



# **sUAS for Biomass Monitoring using LiDAR and Multispectral Data**



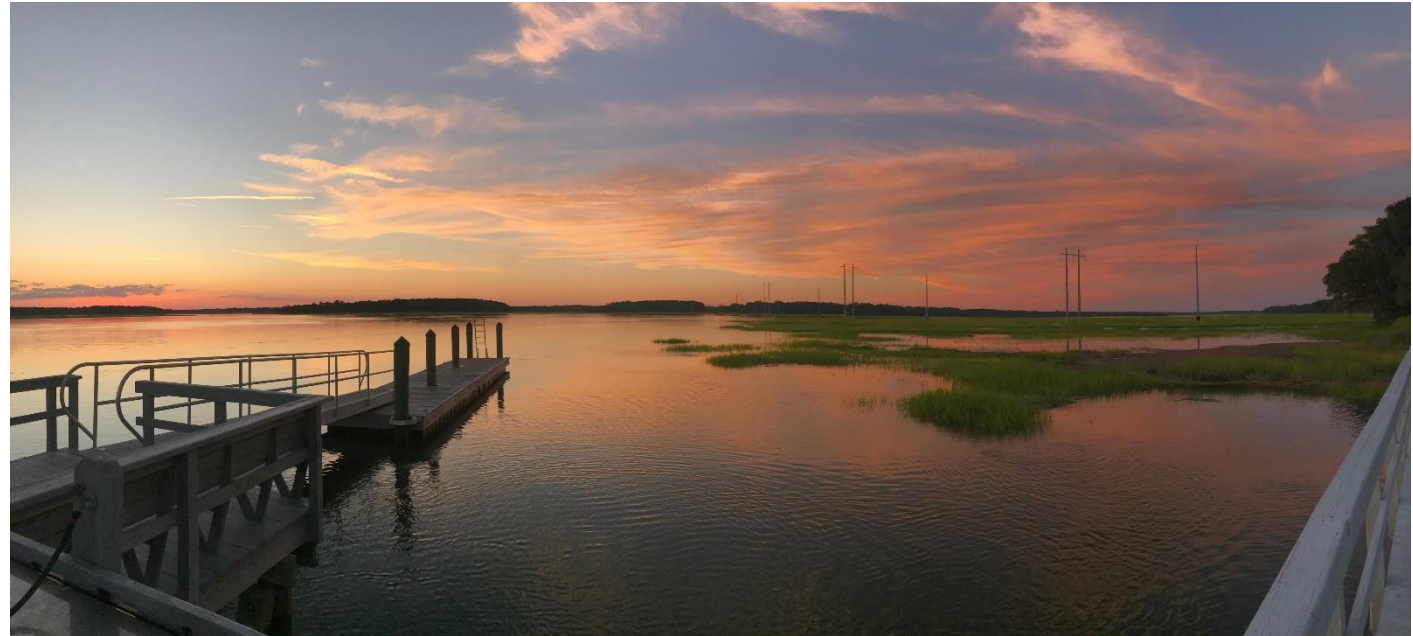
*Grayson R. Morgan*

*Cuizhen (Susan) Wang*

*University of South Carolina, Columbia, SC*



- **Tidal salt marsh** systems are incredibly important
  - **Protection**
  - **Essential habitat for wildlife**
  - **Economic**
  - **Carbon Sink**
- High percentage of Atlantic coast wetlands are found in South Carolina, second to Georgia (Wiegert and Freeman 1990)



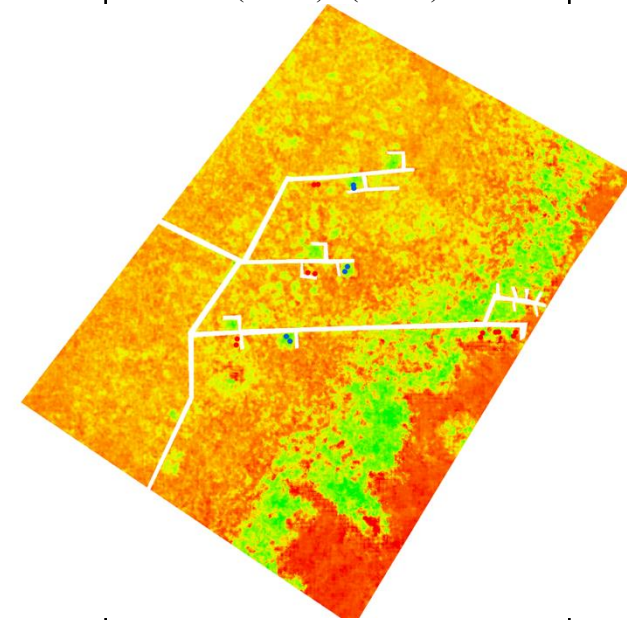


# Experiences on **drone ortho-imagery** for marsh mapping



## RGB Vegetation Indices

Index	Formula	Reference <sup>1</sup>
ExG	$2 \times G - R - B$	Jing et al., 2017
GCC or Green Ratio	$G / B + G + R$	Yue et al., 2017
GRVI	$(G - R) / (G + R)$	Jing et al., 2017
VARI	$(G - R) / (G + R - B)$	Cen et al., 2019
VDVI or GLA	$(2 * G - R - B) / (2 * G + R + B)$	Cen et al., 2019



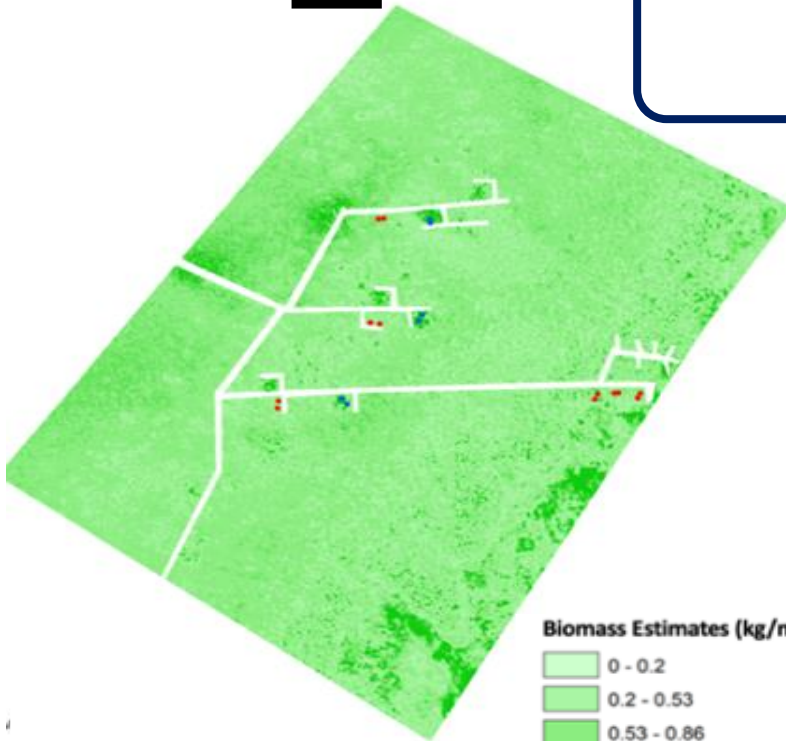


# Experiences on **drone ortho-imagery** for marsh mapping

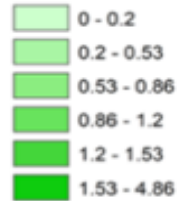
$$\text{Biomass} = 6943.5 * \text{ExG}^2 - 226.82 * \text{ExG} + 2.3477$$

$$\text{RMSE} = 0.598 \text{ kg/m}^2; R^2 = 0.376; P < 0.05$$

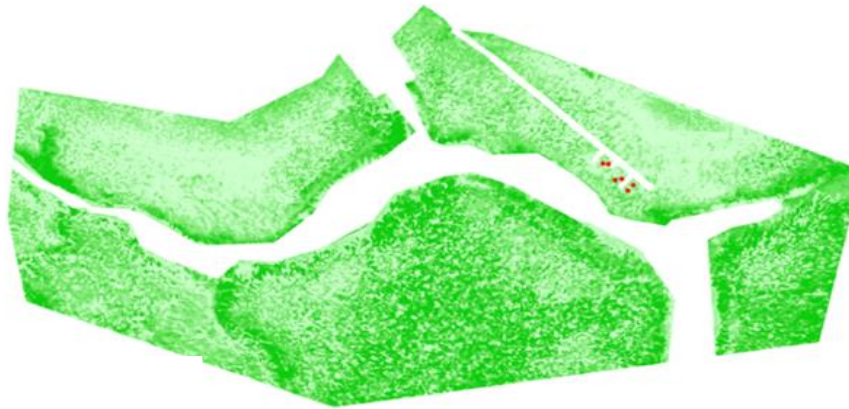
**GI**



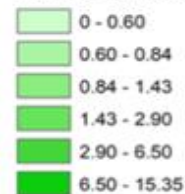
Biomass Estimates (kg/m<sup>2</sup>)



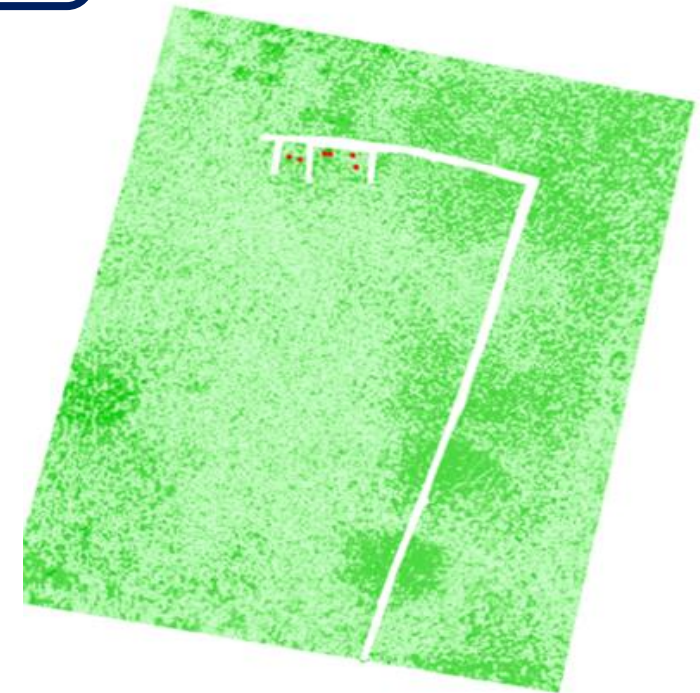
**OL-LM**



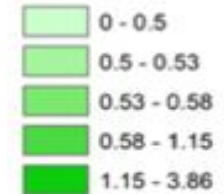
Biomass Estimates (kg/m<sup>2</sup>)



