

# Estimating biodiversity of dry forests and coral reefs with hyperspectral data: a NASA EPSCOR project at UPRM

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# Outline

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- Project goals
- Student education and research
- Study area
- Hyperspectral Image Processing
- Analysis and validation
- Anticipated Results

# Objectives

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- Develop spectrally-derived protocols for designing sampling schemes to estimate alpha-beta and gamma diversity
- Develop novel approaches integrating remote sensing and field data to assess and model components of ecosystem biodiversity
- Utilize hyperspectral image analysis techniques to estimate forest condition and reef condition

# Education

- Train 6-8 graduate and 8-10 undergraduate students from underrepresented groups in remote sensing, advanced processing tools, hyperspectral imaging and terrestrial and coastal ecology
- Collaboration between faculty in Science, Agriculture and Engineering at UPRM and with NASA, GRC, GSFC, ARC and JPL, US Forestry service in PR and PR department of Natural Resources



# What is ecosystem biodiversity ?

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- Defined as variety of all forms of life from genes to broad scale of ecosystem
- Alpha diversity – diversity at a place or community
- Beta diversity – diversity between places or changes in alpha diversity
- Gamma diversity – overall diversity of a region; cumulative species of different communities

# Estimators of biodiversity

- Estimates of similarity or difference of biodiversity from place to place – biodiversity surrogates (Sarkar, 2002)
- Difference in composition between places
- Species richness – total counts of species; is estimator surrogate for species diversity
- Other estimators – environmental parameter composition (NDVI and NPP), vegetation types, species composition
- Heterogeneity or spectral diversity in satellite imagery (Rocchini, 2009)

## Previous Work

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- Spatially estimate structure and diversity of forest using Hyperion imagery applying wavelet decomposition image processing methods (Kalackska, 2007)
- Retrieving vegetation cover using Hyperion and Quick Bird data (Huang et al, 2007)
- Estimating biodiversity at Yellowstone ecosystem using Landsat and GIS data (Debinski et al. 1999)

## Study Area

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- Coral reef – La Parguera (Enrique reef)
- Dry tropical forest - Guanica
- La Parguera ecosystem: shallow coral reef, sand, seagrass, mudflats, deeper areas of coral, sand and mud, deep offshore wall on western edge
- Guanica ecosystem: inland forest with zones of grassland, shrubs and cactus, mangrove forests.



