



Conservation Innovation Center

AI-Powered Wetland Mapping

Dr. Michael Evans
Deputy Director & Lead Data Scientist



12/16/2025



To conserve and restore the natural and cultural resources of the Chesapeake Bay watershed for the enjoyment, education and inspiration of this and future generations.





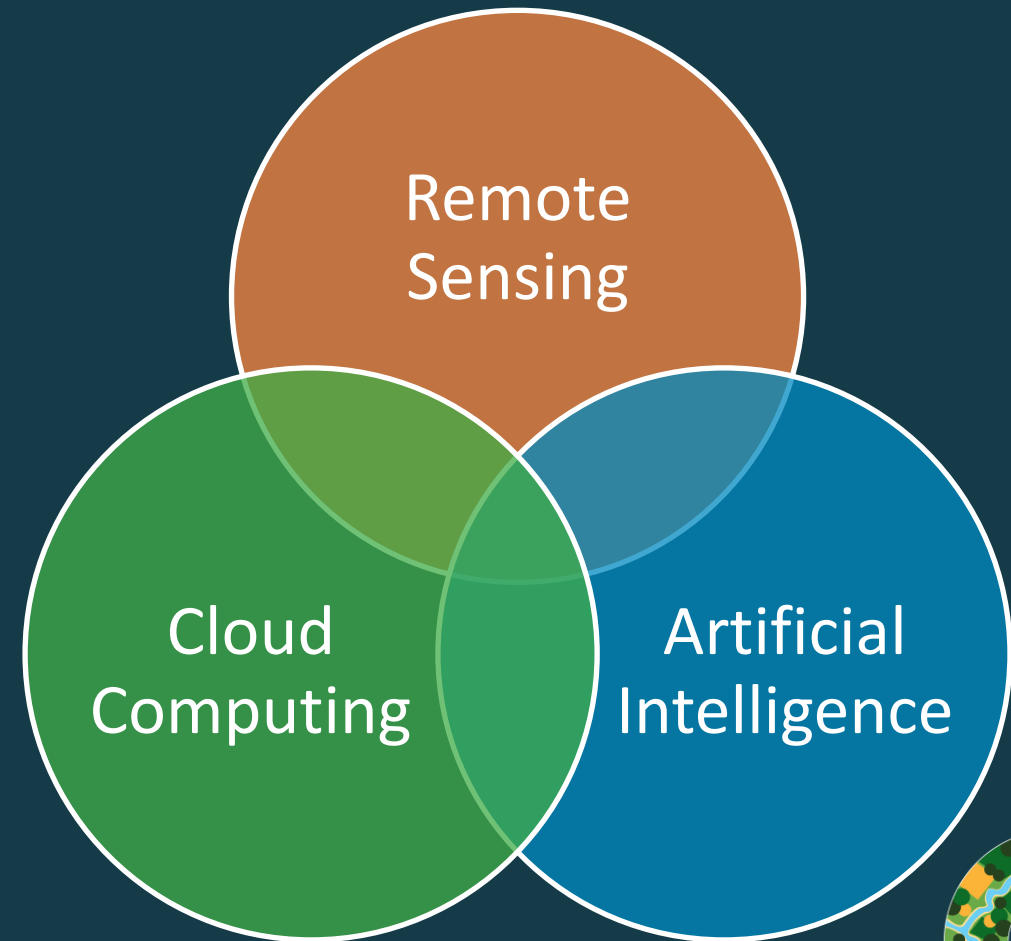
The Challenge

1. Accurate wetland maps needed for compliance and modeling
2. Existing wetland data has gaps
3. Wetlands are hard to map





The Opportunity



Goal

Make water quality modeling and regulation more effective with accurate, up-to-date wetland data.

Objectives

- Automated and repeatable
- Precise (e.g., high resolution)
- Use consistent, publicly available data
- Create data in 12 test counties



Remote Sensing Data: NAIP



Variables: Red, Green, Blue, Near Infrared

Resolution: 0.6 - 1 m

Frequency: ~ 2 – 3 years

Time: 2013 - 2022



Remote Sensing Data: Elevation



Variables: Elevation (m)

Resolution: 1 m

Frequency: ~2 -3 years

Time: 2016 - 2022



Remote Sensing Data: Sentinel-1

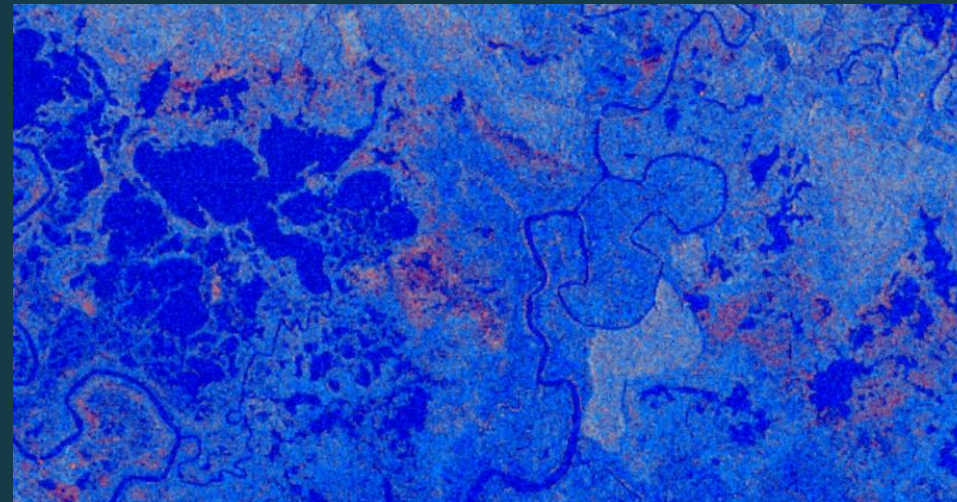
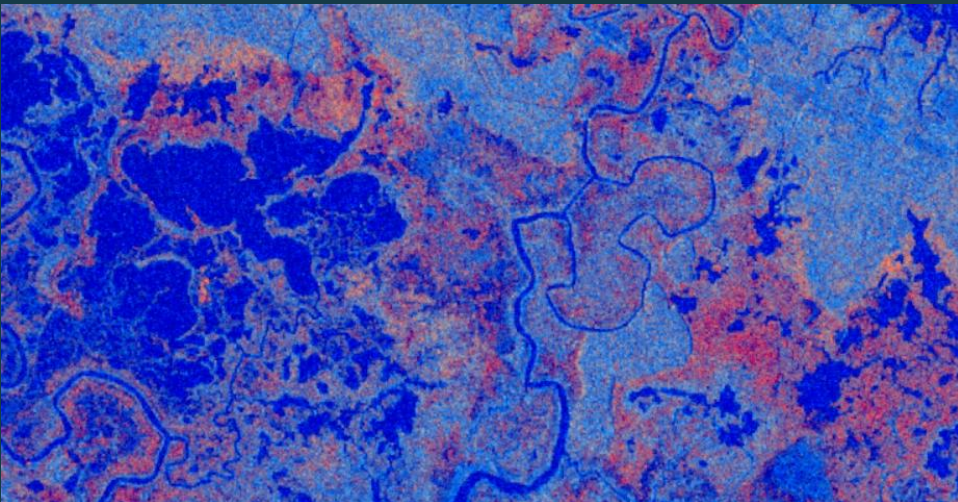
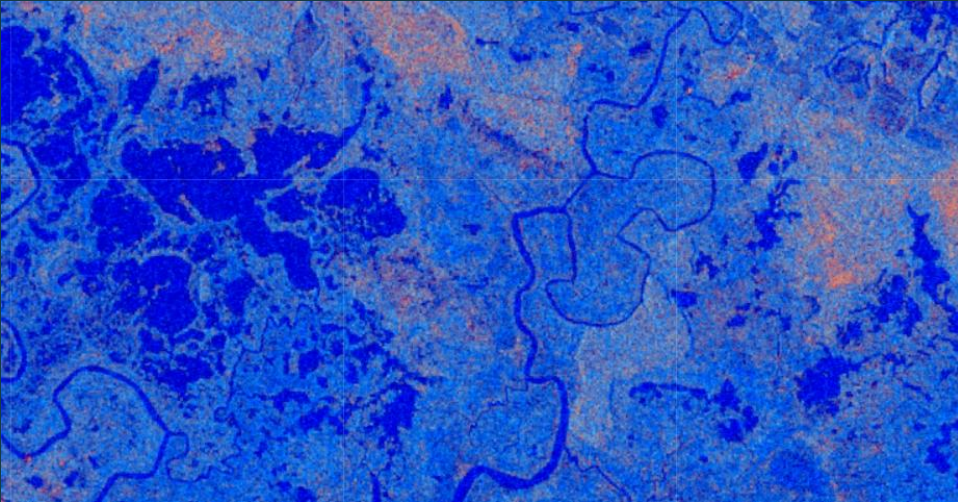


