

# UAV/Drone GIS processing

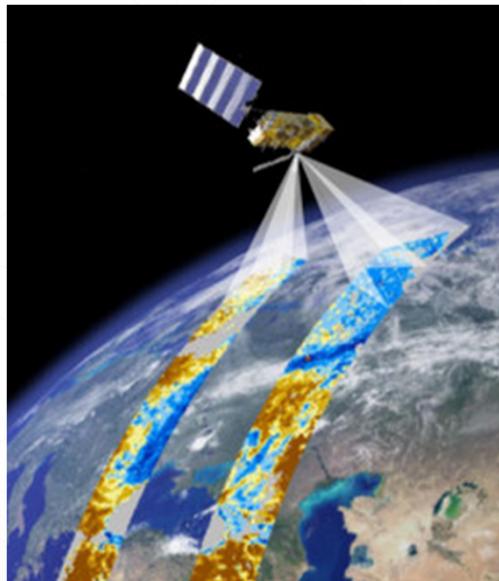
Dr. Bo Yang

[Bo.Yang@ucf.edu](mailto:Bo.Yang@ucf.edu)

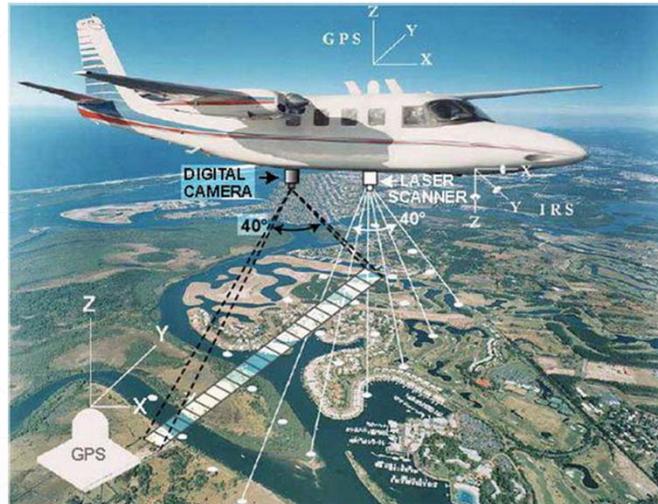
Department of Sociology, University of Central Florida;

Citizen Science GIS, University of Central Florida

# Types of remote sensing data



Satellite remote sensing

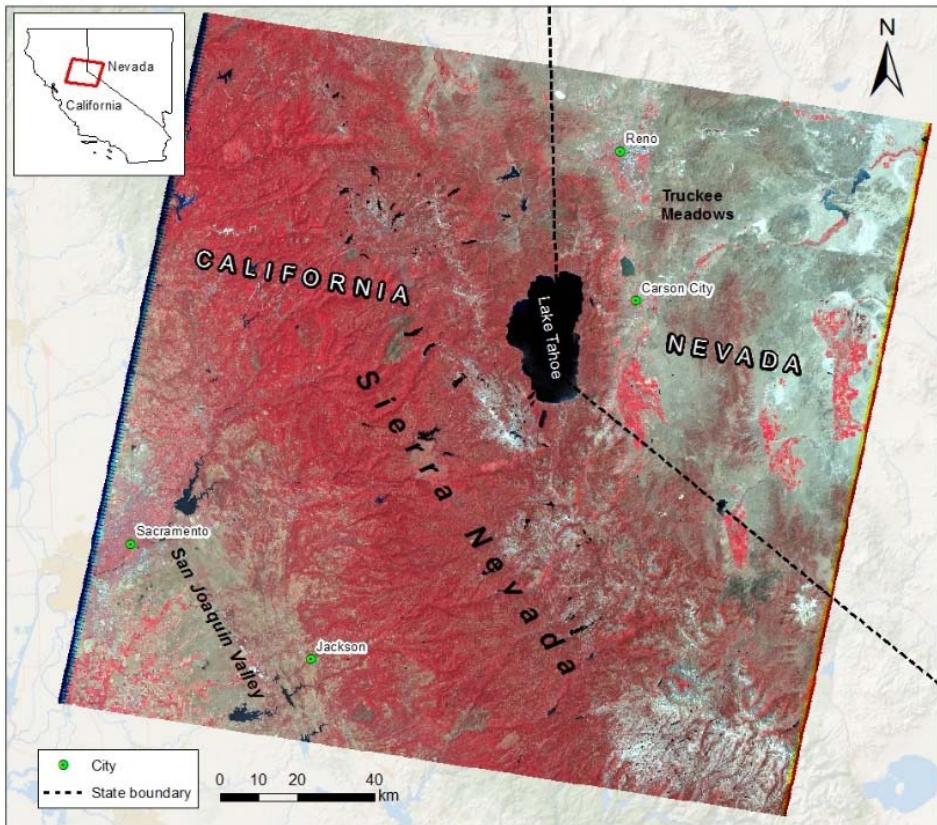


Aerial remote sensing



UAV remote sensing

# Satellite remote sensing



Landsat NIR false color composition multi-spectral imagery

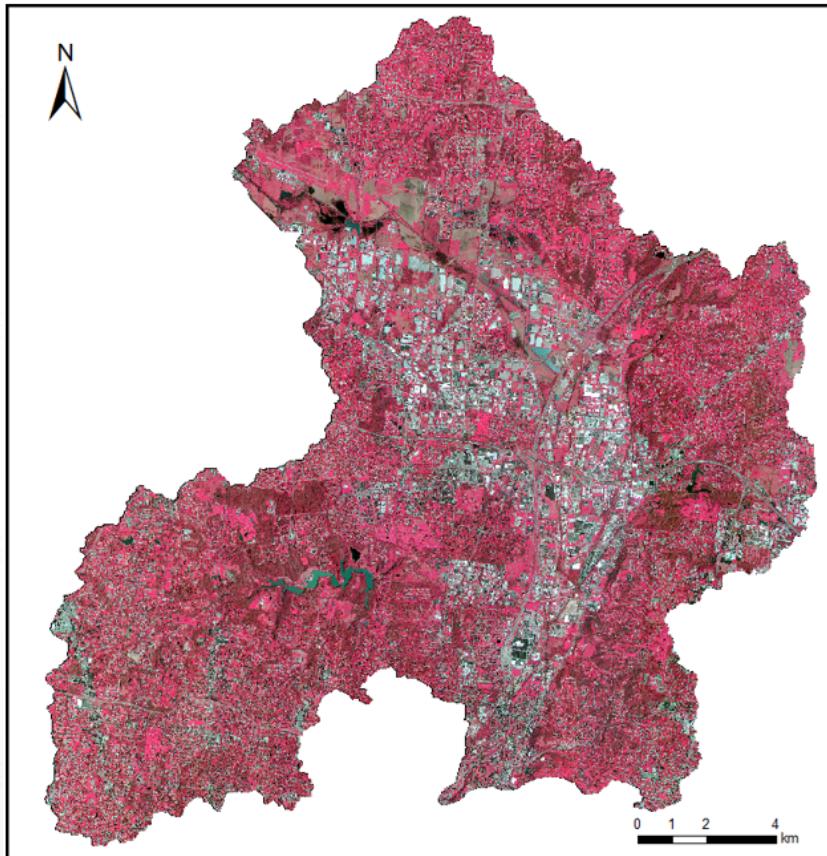
## ➤ Pros

- Temporal revisiting cycle, historical data
- Easier to obtain
- Multi-spectral data

## ➤ Cons

- Moderate spatial resolution
- Can't obtain the target day
- Cloud coverage

# Aerial remote sensing



OSIP aerial imagery NIR false color combination at 1 m resolution

## ➤ Pros

- High spatial resolution
- Date can be control
- Multi-spectral, LiDAR, Thermal sensor optional

