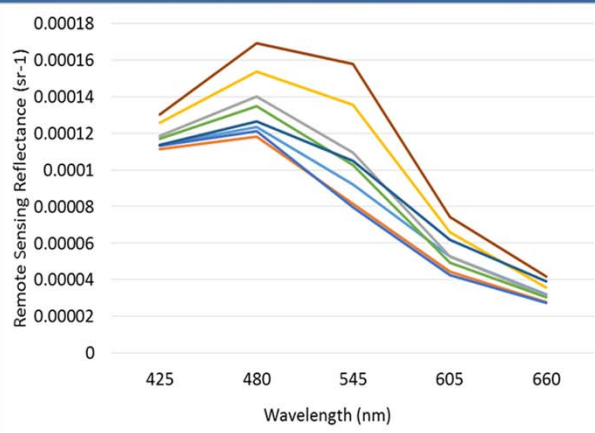
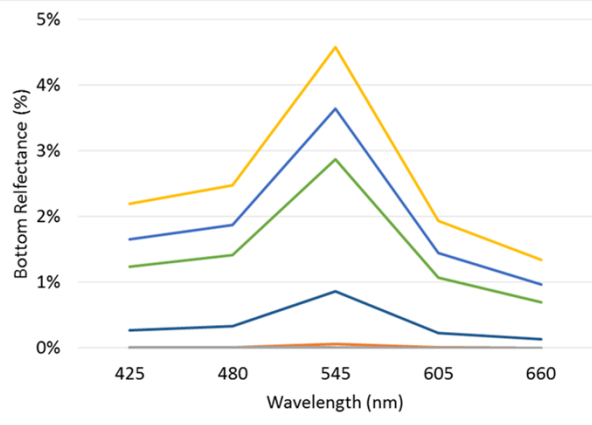
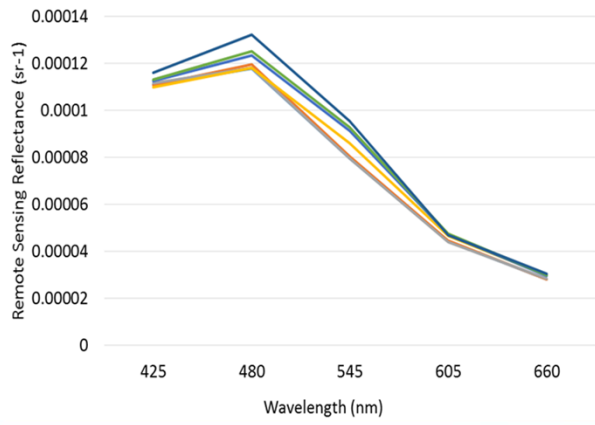
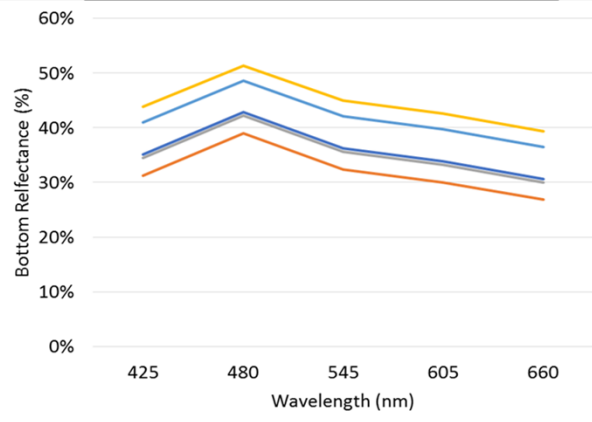