**OL-HM**



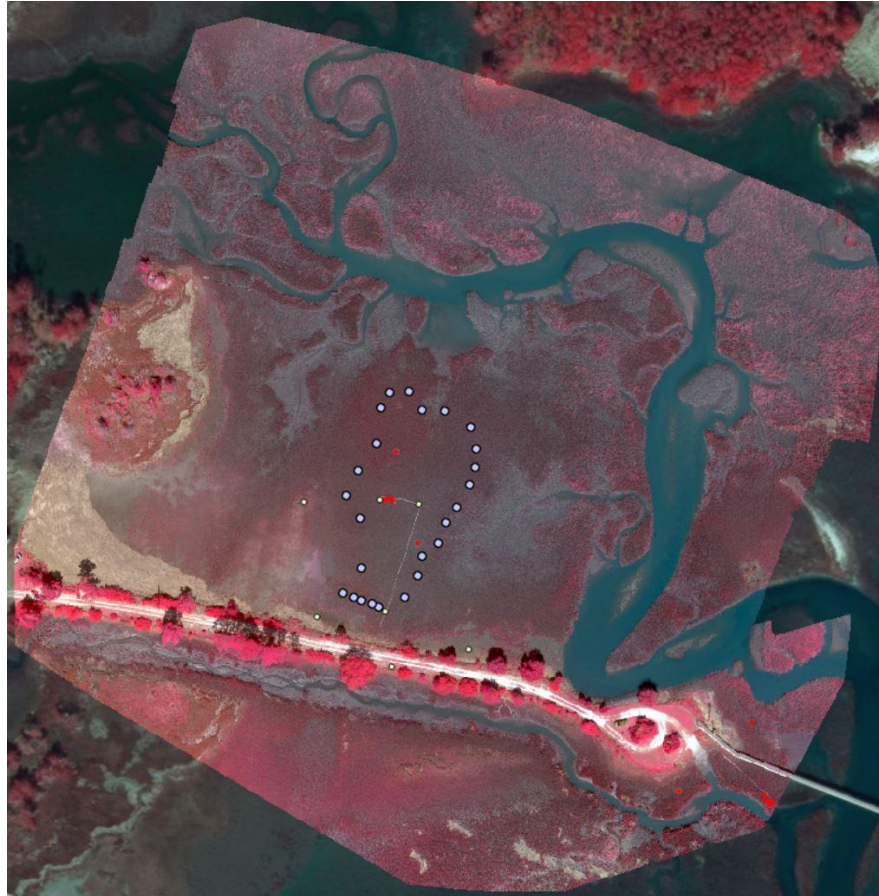
Biomass Estimates (kg/m<sup>2</sup>)



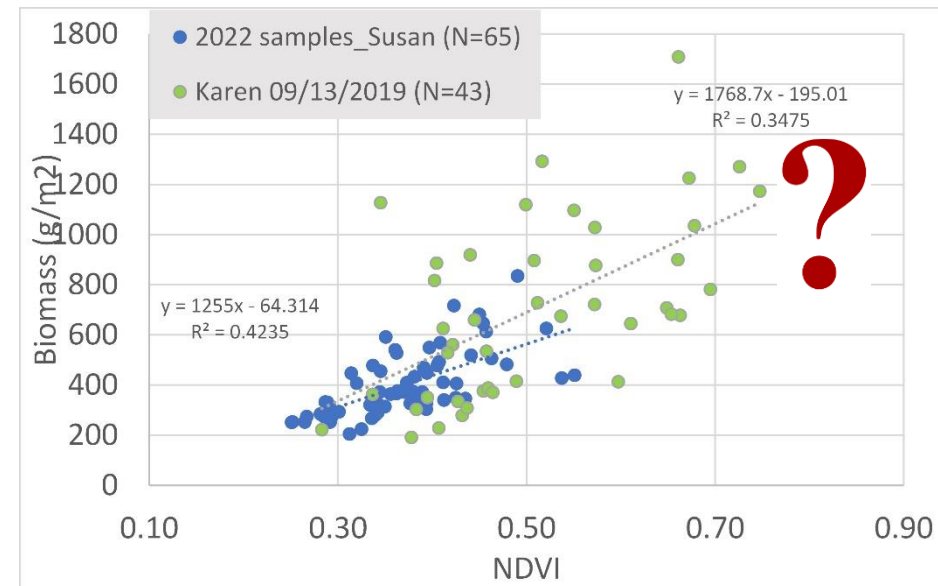


# Experiences on **drone ortho-imagery** for marsh mapping

Matrice100/RedEdge-M, 09/22/2022  
(Goat Island)



Drone-assisted marsh biomass experiments, North Inlet



(Credits: Dr. James Morris and Karen Sundberg at BMFL, USC)

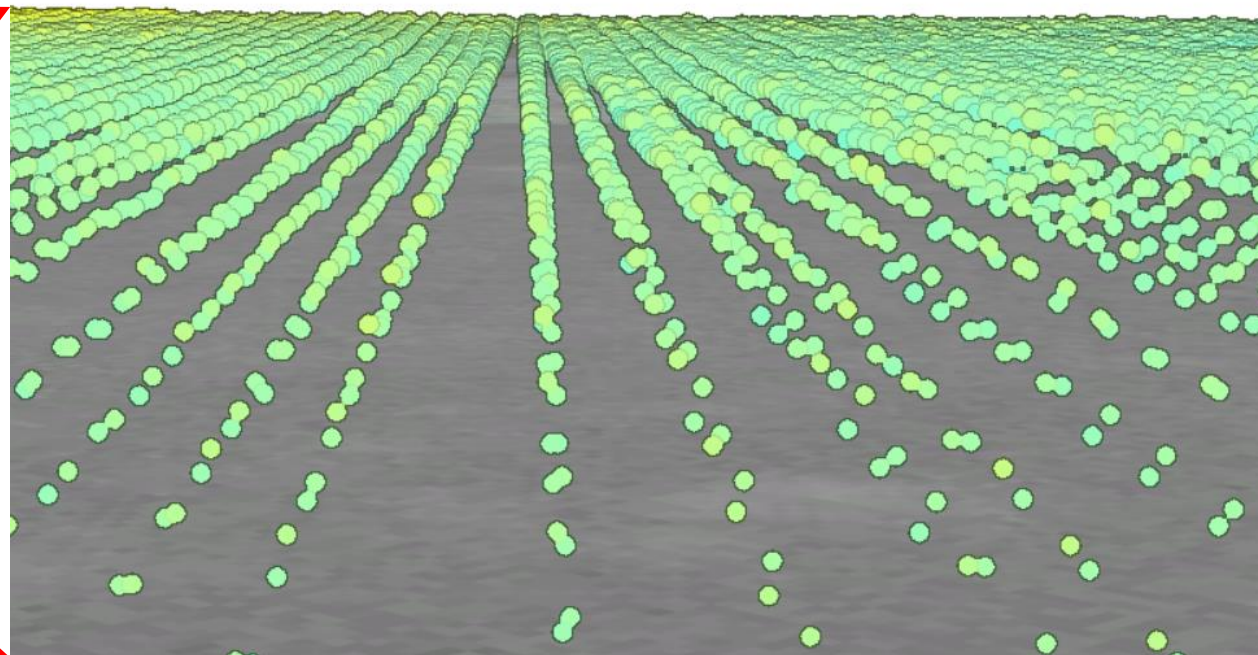
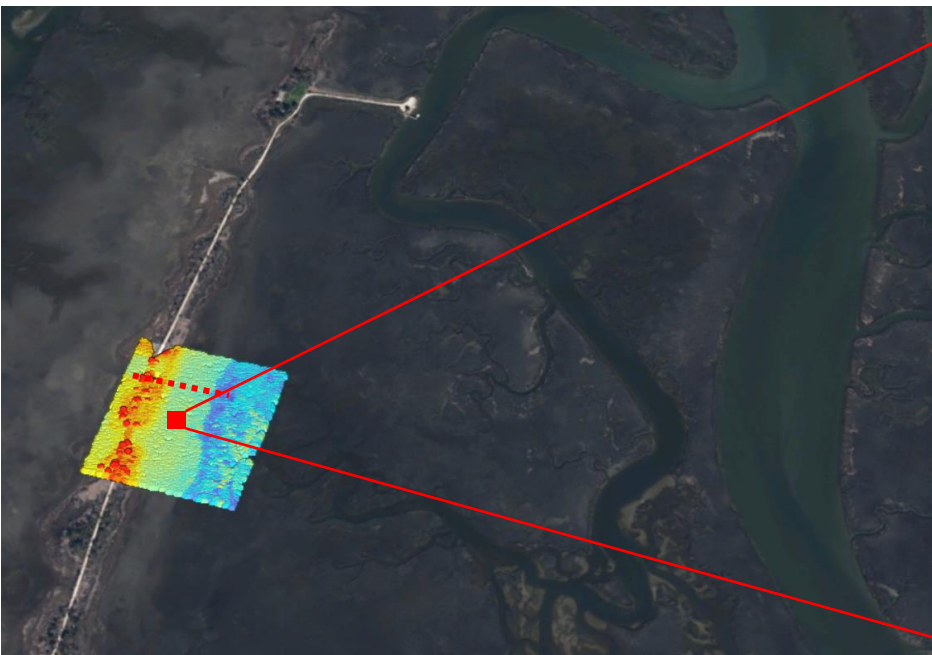