# Guanica forest



# Benthic Habitat



**Seagrass**



**Head coral**



**Algae and Coral**



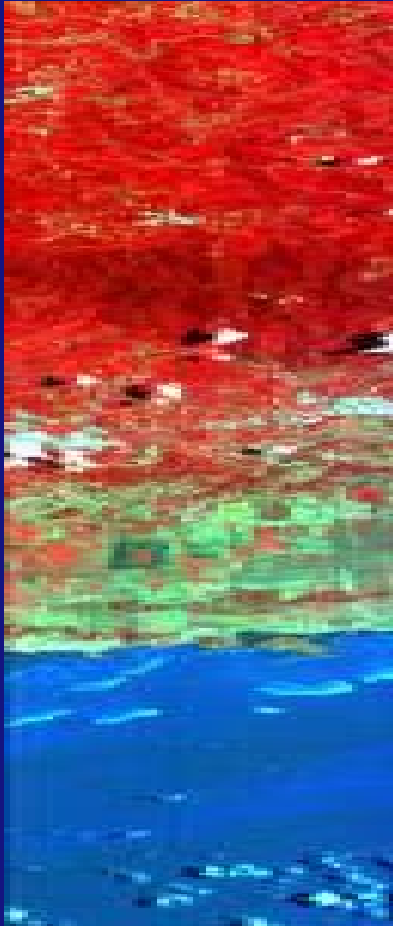
**Aerial Image**

## Existing Data

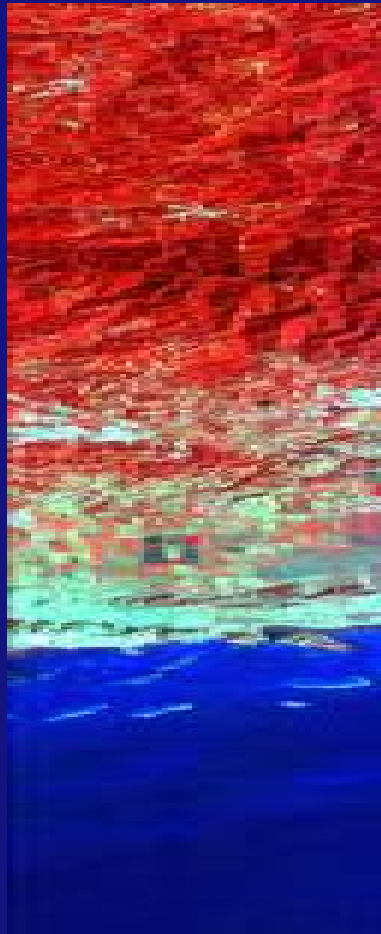
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- La Parguera: field data collected between 2001-2008 with a spectral library of benthic components
- 52 permanently marked study plots on tree growth, forest structure and phenology
- Hyperion hyperspectral imagery of both areas: 196 bands from 400 to 2500 nm at 10 nm spectral sampling interval. Spatial resolution 30 m

