

Assessing Ecosystem Health and Carbon Capture of Coastal Restoration

USING DRONES, REMOTE SENSING
AND MACHINE LEARNING

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**CDM
Smith**

Agenda



Blue Carbon

Study Overview

Sensors

Structure from Motion (SfM)

Elevation Models

Multispectral Indices

Biomass Model Parameters

Biomass Model Results

Blue Carbon

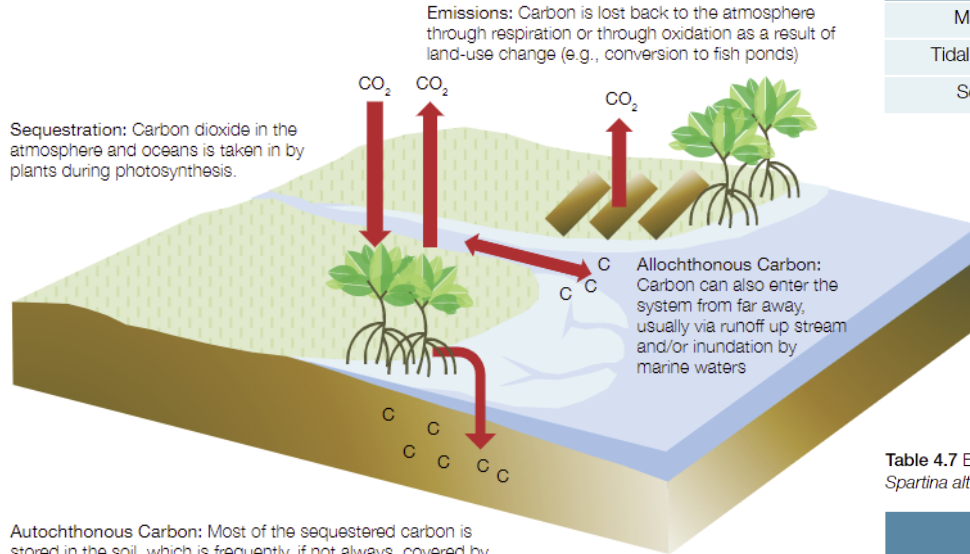


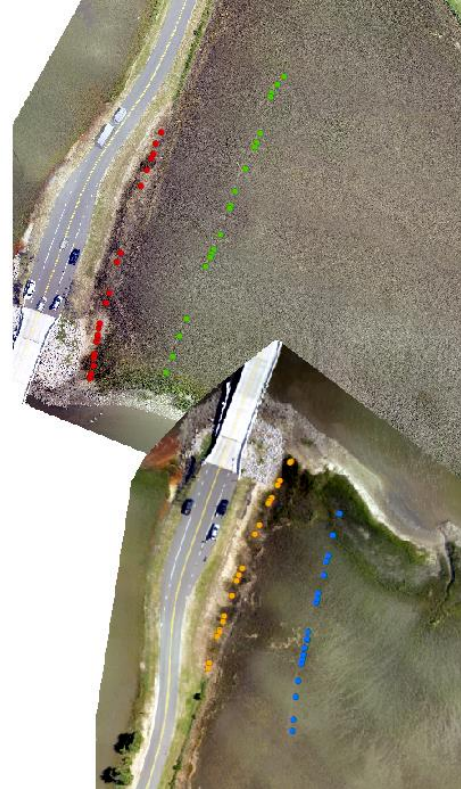
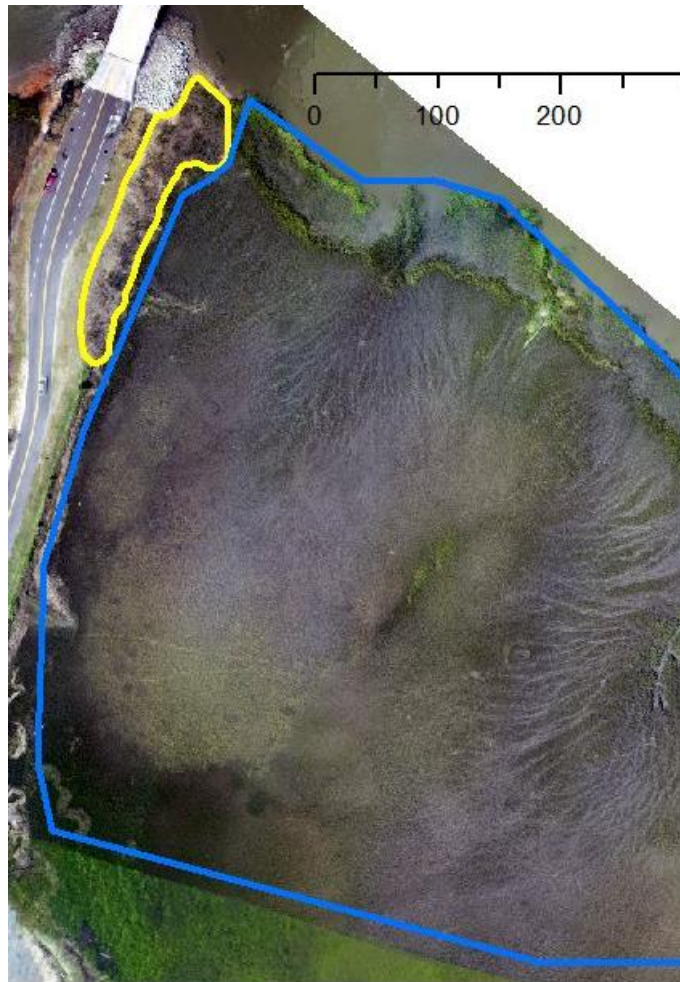
