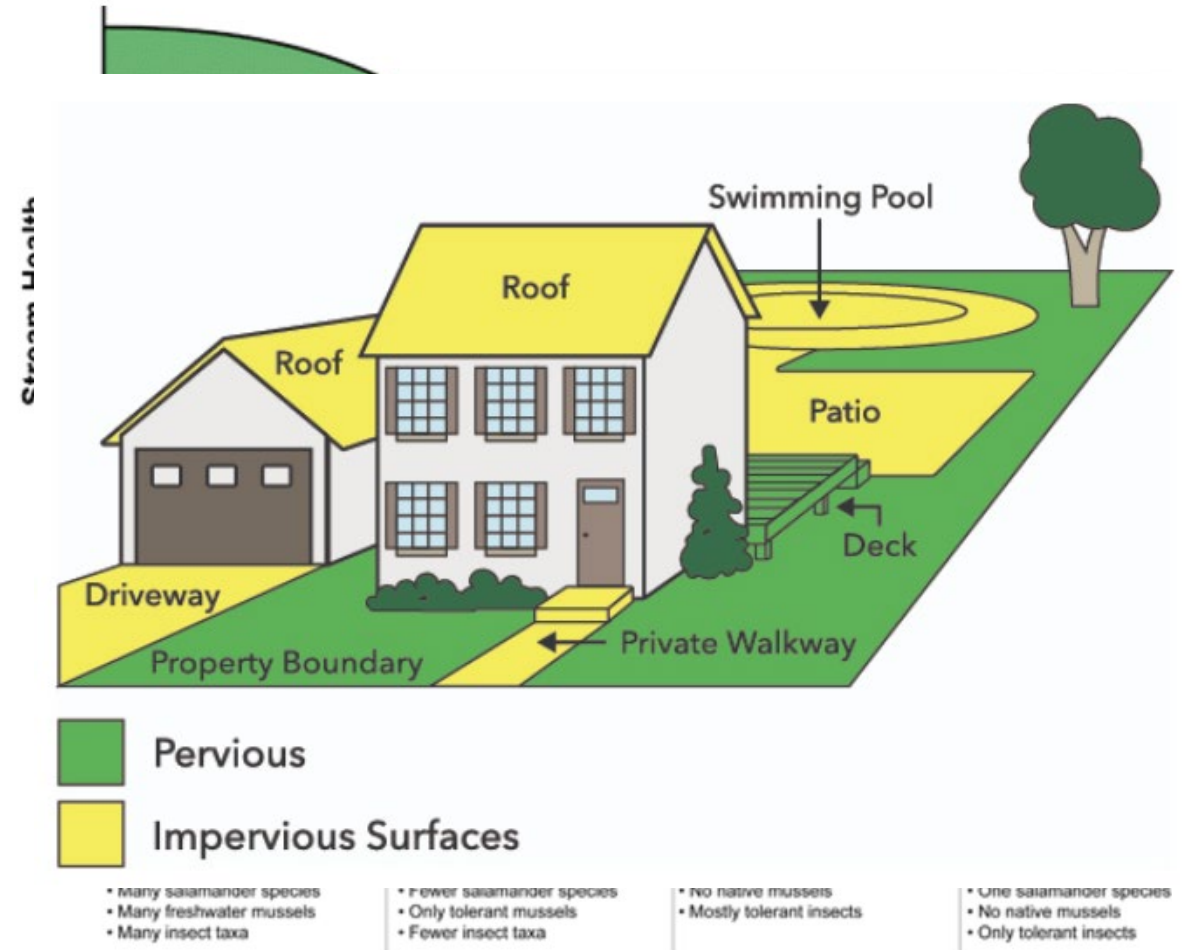




Towards a Fair and Effective Stormwater Management System in Athens, OH:
Experimenting with Lidar Intensity to Improve Impervious Surface Classification in areas of Dense Tree Canopy