Variables: VV, VH C-band radar returns

Resolution: 10 m

Frequency: ~ every 5 days

Time: 2014 - present



Input data – SSURGO soil characteristics

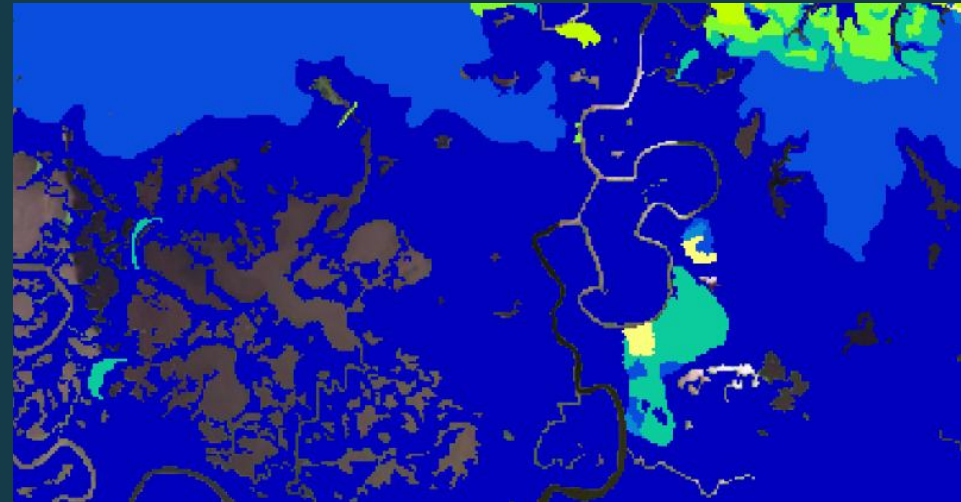
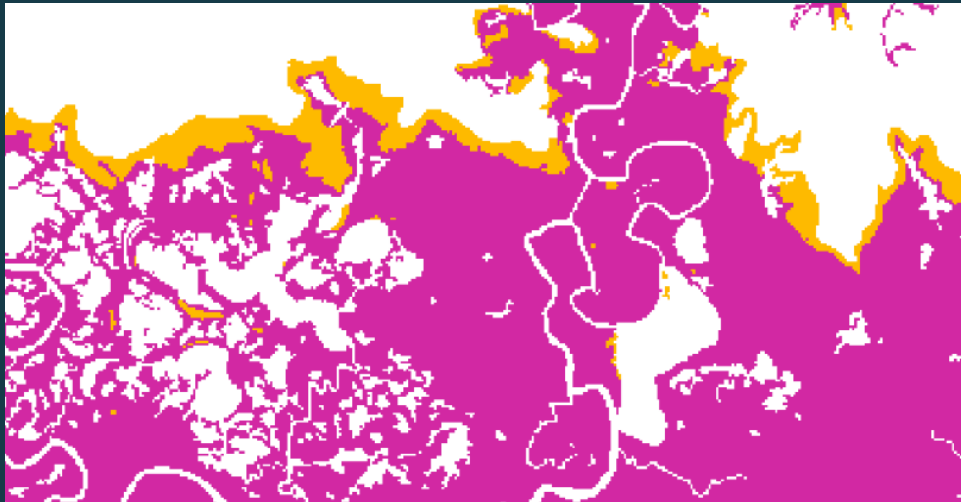


Variables: Wetness depth, Drainage class, Flood frequency, Hydric class

Resolution: 10 m

Frequency: Static

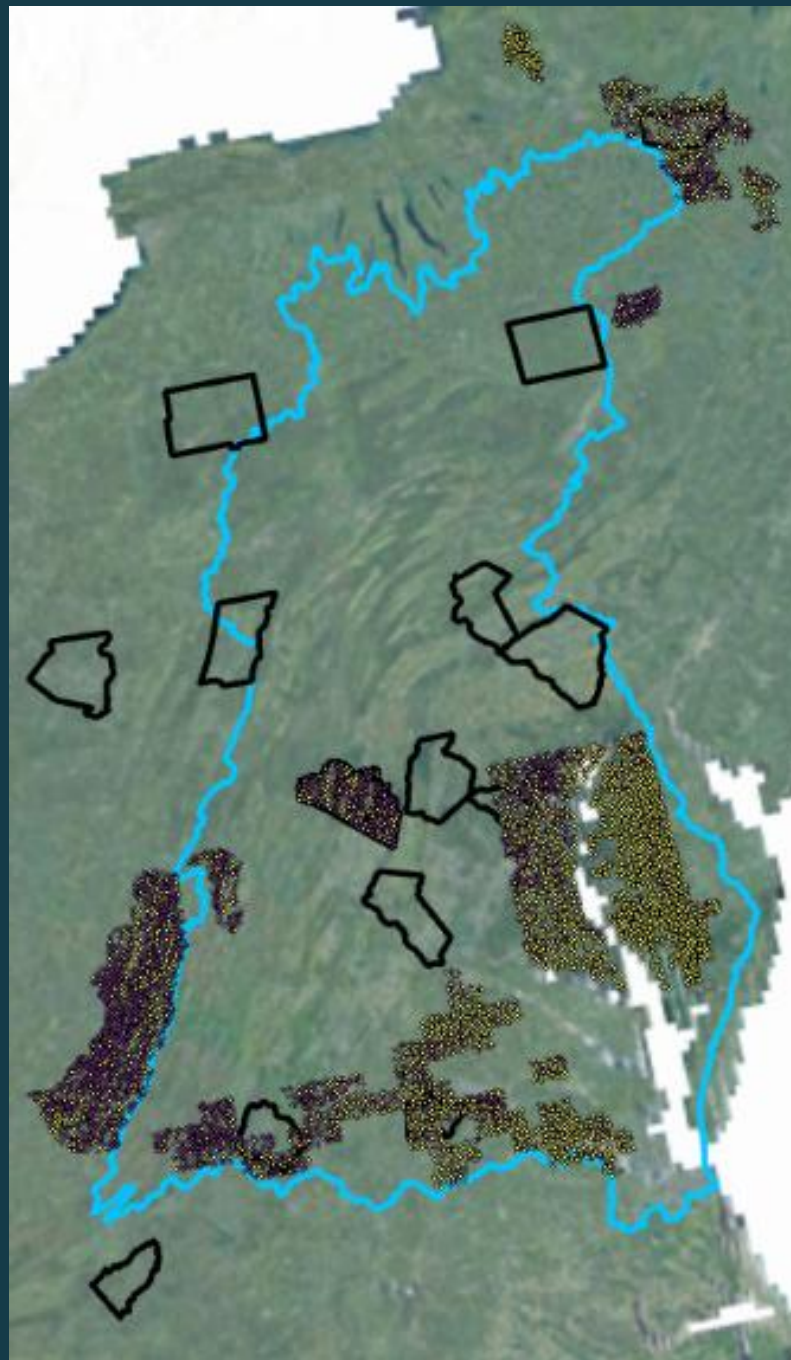
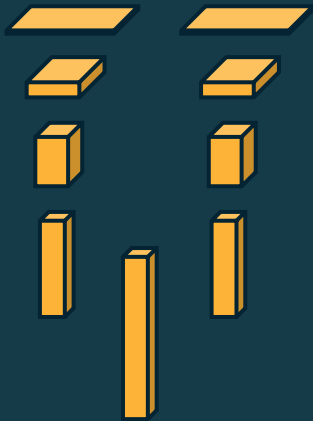
Time: Static



