# Flood Mapping of Recent Major Hurricane Events with Synthetic Aperture Radar, Commercial Imaging, and Aerial Observations

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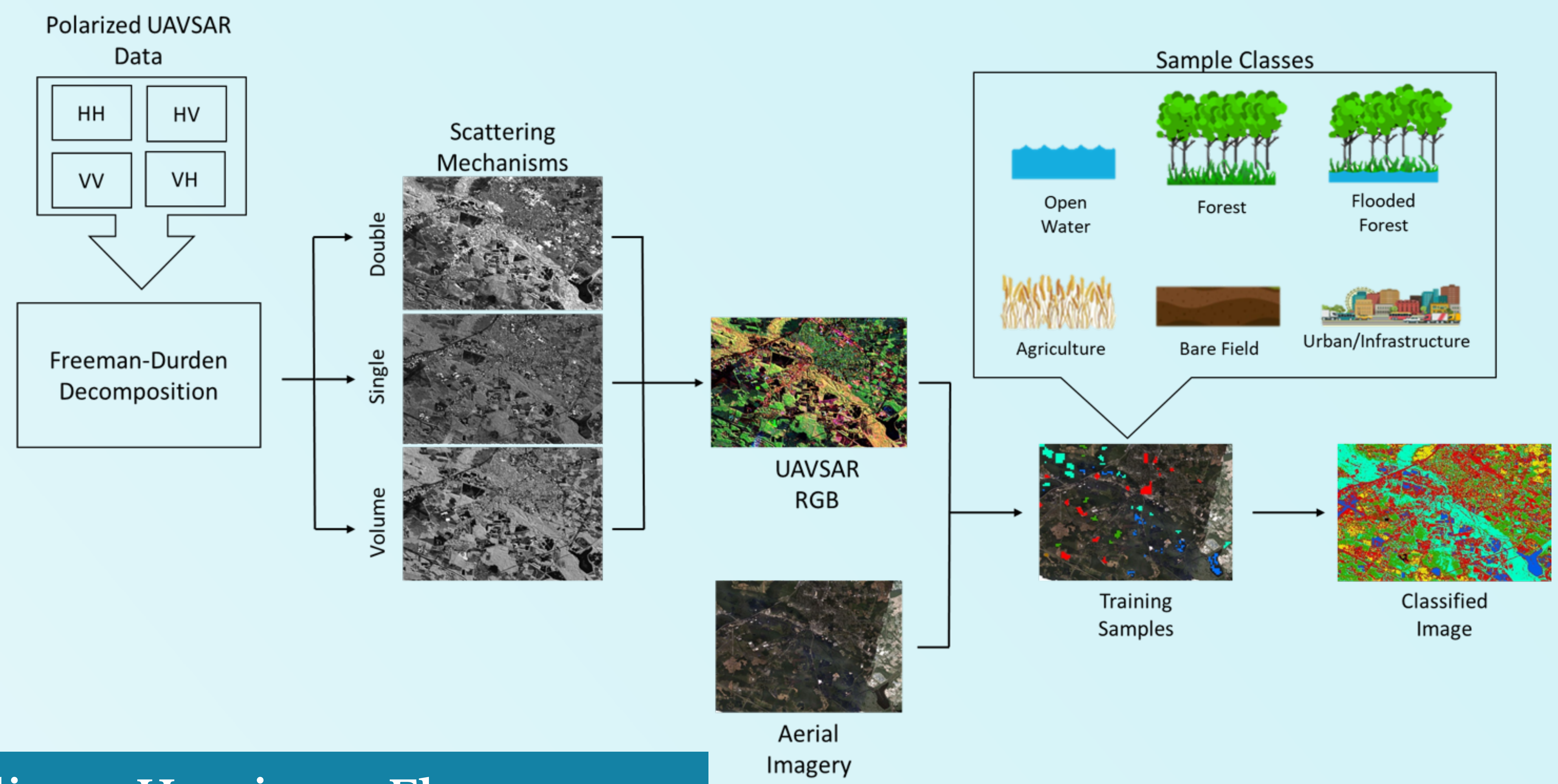
## Introduction

Floodwater mapping is an important remote sensing process that is used for disaster response, recovery, and damage assessment practices. Developing a system to read in Synthetic Aperture Radar (SAR) data and perform land cover classification will allow for the production of near real-time inundation mapping, enabling government and emergency response entities to get a preliminary idea of the situation.