# Airborne Lidar in **coastal marshes**: not as good as we had hoped...

## Challenges on Lidar classification:

- ✓ *Tidal effects*
- ✓ *Gentle topography*
- ✓ *Short, sparse marsh plants*

**USGS Lidar Point Cloud (LPC):**  
Spacing = 2-3 ft





# Question: Could Drone Lidar play a better role in marshes?

- ✓ *Affordable: significantly reduced prices*
- ✓ *Flexible: flight parameters*
- ✓ *Mass points: much denser point clouds*

Low-cost drone Lidar systems, ready-to-go package (<20k):



DJI Zenmuse L1  
(Lidar +RGB)  
\$13,000



ROCK Robotic R2A  
(Livox Avia + RGB)  
\$19,000

(240,000 points/s)



# ❑ **Field experiment:** August – September 2022, North Inlet *(Baruch Marine Field Laboratory, USC)*

## ✓ **Drone Lidar missions:** 08/31 – 09/01

Vendor: Back Forty Aerial Solutions, Columbia, SC



## ✓ **Multispectral drone missions and field survey:** 09/20 – 09/24

NASA EPSCOR Project team:

*Susan Wang,  
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Alex Fullham,  
Naser Lessani*

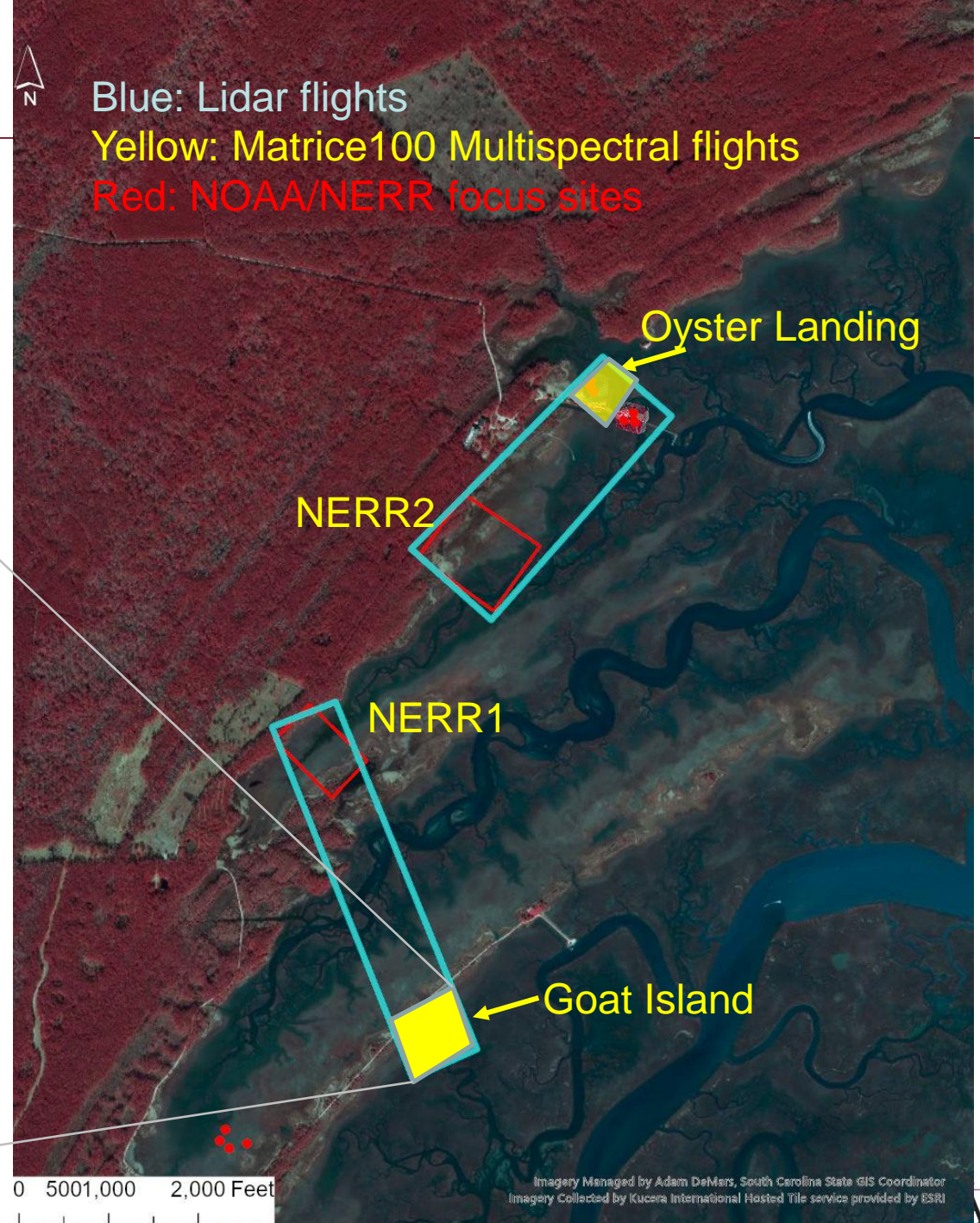
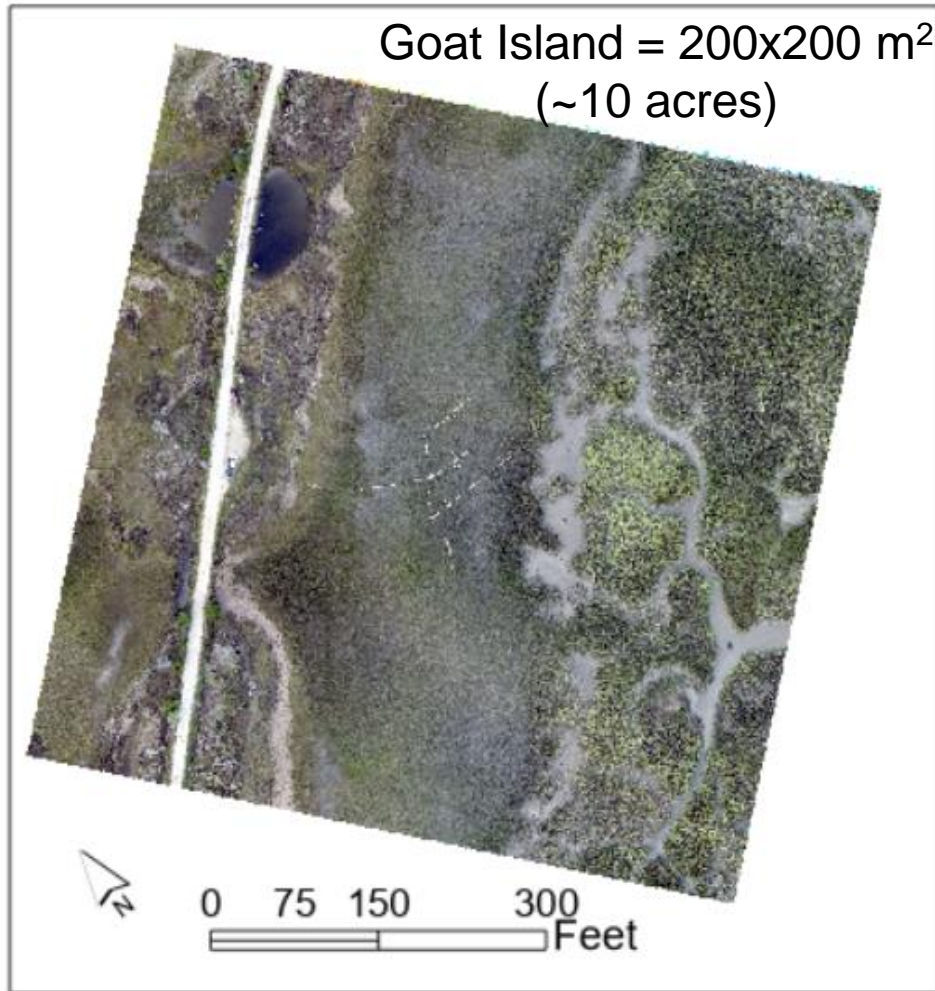