Model Training: Data

25,000 sample points

600 x 600 pixel 'chips'



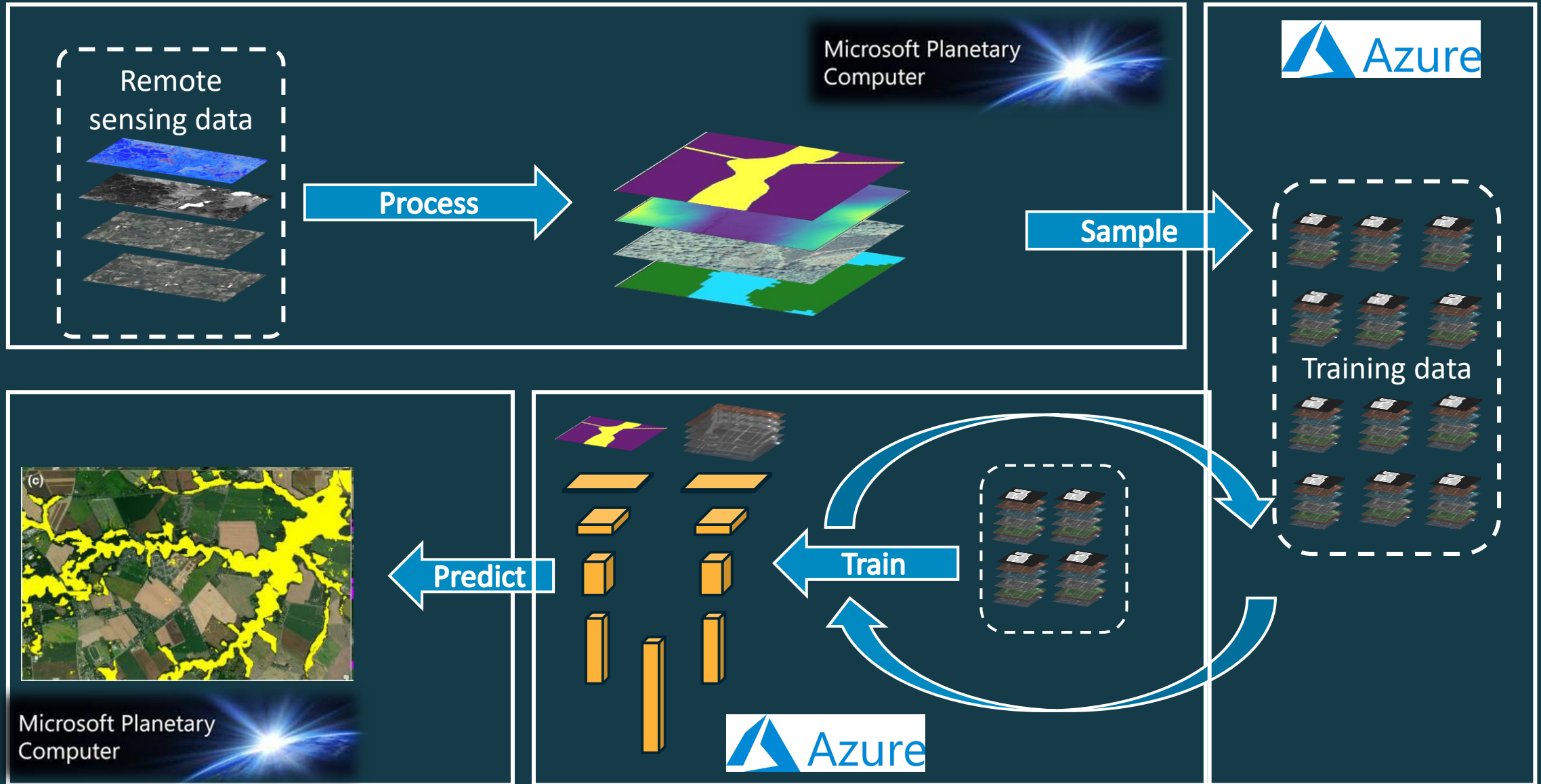
> 2016 NWI Polygons

● Inside wetlands

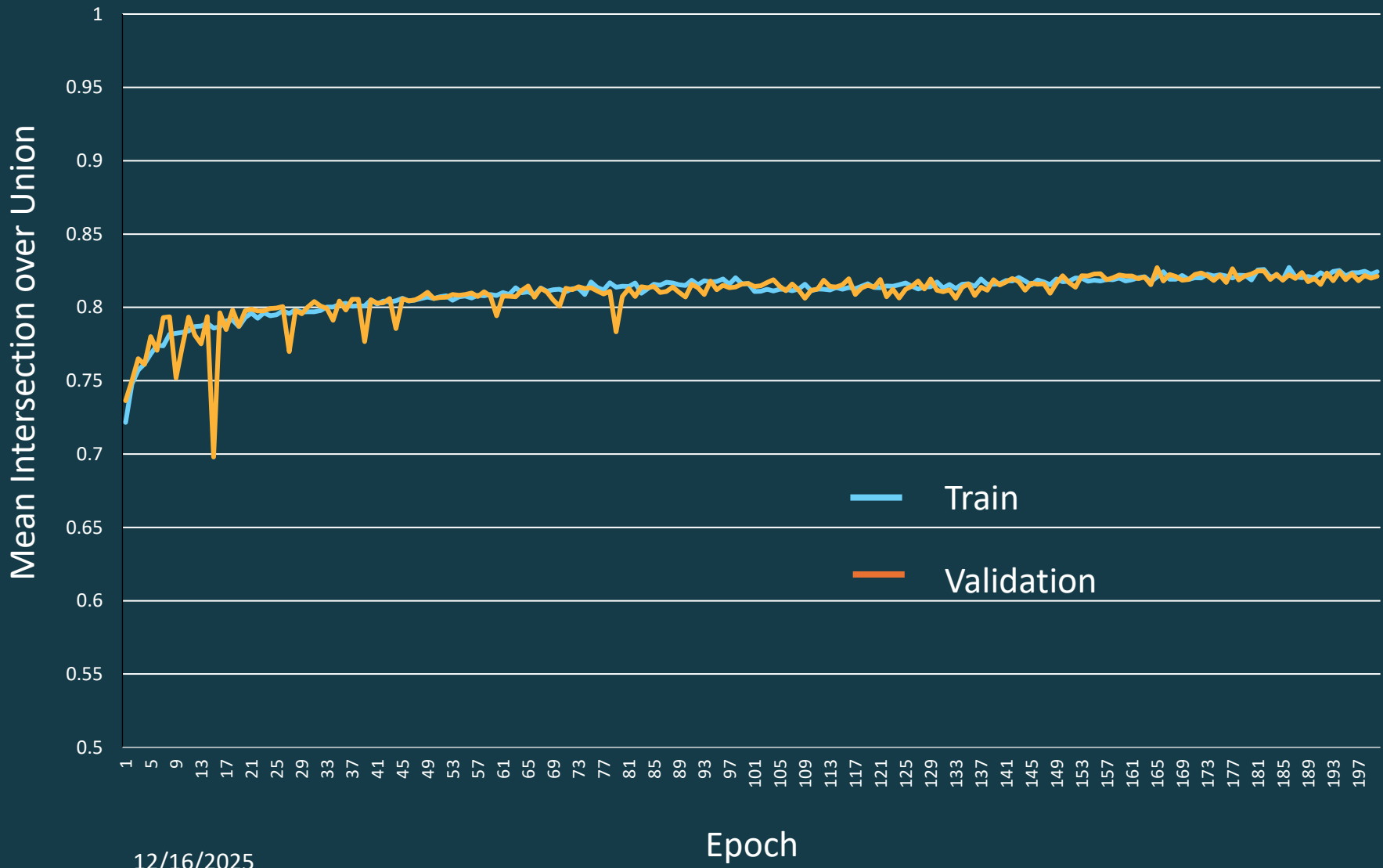
● Outside wetlands



Model Training: Workflow



Model Training: Performance



