Machine Learning Classification of Flood Waters from Hurricanes Harvey and Florence as Captured by Synthetic Aperture Radar and Optical Remote Sensing

A. Melancon<sup>1</sup>, A. Molthan<sup>2</sup>, J. Bell<sup>2</sup>, L. Schultz<sup>1</sup>

1 - University of Alabama in Huntsville, Huntsville, AL 2 - NASA Marshall Space Flight Center, Huntsville, AL

# What is Synthetic Aperture Radar?

- Synthetic Aperture Radar (SAR) is an active remote sensing instrument that relies on reflected energy from Earth's surface (backscatter)
  - Microwave radiation emitted by the radar can penetrate clouds and vegetation, available day or night
- SAR data for this project was obtained by NASA JPL's UAVSAR instrument
  - Side-looking L-band synthetic aperture radar mounted on a Gulfstream-III jet
  - Quad-pol radar outputs data according to its transmitted and received orientation
    - ► HH Horizontal transmit, Horizontal receive
    - ▶ HV Horizontal transmit, Vertical receive
    - > VV Vertical transmit, Vertical receive
    - > VH Vertical transmit, Horizontal receive



Images: NASA Jet Propulsion Laboratory



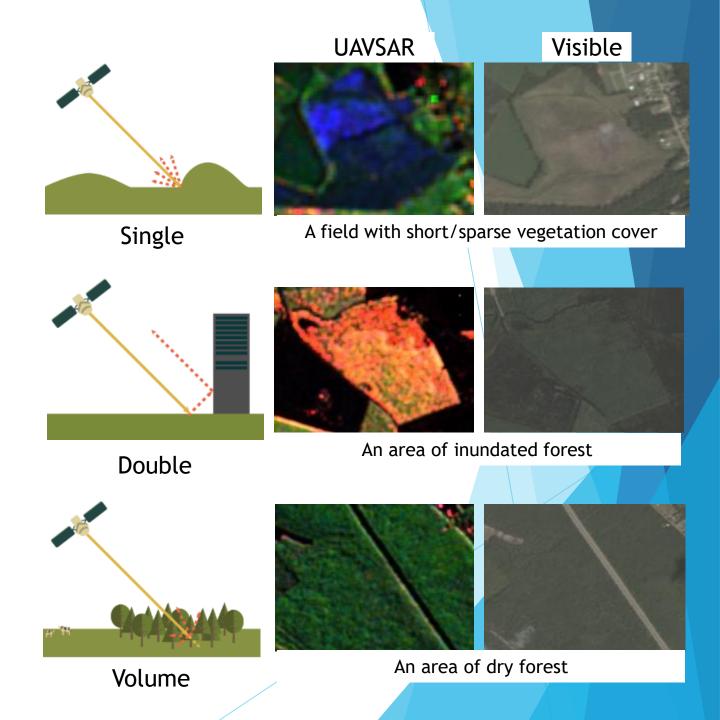
# Visualizing UAVSAR Data

- UAVSAR and other quad-pol data can be visualized through many different decomposition methods. The Freeman-Durden decomposition method is used here.
  - Produces layers that show the fraction of returned radar power resulting from single, double, and volume scattering mechanisms