## ➤ Cons

- Expensive
- Need pilot and plane
- Long repeating cycle

# UAV remote sensing



USGS Drone imagery

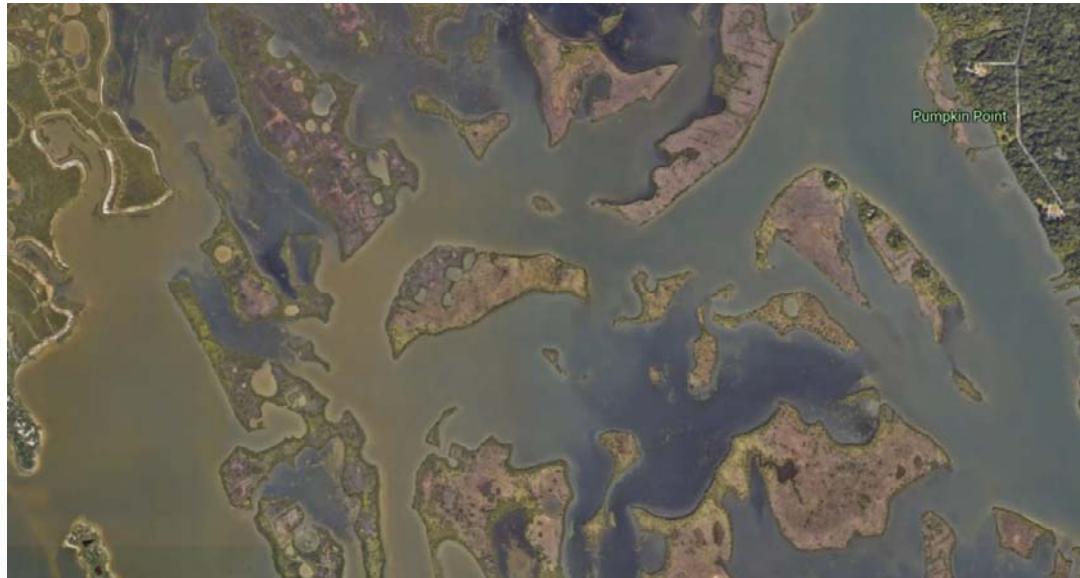
## ➤ Pros

- High spatial resolution
- Real-time data acquiring
- Cheap and convenient

## ➤ Cons

- Accuracy
- Image stitching and processing
- Data validation

## Drone vs. satellite mapping



- In addition to providing spectacular photos and videos for their aesthetic value, drone imagery can complement satellite imagery to improve scientific analysis and resource management.
- Drones have the power to obtain real-time, high-resolution images, as well as footage features.
- You can see how the drone image (right) is more clear and up-to-date than the satellite imagery (left).

# UAVs

DJI phantom 4 Pro



- Budget: \$3000
- Fly time: 28 min
- Payload: 600g

Parrot Bluegrass



- Budget: \$2000
- Fly time: 26 min
- Payload: 500g

DJI Matrice 600 Pro



- Budget: \$8000
- Fly time: 42 min
- Payload: 1000g

# Drone image processing tools

## ➤ Esri Drone2Map

- Orthomosaic Imagery
- Ground Control Points (GCPs)
- Elevation products
- 3D Imagery Products



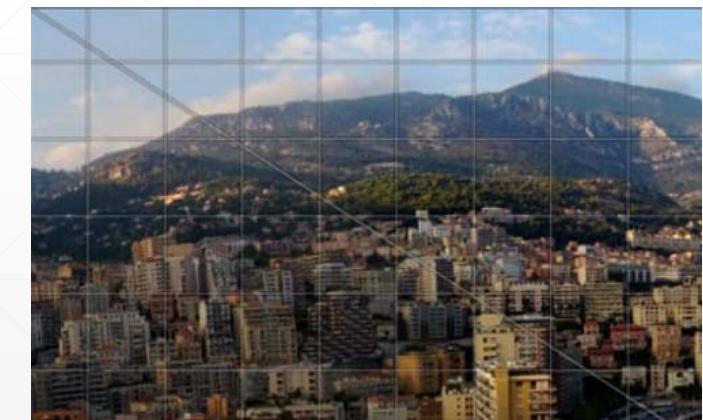
## ➤ Pix4D Mapper

- Purely from images
- Machine-learning algorithm
- Point cloud classification
- Camera self-calibration



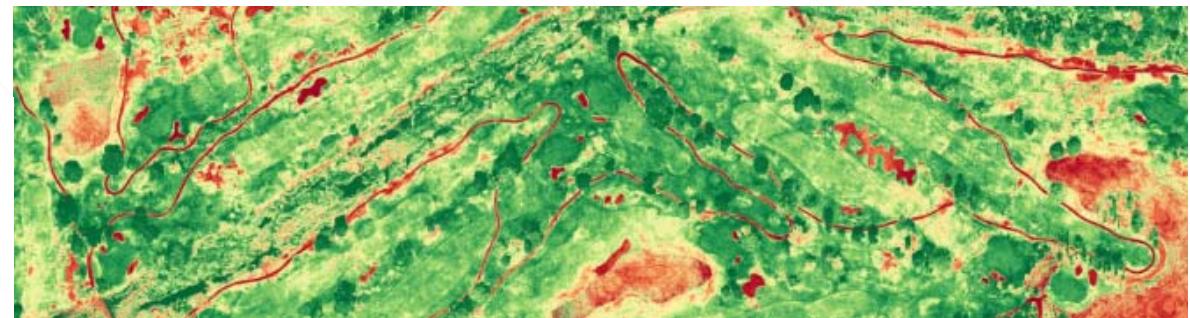
## ➤ Agisoft PhotoScan

- Georeferenced orthomosaic export
- Photogrammetric triangulation
- Dense point cloud:  
editing and classification



# Sensors

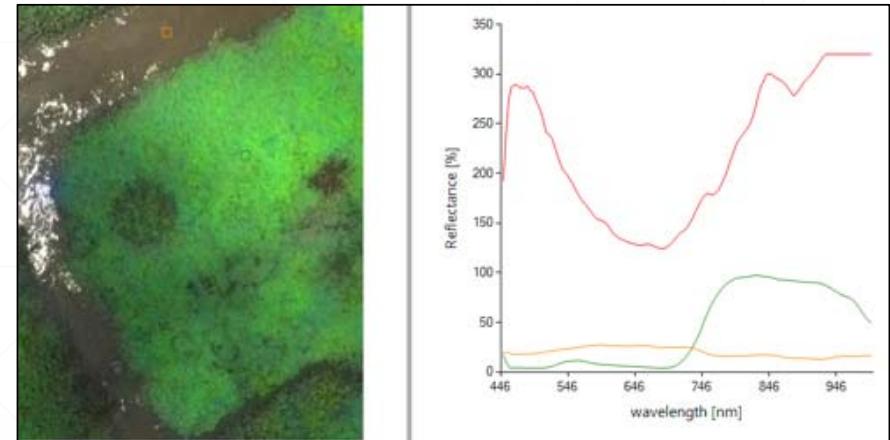
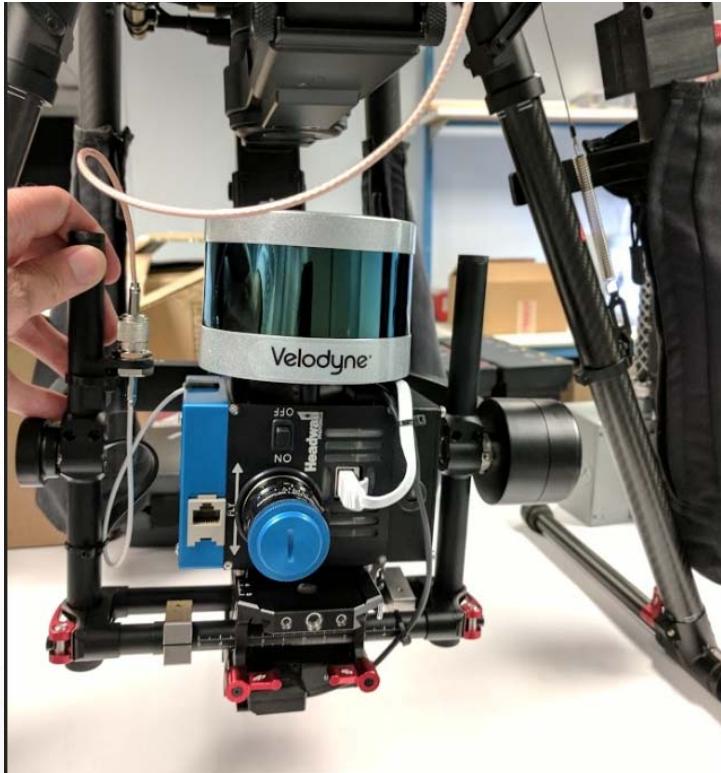
- Bluegrass multi-spectral camera