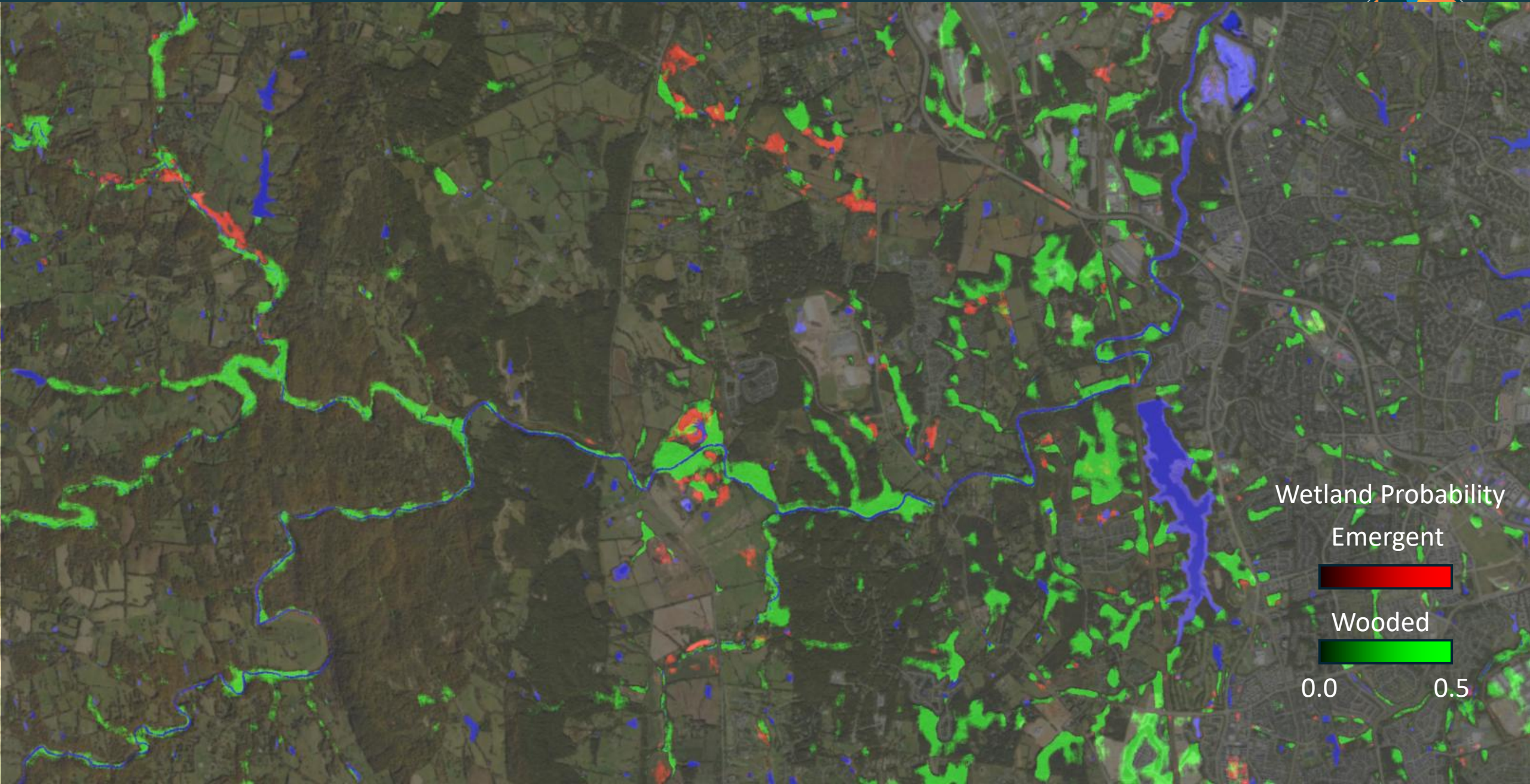
Metric	Model	
	Basic	Full
IoU	83.3%	87.3%
Accuracy	91.6%	94.0%
Precision	90.5%	96.5%
Recall	91.3%	90.2%



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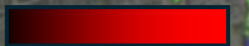
Epoch