WV2

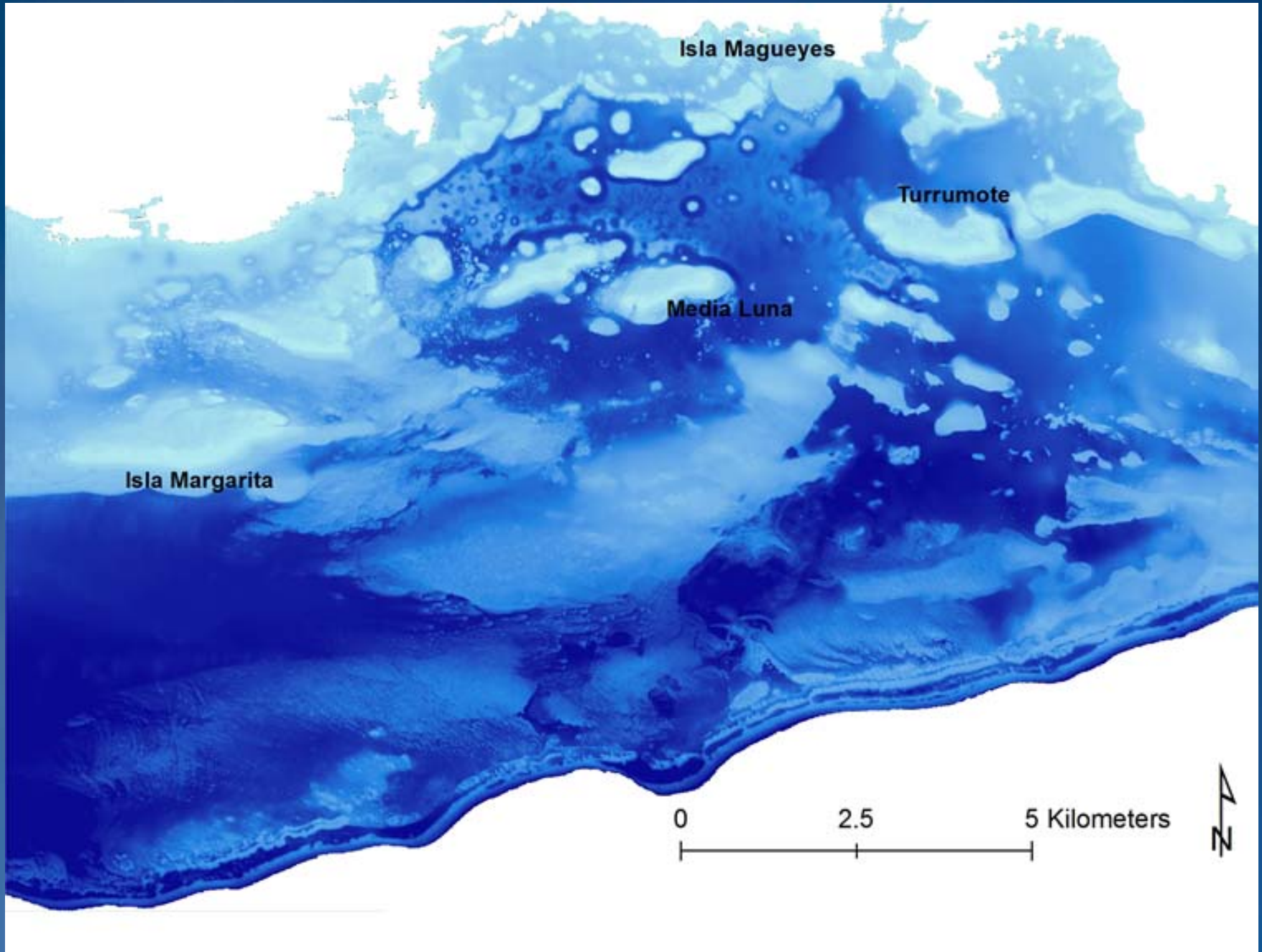
Before Water Column Correction



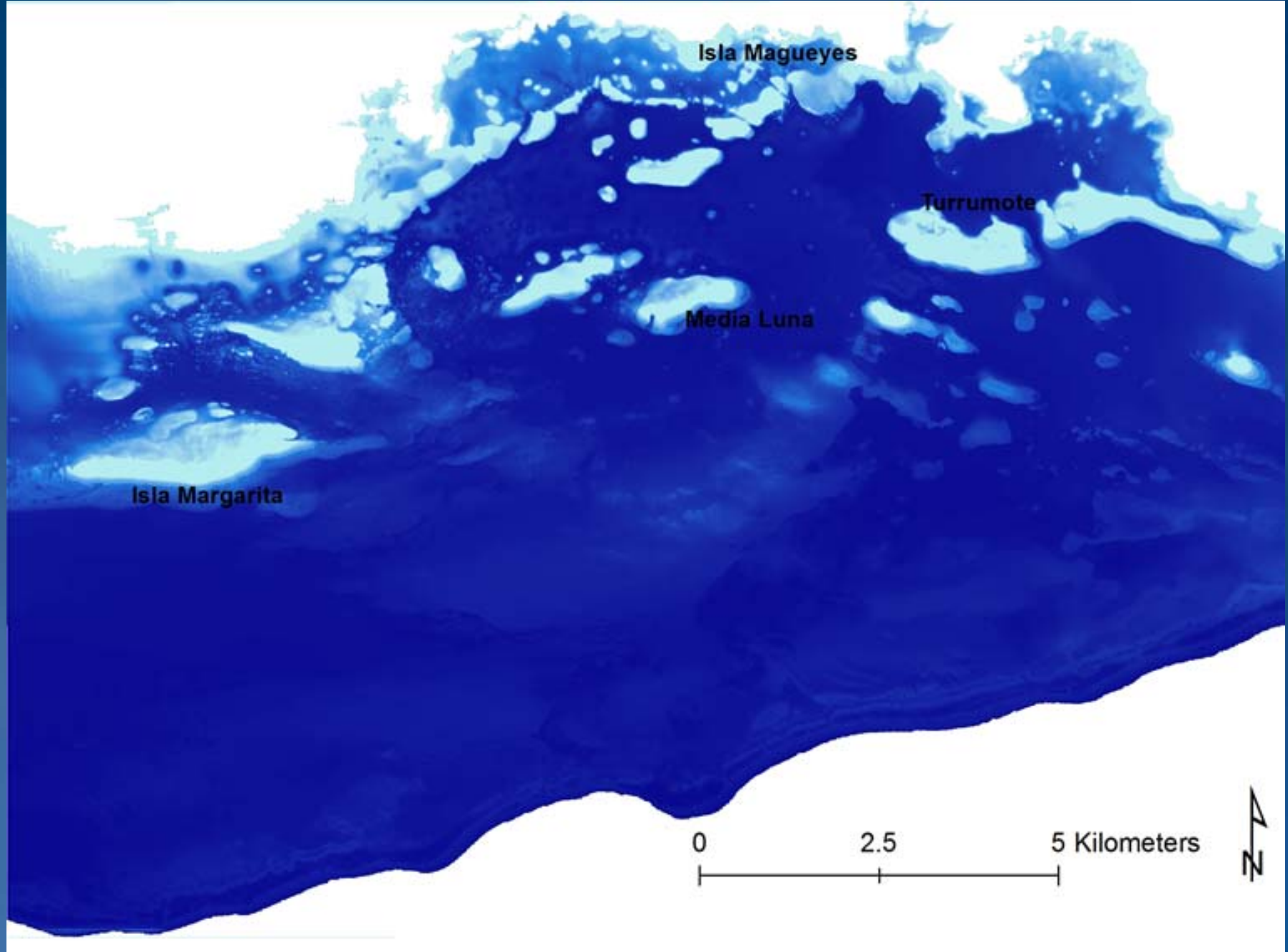
After Water Column Correction



AVIRIS Bottom Albedo Image



WV2 Bottom Albedo Image



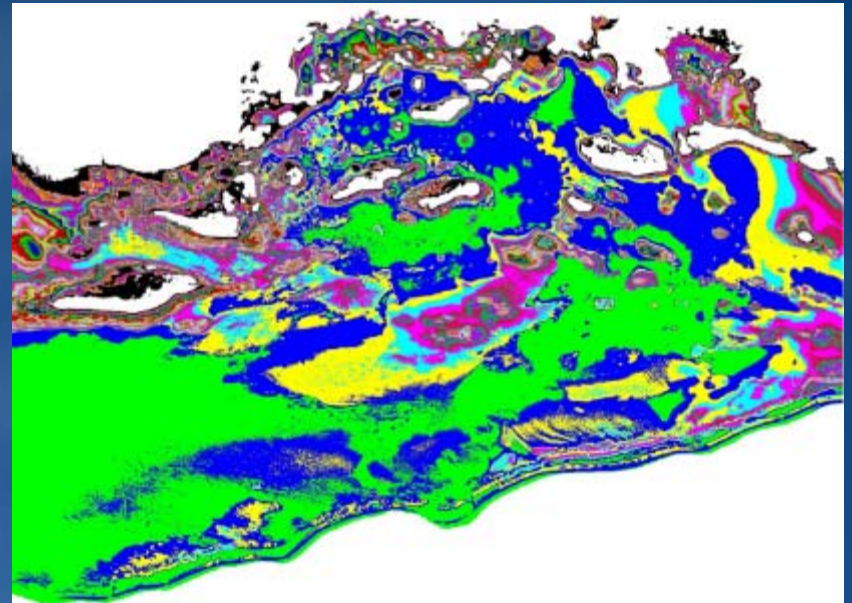