SAR is a unique remote sensing tool. Data in this project was obtained by NASA Jet Propulsion Laboratory's Uninhabited Aerial Vehicle SAR (UAVSAR), an L-band radar mounted to a Gulfstream III jet. Data collected by UAVSAR is similar to what will be available from the NASA-Indian Space Research Organization (NISAR) mission starting in early 2022.

Using Python and ArcGIS applications, a model was developed using training samples taken from NOAA post-event aerial photography and UAVSAR data gathered in the aftermath of Hurricane Florence in September 2018.

## Methodology




## Legend

Scattering Mechanisms

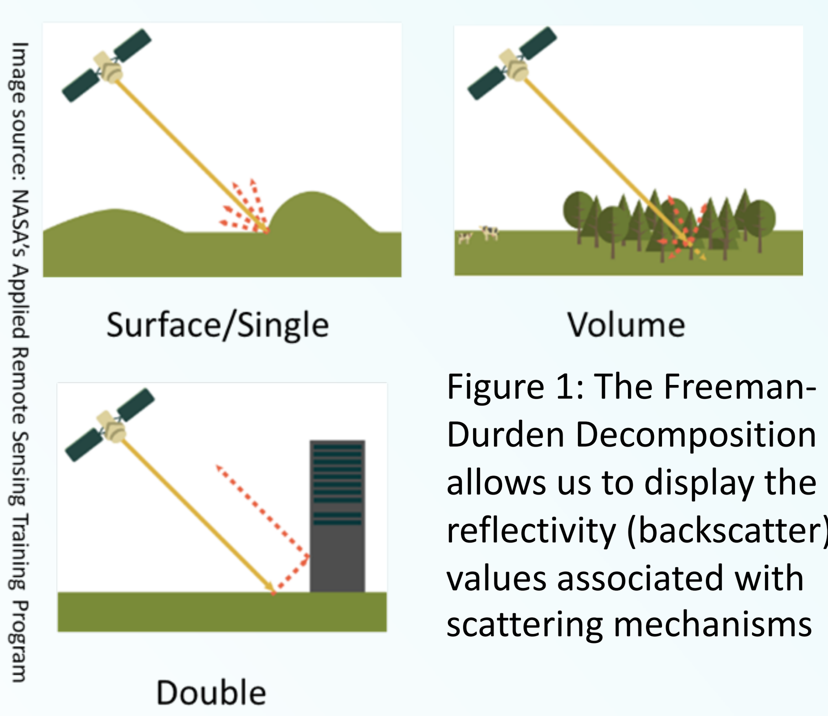



Figure 1: The Freeman-Durden Decomposition allows us to display the reflectivity (backscatter) values associated with scattering mechanisms

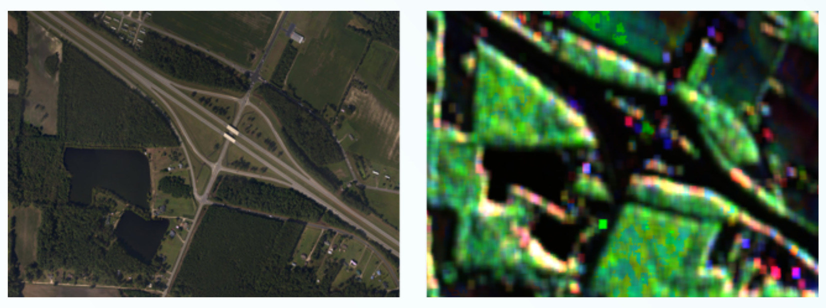


Figure 2: A comparison between NOAA aerial imagery and UAVSAR RGB showing the similar backscatter signatures of open water and roads.

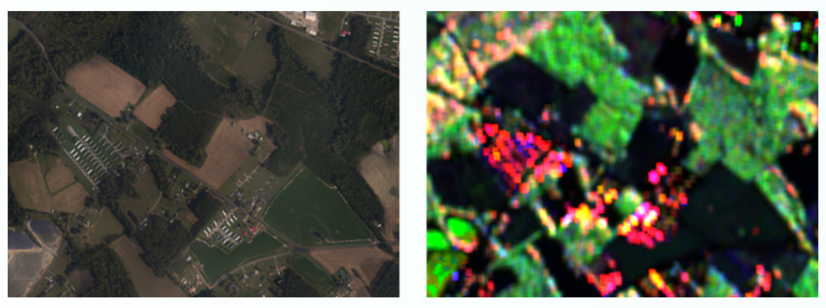


Figure 3: A comparison between NOAA aerial imagery and UAVSAR RGB showing similar backscatter signatures of bare and agricultural fields.