Precise, Current Wetland Maps

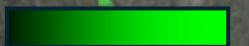


Wetland Probability

Emergent



Wooded



0.0

0.5



Location (UTM 18S):
356006, 4342231

Delineated Wetlands

- BGE
- NWI

Wetland Probability

Emergent

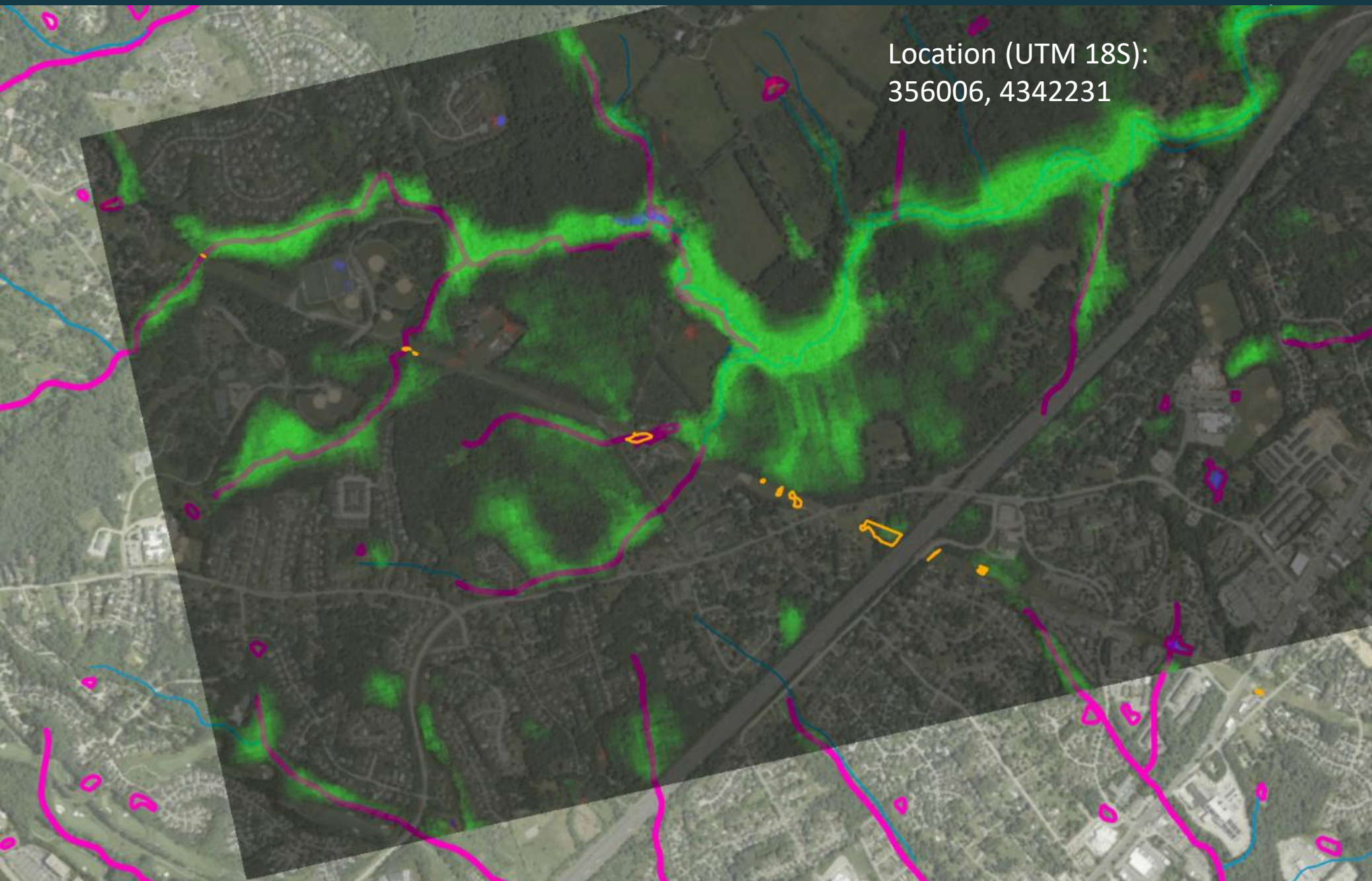


Wooded



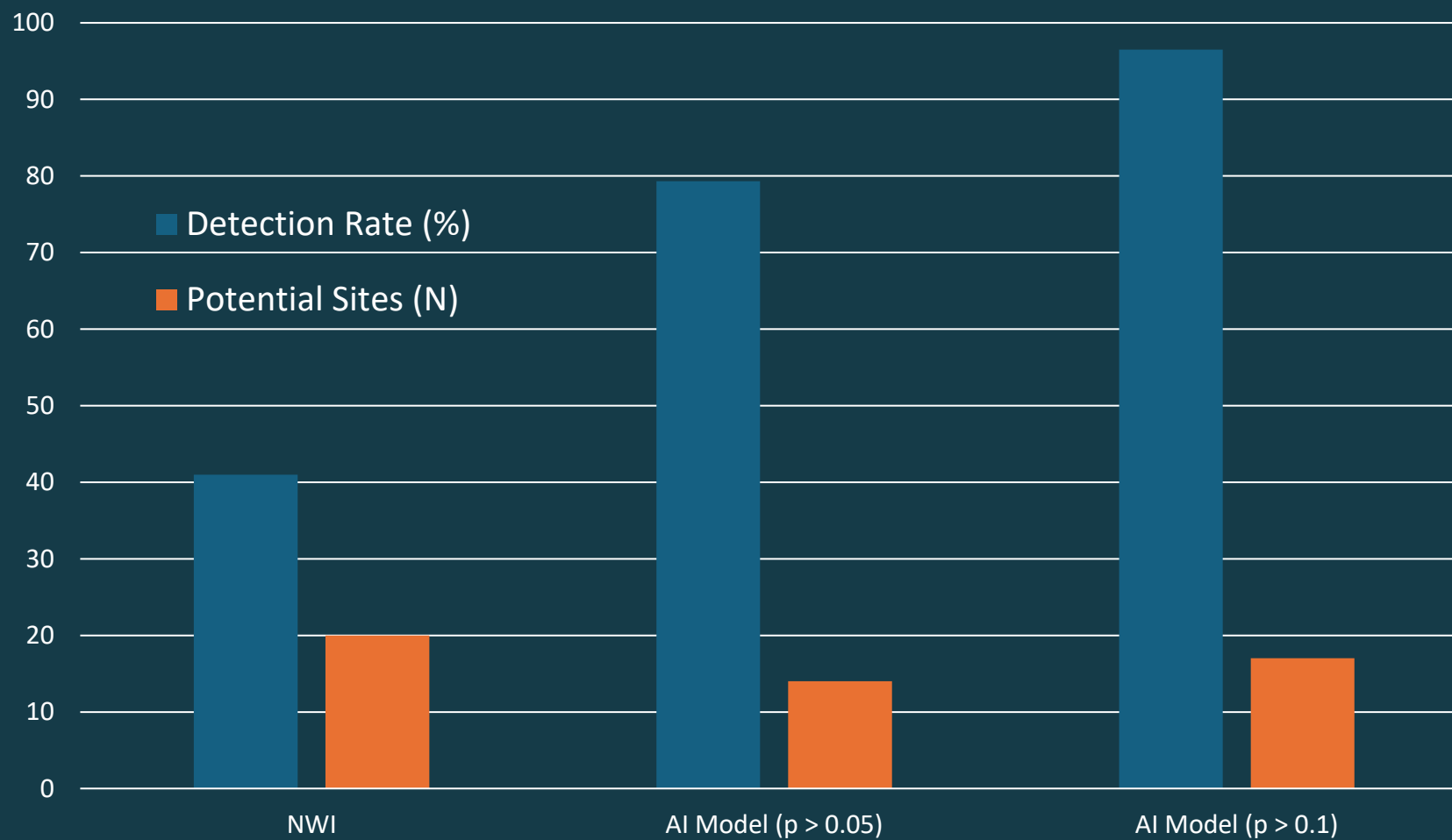
0.0

0.5





AI Model & NWI Wetland Detection Accuracy



Precise, Current Wetland Maps