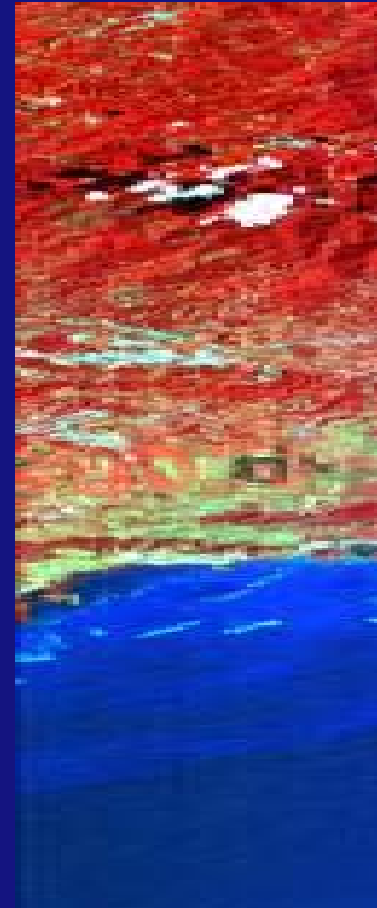
# Hyperion Images of La Parguera



2002/08/15

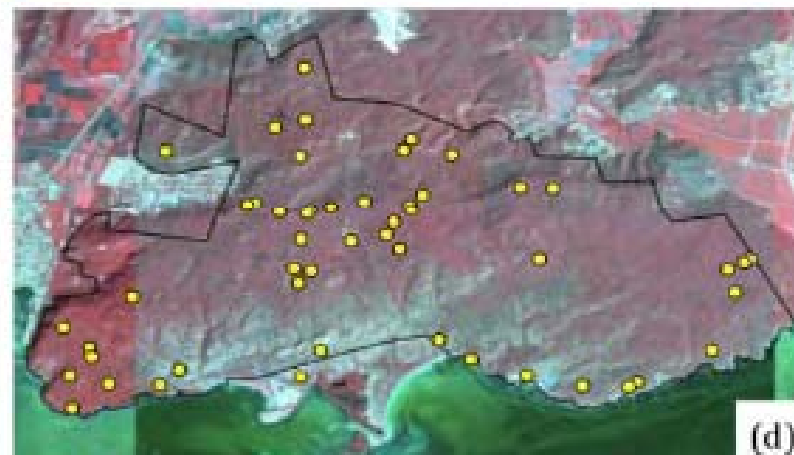
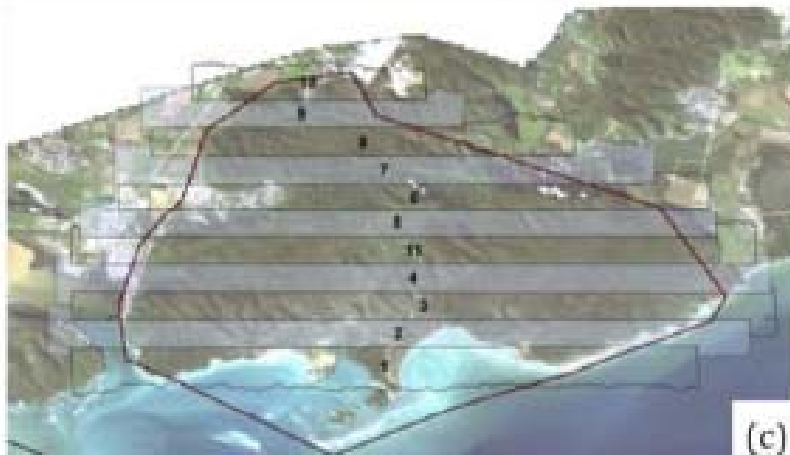
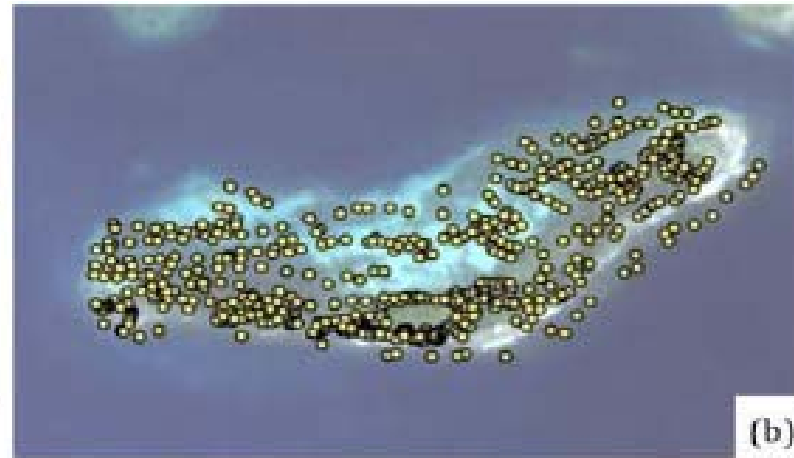
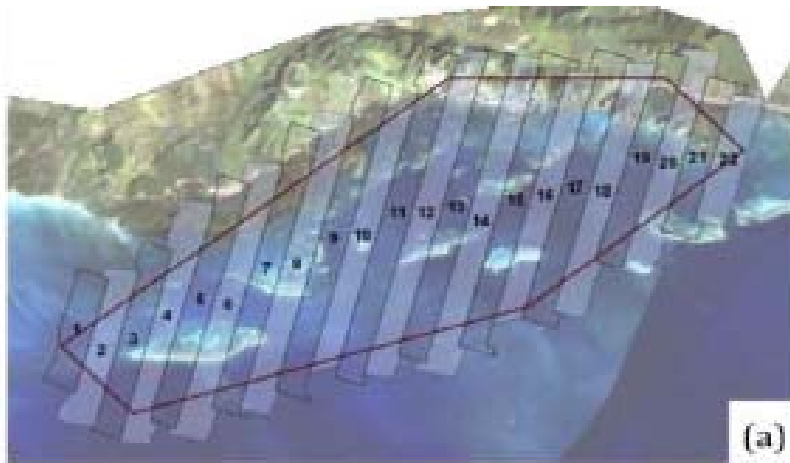