- NIR, Red edge, Red, Green, Blue
- The filters capture 5 channels of light information
- RGN Filter (NIR+ Red+ Green)

# Sensors

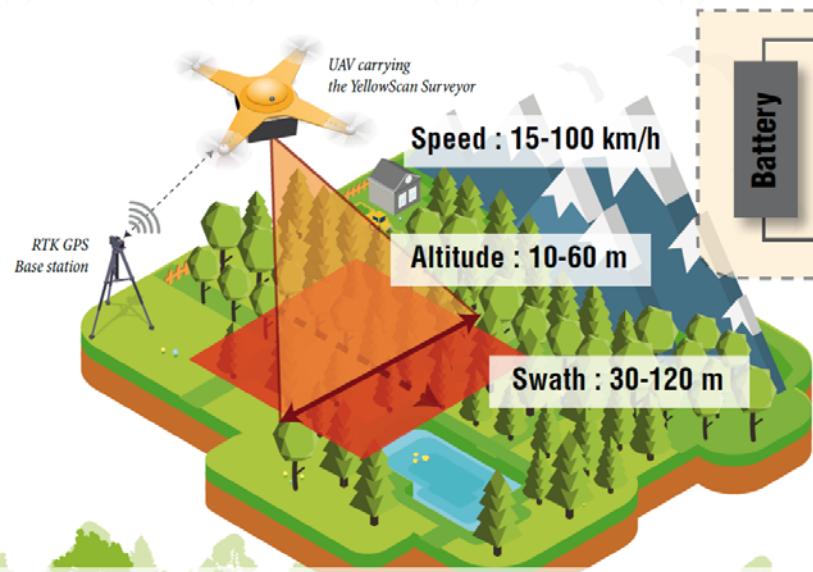
- Headwall Hyperspec VNIR Turnkey Package



- **Hyperspectral** - 270 spectral bands
- Ortho-rectification and RGB mosaic stitching
- Standard GPS included, SpectralView spectral
- Classification and data analysis

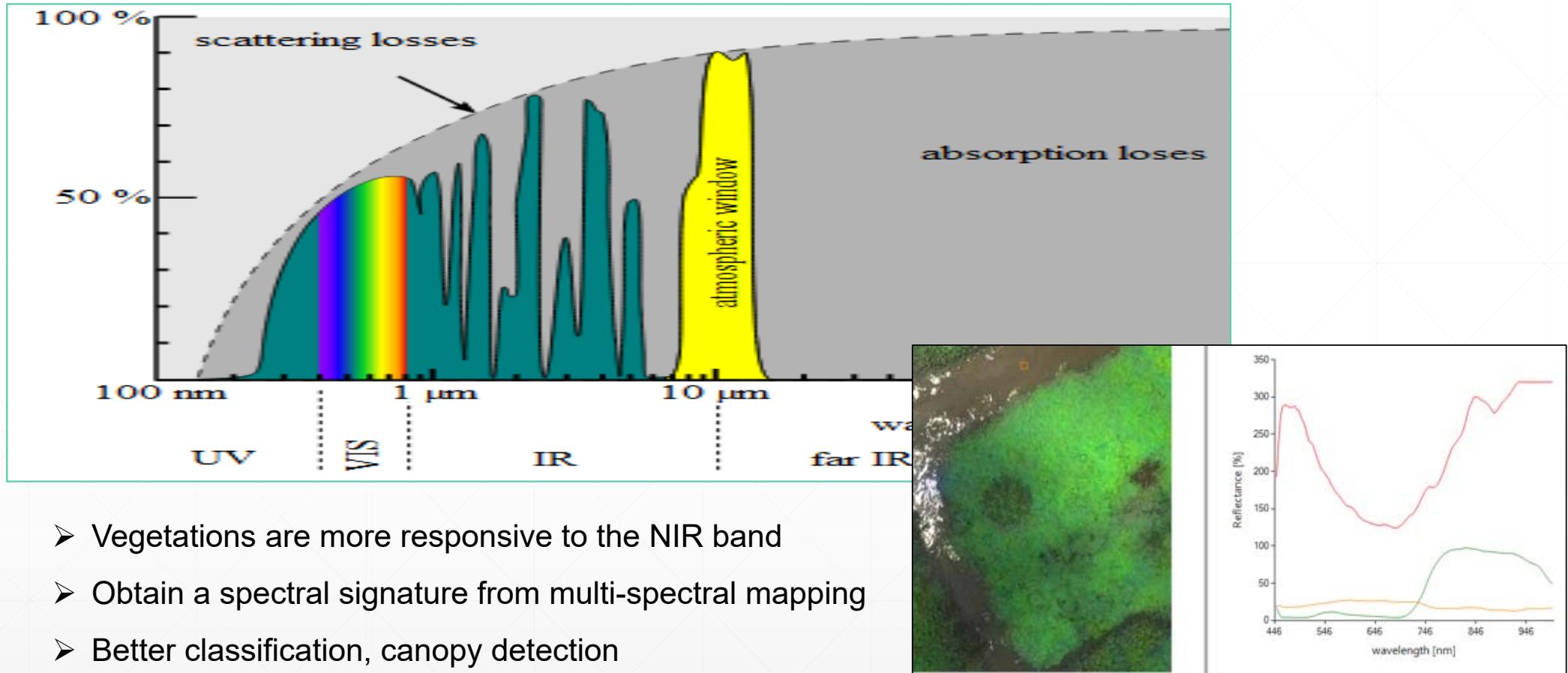
# Sensors

- YellowScan Mapper II UAV LiDAR sensor

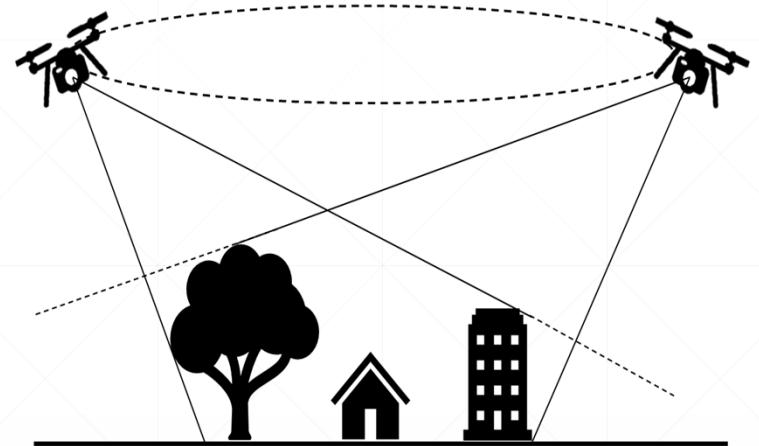
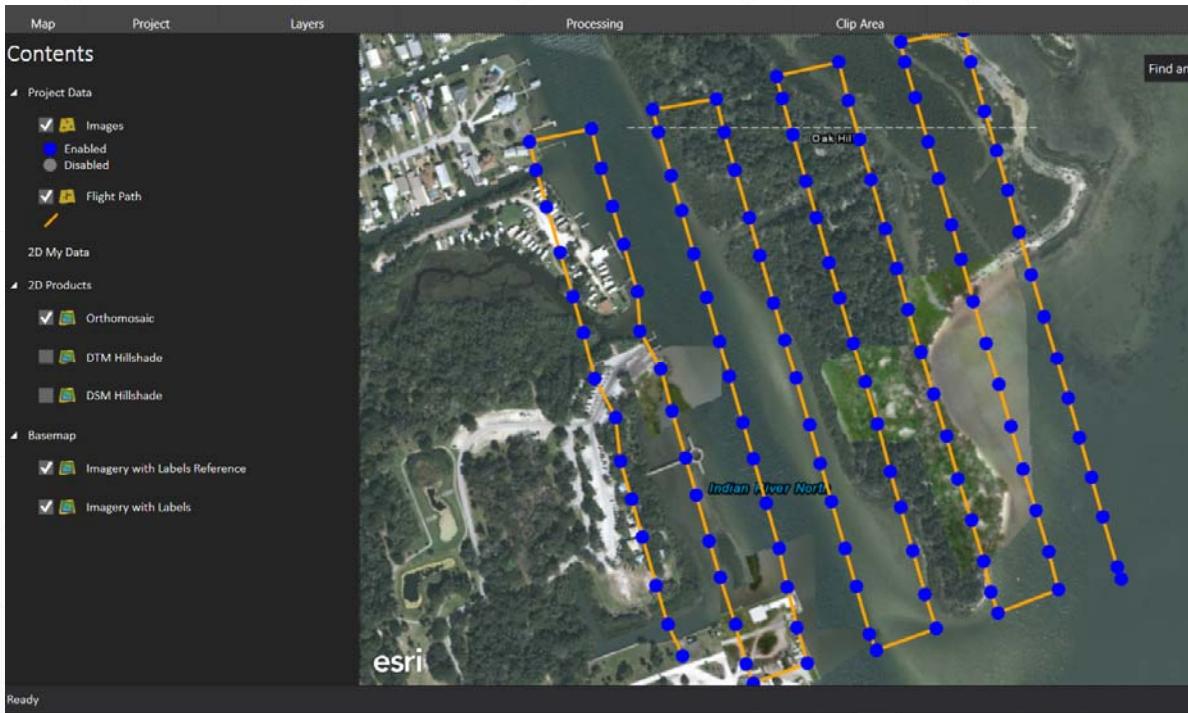