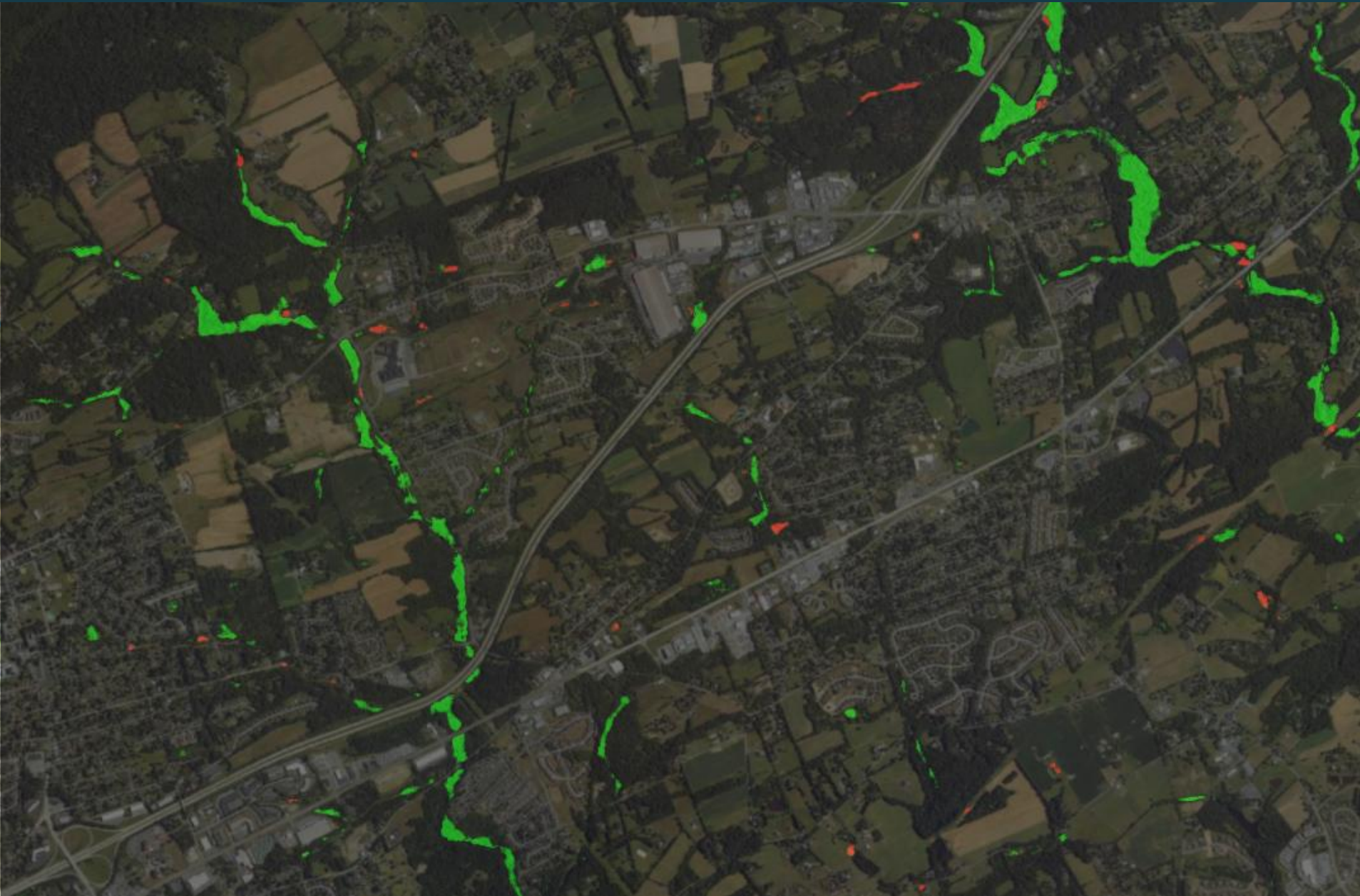


Recall

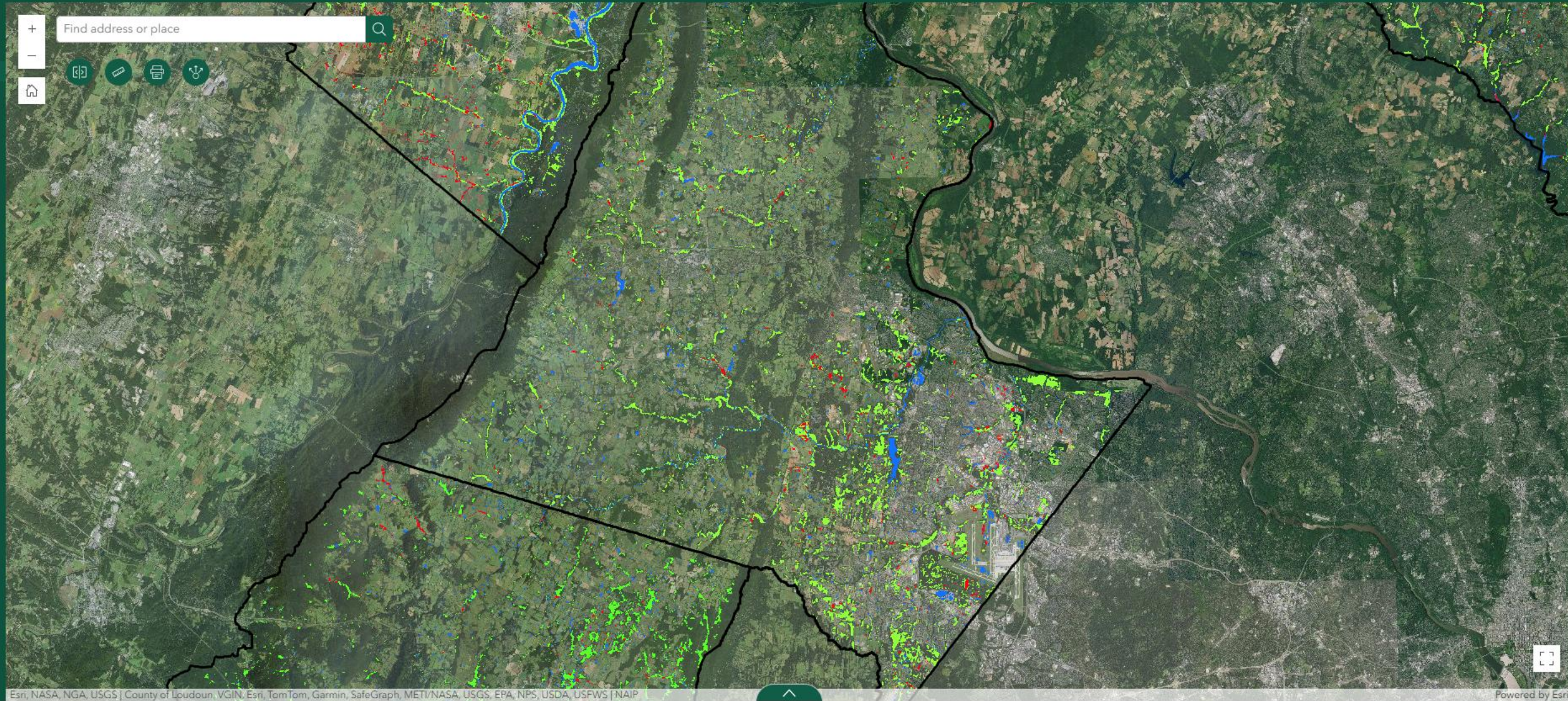
AI Wetland: 80 - 90%
NWI: 41%

Precision

AI Wetland: 93%
NWI: 72%



Chesapeake Conservancy AI Wetlands Web App

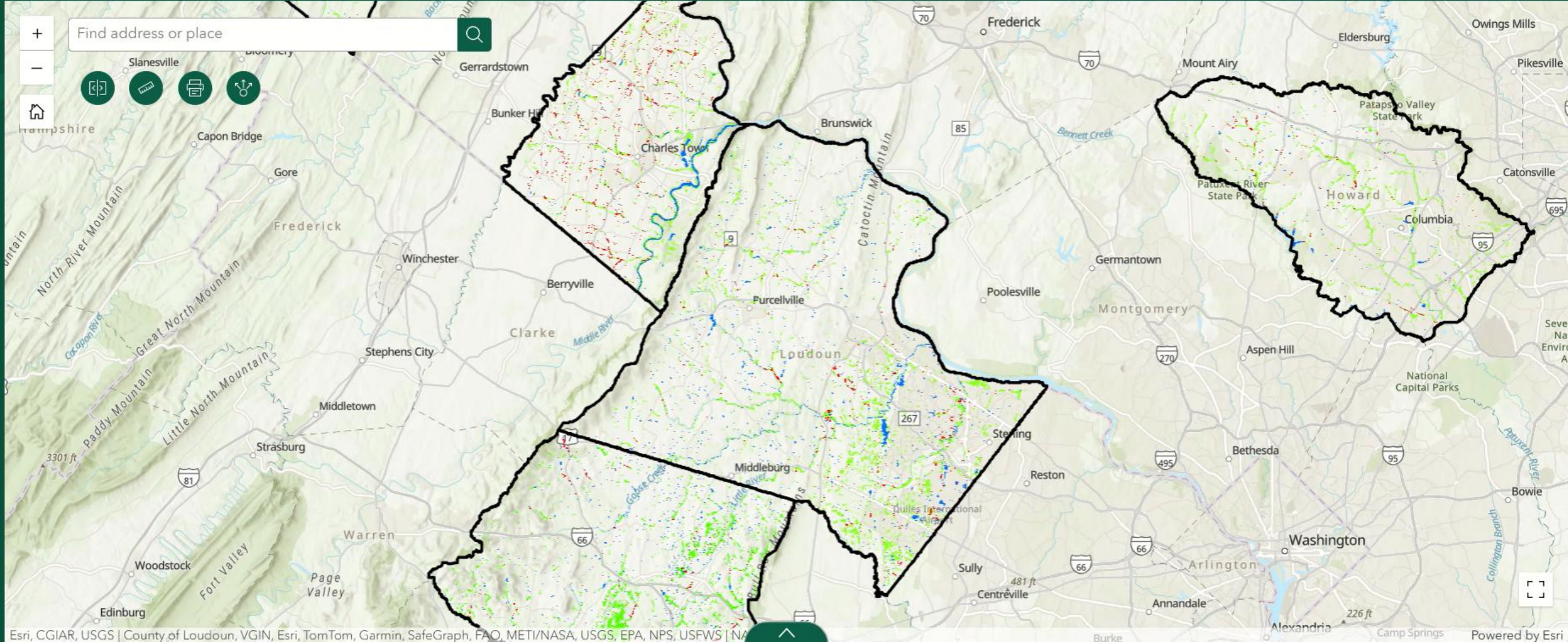


Esri, NASA, NOAA, USGS | County of Loudoun, VGIN, Esri, TomTom, Garmin, SafeGraph, METI/NASA, USGS, EPA, NPS, USDA, USFWS | NAIP