Benthic Habitat Mapping

- **Goals**

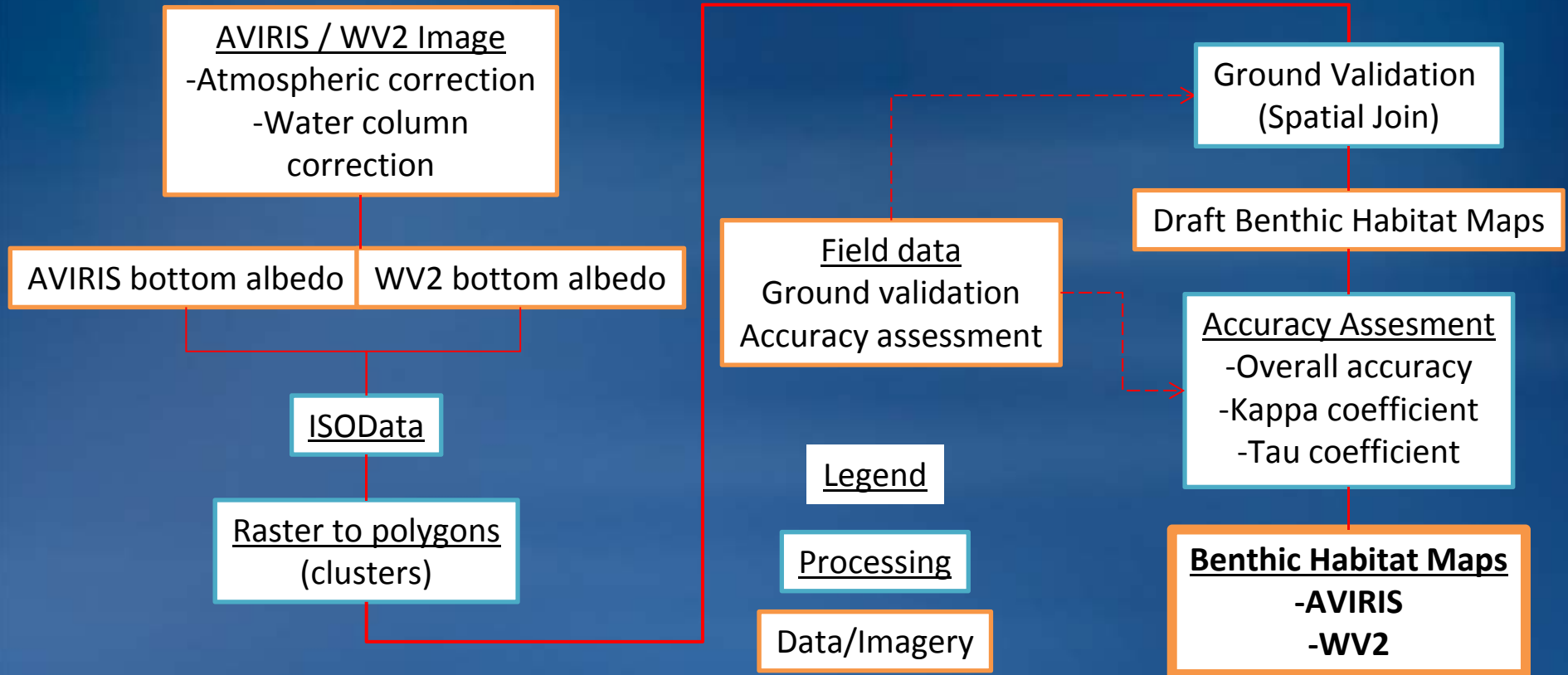
- Develop a high-resolution benthic habitat map
 - AVIRIS and WV2 modeled bottom albedo

- **Importance**

- Identify ecologically important habitats in La Parguera
- Coral reef habitat maps as a management tool.