# Drone Lidar data footprint

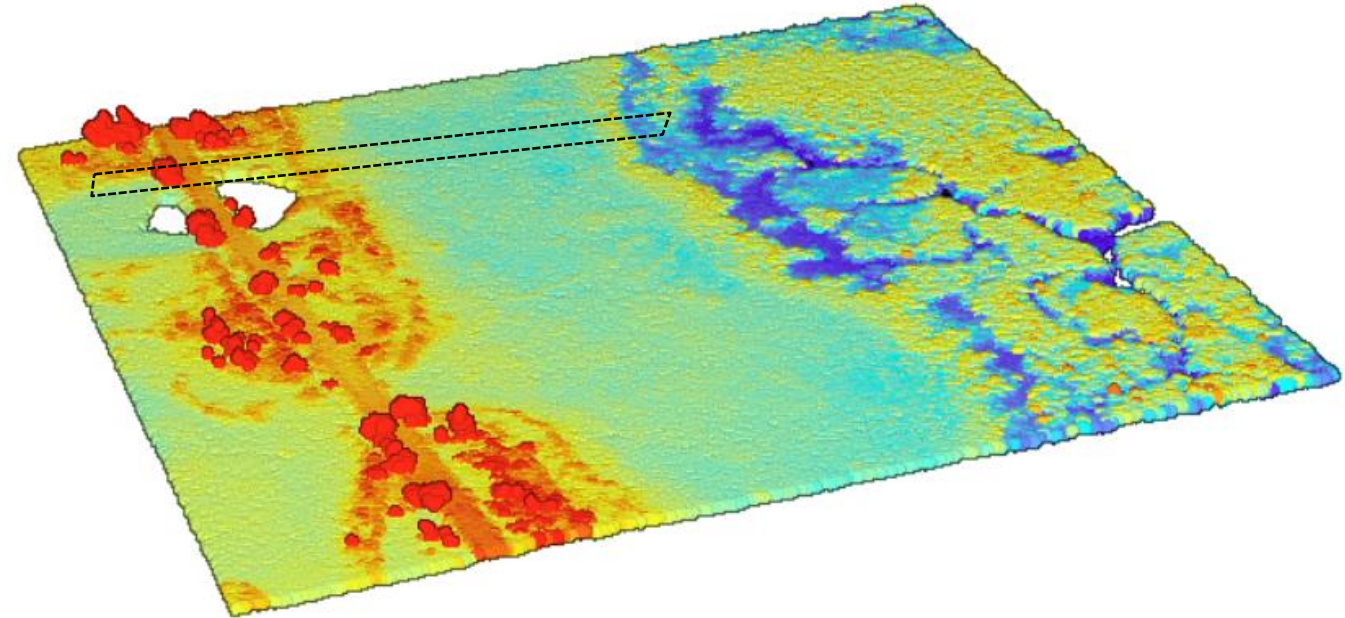
## R2A/RGB ortho image



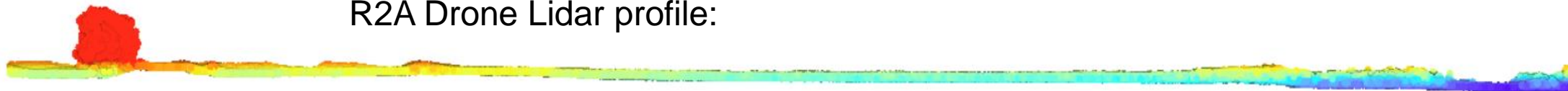


# R2A/Lidar point cloud

Goat Island subset	Drone Lidar	USGS Lidar
Point count	29,561,422	95,269
Spacing	0.036 m	0.639 m
Point density (#/m <sup>2</sup> )	771	2.43



**Caution: R2A Lidar points only have one return in marsh fields!**





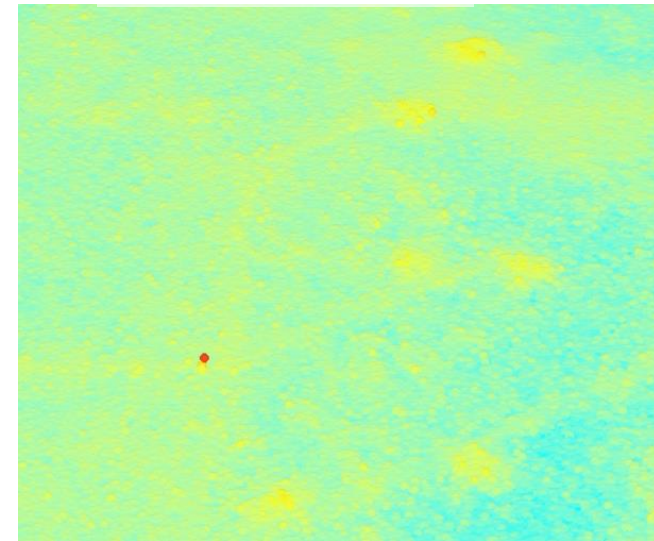
# Visual comparison: Drone Lidar vs. USGS Lidar

High marsh

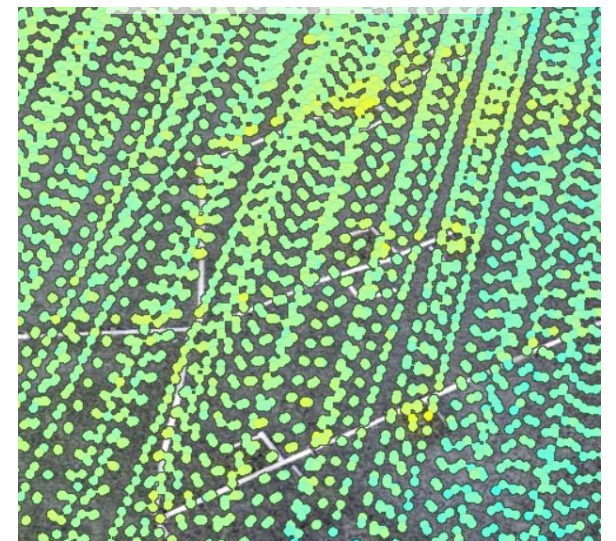
Ortho-image



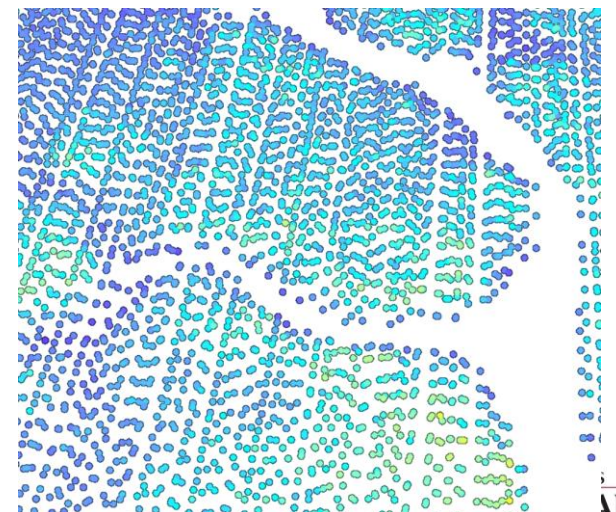
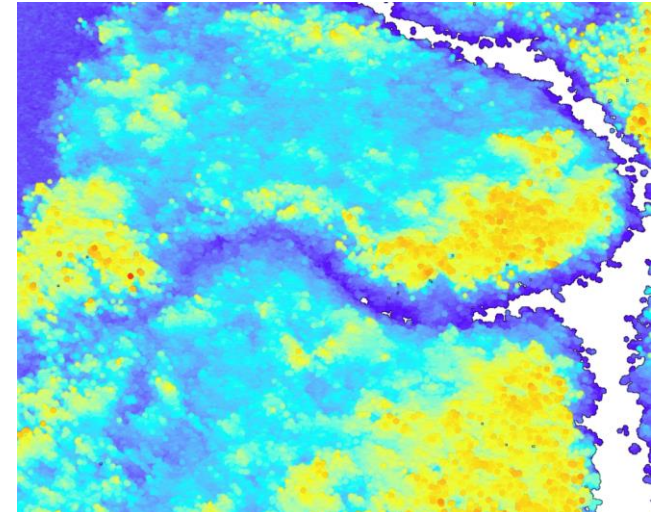
Drone Lidar



USGS Lidar



Low marsh

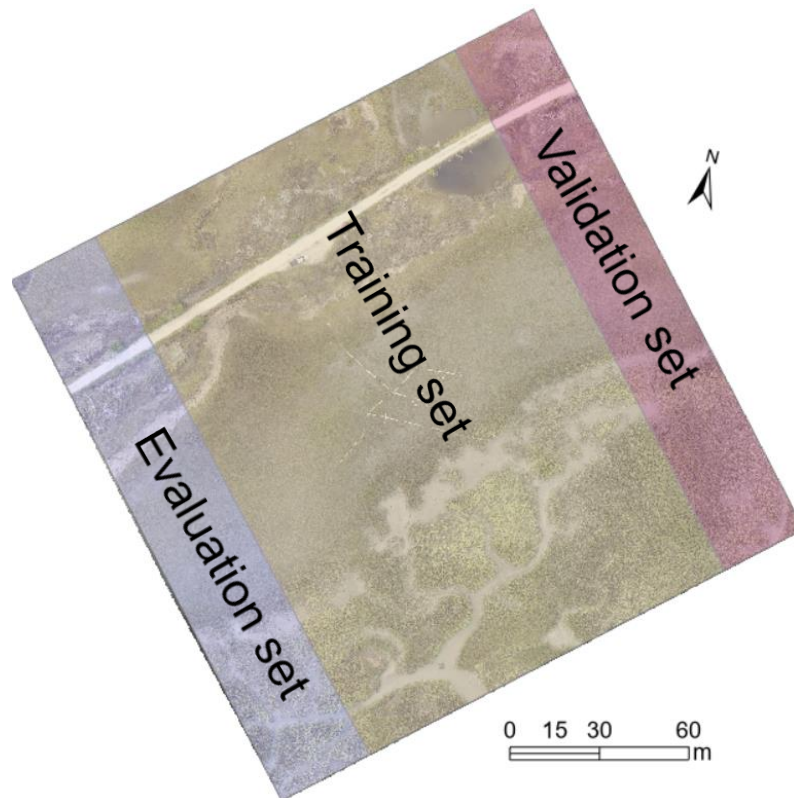




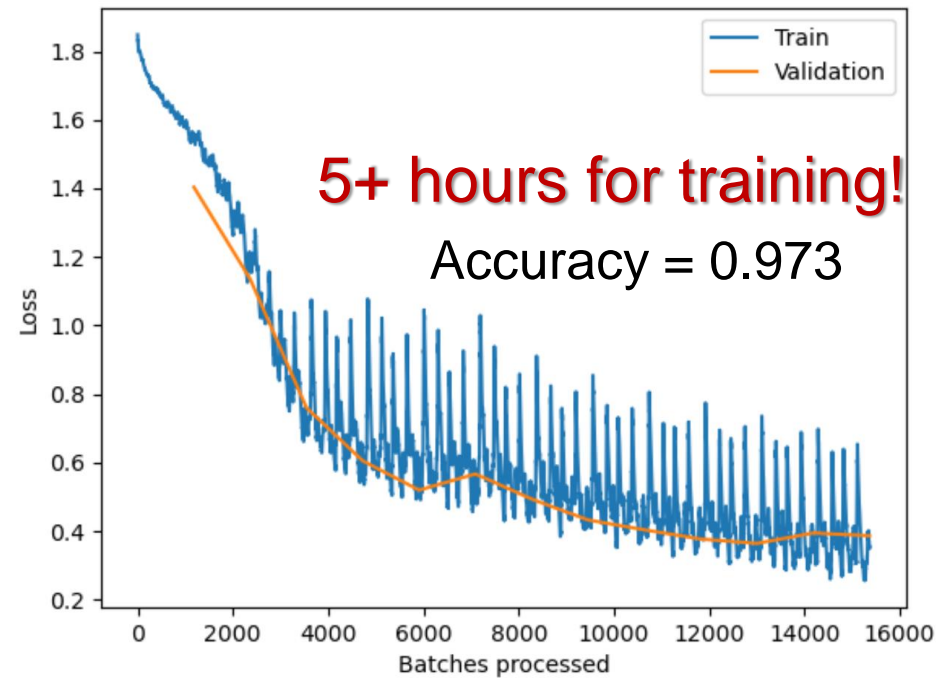
# PointCNN: Deep learning for point cloud classification



We only tested two classes:  
**Vegetation** and **Ground**.