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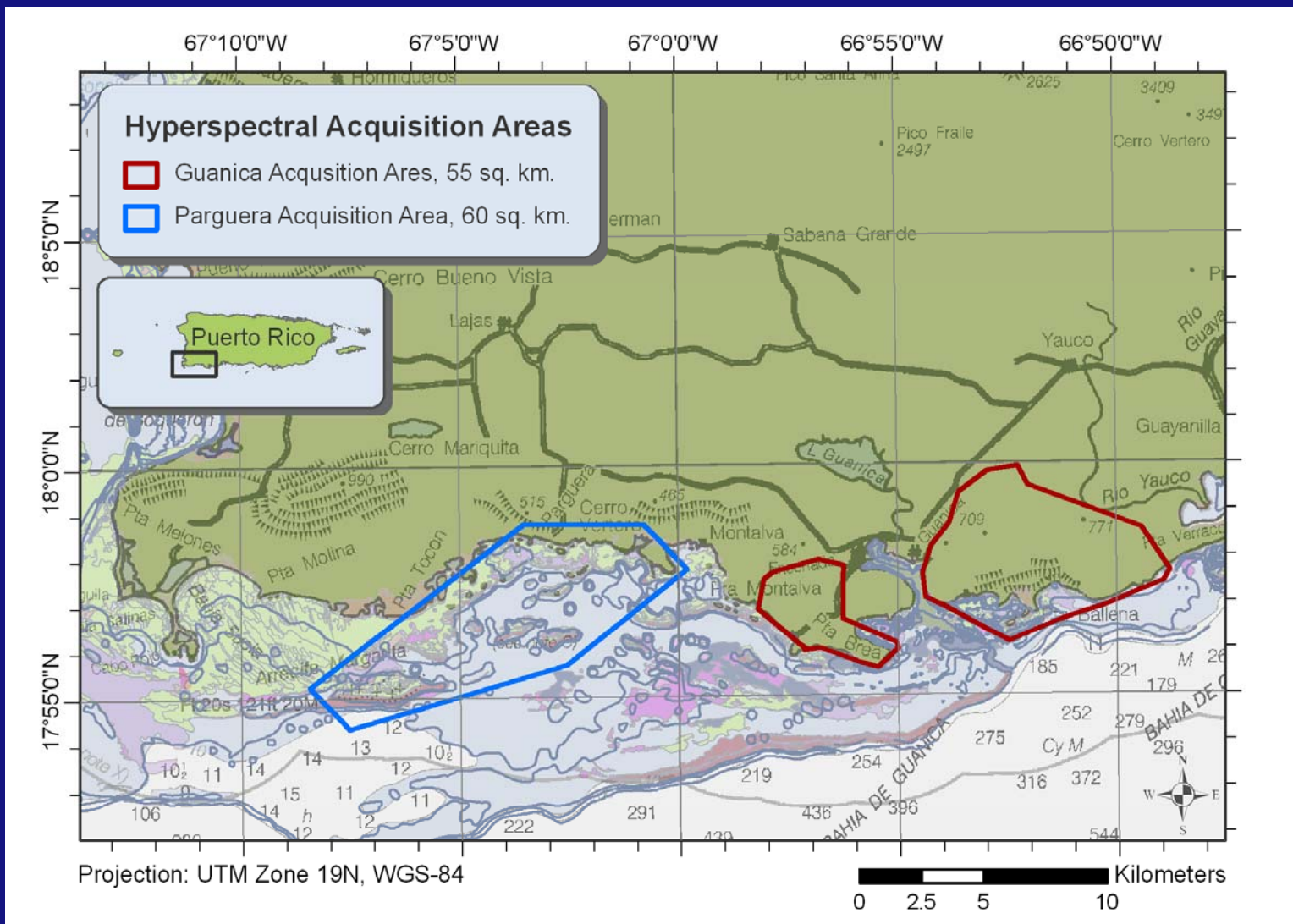
2006/02/17

# AISA Imagery of La Parguera, Guanica, December 2007



(a) 1m imagery of La Parguera, 22 flightlines (b) georeferenced photos  
For Enrique reef – 622 points (c) Guanica 11 flightlines, (d) georeferenced 51 plots