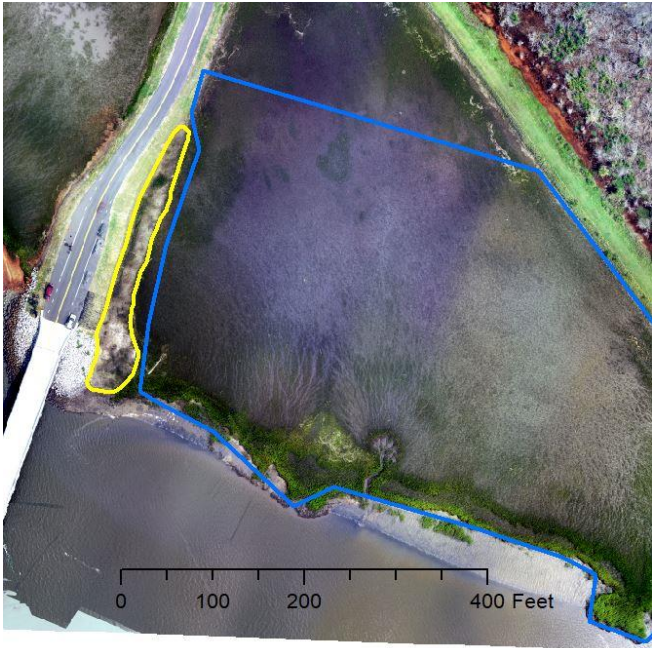
Table 1.2 Mean and range values of soil organic carbon stocks (to 1 m depth) for mangrove, tidal marsh, and seagrass ecosystems and CO₂ equivalents. Examples of how carbon is distributed amongst the different ecosystems and the variation within each ecosystem (IPCC 2013)

ECOSYSTEM	CARBON STOCK Mg/ha	RANGE Mg/ha	CO ₂ Mequiv/ha
Mangrove	386	55 – 1376	1415
Tidal salt marsh	255	16 – 623	935
Seagrass	108	10 – 829	396

