

Use of CIR aerial photography at Huntsville Botanical Garden

TRACY COOK, MASTERS CANDIDATE

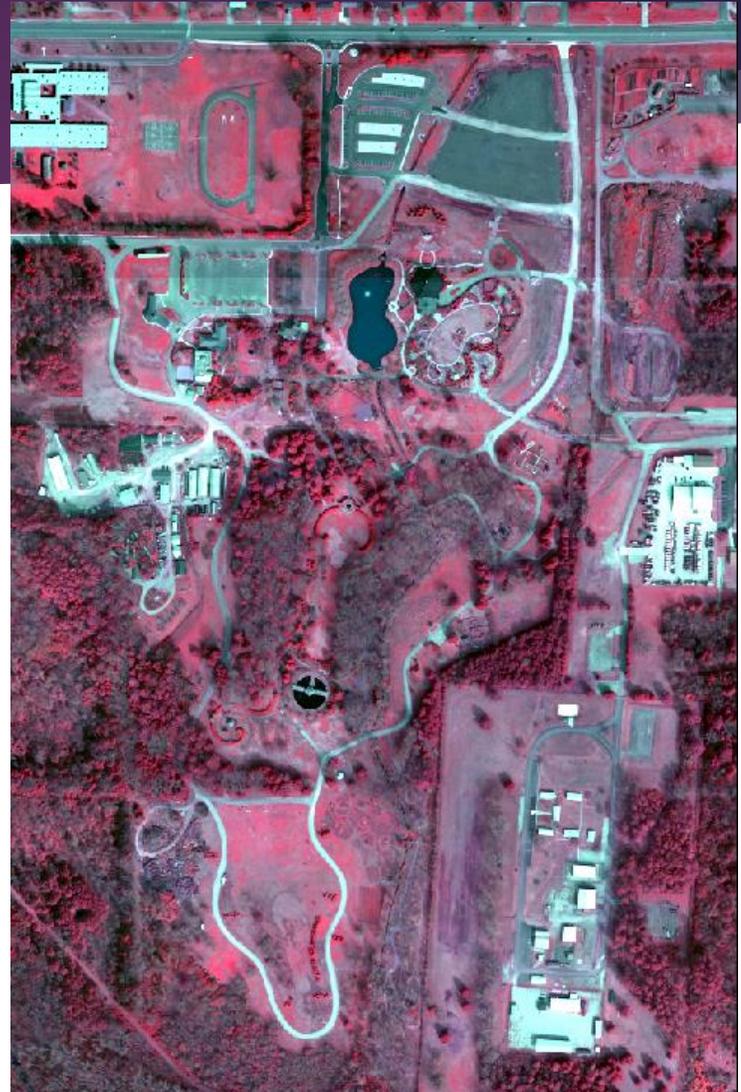
DEPARTMENT OF BIOLOGICAL AND ENVIRONMENTAL SCIENCES

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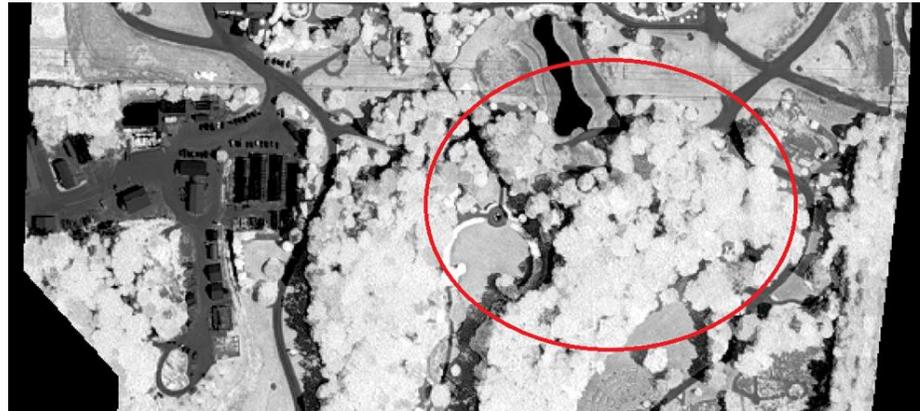
Introduction:

- Mapping and plant collections curation in public gardens are often critically underfunded.
- Efficient methods of mapping changes that take place in vegetation performance is necessary to support the mission.
- Remotely sensed data analysis in a GIS platform is a cost-effective alternative for delineating features, detecting multi-year changes, and broadly assessing the vigor or decline of vegetation over time.