## Lumberton, North Carolina – Hurricane Florence

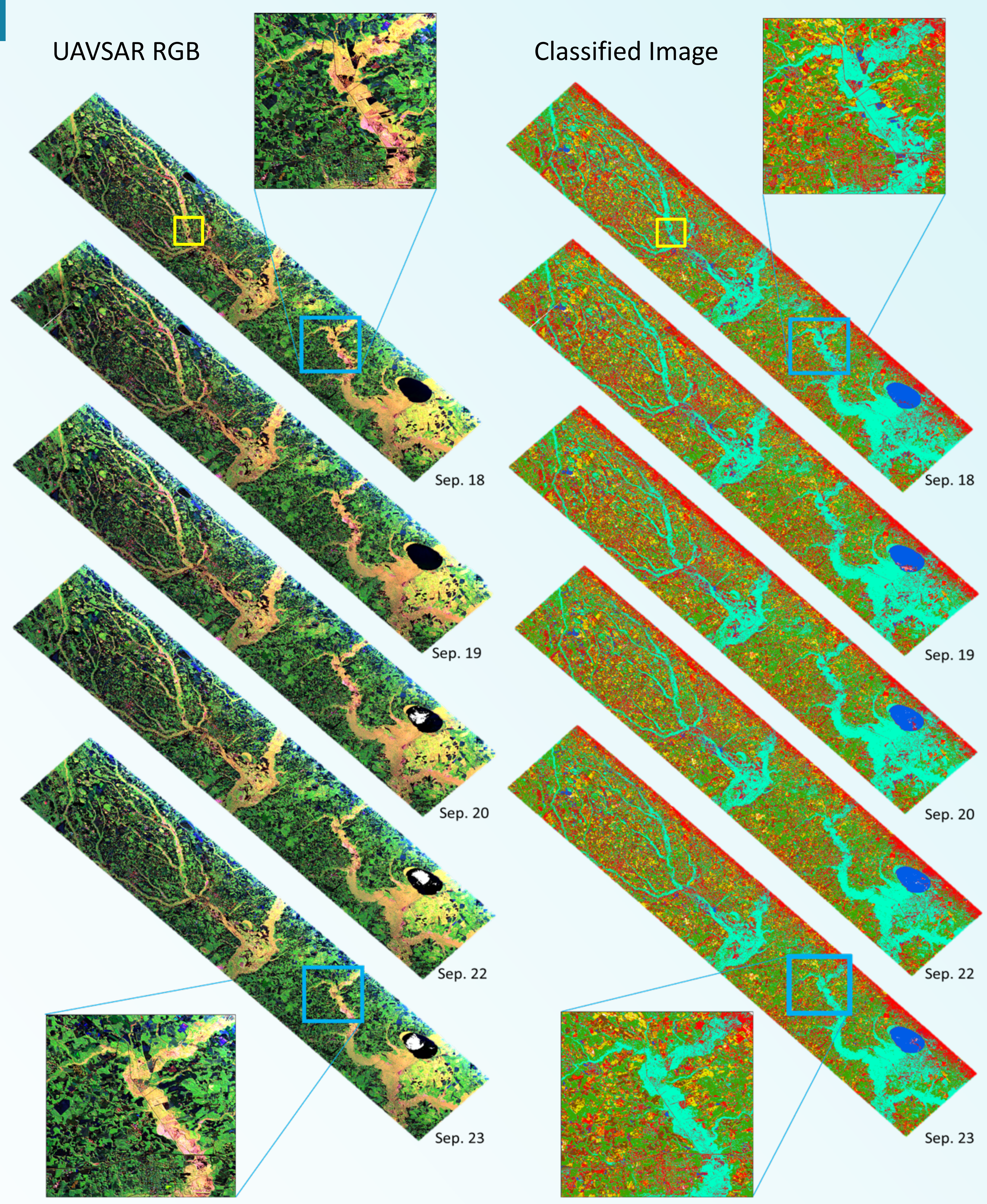



Figure 4: A time series comparison between UAVSAR RGB imagery (left) and the classified image created by the model (right). Blue boxed portions show a noticeable decrease in inundation extent from the beginning to end of the time period.