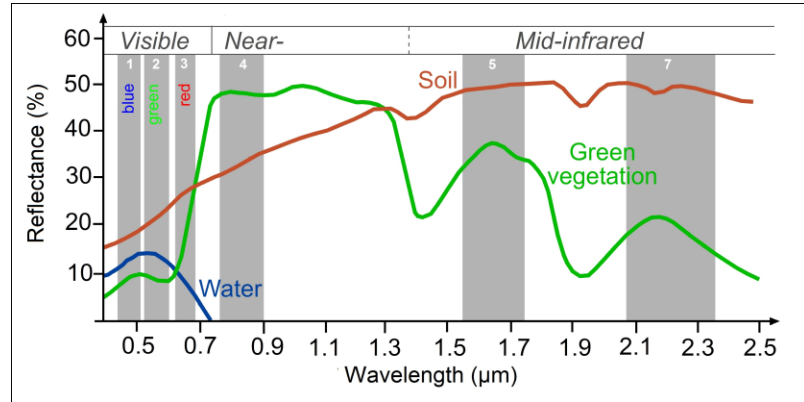
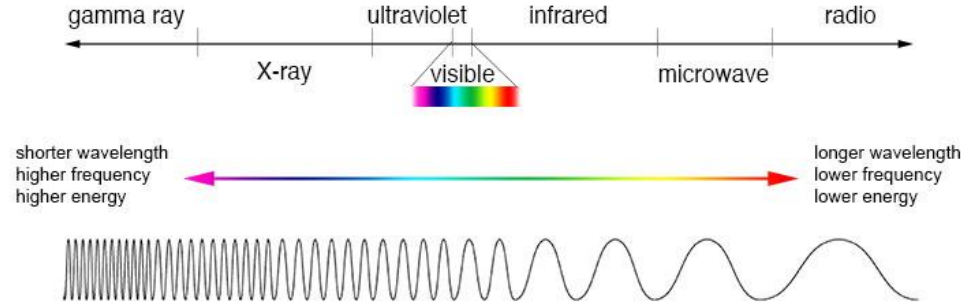
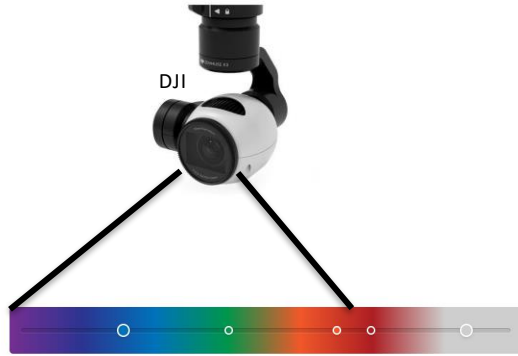
Table 4.7 Equations to predict belowground root and rhizome biomass on the basis of aboveground biomass of *Spartina alterniflora* (Gross *et al.* 1991)

EQUATION	ABOVE-GROUND COMPONENTS INCLUDED (FOR AN ENTIRE PLOT)	r ²
$\ln(\text{Live belowground biomass, g}) = 0.718 \times \ln(\text{Live Aboveground biomass, g}) + 2.646$	Living leaves and stems	0.86
$\ln(\text{Live Belowground biomass, g}) = 0.700 \times \ln(\text{Live Aboveground biomass, g}) + 3.051$	Living leaves only	0.85
$\ln(\text{Live Belowground biomass, g}) = 0.713 \times \ln(\text{Total Aboveground biomass, g}) + 2.235$	All live and dead aboveground	0.86

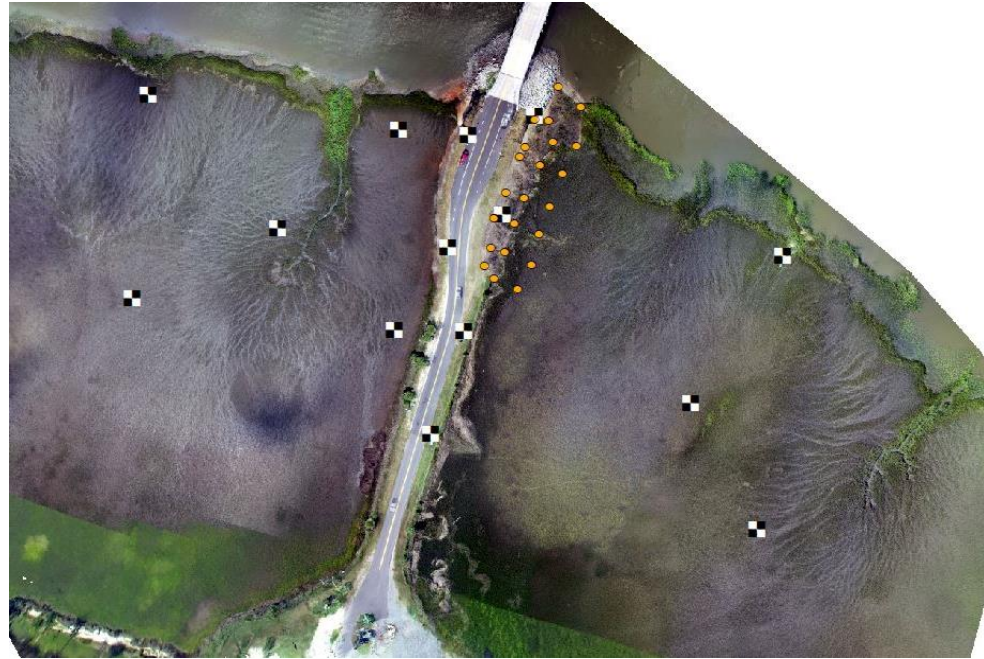
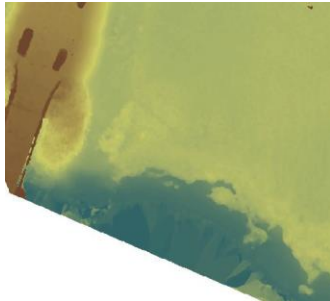
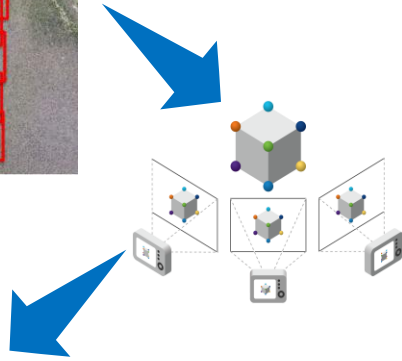
Study Overview