Objectives:

- 1) Delineate features and vegetation types into broad categories via image classification tools in the Spatial Analysis toolbox in ArcMap 10.3.1.
- 2) Apply NDVI and Change Detection to CIR aerial photos with an elapsed time to see if vegetation decline is detectable in areas that were damaged by a storm that felled 17 mature trees in June 2016.



Data:

- 1) A 4 band RGB-CIR NAIP Digital Ortho Photo Image depicting the Huntsville Botanical Garden. (USDA-FSA-APFO Aerial Photography Field Office. Publication Date: 20151125).
- The images were taken in February 2013 and have been orthorectified using 9 control points. This image was used for the supervised classification.
 - Spectral resolution: .48 -.85 μm
 - Radiometric resolution: unsigned 8 bit
 - Spatial resolution: 0.3 meter pixels.
- 2) CIR NAIP Digital Ortho Photo Images depicting the Huntsville Botanical Garden for 2011, 2013, & 2015. (USDA-FSA-APFO Aerial Photography Field Office. Publication Date: 20151125).
- These images were used to produce NDVI surfaces for comparison.
 - Temporal resolution: approximately 2 years (August)
 - Spectral resolution: .48 -.85 μm
 - Radiometric resolution: unsigned 8 bit
 - Spatial resolution: 1 meter pixels.

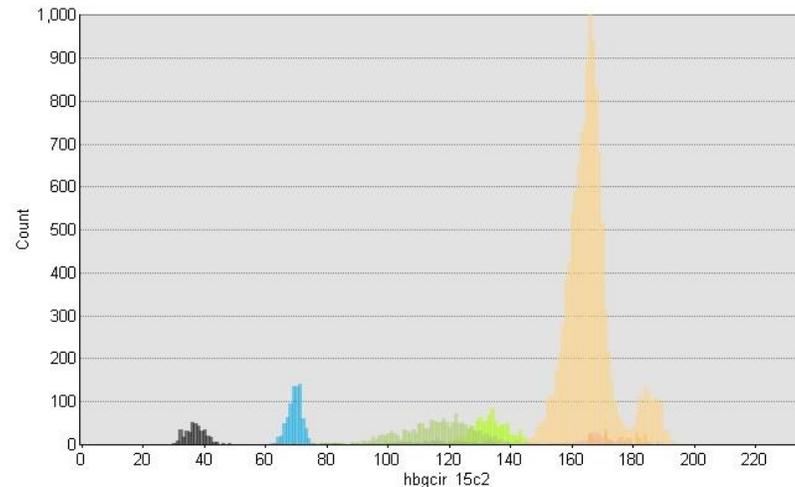
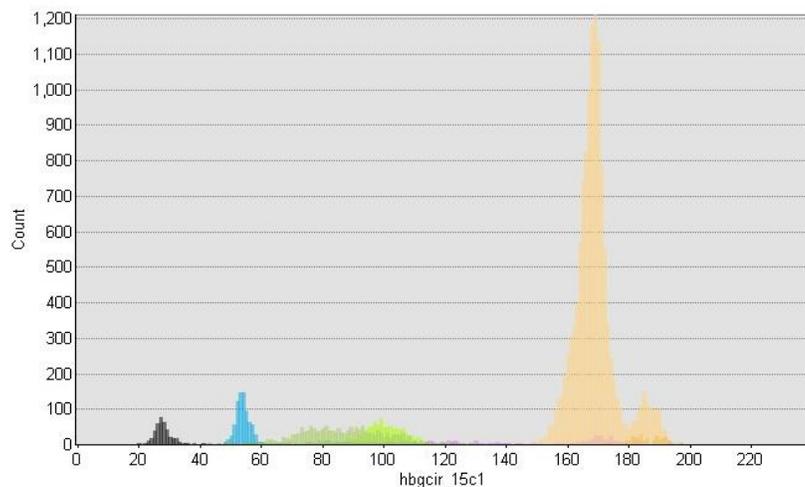