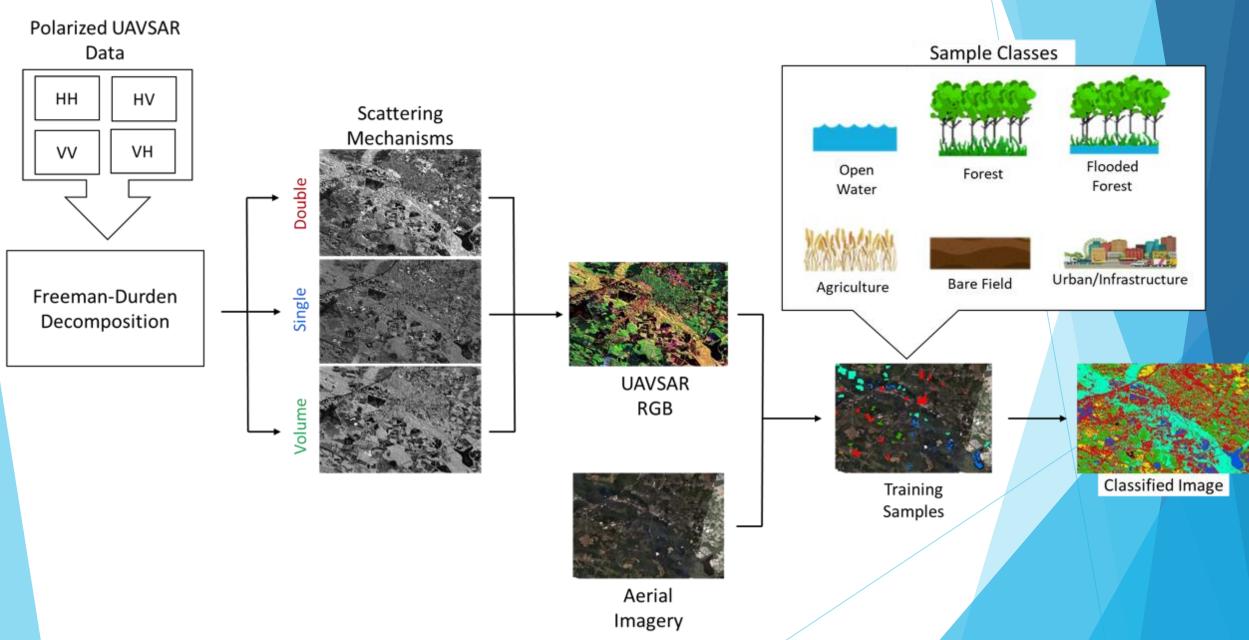
**Single** scattering - smooth, relatively flat surfaces (e.g. open water, paved areas, bare/sparsely-vegetated ground)

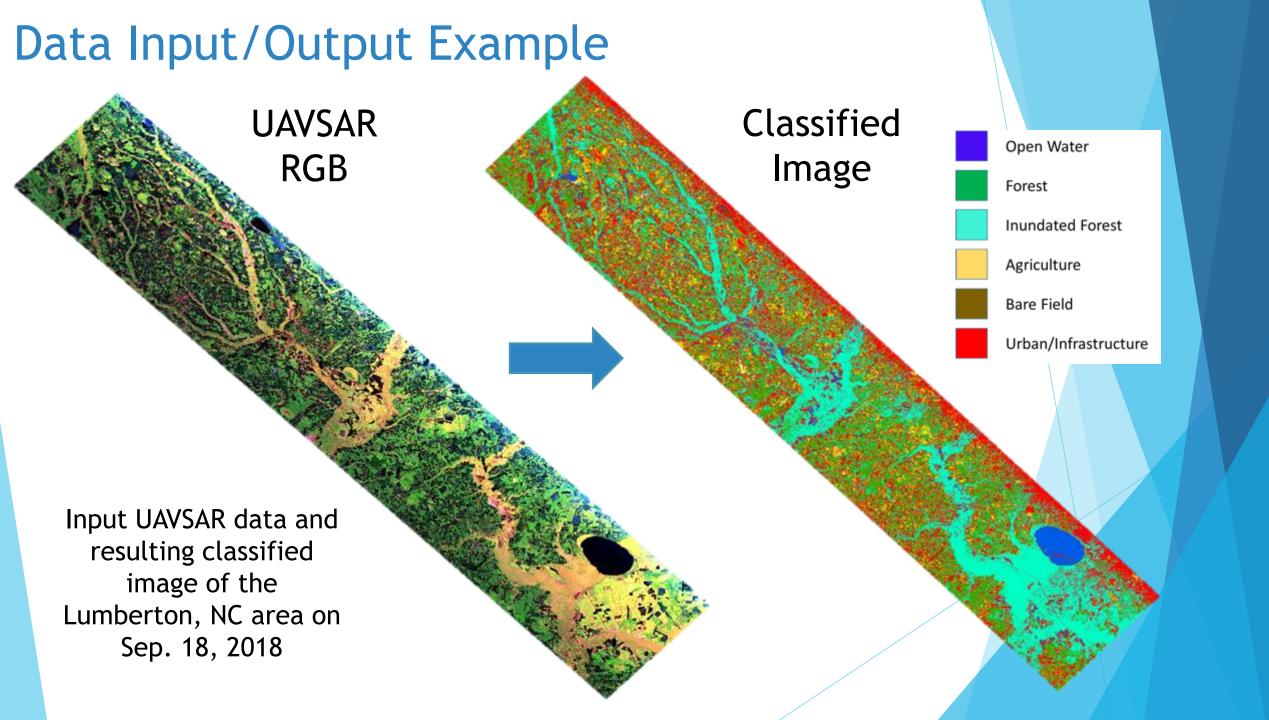
**Double** scattering - perpendicular vertical structures (tree trunks in standing water, buildings, power poles)