Sensors

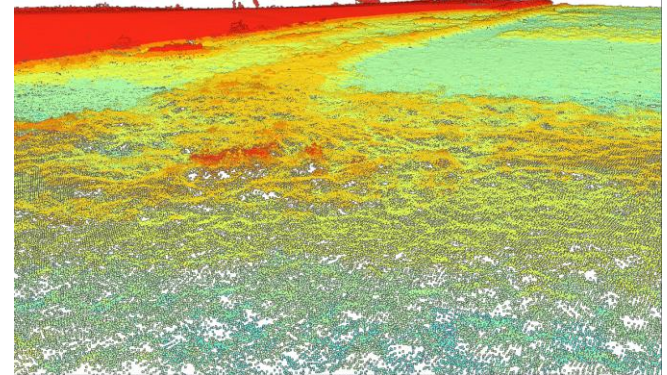
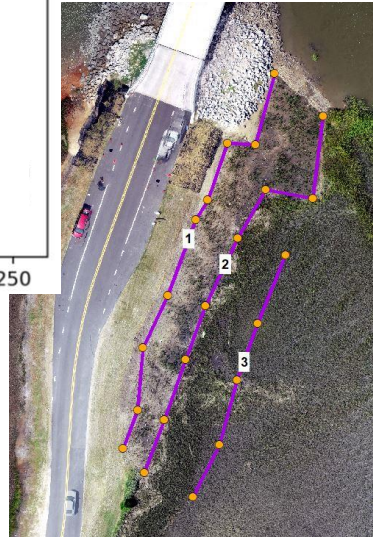
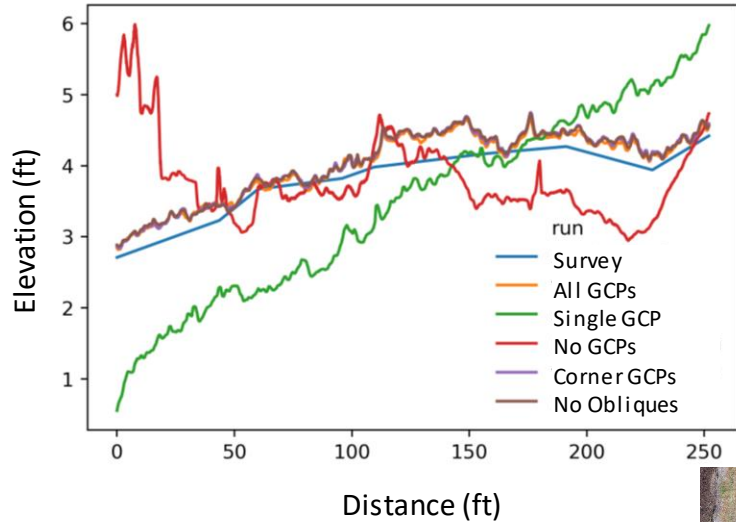