Methods: Supervised Classification

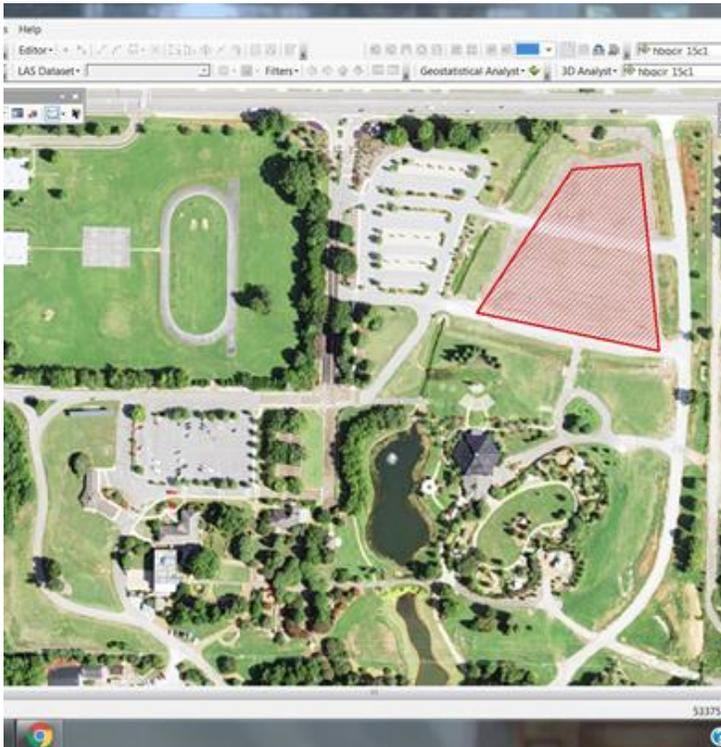
- The *Maximum Likelihood Classification Tool* in the *Multivariate Toolset* in ArcMap 10.3.1. was used to classify the 1 meter resolution image for 2015 using the *Training Sample Manager* to create the signature files.
- The predefined information classes are shown in the classification scheme on the right.

Classification Scheme

1. Paving
2. Buildings
3. Shadows
4. Water
5. Bare Earth
6. Vegetation
 - a. Evergreen Trees
 - b. Deciduous Trees
 - c. Low Vegetation
 - d. Lawns