- UAV **LiDAR** scanner
- High resolution DEM/DSM
- Precision: 4 cm
- Accuracy: 5 cm

# Advance of multi-spectral drone mapping



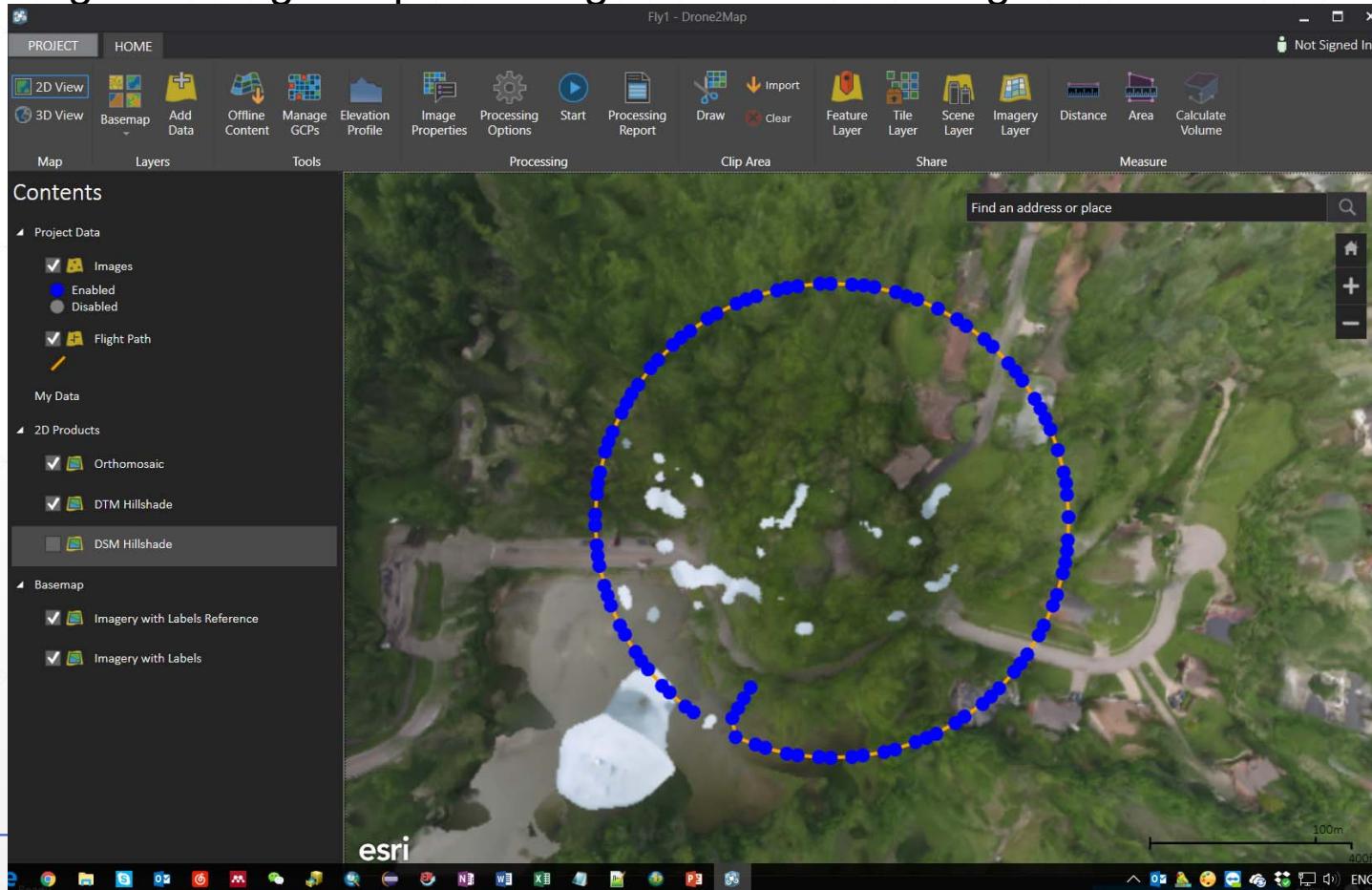
# UAV high resolution data



- Flying plan
- GCPs and fly path

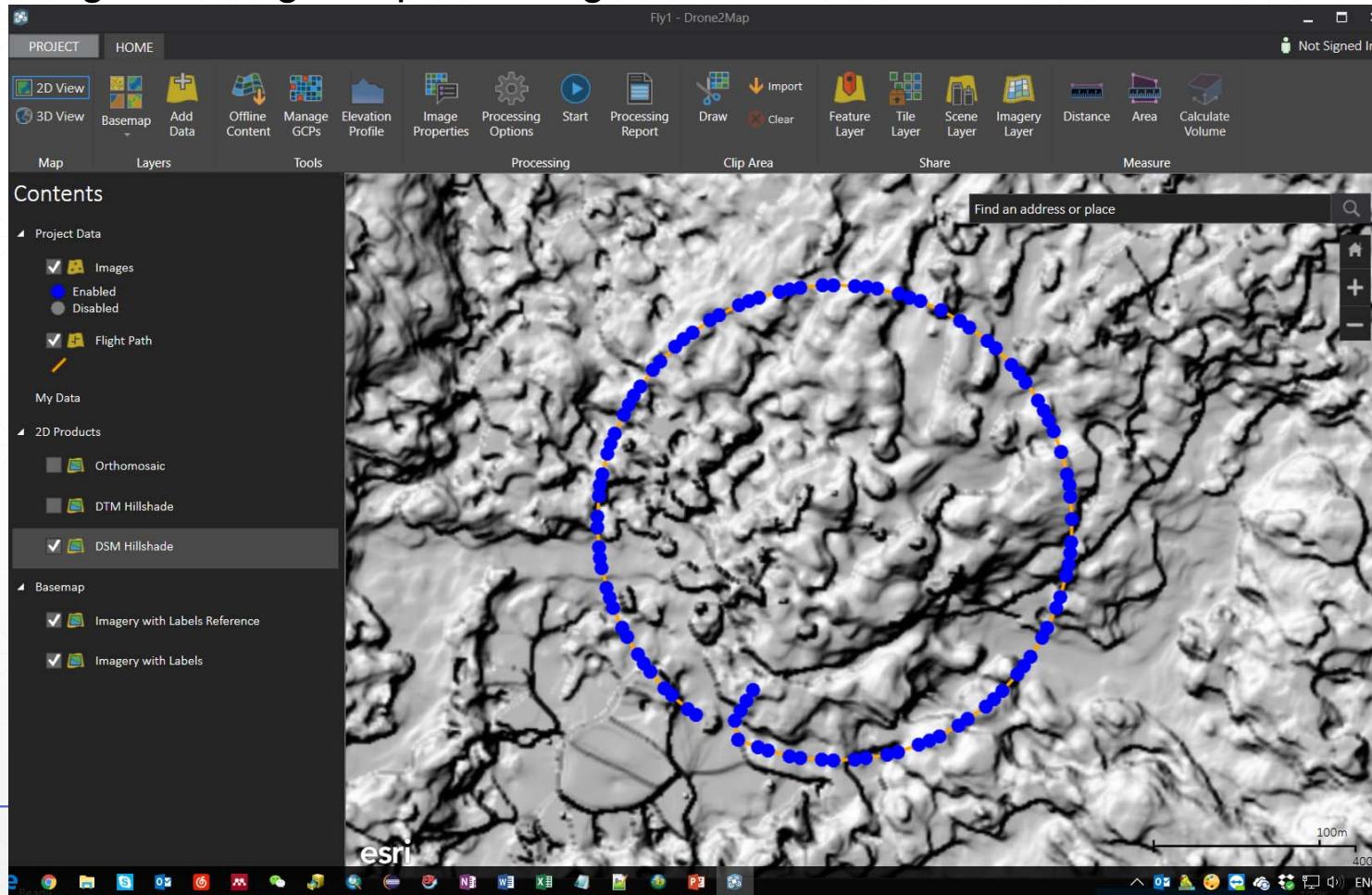
# UAV high resolution data

- UAV image stitching and processing - Orthomosaic image