Impervious Surfaces and Why They Matter

- ▶ Impervious surface refers to all hard surfaces such as paved roads, parking lots, roofs, and highly compacted soils, that prevent the natural soaking of rainwater into the ground.
- ▶ Impervious Surfaces effect several key environmental factors, including:
 - ▶ Flooding and Erosion: “For every percentage point increase in roads, parking lots and other impervious surfaces that prevent water from flowing into the ground, annual floods increase on average by 3.3%” (Blum et al. 2020)
 - ▶ Pollution: Heavy metals from roof shingles and motor vehicles, as well as pesticides, bacteria from sewage, and road treatment chemicals are just some of the pollutants that run untreated into the water table.
 - ▶ Heat Retention: Urban heat islands and killing of aquatic life



Images courtesy of the Maryland Department of Natural Resources: <https://dnr.maryland.gov/streams>

A Brief Overview of Stormwater Management Systems

Stormwater Utility Fees (SUF) are one method to generate revenue for stormwater management.

- 1972 – Creation of the Clean Water Act

- 1987 – Creation of the National Pollutant Discharge

The most common types of SUF—Equivalent Residential Units Elimination System (NPDES) stormwater program

(ERU), Equivalent Hydraulic Area (EHA), and Intensity of Development (ID) – use the amount of impervious surface per

The NPDES requires municipalities to develop a Stormwater

Management Plan to reduce pollutants in the stormwater

discharge to the “maximum extent practicable (USEPA, 1999).

In their 2017 case study *The Financial Impact of Different*

Stormwater Fee Types: A Case Study of Two Municipalities in

Virginia, NPDES places responsibility for funding of stormwater

management policies on local towns and municipalities. As a result, the based

development impervious surface of stormwater utilities are the

primary source of funding. In the case of the two municipalities, the

paid to offset the burden. This is not the case with existing industries

stormwater utilities. The funds often used to offset the effects of

funding to the surface. However, many locations chose to

implement a flat rate SUF structure, due to the financial cost

of providing accurate impervious surface calculations.

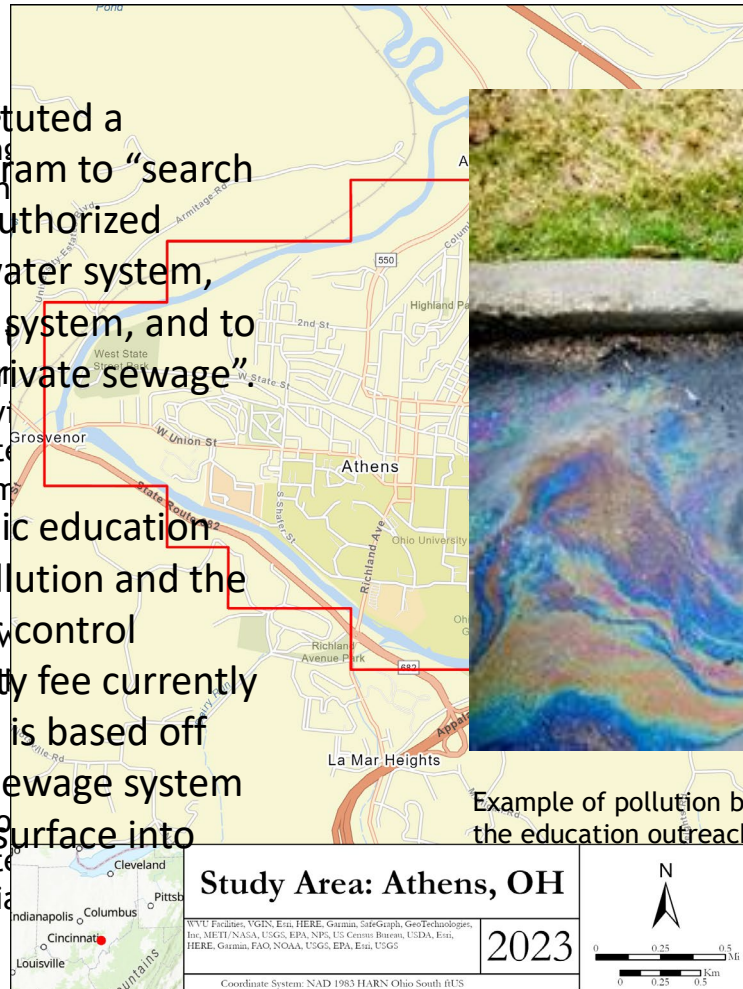
A Clear and Present Need: A look at Stormwater Management in Athens, OH