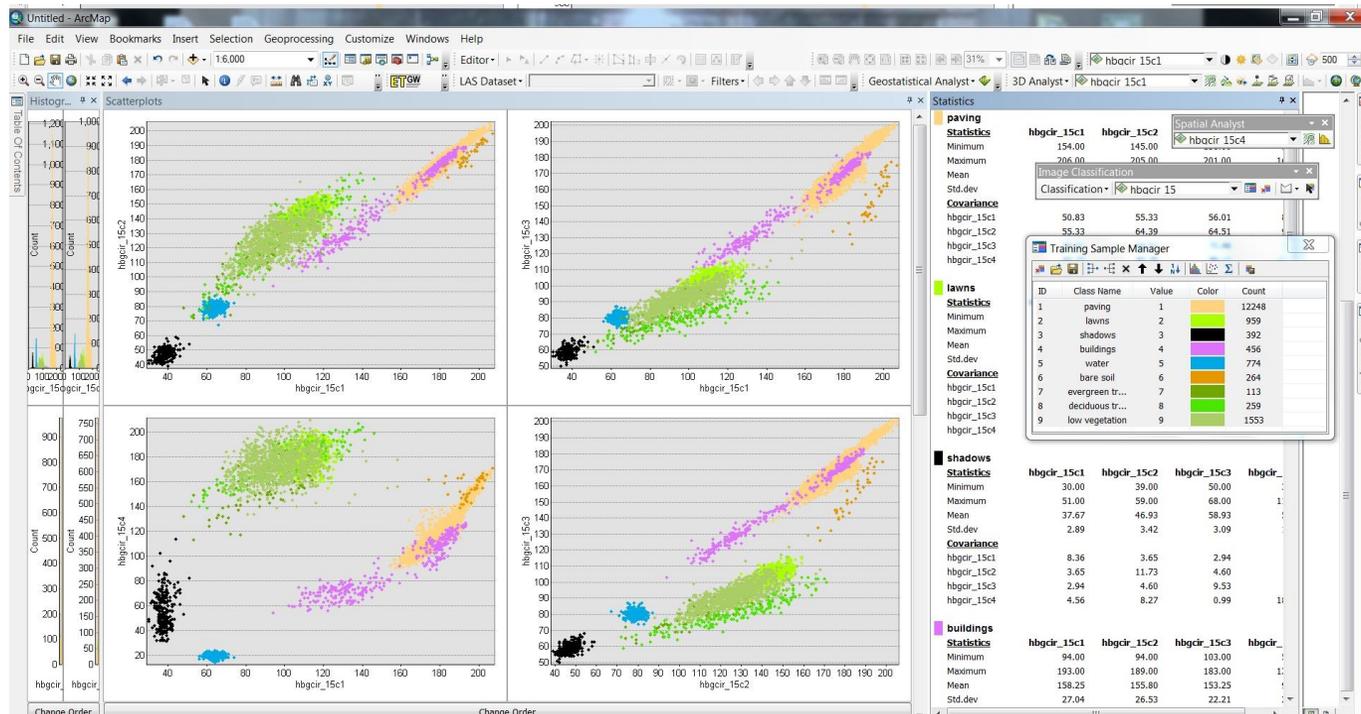
Methods: Supervised Classification



- The polygon drawing tool was used to create a multiband signature file to assign class names to polygon areas.
- 23 signatures were created throughout the image to capture both distinct areas and the variability in those areas across the image.
- The signatures were chosen based on both on personal knowledge of the area and the way pixels appeared in the multiband image.
- The signatures were then assessed to be sure the classes were explicit.

WAIT! It didn't work....

- Several information classes were “spectrally confused”!
- Many attempts were made to separate them using different tools available in the *Training Sample Manager*.



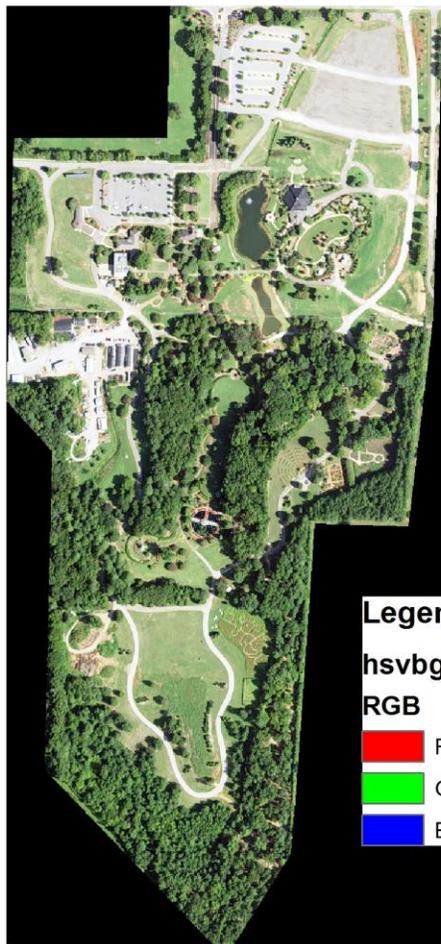
- Histogram comparisons of multiple signatures
- Scatterplots for multiple signatures
- Statistical filtering
- Stretch tool for histogram equalization in bands
- Re-drawing and deletion of signatures in different band combinations

Let's try this again...

- The image used could not be separated into the classes desired based on their spectral signatures in part because the image was taken during the dormant season.
- Data: 2015 1m resolution image from August 2015.
- 36 new signatures were created for this image.
- The evaluation process resulted in 7 signatures deleted and the rest merged.
- Final result: 10 classes; only 2 classes noticeably overlapping.

Comparison of CIR Aerial Photo to Classified Image

Results:
Eureka!

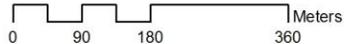


Legend

hsvbg_2015_cir.img

RGB

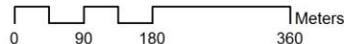
- Red: Layer_1
- Green: Layer_2
- Blue: Layer_3



Legend

HSVBG Classified
Class_Name

- Bare Earth
- Deciduous Trees
- Evergreen Trees
- Lawn
- Low Vegetation
- Paving
- Shadow
- Shadow Buildings
- Tree Shadow
- Water



Methods: NDVI (Normalized Difference Vegetation Index)

- NDVI is a measure of vegetation productivity – healthy vegetation yields a higher pixel value than unhealthy vegetation.