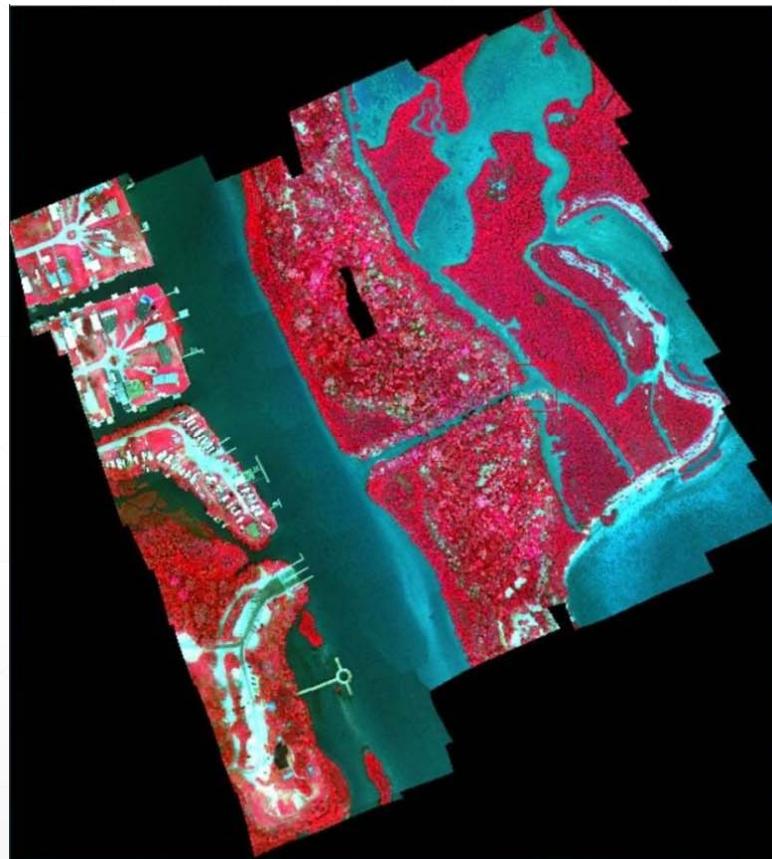


# UAV high resolution data

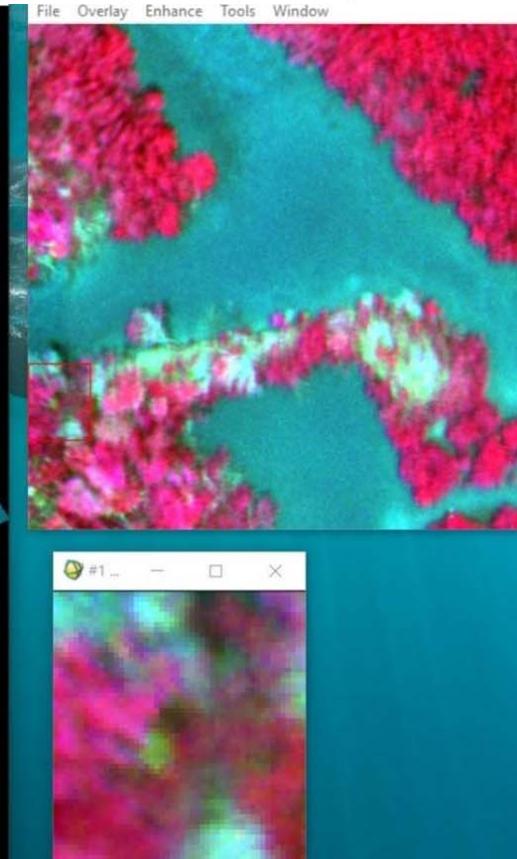
- UAV image stitching and processing - DSM hillshade



## Multi-spectral and RGB data

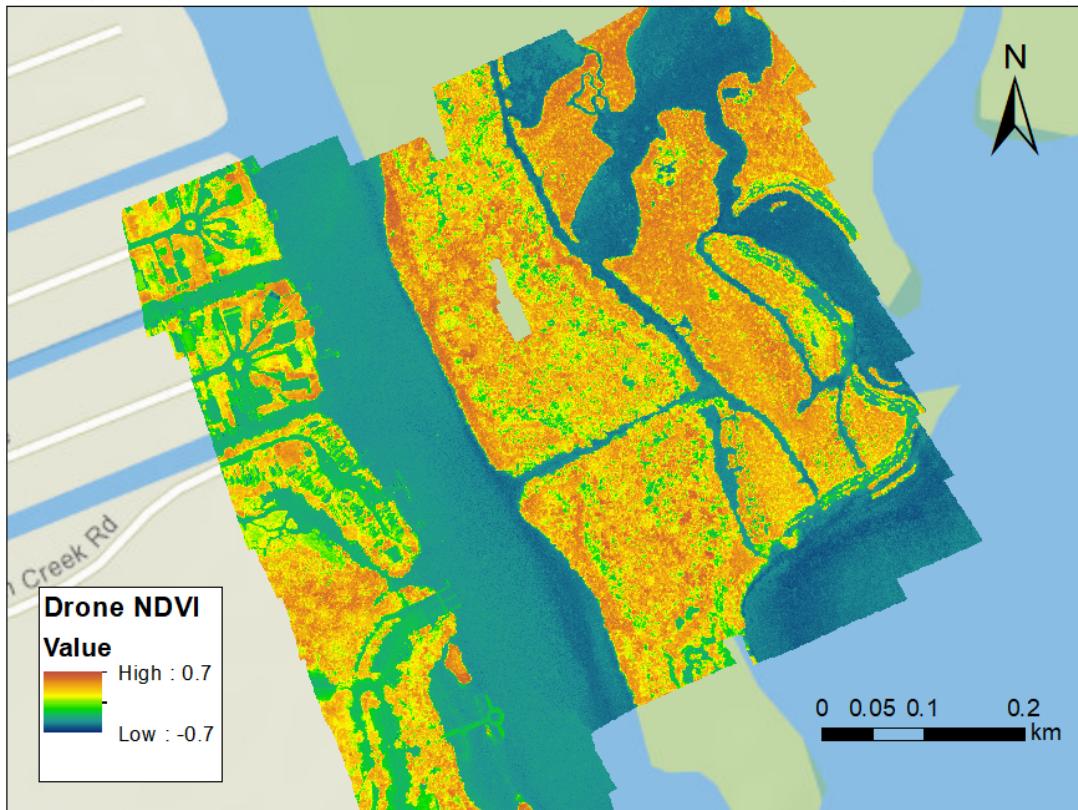


Parrot multi-spectral



DJI RGB

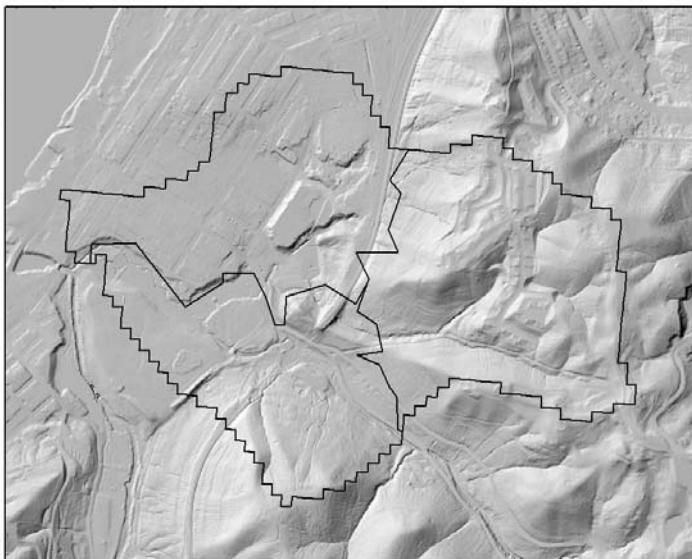
## Normalized difference vegetation index (NDVI)