## Results

| Overall Accuracy | 61.4% |
|---|---|
| Open Water | 35.9% |
| Forest | 75.0% |
| Inundated Forest | 81.8% |
| Agriculture | 59.8% |
| Bare Field | 49.5% |
| Urban/Infrastructure | 64.8% |

*Individual class values are calculated user accuracy values

## Class Probabilities

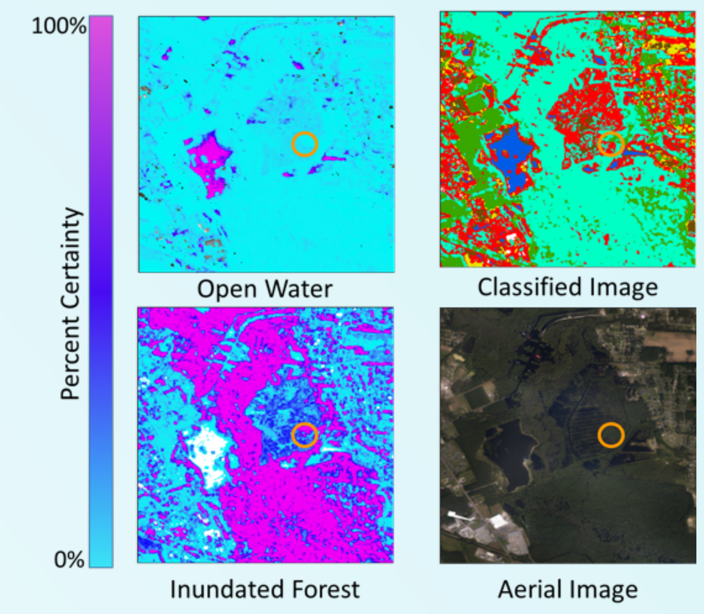



Figure 5: A closer look at the yellow boxed portions from Fig. 4 showing a comparison between class probabilities and the aerial image from Sep. 18.

Users may find that probabilities are more useful than deterministic classes for decision making.

For example, the orange circled area in Fig. 5 is classified as "urban" despite being surrounded by water and inundated forest. A user may be able to infer that this area is likely inundated as well. Fig. 6 shows that there was only a 0.5% difference between the pixel being classified as "urban" or "inundated forest" – something that can be discounted for the sake of continuity.

| Class | Percent Certainty |
|---|---|
| Open Water | 3.8% |
| Forest | 0.3% |
| Inundated Forest | 42.6% |
| Agriculture | 5.2% |
| Bare Field | 5.0% |
| Urban/Infrastructure | 43.1% |
| Classification | Urban/Infrastructure |

Figure 6: Table of class certainties for area within orange circle of Fig. 5.

## Conclusions

- Floodwaters extending below the forest canopy have high backscatter values in the double component, making them easily discernable in L-band imagery
- The model's rapid classification of the swath offers a reasonable depiction of floodwater extent in a short runtime, making it useful for near-real time applications
- Similarities in backscatter signatures for different classes leads to model "confusion," as in the misclassification of roads as open water and similar issues
    - Swath data should be constrained by radar incidence angle in order to reduce noise and retain highest quality data

## Future Work

- Overall and individual class accuracy may be improved with a merging of the Agriculture and Bare Fields classes, which often get confused for each other
    - Must look closely at what classes are best characterized by radar (i.e. a unique backscatter signature) as well as complementary information to improve final mapping
- Assessment and improvement of reference data points, which are misaligned due to a discontinuity in spatial resolution between NOAA aerial imagery and UAVSAR data
- Collaboration with partners to estimate depth of floodwaters

## Acknowledgements

The author gives special thanks to Andrew Molthan, Lori Schultz, Jordan Bell, Suravi Shrestha, and Esayas Gebremichael, along with Yunling Lou, Naiara Pinto, and the UAVSAR Flight Team.

Methods were adopted from Bruce Chapman at the NASA Jet Propulsion Laboratory.

This project was made possible by the support of the MSFC Earth Science Branch (ST11).

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