**Volume** scattering - vegetative canopies, sometimes urban areas



## Methodology





### **Results - Florence**

#### **Classified Points**

<b>Reference Points</b>		Open Water	Dry Forest	Inundated Forest	Agriculture	Bare Field	Infrastructure	Total
	Open Water	37	6	25	1	6	28	103
	Dry Forest	1	81	9	8	2	7	108
	Inundated Forest	0	13	90	1	0	6	110
	Agriculture	1	3	6	64	19	14	107
	Bare Field	0	1	8	19	54	27	109
	Infrastructure	1	7	13	10	6	68	105
	Total	40	111	151	103	87	150	642

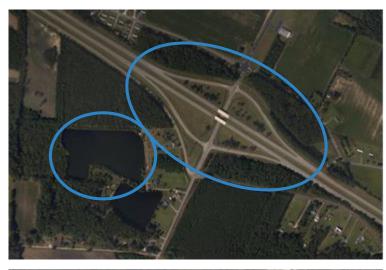
Overall Accuracy: 61.37%

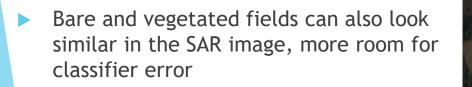
Kappa Coefficient: .536

### Limitations

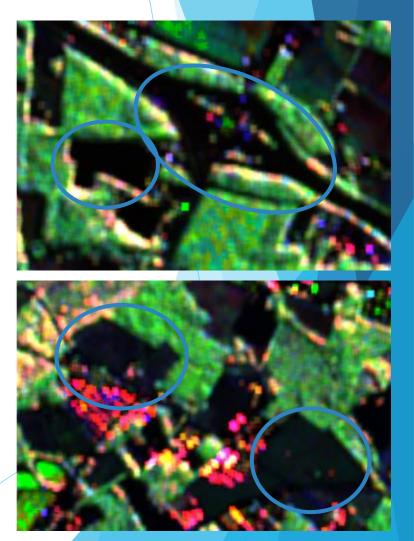
- Since UAVSAR provides a measure of surface roughness, different land cover types with similar textures tend to look similar in the imagery.
  - Creates many opportunities for the classifier to be confused

An example of the similarity of water and road networks in SAR imagery, both relatively smooth/flat surfaces









#### **Conclusions/Future Work**

- This method allows for the rapid classification of UAVSAR swaths with reasonable accuracy, making it potentially useful for near-time applications
- Moving forward, the main goal is to increase the accuracy of the classifier through a combination of methods:
  - The adjustment of classes and their respective training samples by focusing on classes with a distinct backscatter signature in the SAR image
  - Differing spatial resolutions between UAVSAR, NOAA aerial imagery, and Planet visible imagery cause a misalignment of accuracy assessment points, likely influencing the results of the assessment
  - Addition of a building mask may eliminate the need for an Urban/Infrastructure class, also eliminating the confusion associated with the class