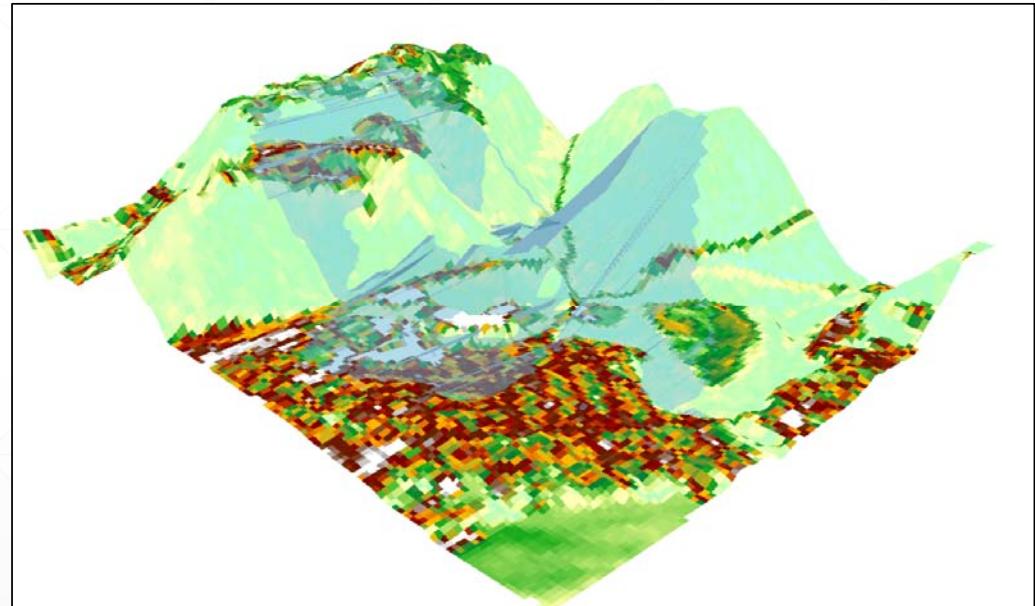
- NDVI could be calculated from the drone mapping imagery.
  - NDVI is utilized to analyze whether the area contains live green vegetation or not.
  - water bodies (from -0.0175 to -0.328)
  - Built environment (-0.019 to 0.060)
  - bare soil (-0.001 to 0.166).
  - dense vegetation (0.500 to 0.575)

## UAV high resolution data

- UAV image stitching and processing



**DEM hillshade (1 m)**



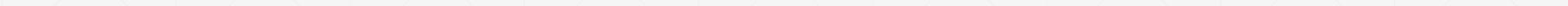
**3D View of DEM**

## The aims of field survey

- 1. To identify each feature of interest (e.g. each habitat type).
  - 2. To locate representative areas of each feature in order to generate spectral signatures (spectra) from the imagery.
  - 3. To generate adequate additional data to test the quality or accuracy of the image classification (i.e. habitat map).
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## Global Positioning Systems (GPS)

1. to measure the position of prominent features on an image *in situ* which can be used to provide ground control points for geometric correction
2. to assign positions to field data. These field data can then be correlated with spectral information at the same point on a geometrically corrected image.



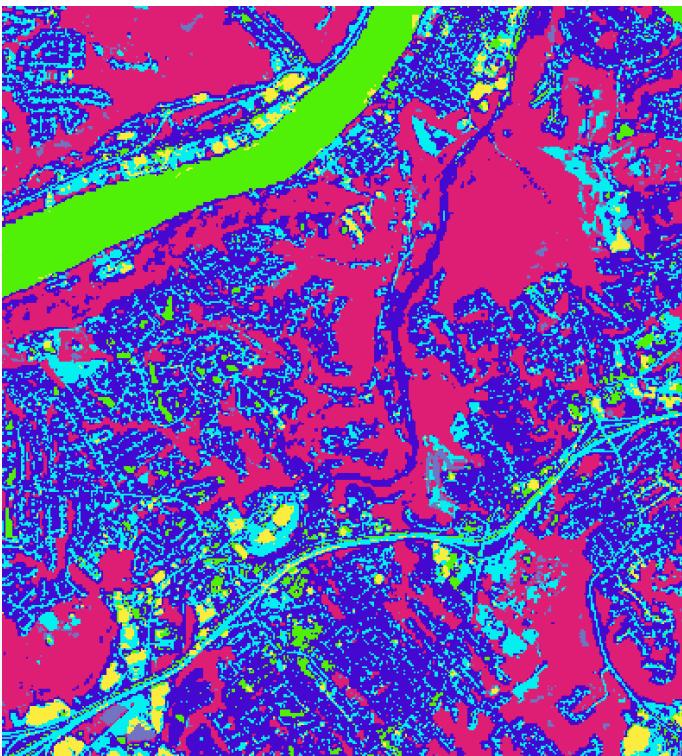
# Ground Control Points (GCPs)



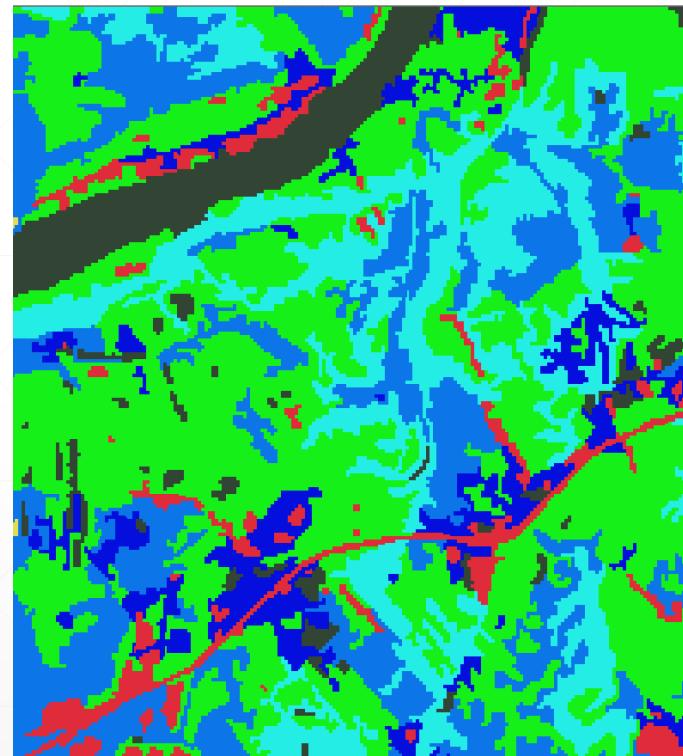
- Use GCPs to calibrate the Drone imagery
- Collected by Garmin R1 GNSS system
- GPSs located in intersection/coner of roads

# UAV imagery classification

- Object oriented classification



Pixel based classification (ISODATA)