Methods



Segmentation

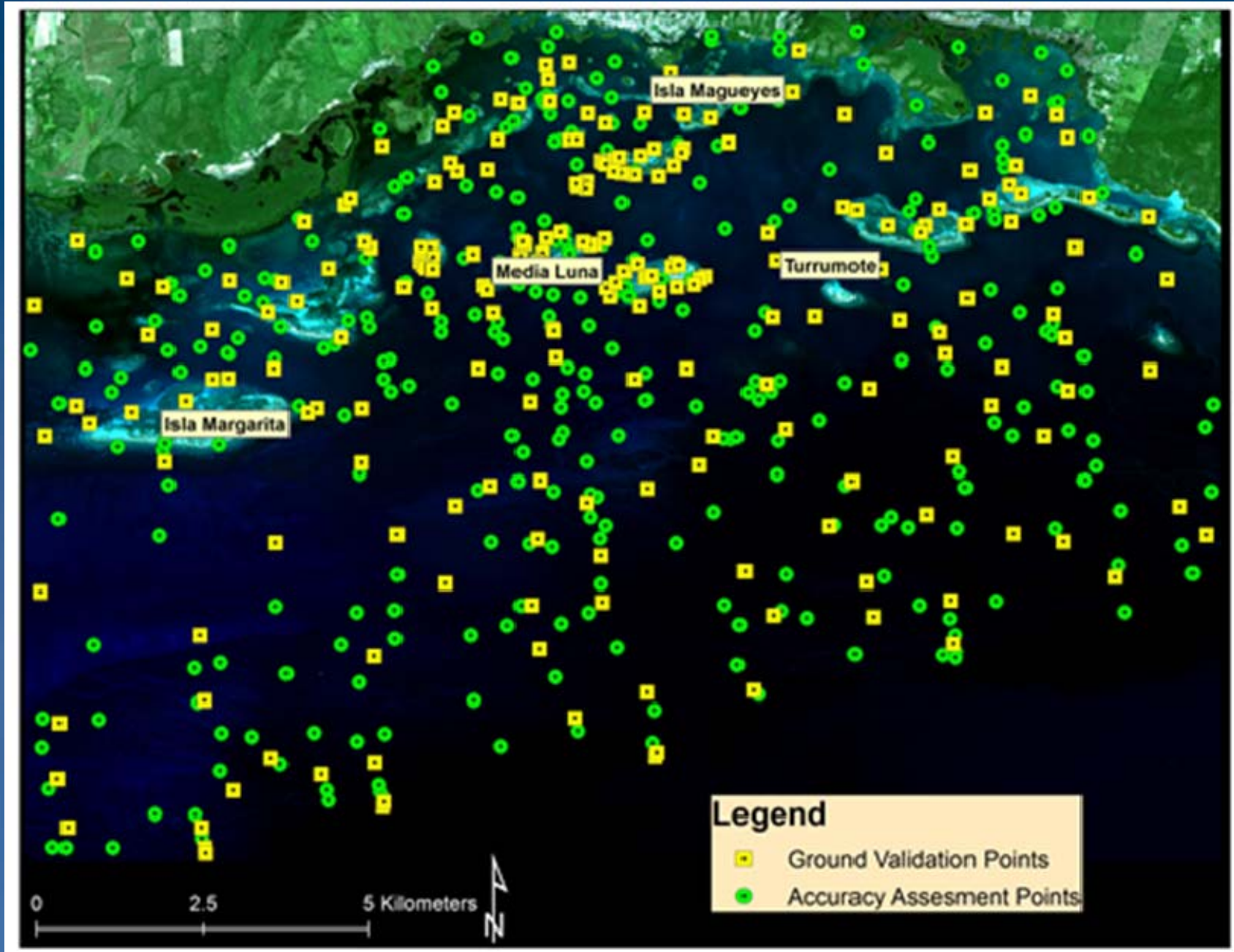
- **The Iterative Self Organizing Data (ISODATA) algorithm**
 - **unsupervised classification**
 - **statistical clustering algorithm**
 - **Various iterations and combinations of maximum clusters were evaluated.**
 - **150 clusters with 5 iterations were selected as the maximum for the ISODATA classification**
 - **spectral clusters that appeared to belong to multiple class / benthic habitat (confused pixels) were also identified.**