# Data Acquisition: Jan 25 –Feb 3 2010



# Field work



# Remote Sensing using Hyperspectral Images

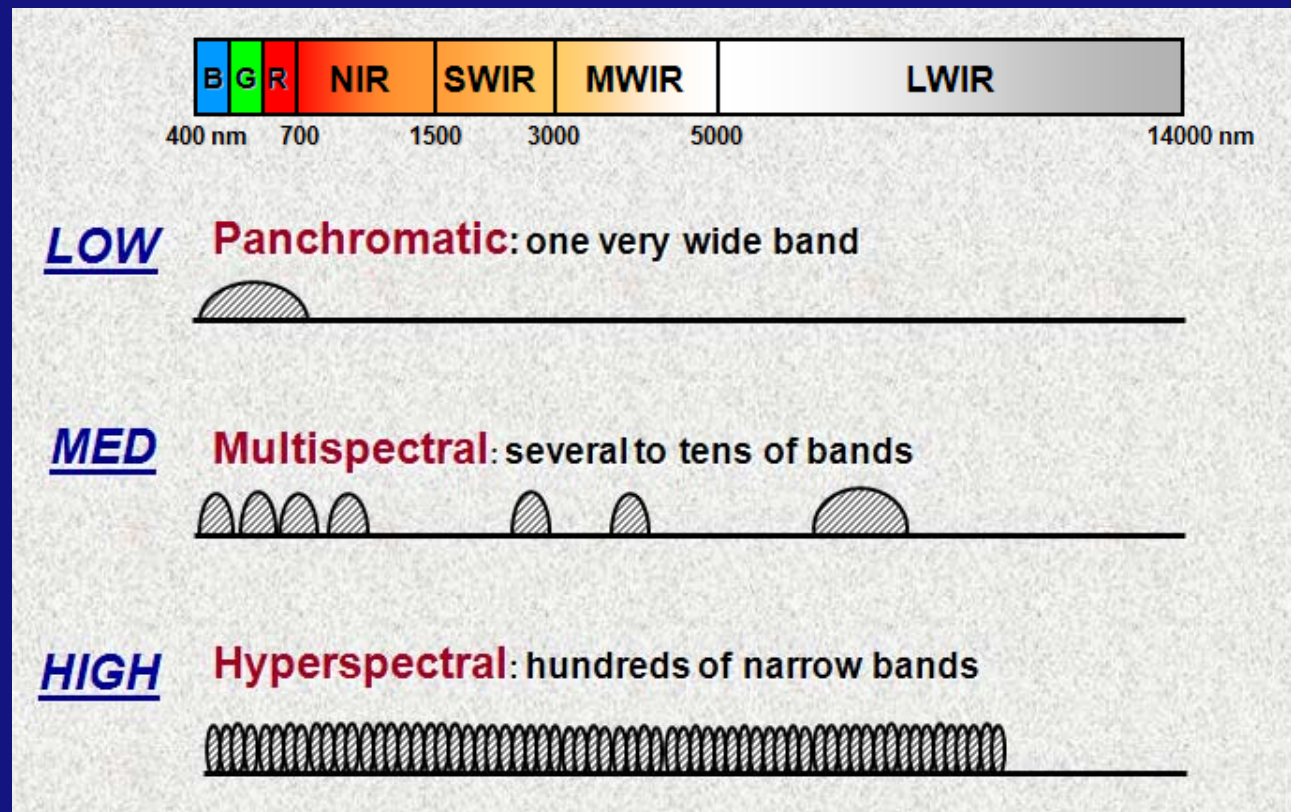
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- Quantitative analysis of biodiversity using hyperspectral images
- **Goal:** Use high spectral resolution imagery to estimate the presence or abundance of different species by their spectral response patterns



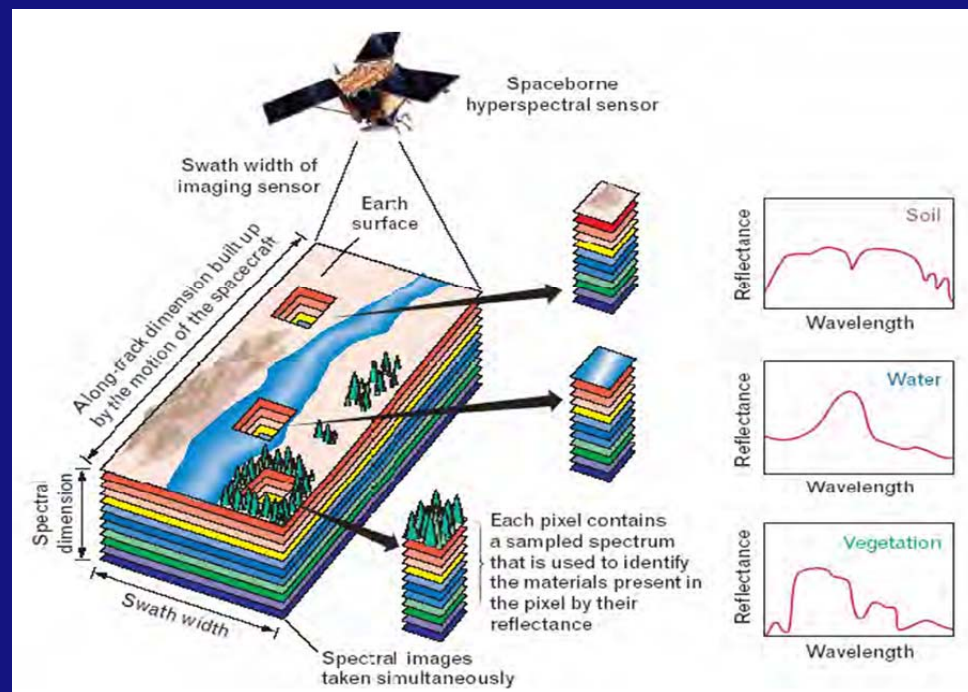
# Spectral Resolution

- Number and dimension of specific wavelength intervals, referred to as bands or channels.