Object oriented classification

## Object scale in training data sets

- Object oriented classification



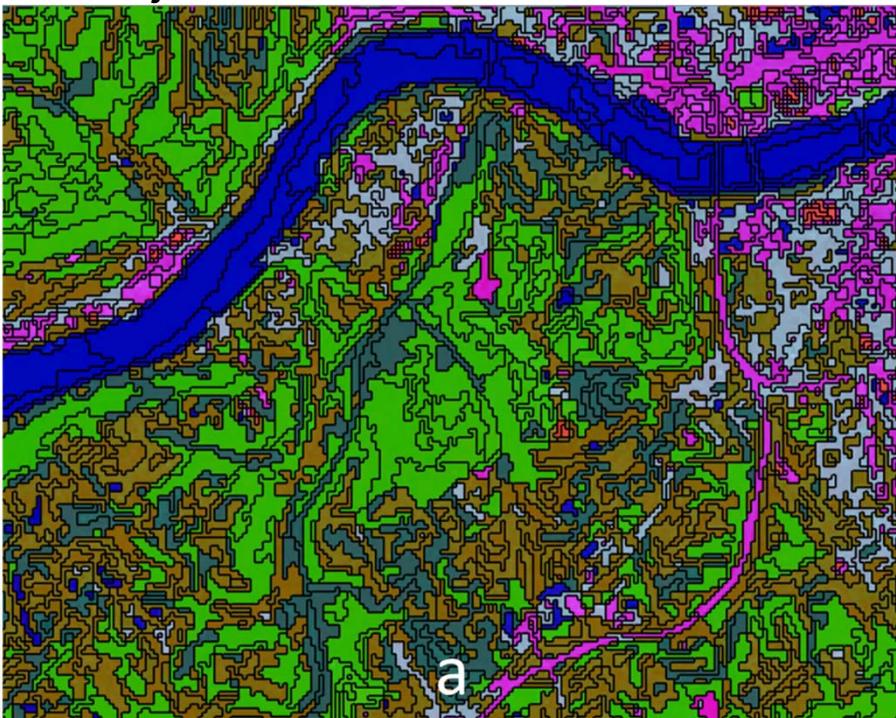
Object scale 300



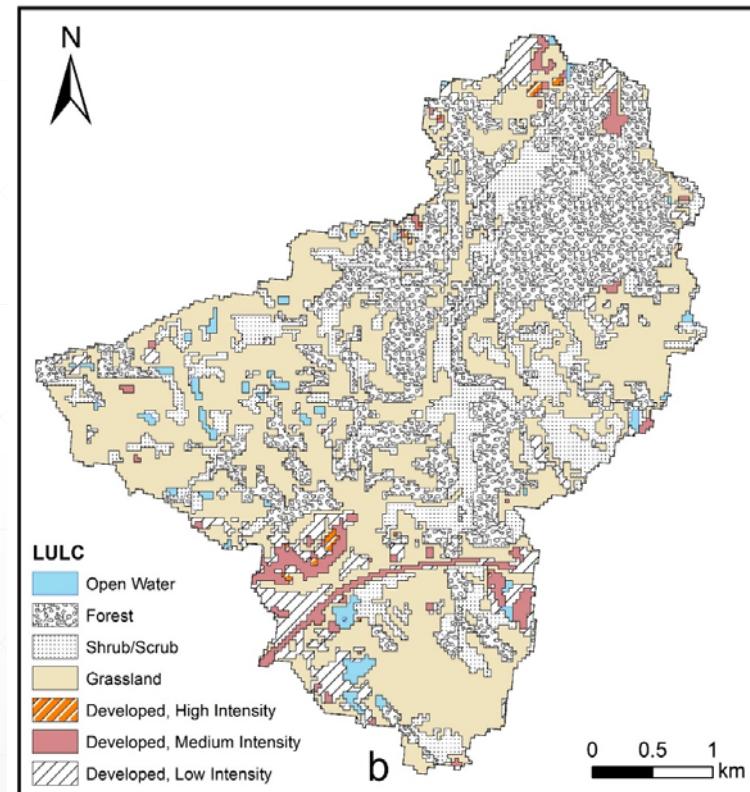
Object scale 200

# UAV imagery classification

- Object oriented classification



object training samples

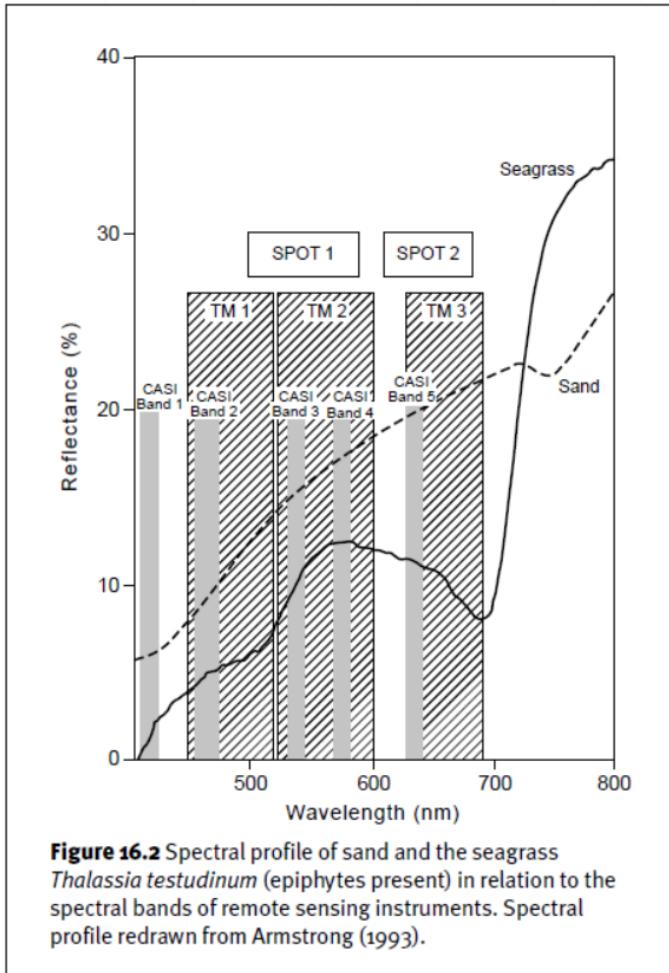


Object-oriented classification results

# UAV Applications for seagrass

- **Seagrass standing crop**
  - **Mangrove Leaf Area Index (LAI) and canopy closure**
  - **Assessing of seagrass abundance**
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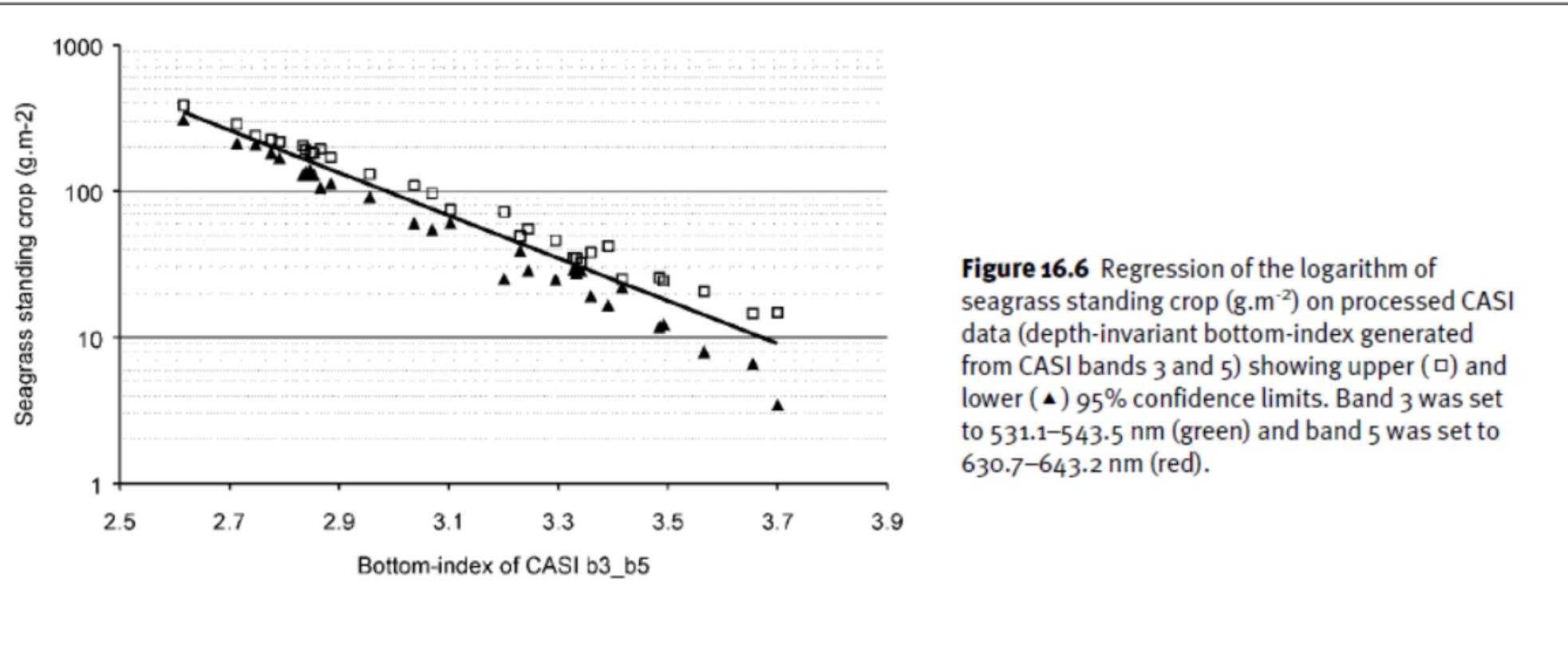
# Assessing Seagrass Standing Crop