Sampling Sites

- **Delta Vision Pro**
 - Drop Camera HD Video (1080p)
 - 10-second video collected
 - DVR
- **Trimble Juno GPS**
 - 10-second averaging
 - dGPS
 - 2 meters
- **Synchronized GPS and video**



Ground Validation and Accuracy Assessment Points



Benthic habitat classification scheme

(1) Coral Reefs



(2) Seagrass



(3) Hardbottom



**(4) Mix: Sand/
Hardbottom/Coral**



(5) Mud



(6) Sand

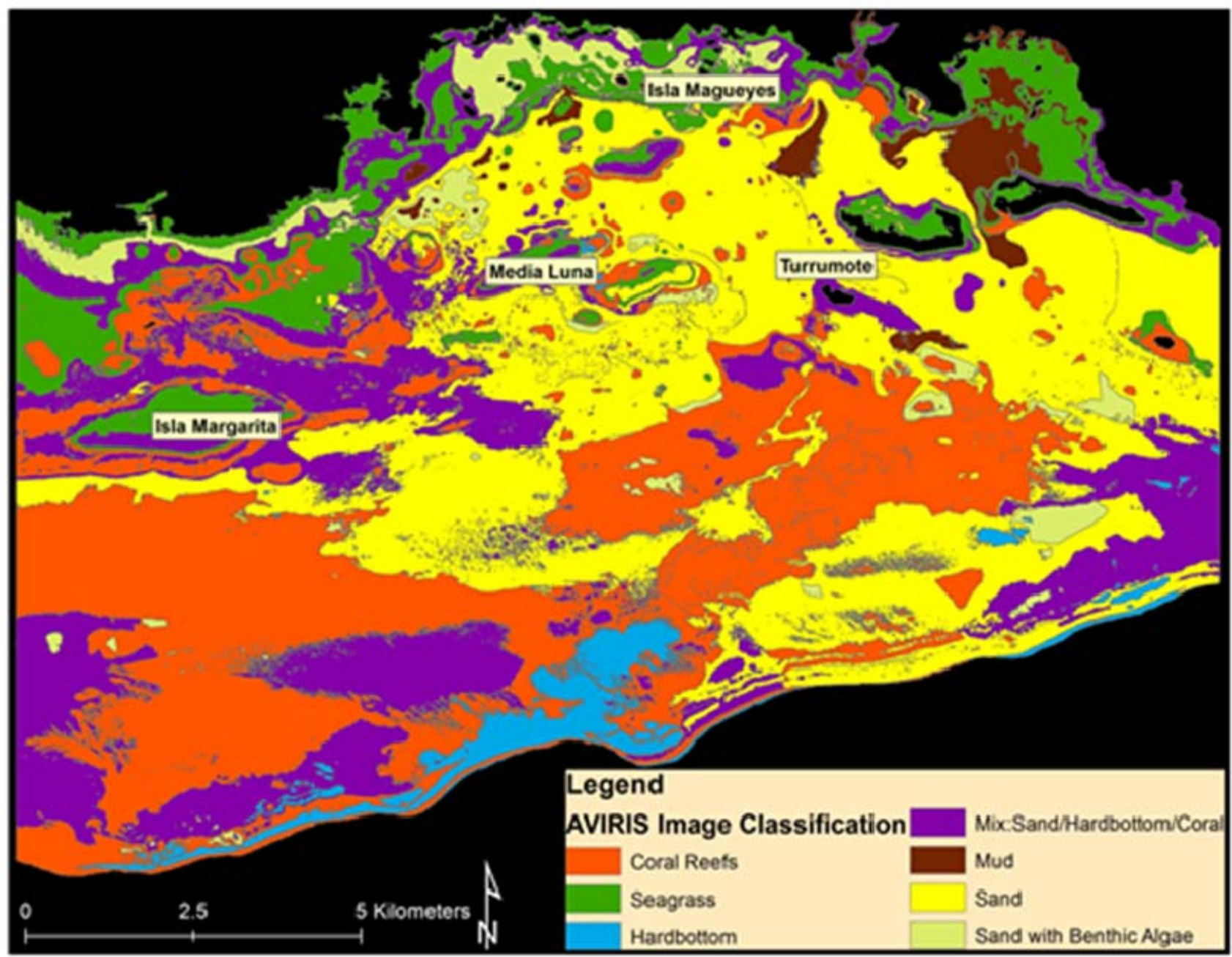


**(7) Sand with
Benthic Algae**

Classification

- Clusters obtained from ISODATA classification
- Converted to polygons in ESRI ArcMap 10.3.
- Spatial Join Tool
 - Polygons assigned to a class based on ground validation.
 - Joining based on spatial location.
 - Attribute of the nearest point is collected and a distance value is recorded.
 - All polygons were aggregated and assigned to a specific benthic habitat category.
 - The final polygon layer were aggregated based on the benthic class using the Dissolve Tool from ESRI ArcMap 10.3.

AVIRIS (Airborne Visible Infrared Imaging Spectrometer) macroecology



Accuracy Assessment

- **Confusion matrix (Jensen 1996)**
- **Individual classes**
 - **Overall accuracy**
 - **Producer accuracy (classifier)**
 - percent of the time, that benthic type was classified as such.
 - **User accuracy (management)**
 - each time an area labelled as a sand on the map is visited, there is a 89.29% probability that it is actually sand.
 - **Kappa coefficient (“true” versus “chance” agreement)**
 - reduction in error generated by a classification process, compared with a completely random classification.
 - **Tau coefficient**
 - Indicates correct classification of pixels than expected by chance.

AVIRIS Image Classification

Classified Data	Truth Data							TOTAL	PRODUCER ACCURACY
	Coral Reefs	Seagrass	Hardbottom	Mix: Sand/Hardbottom /Coral	Mud	Sand	Sand with Benthic Algae		
Coral Reefs	62	5	6	13	0	0	6	92	67.39%
Seagrass	4	33	1	4	0	0	2	44	75.00%
Hard Bottom	2	0	12	1	0	0	1	16	75.00%
Mix: Sand/Hardbottom/ Coral	13	13	2	40	0	0	7	75	53.33%
Mud	0	4	0	0	13	0	1	18	72.22%
Sand	11	4	0	4	7	25	1	52	48.08%
Sand with Benthic Algae	0	2	0	0	0	3	19	24	79.17%
TOTAL	92	61	21	62	20	28	37	321	
USER ACCURACY	67.39%	54.10%	57.14%	64.52%	65.00%	89.29%	51.35%		

Overall Accuracy

63.55%

Kappa Coefficient

0.55

Tau Coefficient

0.59

WV2 Image Classification

Truth Data

Classified Data	Truth Data							TOTAL	PRODUCER ACCURACY
	Coral Reefs	Seagrass	Hardbottom	Mix: Sand/Hardbottom/ Coral	Mud	Sand	Sand with Benthic Algae		
Coral Reefs	65	5	2	13	0	2	2	89	73.03%
Seagrass	6	36	0	2	0	1	5	50	72.00%
Hard Bottom	1	0	12	1	0	2	0	16	75.00%
Mix: Sand/Hardbottom/ Coral	9	7	2	40	0	0	5	63	63.49%
Mud	0	5	0	0	15	0	2	22	68.18%
Sand	9	5	5	6	5	23	4	57	40.35%
Sand with Benthic Algae	1	7	0	0	0	0	19	27	70.37%
TOTAL	91	65	21	62	20	28	37	324	
USER ACCURACY	71.43%	55.38%	57.14%	64.52%	75.00%	82.14%	51.35%		

Overall Accuracy

64.81%

Kappa Coefficient

0.57

Tau Coefficient

0.60

Findings

AVIRIS classification = 63.55%

WV2 classification = 64.81%.