### Training/validation loss curves

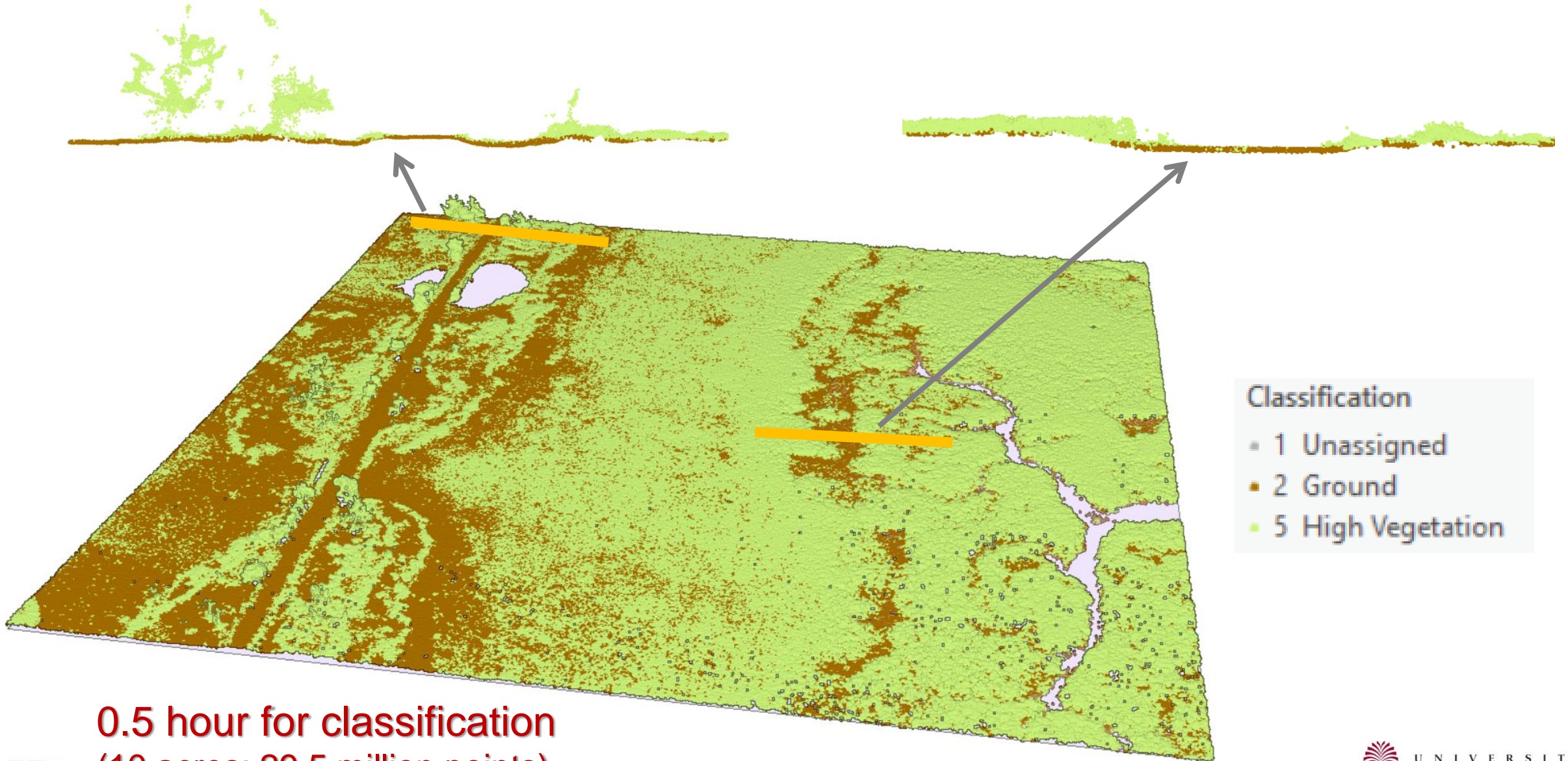


### Evaluation matrix

CLASS_CODE	CLASS_NAME	PRECISION	RECALL	F1_SCORE
1	background	0.002441	0.032227	0.004538
2	Ground	0.969974	0.957472	0.963683
5	High Vegetation	0.992208	0.982159	0.987158



# Point classification results at Goat Island

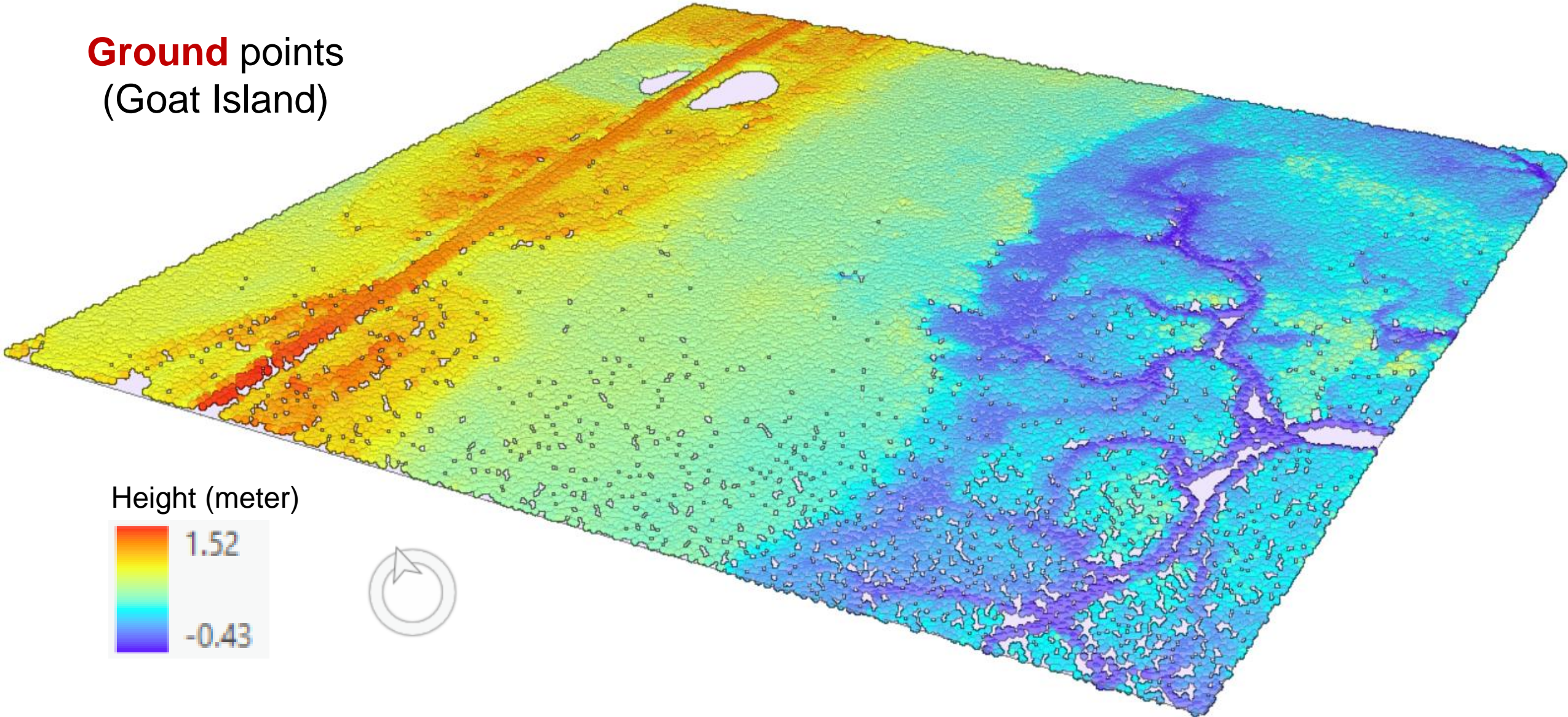


0.5 hour for classification  
(10 acres; 29.5 million points)