Bands of 400–650 nm were designated in the region of the electromagnetic spectrum which best penetrates water

The choice of CASI bands is compared to a reflectance profile of the Caribbean seagrass *Thalassia testudinum* (Armstrong 1993).

Several of the bands have different reflectance characteristics for sand and seagrass which is necessary for discriminating seagrass bottom types using the water column correction method.



**Figure 16.6** Regression of the logarithm of seagrass standing crop ( $\text{g.m}^{-2}$ ) on processed CASI data (depth-invariant bottom-index generated from CASI bands 3 and 5) showing upper (□) and lower (▲) 95% confidence limits. Band 3 was set to 531.1–543.5 nm (green) and band 5 was set to 630.7–643.2 nm (red).

Source: Green, E., Mumby, P., Edwards, A., & Clark, C. (2000). *Remote Sensing: Handbook for Tropical Coastal Management*. United Nations Educational, Scientific and Cultural Organization (UNESCO).

## ***Measurement of canopy transmittance and calculation of LAI***

Canopy transmittance is given by the ratio  $I_c/I_o$  where  $I_c$  = light flux density beneath the canopy and  $I_o$  = light flux density outside the canopy. LAI can then be calculated, and corrected for the angle of the sun from the vertical, using the formula.

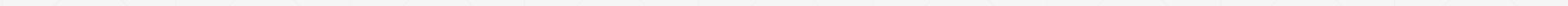
$$LAI = \frac{\log_e \left( \frac{I_c}{I_o} \right)}{k} \times \cos \theta$$

LAI = leaf area index, k = canopy light extinction coefficient, which is a function of the angle and spatial arrangement of the leaves.

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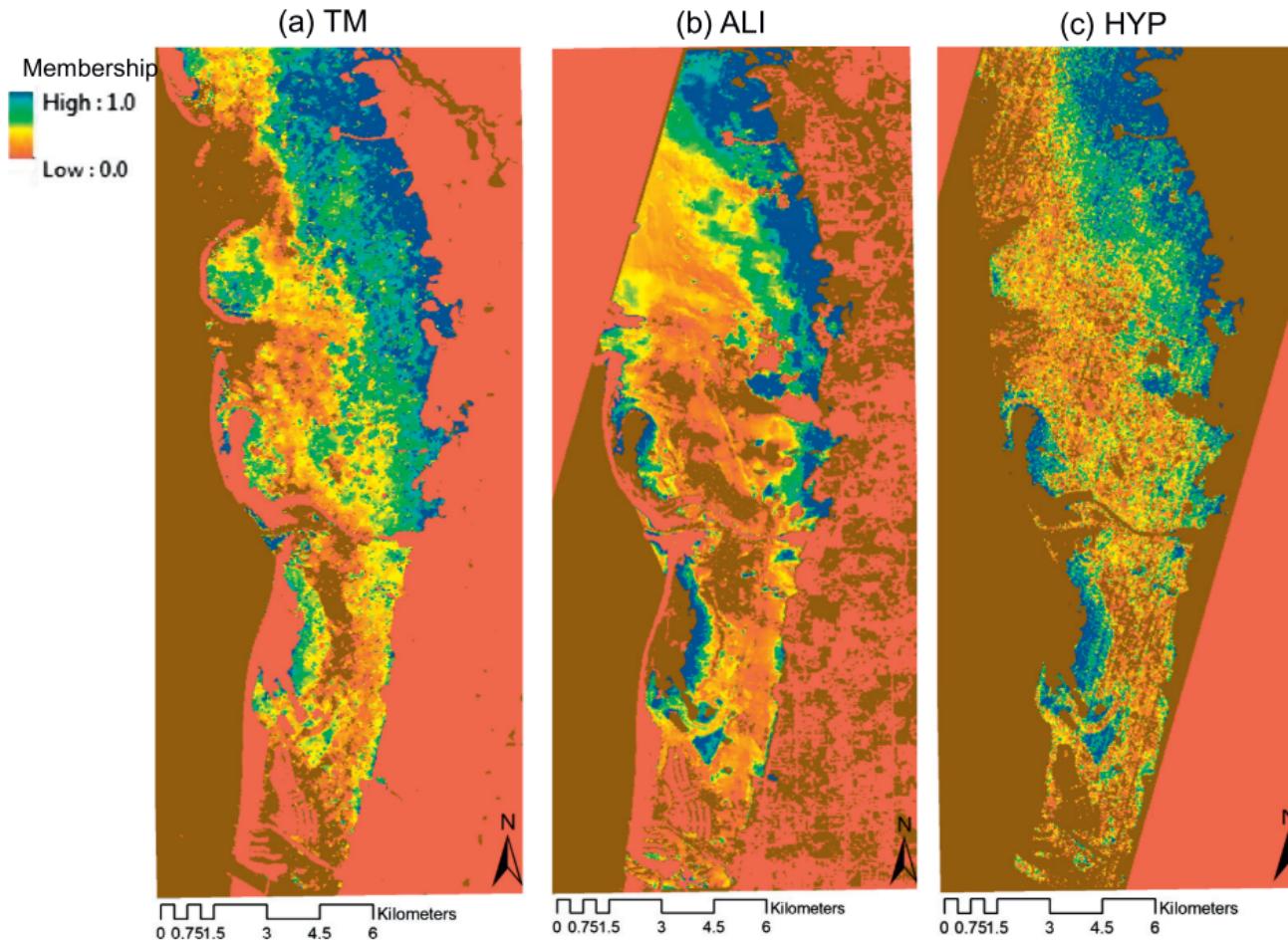
## Assessing of seagrass abundance

- Combining the calculation of NDVI, LAI and classification methods.
- Change detection,
- Two major components:
  - (1)image optimization and atmospheric correction preprocessing for detailed seagrass classification
  - (2) a fuzzy synthetic evaluation model for assessing seagrass abundance.



# Assessing of seagrass abundance

LAI Membership Maps



LAI membership maps created using the three multiple regression models for LAI (Table 5) developed with data from the three sensors [TM (a), ALI (b), and HYP (c)] and field transect survey data

Source: Pu, R.; Bell, S. A protocol for improving mapping and assessing of seagrass abundance along the West Central Coast of Florida using Landsat TM and EO-1 ALI/Hyperion images. *ISPRS J. Photogramm. Remote Sens.* **2013**, *83*, 116–129.

# GIS data sharing platform

- Open reef data sharing on ArcGIS online



<http://ucfonline.maps.arcgis.com/home/item.html?id=a93efd6937a843e4b0fe513696c164f7>

# Outreach and community drone projects



PUBLIC SCHOLARS  
LISTEN-ENGAGE-PARTNER



# Thanks for your attention!

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