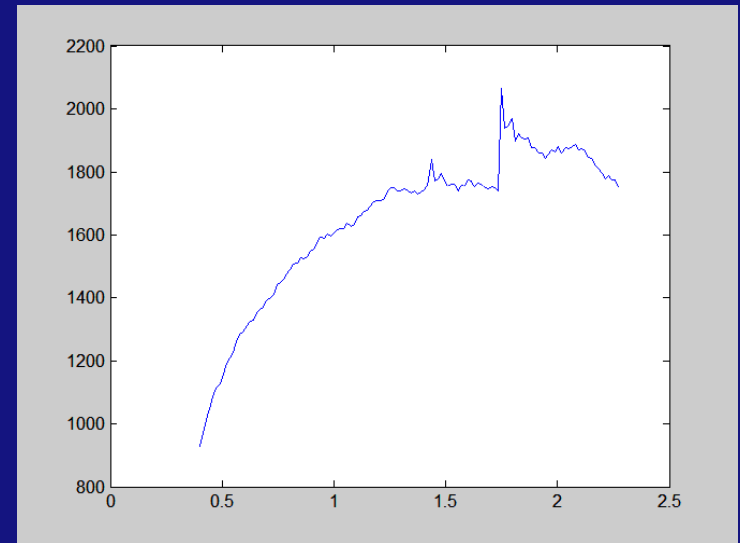
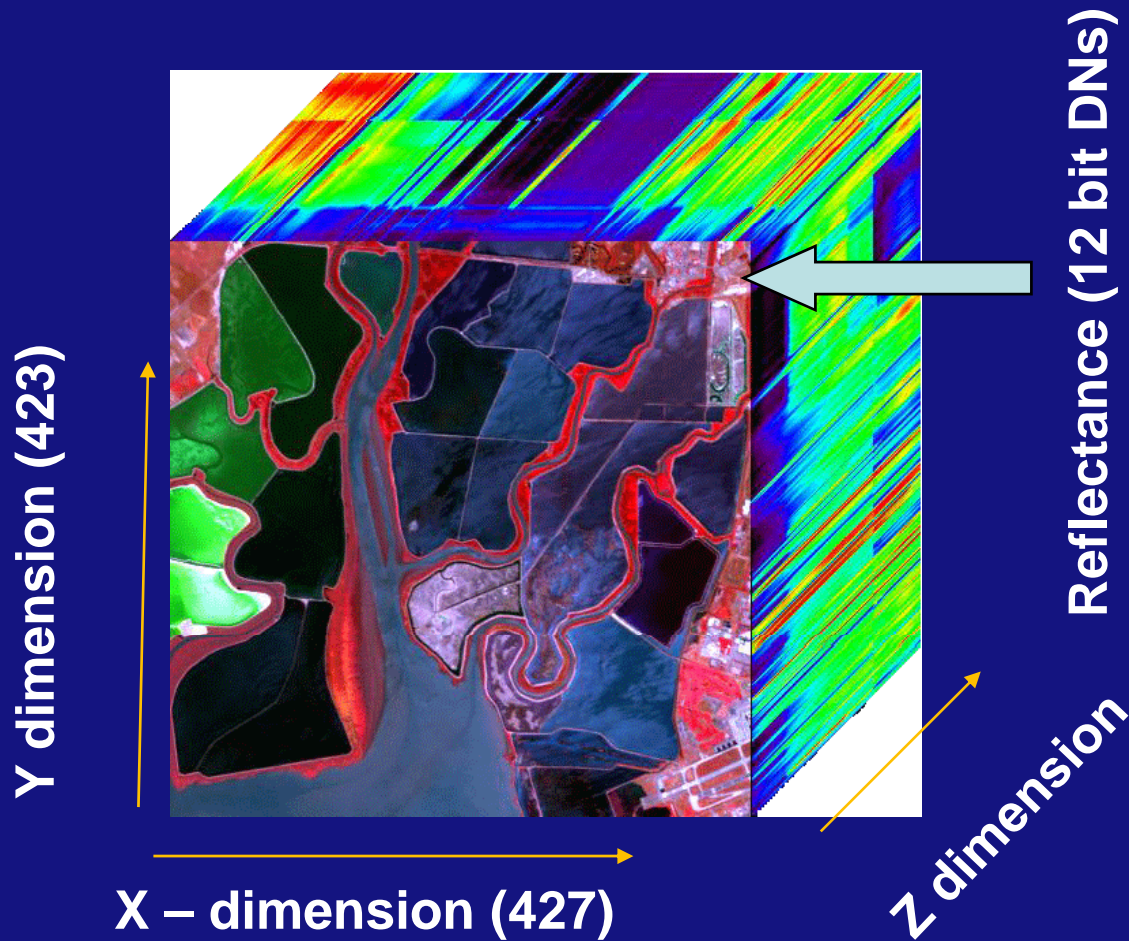