Following severe back-to-back floods in 2017, the City of Athens instituted a stormwater management program to “search for, identify and eliminate unauthorized discharges to the city’s stormwater system, create a storm sewer mapping system, and to map all residential and private sewage”.

These improvements consisted of a 400 ft of the channel to a bottom width of 10 feet...the improved channel shortened the length through Athens by approximately 1/2 mile.

The 400’ also encourages public education about non-point sources of pollution and the need for increased stormwater control projects. No stormwater utility fee currently exists as such, instead, funding is based off amount of water put into the sewage system and does not take impervious surface into account at all.

According to the Hawking River Conservancy website current hydrologic records indicate a year frequency is a more appropriate



Example of pollution build-up from impervious surface used as part of the education outreach by the City of Athens

Benefits of A GIS System for Impervious Surface Calculation

Low Cost

- Object Based Image Extraction (OBIA) reduces cost in time and labor, allowing for an increase of funds being spent on managing storm water
- Uses publicly available data generated from state or national funding

Potential Increased Revenue

- Lower cost of production could allow the SUF to be based on impervious surface per parcel, a method which has been shown to increase total revenue the municipality receives (Fedorchak et. Al 2020). Costs are distributed in correlation with burden provided.

Process is Largely Automated

- Cost of upkeep is lowered and a system for updating is built in

However...

These benefits can only be realized if the impervious surface can be reliably extracted throughout the entire study area, including areas occluded by the tree canopy.

June 2021, Leaf-on NAIP



November 2017 Leaf-off NAIP



lidar
arly
accuracy
action.
a to keep
future

Methodology


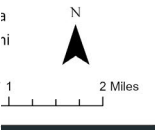
Create Classification Ruleset in eCognition

- Imagery:
 - Select Test Area to Develop eCognition Ruleset
 - In 2017 NAIP, 60cm 2021 NAIP
 - 6" 2014 CIR, 6" 2014 RGB
 - Perform Segmentation
 - Contrast Split (L1)
 - Multi-threshold (L2)
- Lidar: 2021 point cloud
- Vector: Road centerlines, parcels Separately
- Convert Mr. Sid to .img format
- Classify Data
- Develop Lidar Derived Raster Products Based on Ground and Final Returns:
 - Intensity Threshold ≤ 3450
 - $NDVI > 0$
 - Area and Region Grow
 - $nDSM, slope, nDSM, Intensity$
- Export Data
 - Export to ArcMap to calculate the percent impervious surface per parcel

Aerial Imagery						
	2017 NAIP	2021 NAIP	2014 CIR (OSIP* ^{1,2})	2014 RGB (OSIP)	2020 RGB (OSIP)	2007 RBG (OSIP)
Projection	NAD 1983 UTM Zone 17N	NAD 1983 UTM Zone 17N	NAD_1983_HARN_Ohio_South_ftUS	NAD_1983_HARN_Ohio_South_ftUS	NAD 1983 HARN StatePlane Ohio South FIPS 3402 (US Feet)	NAD 1983 HARN StatePlane Ohio South FIPS 3402 (US Feet)
Resolution	1m	.6cm	6inch	6inch	6inch	1ft
Bands/Classes	4 bands	4 bands	3 (NIR-G-R)	3	3	3
Bit Depth	8	8	8	8	8	8
Sensors	CNIR	CNIR	Leica ADS80-SH81/82 Airborne Digital Sensor	Leica ADS80-SH81/82 Airborne Digital Sensor	VisionMap A3 Edge digital camera systems	Leica ADS40/51/52 digital cameras
Collection Date	Jun-17	Nov-21	Spring 2014	Spring 2014	2020 - no month	March-April leaf off