Structure from Motion (SfM)



Elevation Models

Ground Control Processing Comparison
(Transect 1)



SfM Point Cloud Density	
Flight Height (ft)	Density (pts/0.25m*0.25m)
125	40
400	4

Multispectral Indices

Normalized Difference Vegetation Index (NDVI)

$$\frac{\text{NIR} - \text{Red}}{\text{NIR} + \text{Red}}$$

Fractional Vegetation Coverage (FVC)

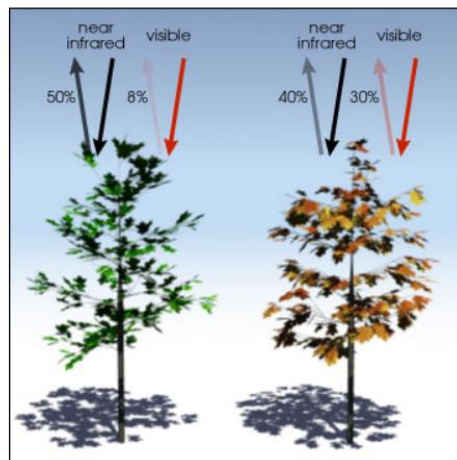
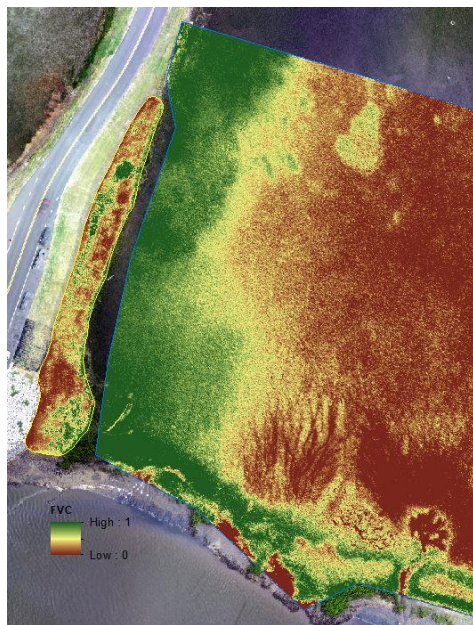
$$\frac{\text{NDVI} - \text{NDVI}_{\text{soil}}}{\text{NDVI}_{\text{veg}} - \text{NDVI}_{\text{soil}}}$$

Normalized Difference Red Edge (NDRE)

$$\frac{\text{NIR} - \text{Red Edge}}{\text{NIR} + \text{Red Edge}}$$

Normalized Difference Water Index (NDWI)

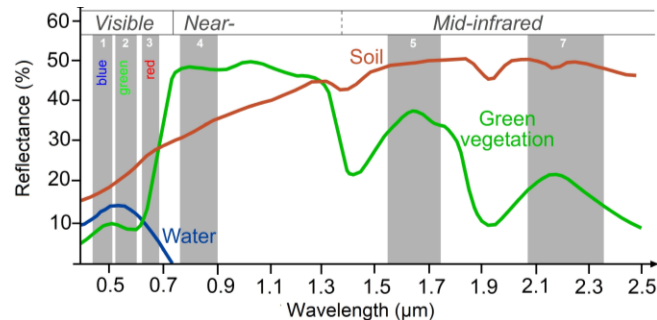
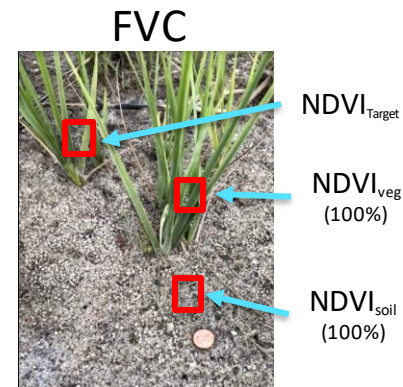
$$\frac{\text{Green} - \text{NIR}}{\text{Green} + \text{NIR}}$$