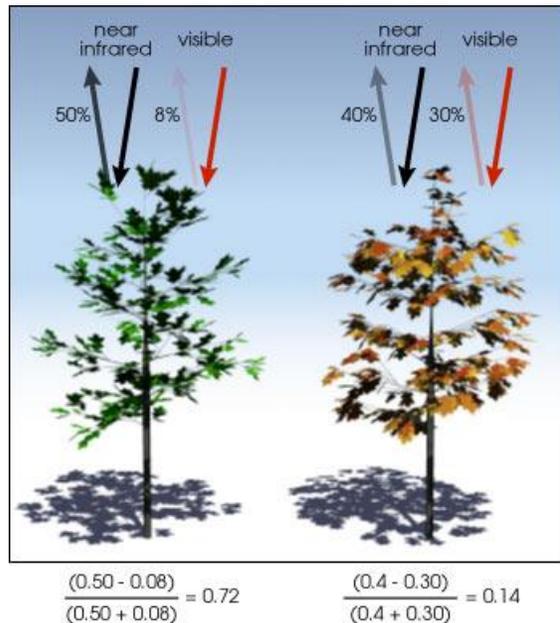
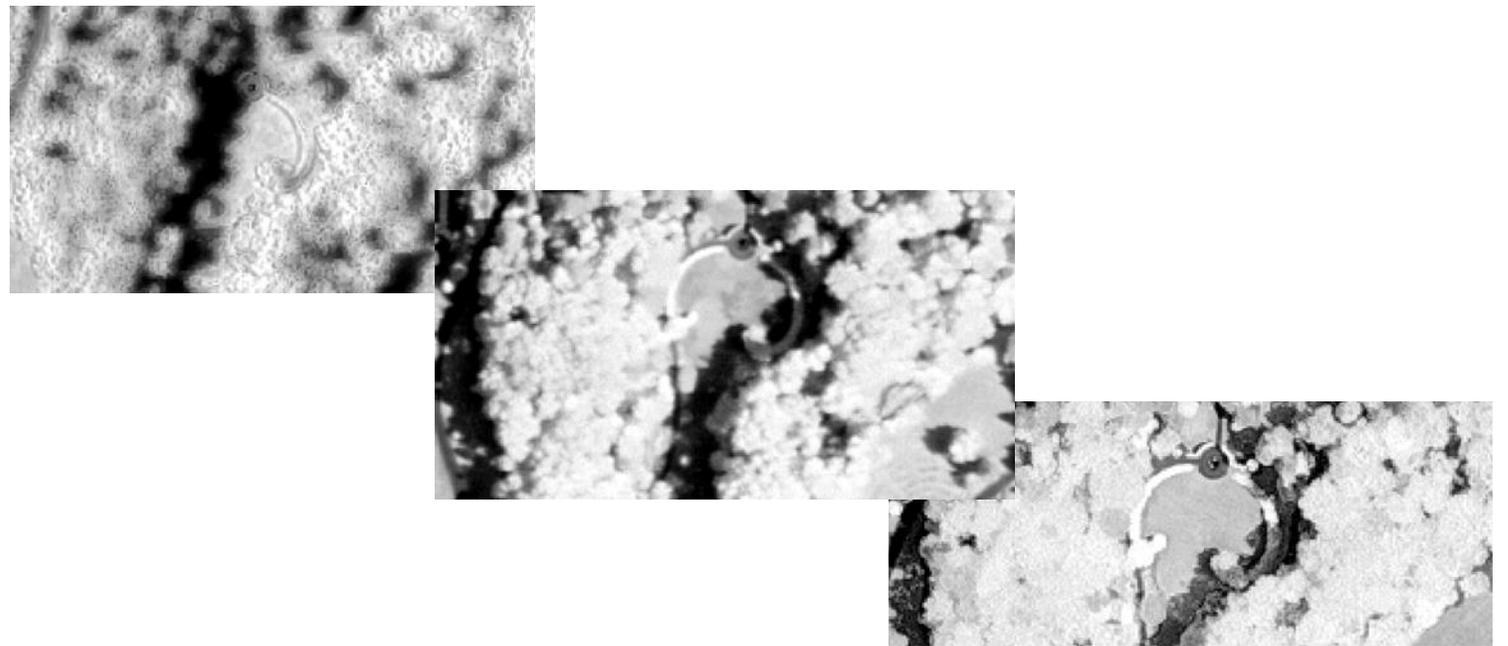