# Bare Earth Surfaces

**Ground** points  
(Goat Island)

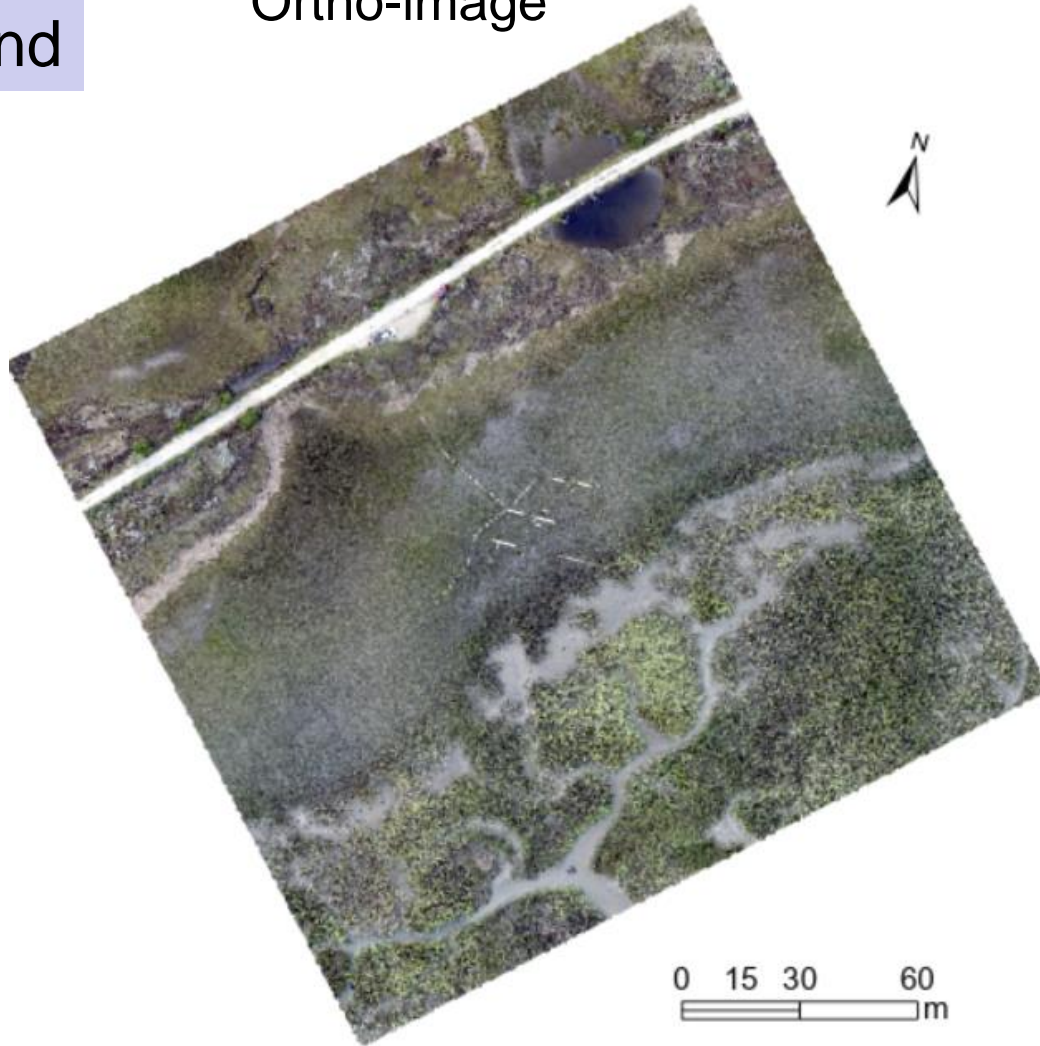




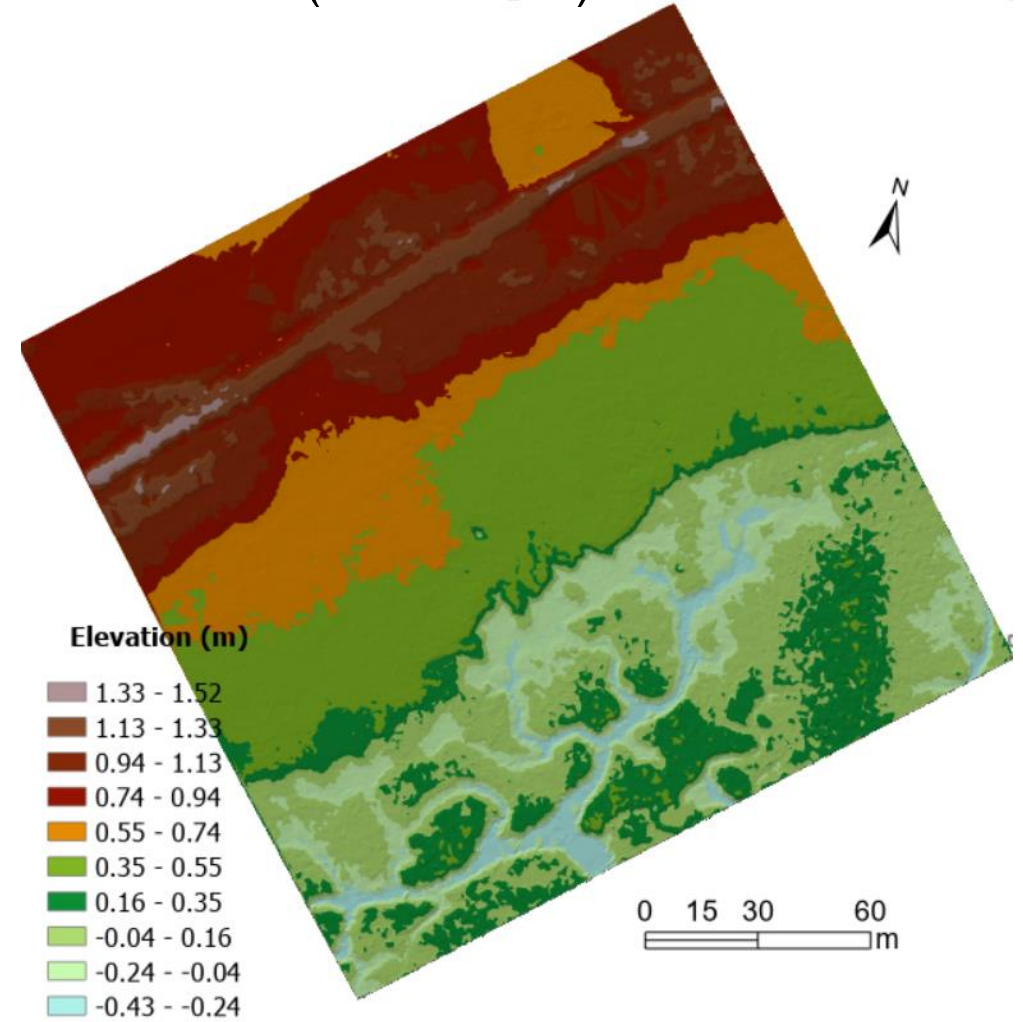
# Topography in marshes: bare Earth surfaces

Goat Island

Ortho-image

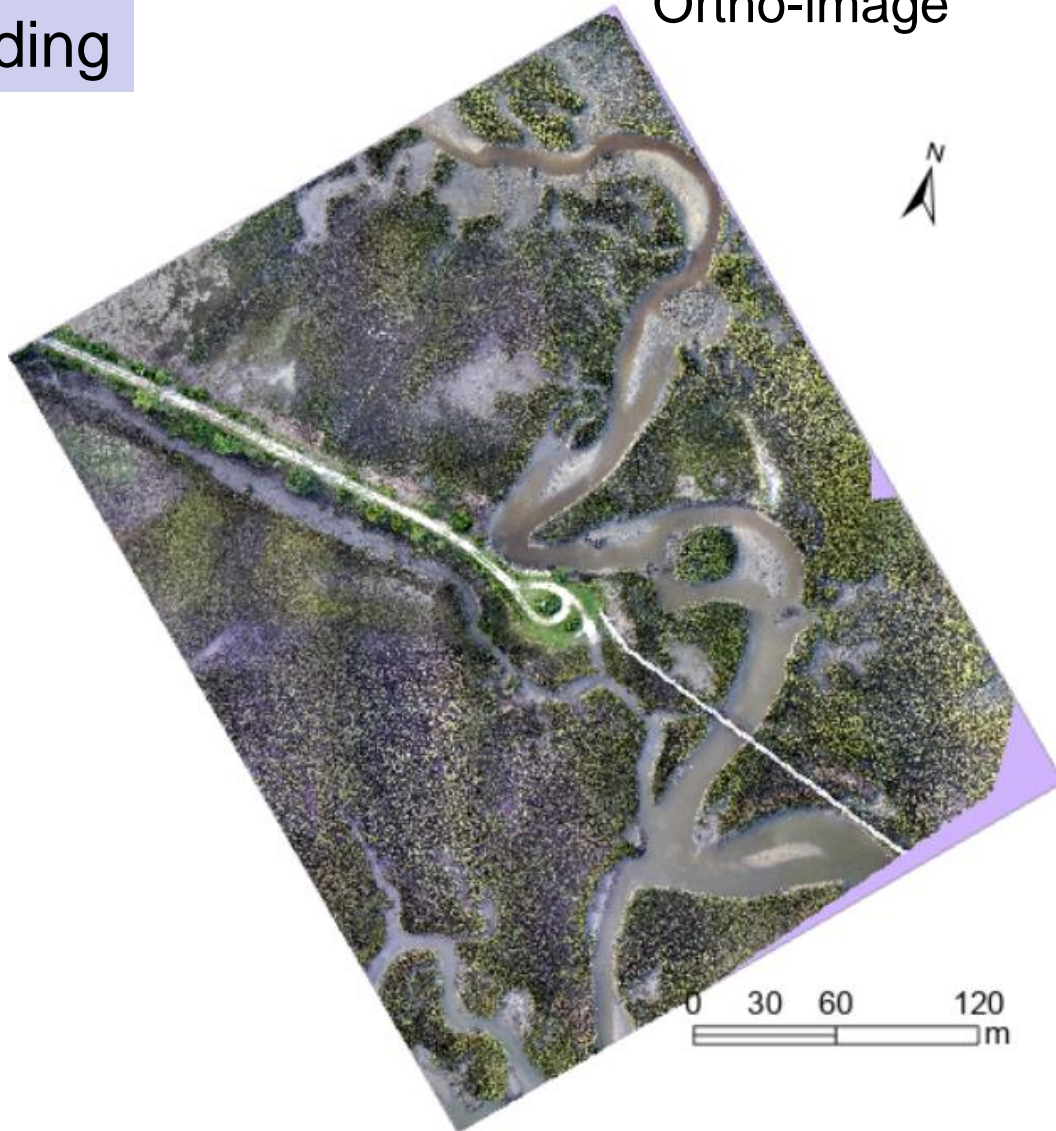


TIN (Bare Earth)