- **Total classified area**
 - AVIRIS = 168.24 km², WV2 = 168.39 km²
- **Classes**
 - Very similar total area per class for AVIRIS and WV2.
 - Coral reefs class total
 - **AVIRIS = 30% (50.32 km²) (10,822 polygons)**
 - **WV2 = 14% (22.89 km²) (11,652 polygons)**
 - Sand class total
 - AVIRIS = 32% (53.50 km²) (1,539 polygons)
 - WV2 = 40% (67.27 km²) (1,452 polygons)

Findings

- **Classification Remarks**

- Sensors performs equally well in deep and shallow environments.
 - Even with WV2 bottom albedo limitations.

- **Image acquisition dates.**

- Massive bleaching event occurred during the AVIRIS image acquisition followed by a coral reef mass-mortality (Eakin et al. 2010).
- Detrimental to *Montastraea (Orbicella) annularis* complex resulting in mortalities in the order of 50% (Garcia-Sais et al. 2008).
- These factors may explain the difference in the total area covered of the coral reef class between the AVIRIS image (50.32 km²) and the WV2 (22.89 Km²).

Photo Interpretation vs Object-based Classification

- **Bauer et al. (2012) used photo interpretation for the classification for the southwest Puerto Rico area, including La Parguera.**
 - Total coral reefs class was 44.1 km² (662 polygons including aggregated reefs, aggregated patch reefs, patch reefs and spur and groove).
- **Total area for AVIRIS (50.3 km²) and WV2 (22.9 km²)**
 - (~10,000 polygons)
- **Total overlapped areas between the studies**
 - AVIRIS was 19.1 km² (38%), WV2 was 12.4 km² (54%).
- **1 kilometer vs 4 meters**
- **Subjective vs Objective**

Conclusions and Remarks

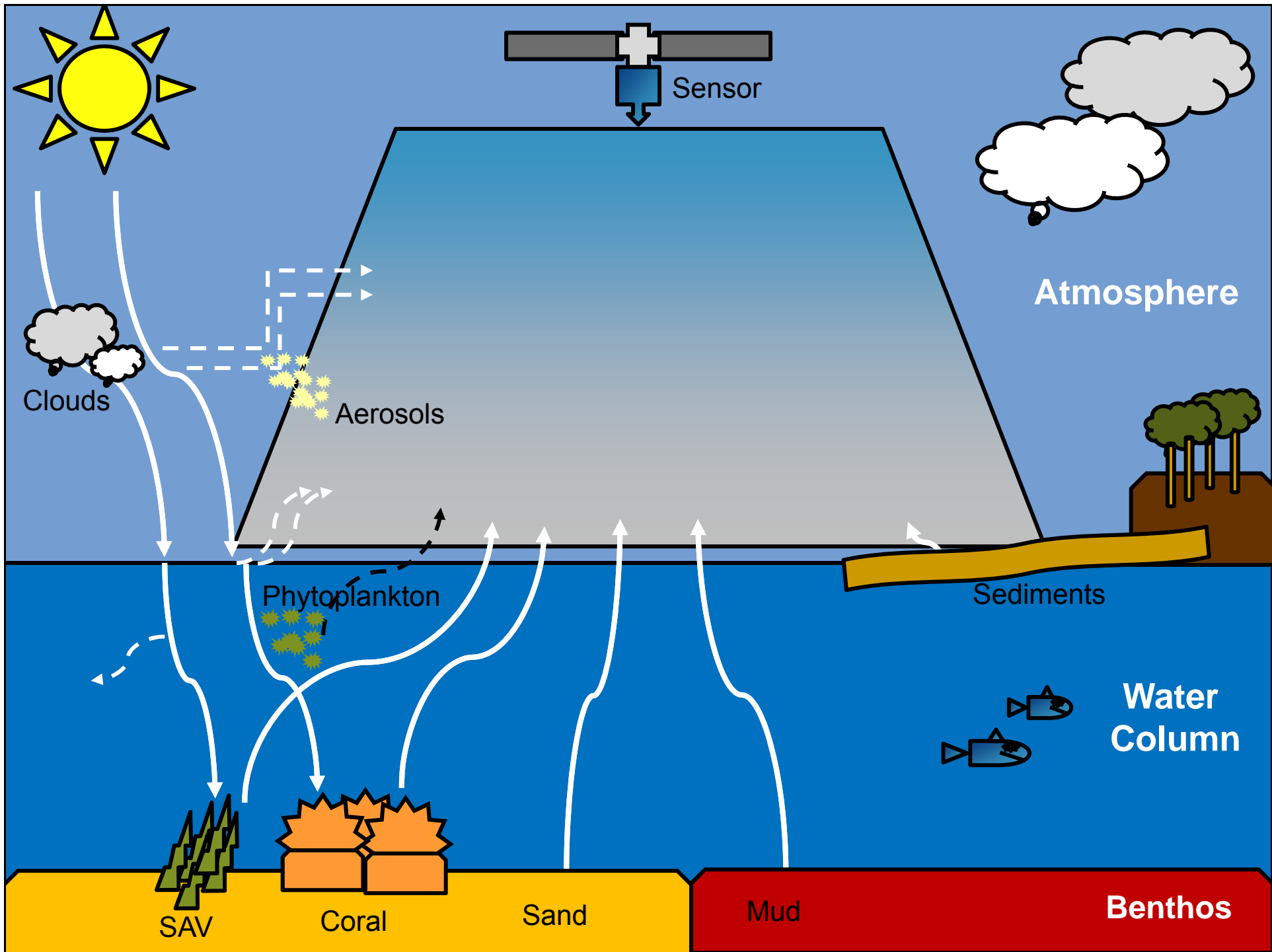
- **Benthic habitat maps developed from bottom albedo images of both AVIRIS and WV2 sensors.**
- **Atmospheric and water column corrections (with LiDAR) improve the benthic habitat mapping.**
- **Reduction in the coral reefs class total could be attributed to temporal differences of the images depicting the changes in habitat types within the reserve.**
- **A major contribution of this study was that no previous benthic habitat map was available for La Parguera Reserve that provided:**
 - **Spatial scale (4 square meters).**
 - **Covered the extent of the reserve (deep areas).**
 - **Utilized the full spectral range of the imagery.**
 - **Methods extrapolated to other areas.**
 - **Change detection.**