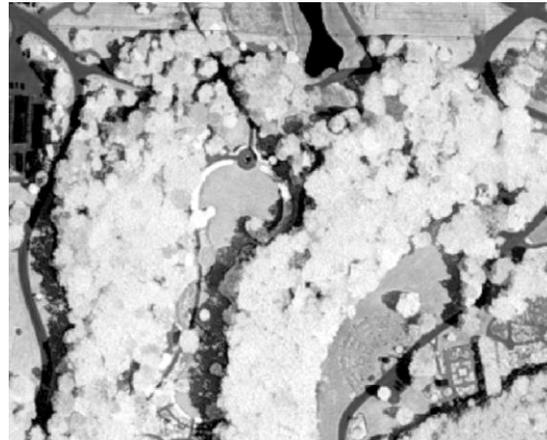
Image Credit:
<http://earthobservatory.nasa.gov/Features/MeasuringVegetation>



Comparison of 2011, 2013, & 2015 NDVI images

Methods: NDVI (Normalized Difference Vegetation Index)

- NDVI was generated for the area of interest from the Aug 2013 and Aug 2015 1 m images using the *Image Analysis* window and produces a panchromatic image with the pixel values ranging from 1.0 to -1.0.



$$\text{NDVI} = \frac{(\text{NIR} - \text{Red})}{(\text{NIR} + \text{Red})}$$

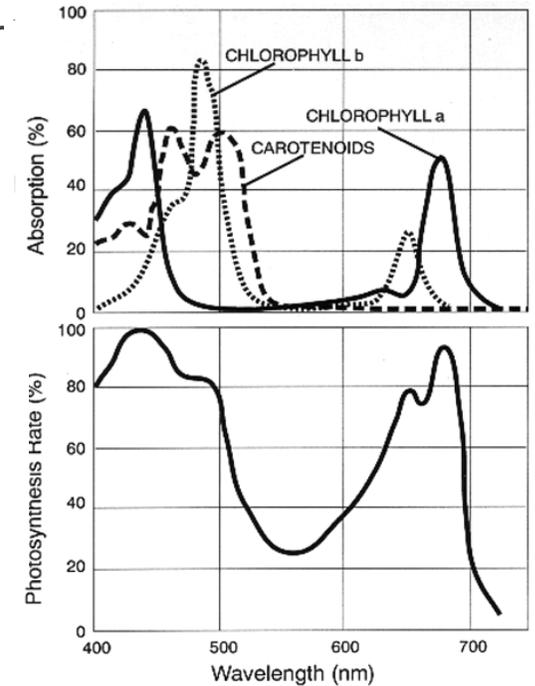


Image credit: By John Whitmarsh and Govindjee.
<http://www.life.uiuc.edu/govindjee/paper/gov.html>
from "Concepts in Photobiology: Photosynthesis and Photomorphogenesis"