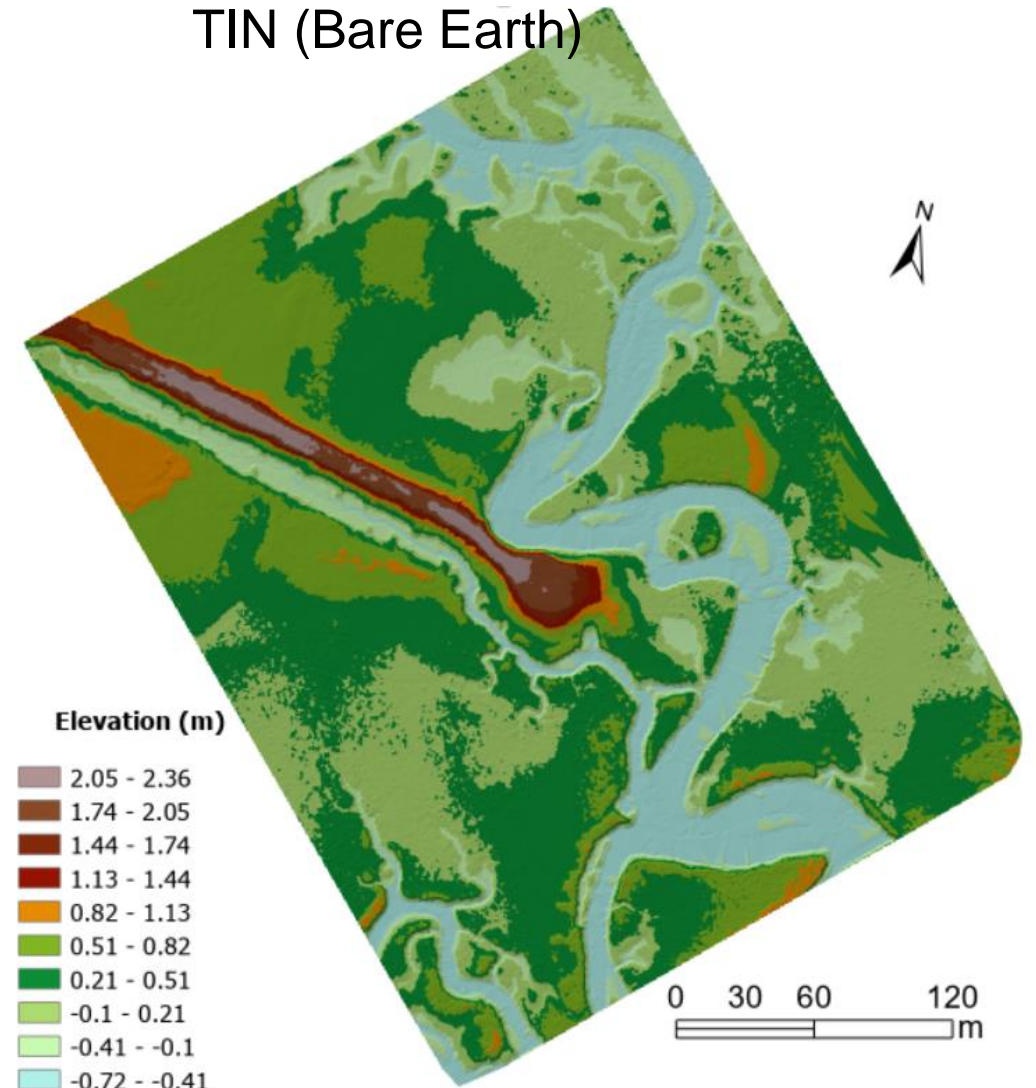


# Oyster Landing

## Ortho-image



## TIN (Bare Earth)



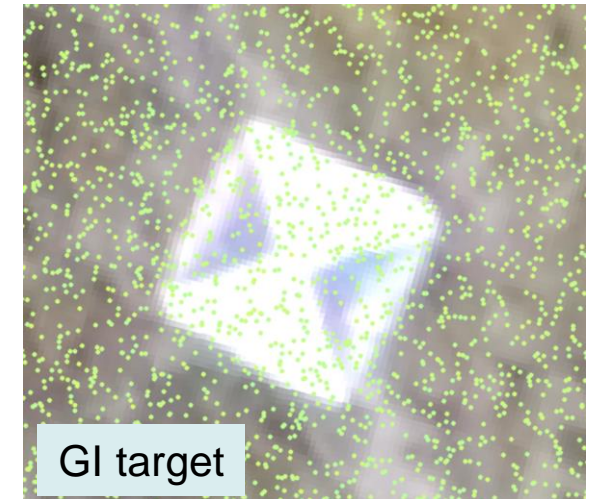
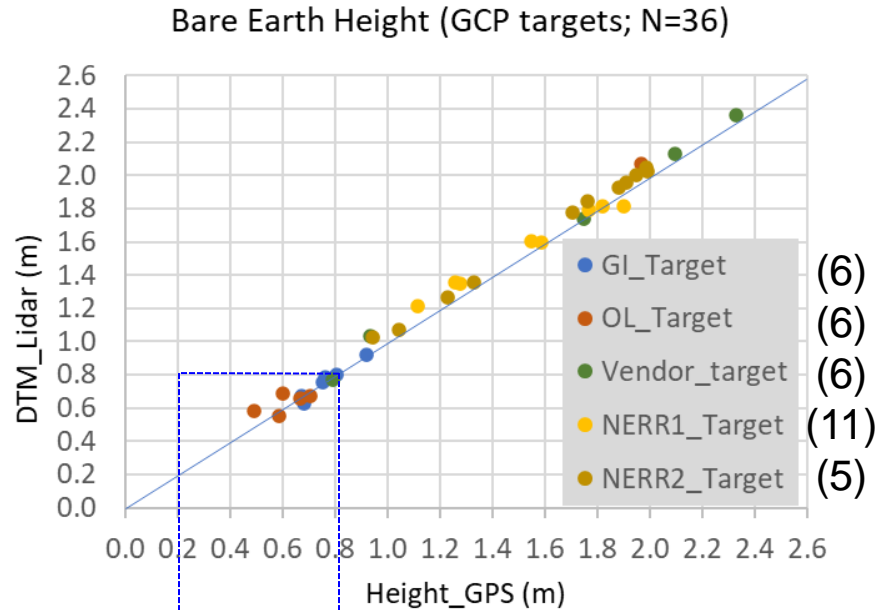




# How good is drone Lidar on extracting Bare Earth surface (DTM)?

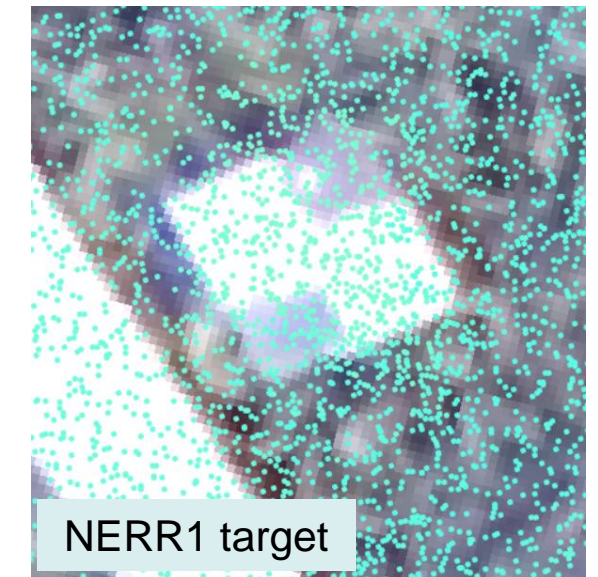
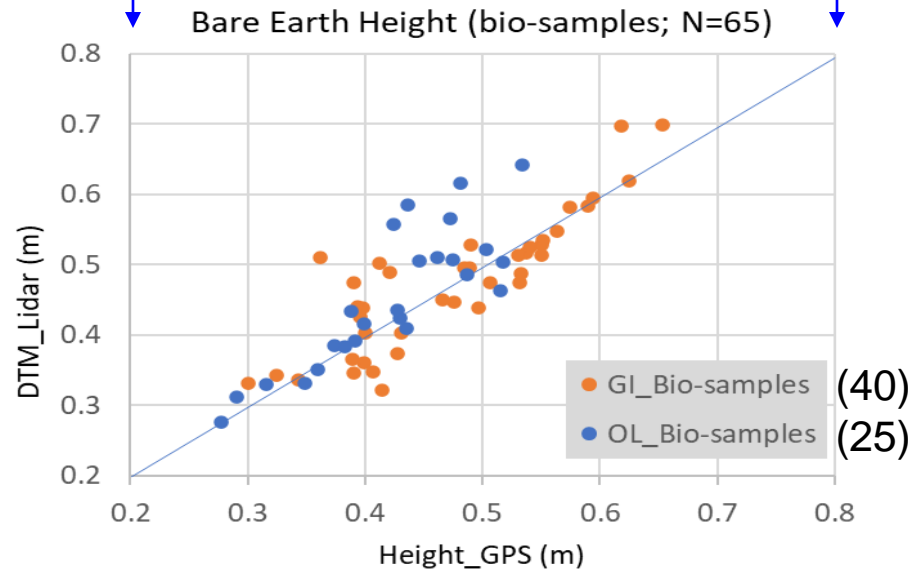
DTM  
at Ground Control Targets (36)

**RMSE = 5.55 cm**



DTM  
at Biomass samples (65)