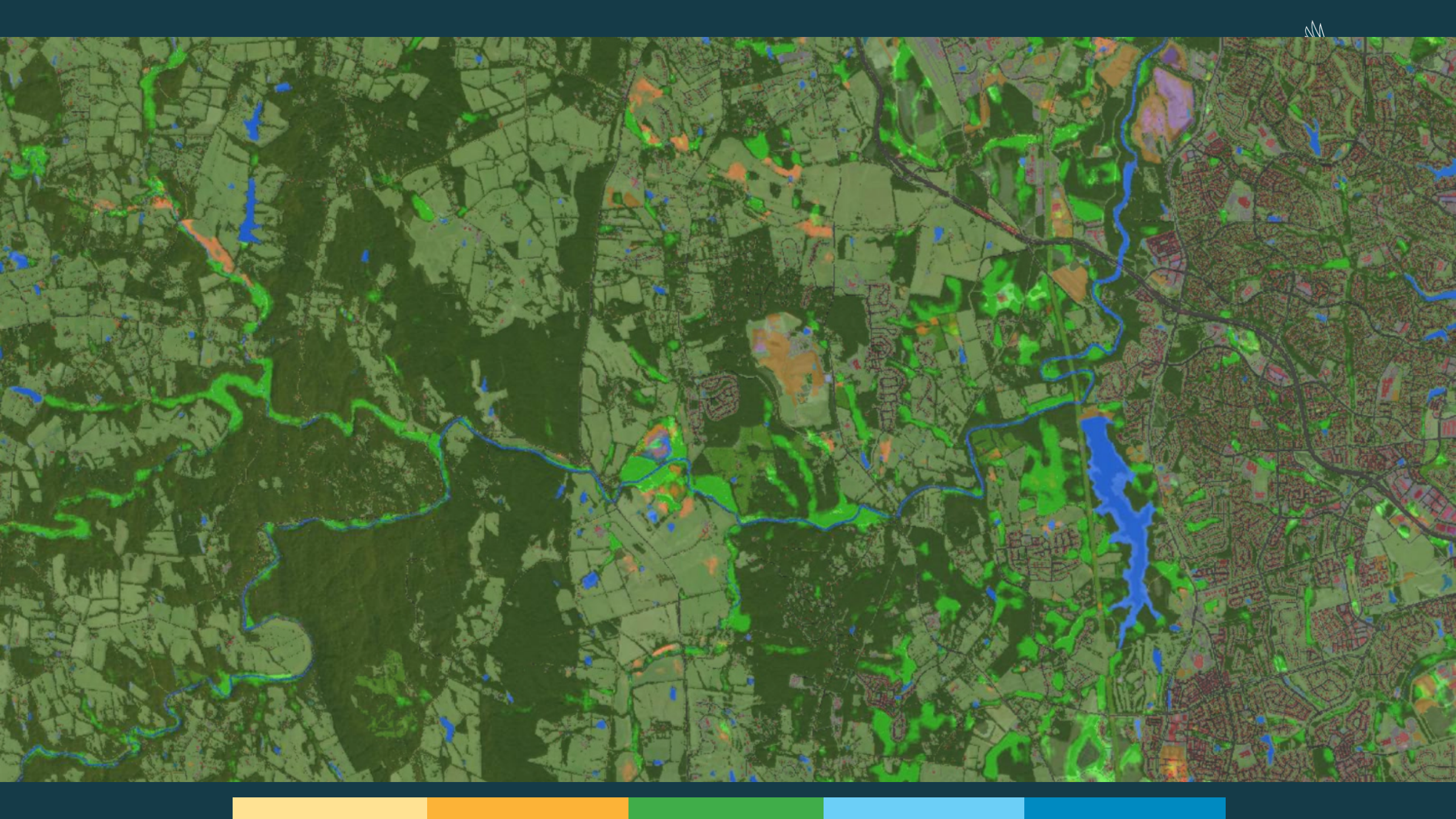
Powered by Esri

<https://experience.arcgis.com/experience/4c220be8365b47d19f3effbdc98f005b>

Chesapeake Conservancy AI Wetlands Web App



<https://experience.arcgis.com/experience/4c220be8365b47d19f3effbdc98f005b>



Restoration Opportunities





Chesapeake
Conservancy



Conservation
Innovation Center

EPRI



Chesapeake Bay Program
Science. Restoration. Partnership.