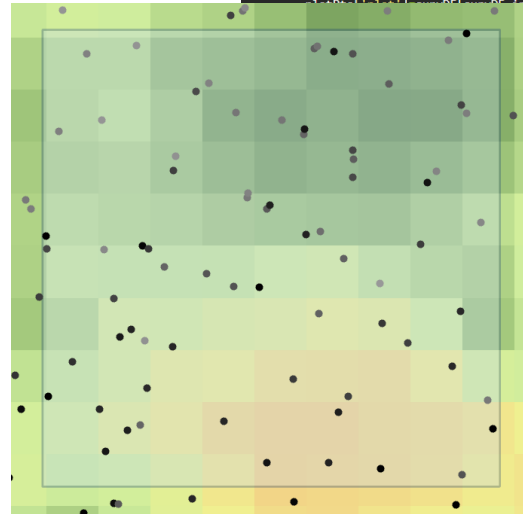
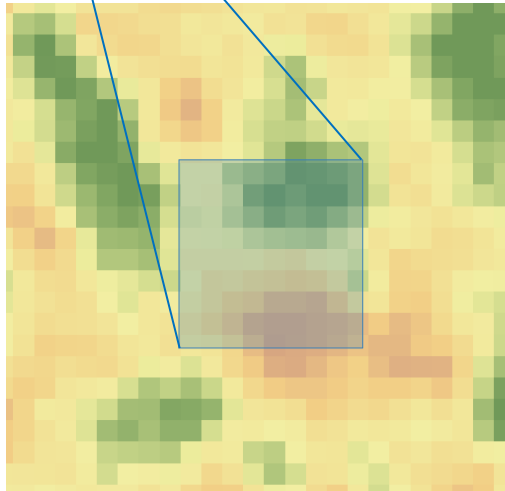
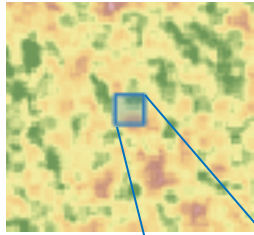
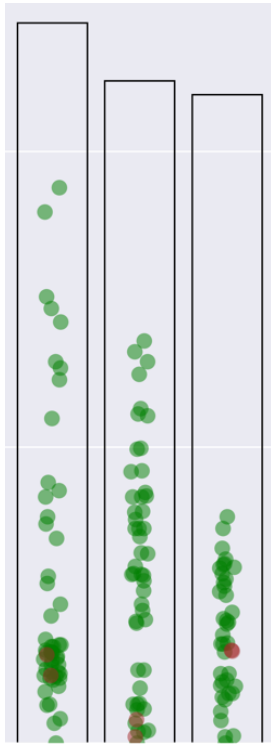
$$\frac{(0.50 - 0.08)}{(0.50 + 0.08)} = 0.72$$