Web Mapping Application



References:

- Bierwirth, P. N., Lee, T. J., and Burne, R. V. Shallow sea-floor reflectance and water depth derived by unmixing multispectral imagery. *Photogramm. Eng. Remote Sens.*, vol. 59, pp. 331–338, 1993.
- Costa, B.M., Battista, T.A., Pittman, S.J. 2009. Comparative evaluation of airborne LiDAR and ship-based multibeam SoNAR bathymetry and intensity for mapping coral reef ecosystems. *Remote Sensing of Environment* 113 (2009) 1082–1100
- Guild, L., Lobitz, B., Armstrong, R. Gilbes, F., Goodman, J., Detres, Y., Berthold, R., Kerr, J. 2008. NASA airborne AVIRIS and DCS remote sensing of coral reefs. Proceedings of the 11th International Coral Reef Symposium, Ft. Lauderdale, Florida, 7-11 July 2008.
- Kendall, M.S., M.E. Monaco, K.R. Buja, J.D. Christensen, C.R. Kruer, and M. Finkbeiner, R.A. Warner. 2001. (On-line). Methods Used to Map the Benthic Habitats of Puerto Rico and the U.S. Virgin Islands URL:
<http://biogeo.nos.noaa.gov/projects/mapping/caribbean/startup.htm>.
- Mishra, Deepak R. Narumalani, Sunil Rundquist, Donald Lawson, Merlin and R. Perk. 2005. Enhancing the detection and classification of coral reef and associated benthic habitats: A hyperspectral remote sensing approach. *Journal of Geophysical Research*, Vol. 112, C08014.
- Purkis, S. J. 2005. A “Reef-Up” Approach to Classifying Coral Habitats From IKONOS Imagery. *IEEE Transactions On Geoscience And Remote Sensing*, Vol. 43, No. 6, June 2005.



Findings

- **Overall Accuracy**

- AVIRIS classification = 63.55%

- WV2 classification = 64.81%.

- Mumby et al. (1998) CASI sensor (81%) and Landsat TM (31%)

- Mishra et al. (2007) AISA of 83.6% and 80.6% from IKONOS

- Purkis (2005) IKONOS of 69% for seven classes

- Our study area

- ~168 Km²

- depth range from 0-41 meters (average depth = ~18 meters).

Findings

- **Producer's Accuracy (classifier)**

- **AVIRIS**

- Sand with benthic algae (79.2%), Seagrass and Hardbottom (75%)

- **WV2**

- Hardbottom (75.%), Coral reefs (73%), Seagrass (72.%)

- **User's Accuracy (manager)**

- **AVIRIS**

- Sand (89.3%), Coral Reefs (67.4%), Mud (65%) class,

- **WV2**

- Sand (82.3%), Coral reefs (71.4%), Mud (75%)

Findings

- **Kappa coefficient**

- AVIRIS (55%) and WV2 (57%). “Moderate” classification (Landis and Koch 1977)
- Mishra et al. (2005 and 2007) IKONOS 77.4% and AISA 80.8% for seven classes.
- Arce Arce (2005) and Zayas-Santiago (2011) very similar kappa coefficient compared to our study.

- **Tau coefficient**

- AVIRIS (59%) and WV2 (60%).
- Purkis (2005) IKONOS 65% for seven classes.
- Mumby and Edwards (2002) reduced in IKONOS (67% to 47%) and Landsat TM (61% to 24%), and CASI hyperspectral image (85%-78%) when classes were increased from 4 to 13 classes.

Coastal and Marine Remote Sensing Instruments