Methods: Change Detection

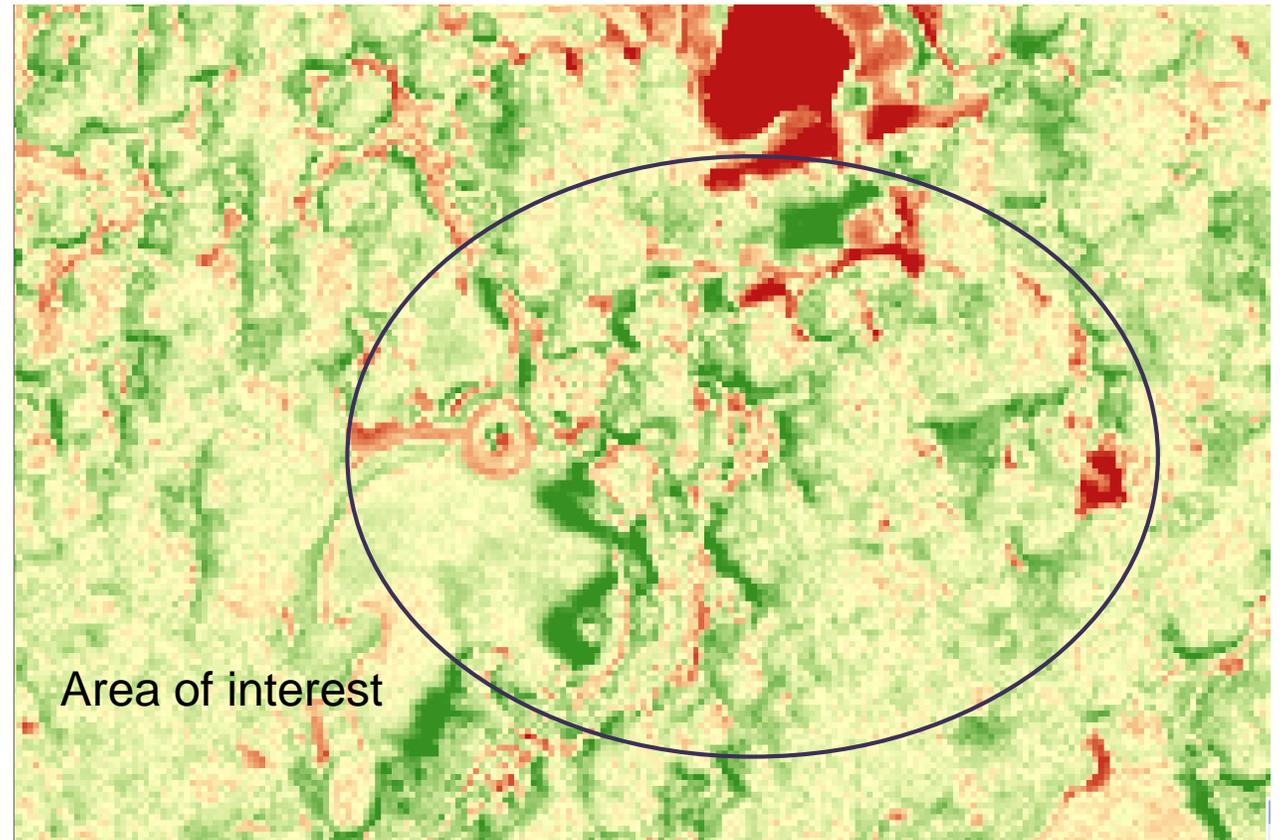
- Change detection of the 2013 & 2015 NDVI images was taken using the *difference* tool in the *Image Analysis* window to determine if declines in vegetation productivity / reflectance were detectable over that period.



Difference 2013 & 2015 NDVI images

Results: NDVI & Change Detection

- This produces a per-pixel change raster with continuous values based off of the input rasters. Green = increased vegetation, red = decreased, tan = no change.
- The very dark red pixels indicate big man-made changes.
- The differences raster produced was visually inspected to see whether the areas identified are correlated with storm damage that occurred in mid-2016.



Discussion: Classification

- Timing and season of the aerial photo dramatically affects its usefulness in classification and NDVI.
- Depending on the spectral qualities present in the photo, a hybrid supervised/unsupervised classification would probably yield as good or better results with less effort.
- A classified image is useful in quickly delineating simple classes, but fails when attempting to extract features that are easily confused i.e. nursery weed fabric & shallow water with pond dye.
- Higher spatial resolution is not always better.



Discussion: NDVI / Change Detection

- The elapsed time between images used was very good at highlighting big man-made changes, but not enough to detect tree decline.
- It is possible that using the *slice* tool on the change detection image could make visible more subtle changes within the data.
- It is also possible that the trees felled in the storm were victims of strait-line winds, and no decline is present.
- It would be interesting to see if duplication of this analysis on images with greater elapsed time would yield more noticeable vegetation effects.



References

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Acknowledgements

- Shelby Aesthetic
- Dr. Wubishet Tadesse
- Dr. Dawn Lemke
- Students and staff of the GREC lab
- This work is funded through the Center for Forest Ecosystem Assessment, Center for Research Excellence in Science and Technology, administered through the National Science Foundation (grant # 1036600).

*Thank
You*



Questions?