$$\frac{(0.4 - 0.30)}{(0.4 + 0.30)} = 0.14$$

NASA

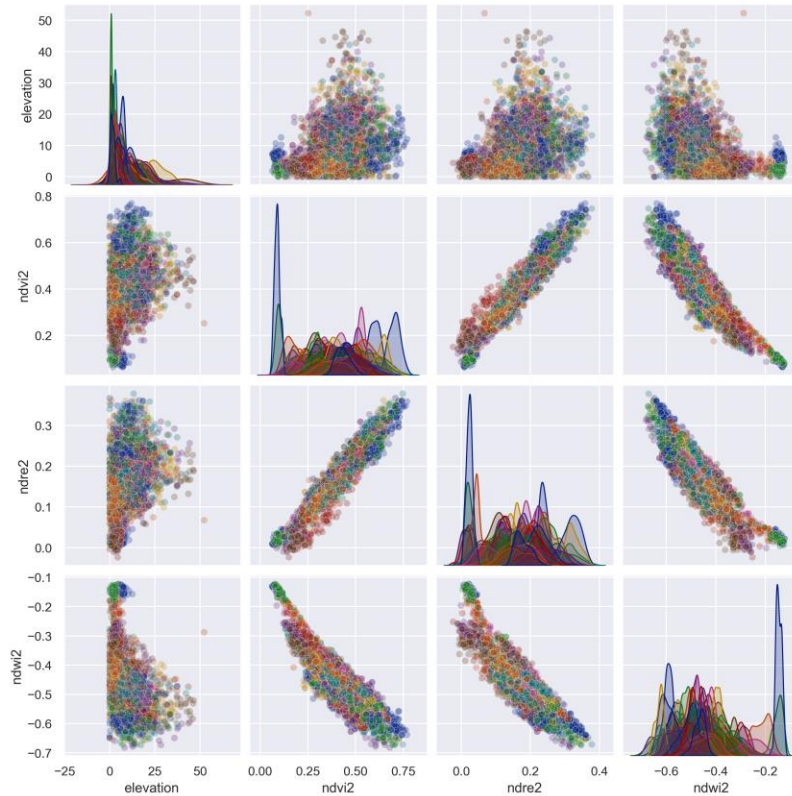


Biomass Model Parameters

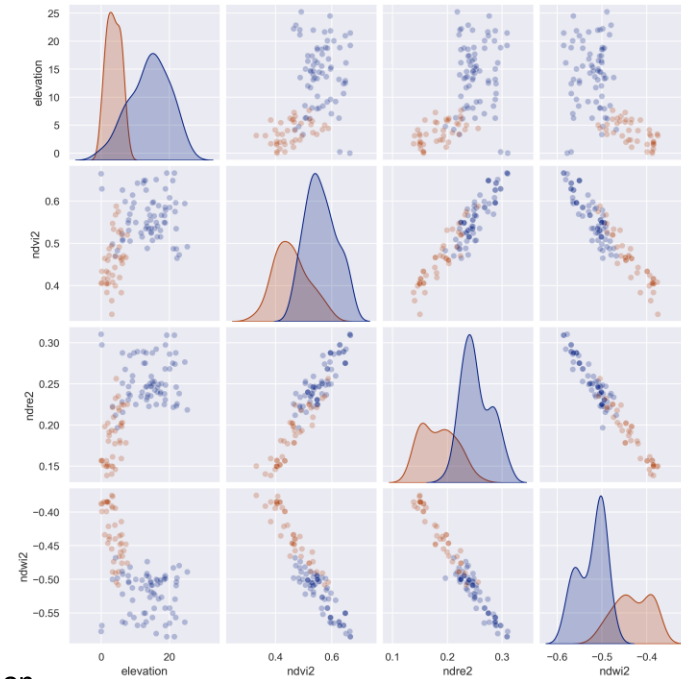


```
for plot in survDF.index:
    print("processing plot: {}".format(survDF[survDF.index==plot]['plot_id'].values))
    ptx=survDF[survDF.index==plot]['x_coord'].values[0]
    pty=survDF[survDF.index==plot]['y_coord'].values[0]
    #get plot extents
    yma=pty+ps
    ymi=pty-ps
    xma=ptx+ps
    xmi=ptx-ps
    #list of all elevations
    plotPts=pcDF[(pcDF['x']<xma)&(pcDF['x']>xmi)&(pcDF['y']<yma)&(pcDF['y']>ymi)][['x',
    #normalize against plot minimum and convert to cm
    plotPts['elevation']=(plotPts['z']-min(list(plotPts['z'].values)))*30.48
    #get corresponding raster values
    #iterate point elev and pull NDVI values
    for i2 in range(0, len(idx)):
        with rasterio.open(os.path.join(inP, idx[i2][i])) as src:
            plotPts[cols[i2]]=plotPts.apply(lambda x: cellVal(x.x, x.y, src), axis=1)
            plotPts[cols[i2]]=plotPts[cols[i2]]+cols[i2]
```