**RMSE = 5.33 cm**





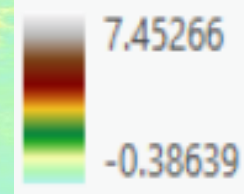
# 3D marsh modeling

Marsh canopy height:

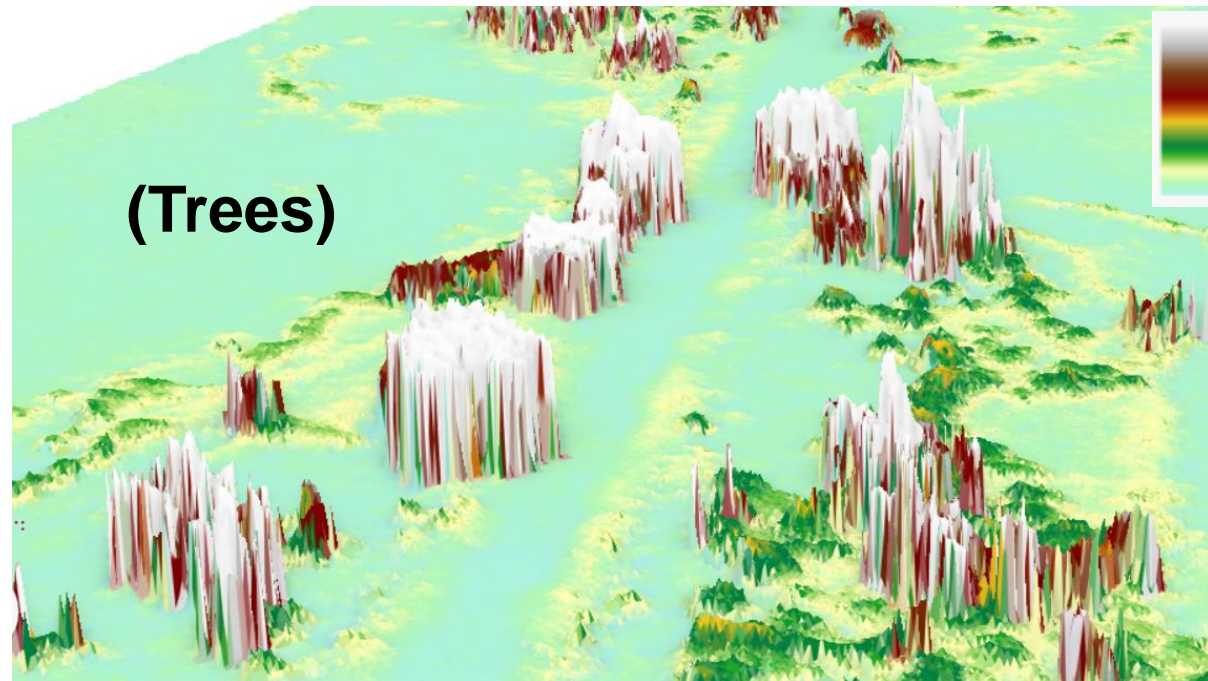
$$H = \text{DSM} - \text{DTM}$$



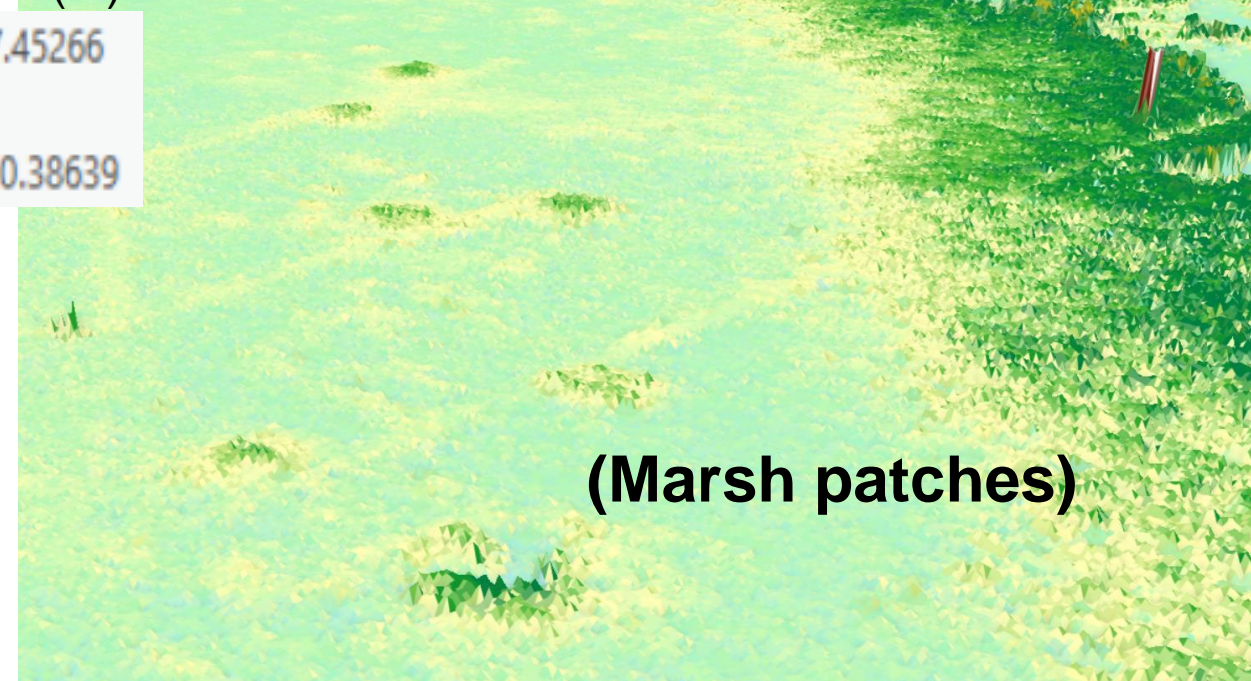
Height (m)



(Trees)



(Marsh patches)



(Goat island)



# Marsh biomass model (preliminary)

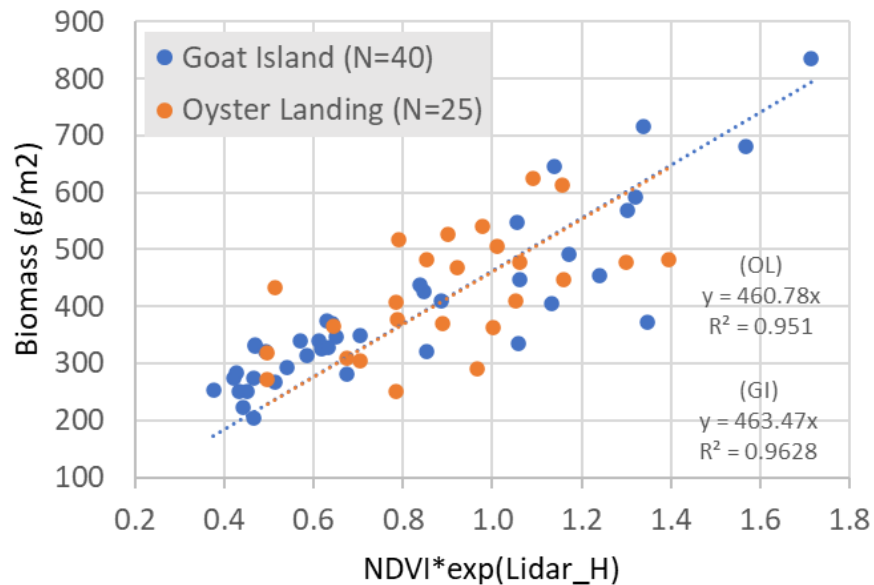
Model calibration (Goat Island, N=40):

$$\text{Marsh Biomass (g/m}^2\text{)} = 463.47 \times \text{NDVI} \times e^H$$

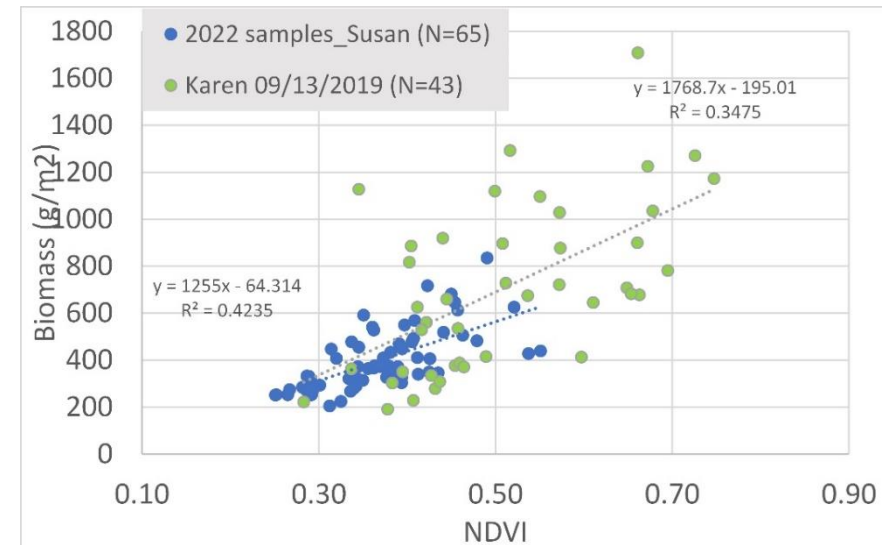
Model validation (Oyster Landing, N=25):

$$\text{RMSE} = 96.81 \text{ g/m}^2$$

Marsh biomass model with Lidar + multispectral camera



Recall what we got without Height information:





# □ Drone Lidar for 3D marsh mapping: Pros & Cons

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## Pros

- Flexible, large-coverage data acquisition
- **5cm** vertical accuracy on DTM
- Much **finer spacing** than airborne Lidar
- Deep Learning: automated mass data analysis
- Broader applications along SC Coast

## Cons

- Hardware/software maintenance
- Rapidly evolving systems
- **Single returns** in marshes
- Financial/operational/data analysis challenges
- Time commitment



# Questions?

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