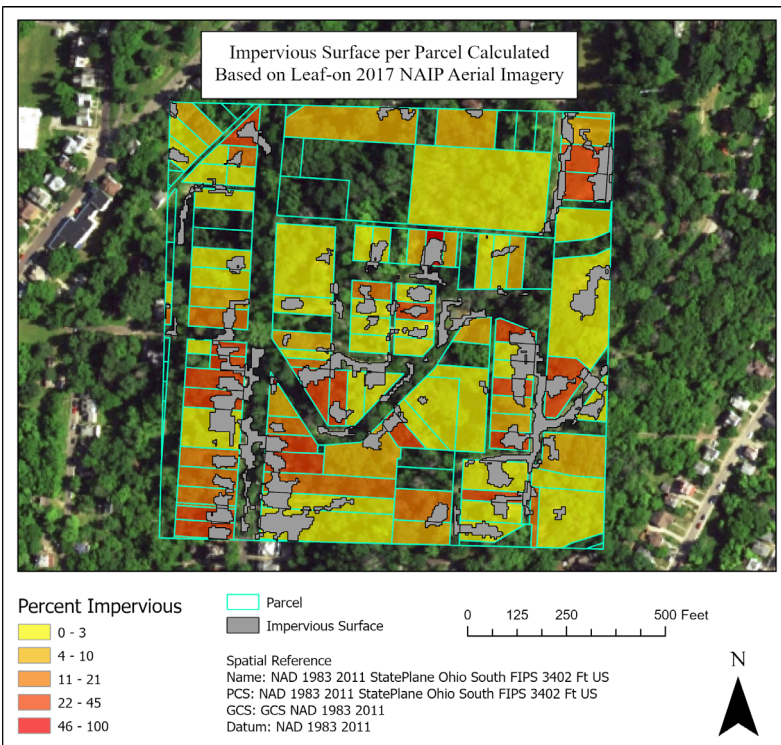
*OSIP = Ohio Statewide Imagery Program

Lidar		
	2007 lidar ⁴	2021 lidar ^{1,2}
Projection	NAD 1983 HARN StatePlane Ohio South FIPS 3402 (US Feet)	NAD 1983 (2011) StatePlane Ohio South FIPS 3402 (US Feet)
NPS	1,828 (6ft)	0.35
Quality	2	1

Version

- Find Building Candidate
- Extract Boundary
 - on main : chess board: 2500 creating 'New Level'
 - unclassified with Num. of overlap: Boundary > 0 at New Level: Working dataset
 - Working dataset at New Level: chess board: 2
- Segment Buildings
 - Working dataset with Mean nDSM_fil > 8 at New Level: Tall
 - edge diff. split nDSM_fil [0-45:+5] (0,0) (contrast>0.4) :2500-> [creating 'BO Level', Potental Building,Not Building, VAR_best_threshold, VAR_best_contrast]
 - Potental Building with Number of pixels < 500 at BO Level: remove objects into Not Building (merge by shape)
 - Potental Building at BO Level: multi-resolution: 12 [shape:0.2 compct.:0.9] creating 'TreeBuildingDifferential'
 - unclassified with NDVI >= 0.1 at TreeBuildingDifferential: Tree
 - unclassified at TreeBuildingDifferential: Building Outline
 - Building Outline at TreeBuildingDifferential: export object shapes to ObjectShapes
- Segment Roads
 - multi-resolution: 40 [shape:0.2 compct.:0.9] creating 'R2'
 - unclassified with Mean Intensity_ground <= 3450 or NDVI < 0 at R2: Roads
 - Roads with Mean nDSM_fil > 1 at R2: unclassified
 - Roads at R2: vector-based segmentation (merge objects) using Road Buffer
 - Roads with Number of pixels < 50000 at R2: unclassified
 - Roads at R2: export object shapes to ObjectShapes
 - Potental Building at BO Level: convert image objects to 'Polygon' to layer 'Potential Buildings'