# Hyperspectral Images

- Large amounts of data taken at narrow and contiguous spectral bands.
- Helps to discriminate better between different objects.
- Frequently used
  - Land cover classification
  - Detection and target recognition
  - Search and rescue operations
  - Biomedical applications



# Example of Hyperspectral Image



Wavelength ( $\mu\text{m}$ )

AVIRIS data-224 spectral  
Bands 0.4-2.4  $\mu\text{m}$  with  
20m spatial resolution

# Hyperspectral Image Processing Approach

## Pre-Processing

Terrestrial  
Atmospheric Correction  
Geographic Correction

Marine (Scenario 1)  
Atmospheric Correction  
Sunglint Removal  
Geographic Correction

Marine (Scenario 2)  
Inversion Model  
Water Property Retrieval  
Benthic Reflectance

## Biodiversity Processing

Image Segmentation  
Unsupervised  
Spatial/Spectral Analysis

Spectral Analysis  
Identify Endmembers  
Subset Endmembers  
Spectral Unmixing

Biodiversity  
Spatial Metrics from  
Endmember Spectra  
and Abundances

## Validation

Terrestrial  
Field Plots  
Biodiversity

Accuracy Assessment  
Correlation of  
Predicted and Observed  
Biodiversity

Marine  
Photo Quadrats  
Biodiversity

# Hyperspectral Image Processing

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- **Goal:** utilize available spectral, spatial and temporal information
- Data Pre-processing: atmospheric correction using FLAASH, sunglint removal, reduce striping, improve SNR
- Apply inversion model of Lee to derive bathymetry and water properties for every pixel

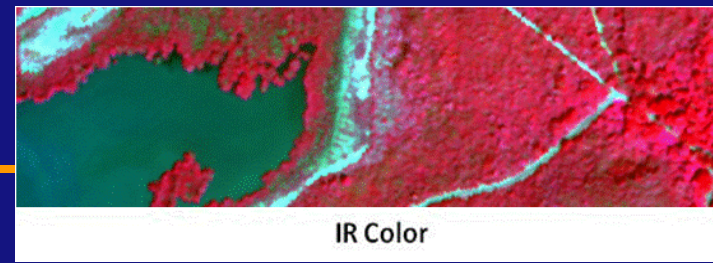
# Unmixing

- The measured spectrum of a pixel is decomposed into a collection of spectra, or endmembers and a corresponding set of fractions or abundances
- Separations of different components in landscapes with mixed communities and variable plant densities
- Linear Mixing Model:  $x = Sa + w$
- Positive Matrix Factorization (Masalmah and Velez-Reyes, 2008)



True color

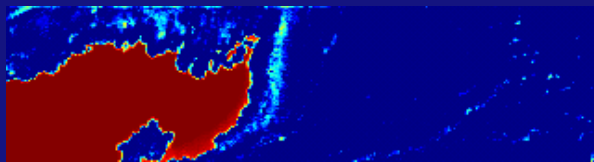
(a)



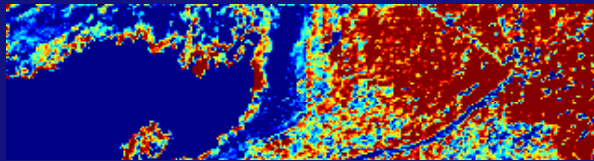
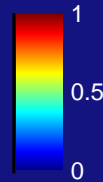
IR Color

(b)

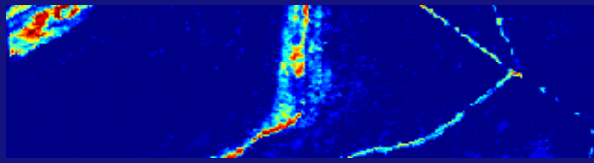
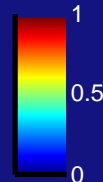
**cPMF Retrieved Fractional Abundance Maps**



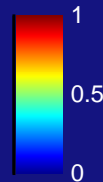
(c)