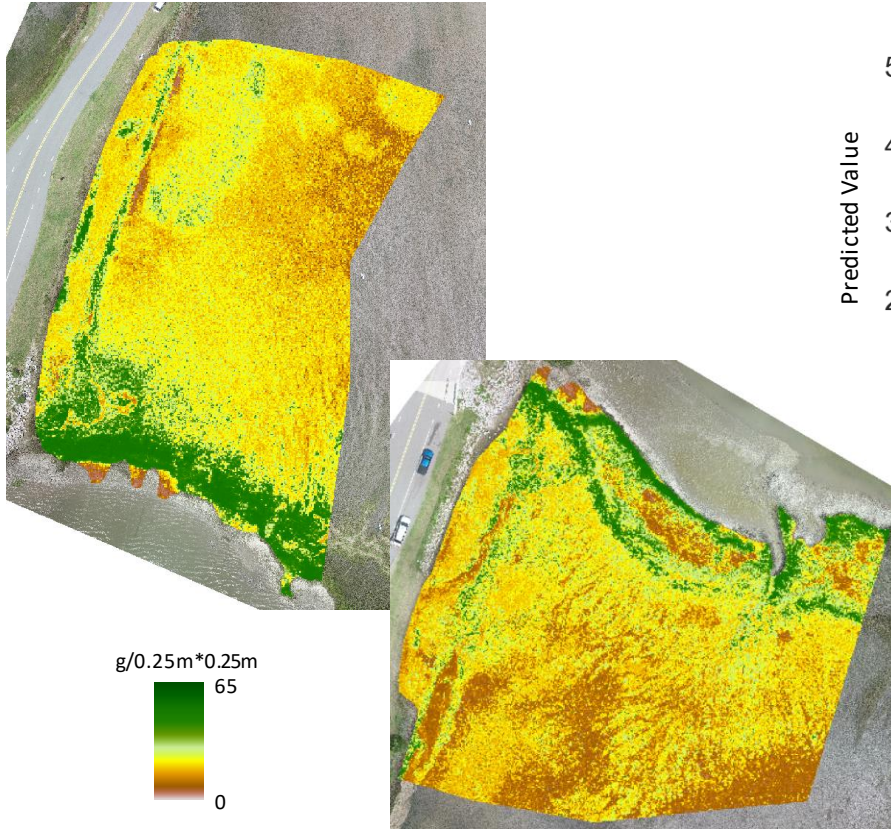
Biomass Model Parameters



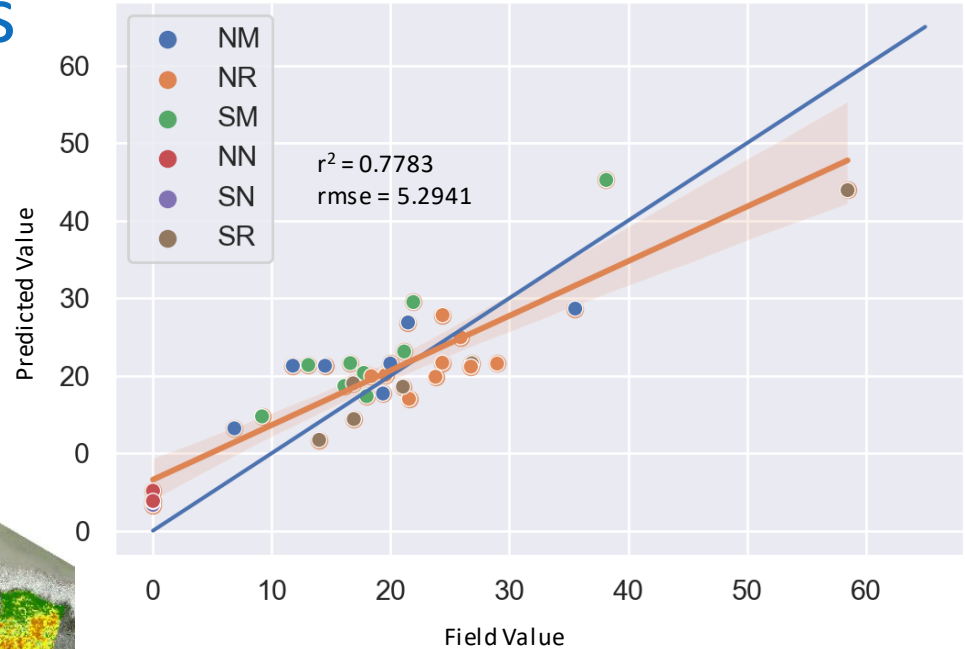
- Minimum
- 5th percentile
- 25th percentile
- 50th percentile
- 75th percentile
- 90th percentile
- Maximum
- Mean
- Standard Deviation
- Interquartile Range
- Kurtosis
- Skewness
- Variance
- Coefficient of Variation
- Mean Absolute Deviation (AAD)
- Median Absolute Deviation (MAD)
- Standard Deviation/Mean



Biomass Model Results



Random Forest AGB (g/0.25m*0.25m)



- scikit-learn
 - random forest
- Training/Testing
 - 60%/40%