RESULTS

Leaf On Accuracy Assessment

Absolute Producer's Impervious Class Accuracy: 27%
Absolute Overall Accuracy: 85%

Model Producer's Impervious Class Accuracy: 96%
Model Overall Accuracy: 96%

Leaf Off Accuracy Assessment

Absolute Producer's Impervious Class Accuracy: 24%
Absolute Overall Accuracy: 85%

Model Producer's Impervious Class Accuracy: 86%
Model Overall Accuracy: 90%

Intensity Accuracy Assessment

Absolute Producer's Impervious Class Accuracy: 55%
Absolute Overall Accuracy: 86%

Analysis and Future Work

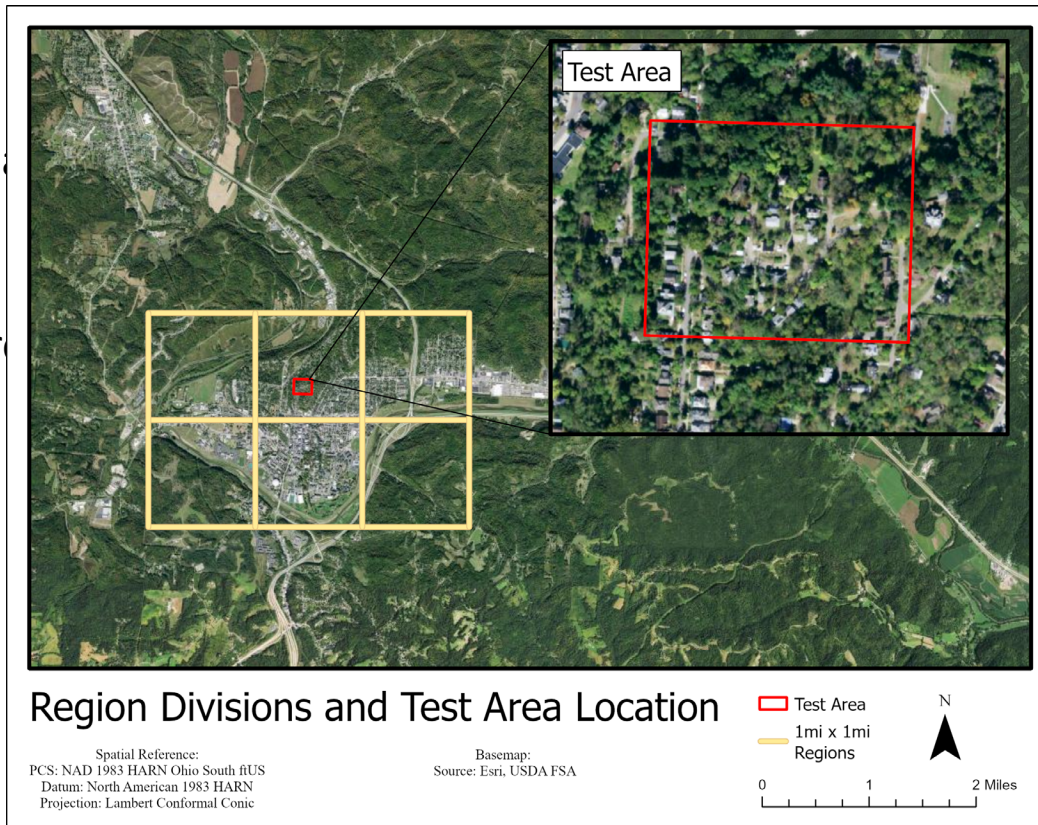
Expand Classification of Test Area Python Interface!

- Allow larger region to be classified without the use of eCognition Server or other alternatives
- Primary strengths
 - Some occluded areas included

• Prim

• Impr

-
-



class

Finale

The background features abstract, overlapping geometric shapes in various shades of green, ranging from light lime to dark forest green. These shapes are primarily located on the right side of the frame, creating a dynamic, layered effect. The rest of the background is plain white.