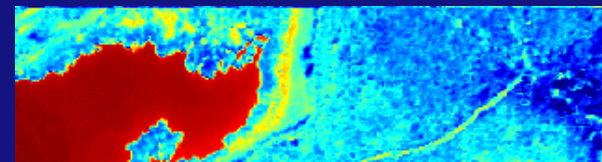
(d)



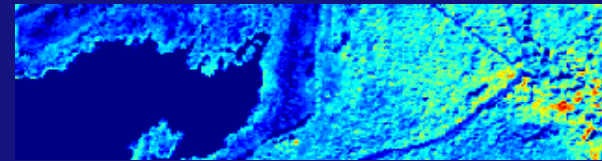
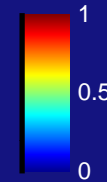
(e)



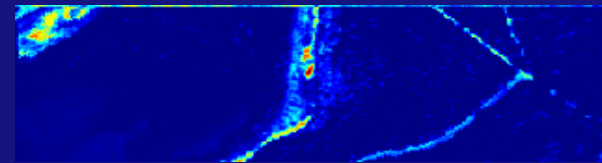
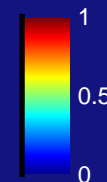
**SMACC Retrieved Fractional Abundance Maps**



(f)



(g)



(h)

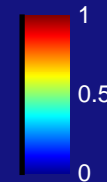


Figure 2. Comparison between SMACC and cPMF of retrieved fractional abundance maps : (a) true color composite, and (b) color infrared composite for the Vieques image; cPMF retrieved fractional abundance maps for (c) water , (d) forest, and (e) dry soil endmembers; SMACC retrieved fractional abundance maps for (f) water, (g) forest, and (h) dry soil endmembers .

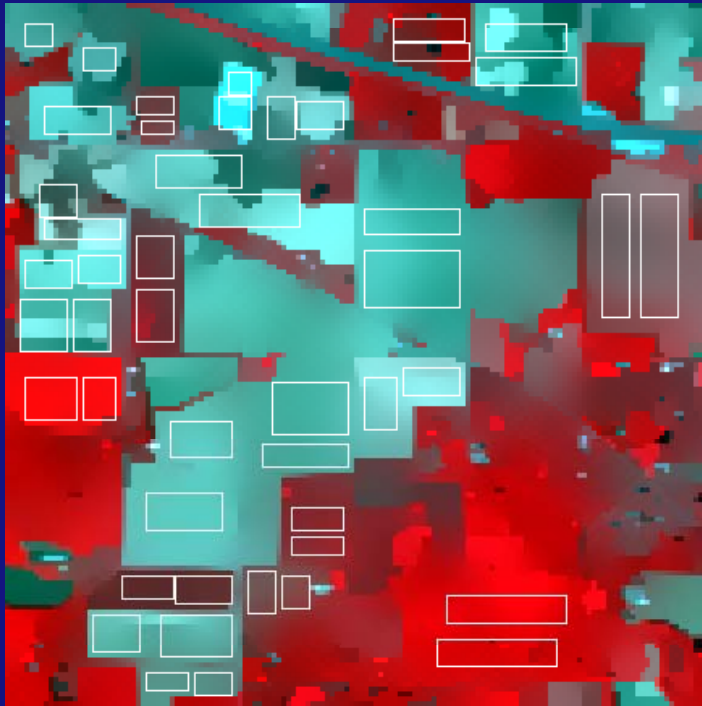
# Geometric PDFs for Hyperspectral Imagery

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- Nonlinear anisotropic diffusion (NAD) reduces variability in regions of similar spectral characteristics and enhances borders
- Improves classification and contrast enhancement (Duarte –Carvajalino, 2007, 2008)



# NAD PDE Application to Classification



Training & testing Samples NW Indian Pines Image



Classification map, smoothed image, 93.7% overall accuracy

# Measures of Spectral Variability

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- Measure jointly spectral and spatial variability using texture metrics to assess biodiversity in hyperspectral images
- Some metrics to be computed using weighted spatial-spectral kernels are kurtosis, homogeneity, skewness and entropy (Velasco-Forero and Manian, 2009)

# Anticipated Results

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- Provide new methods to assess biodiversity
- Assess spatial and temporal changes in biodiversity
- Develop test and validate algorithms for fully automated spectral unmixing using PMF and of new indices to predict species richness using hyperspectral images
- Educate underrepresented students in NASA related areas