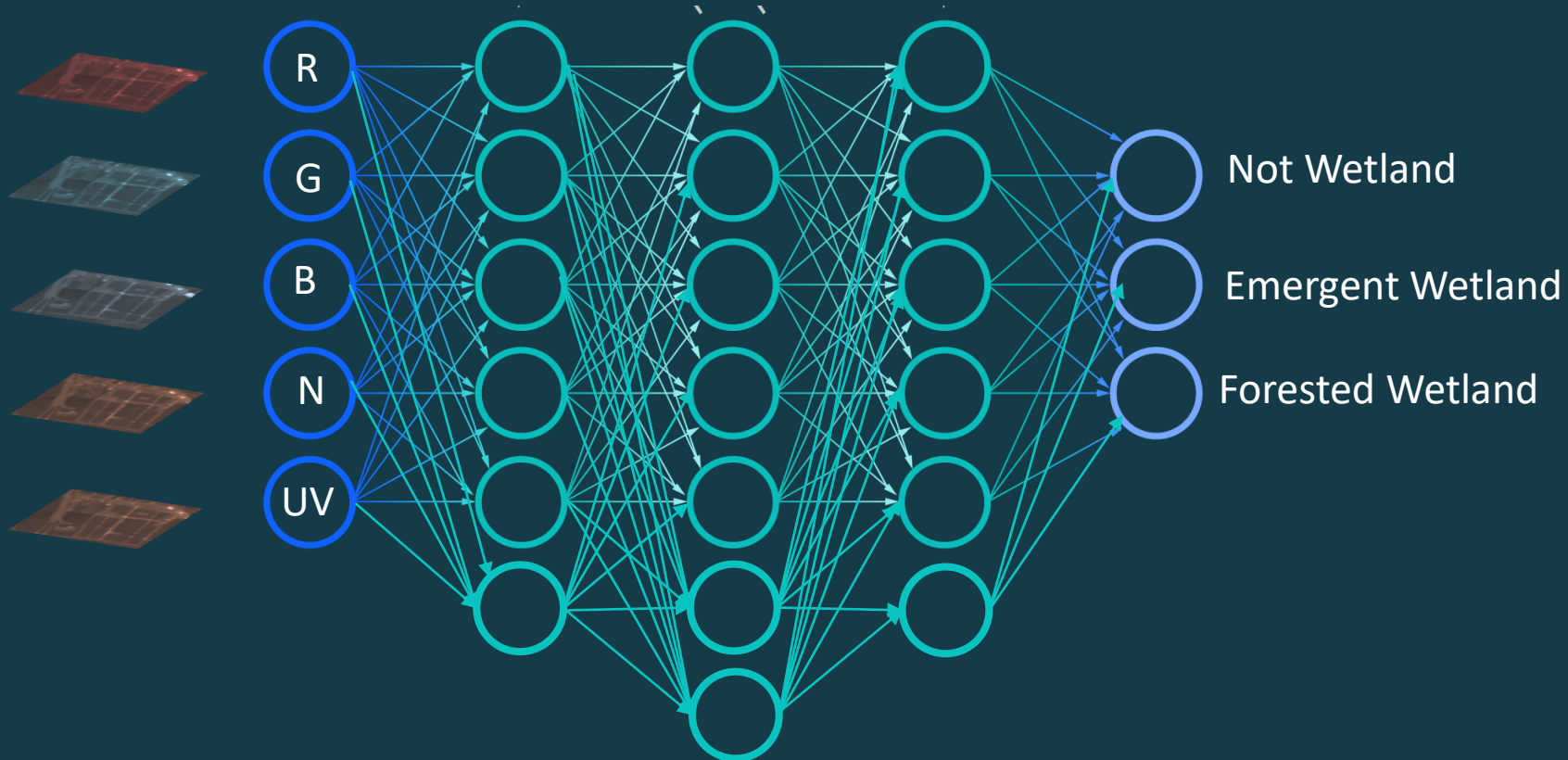
Thank You!

Michael Evans, PhD
mevans@chesapeakeconservancy.org

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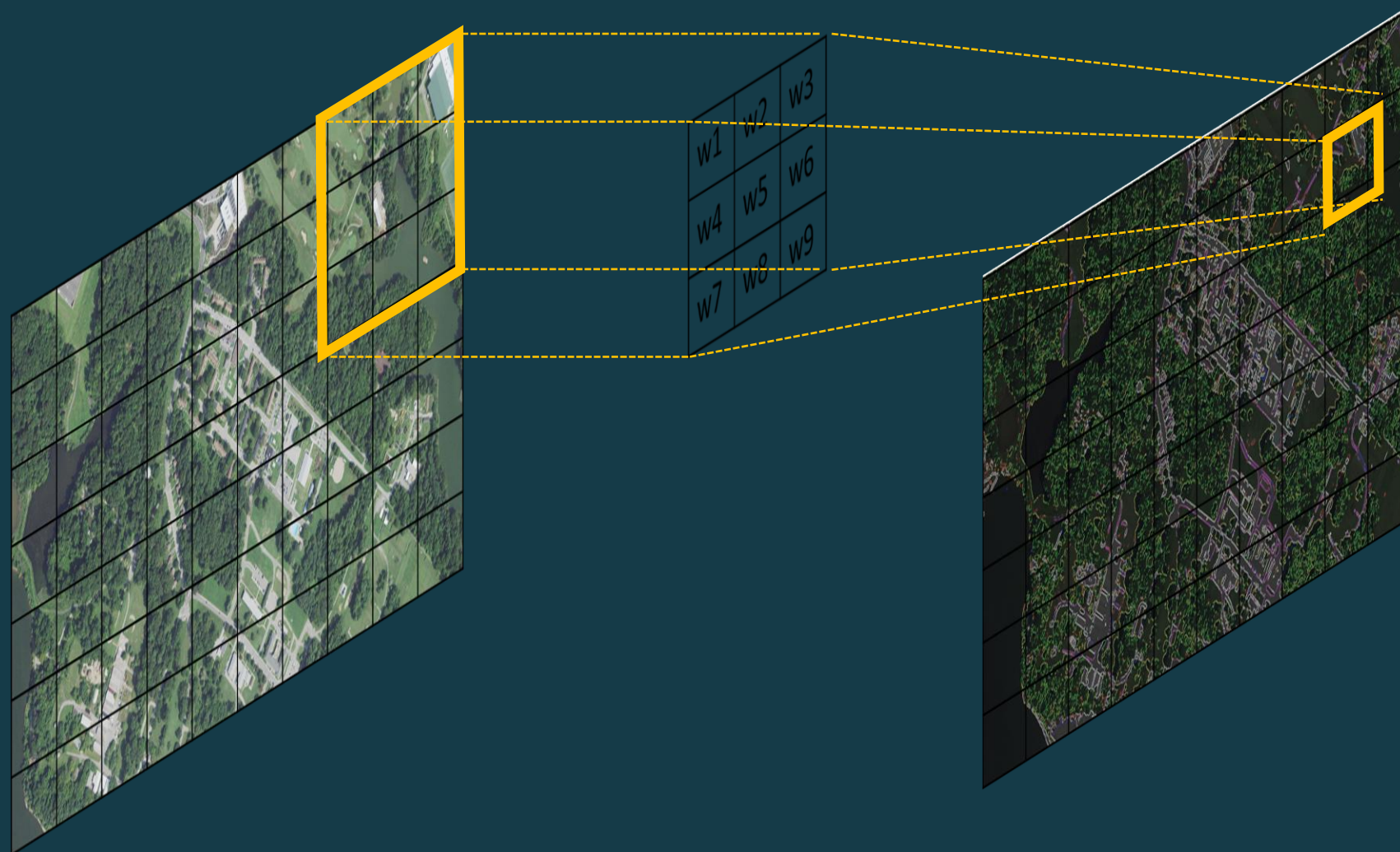
Deep Learning (AI)

Neural network – learn non-linearities, conditionalities, & interactions



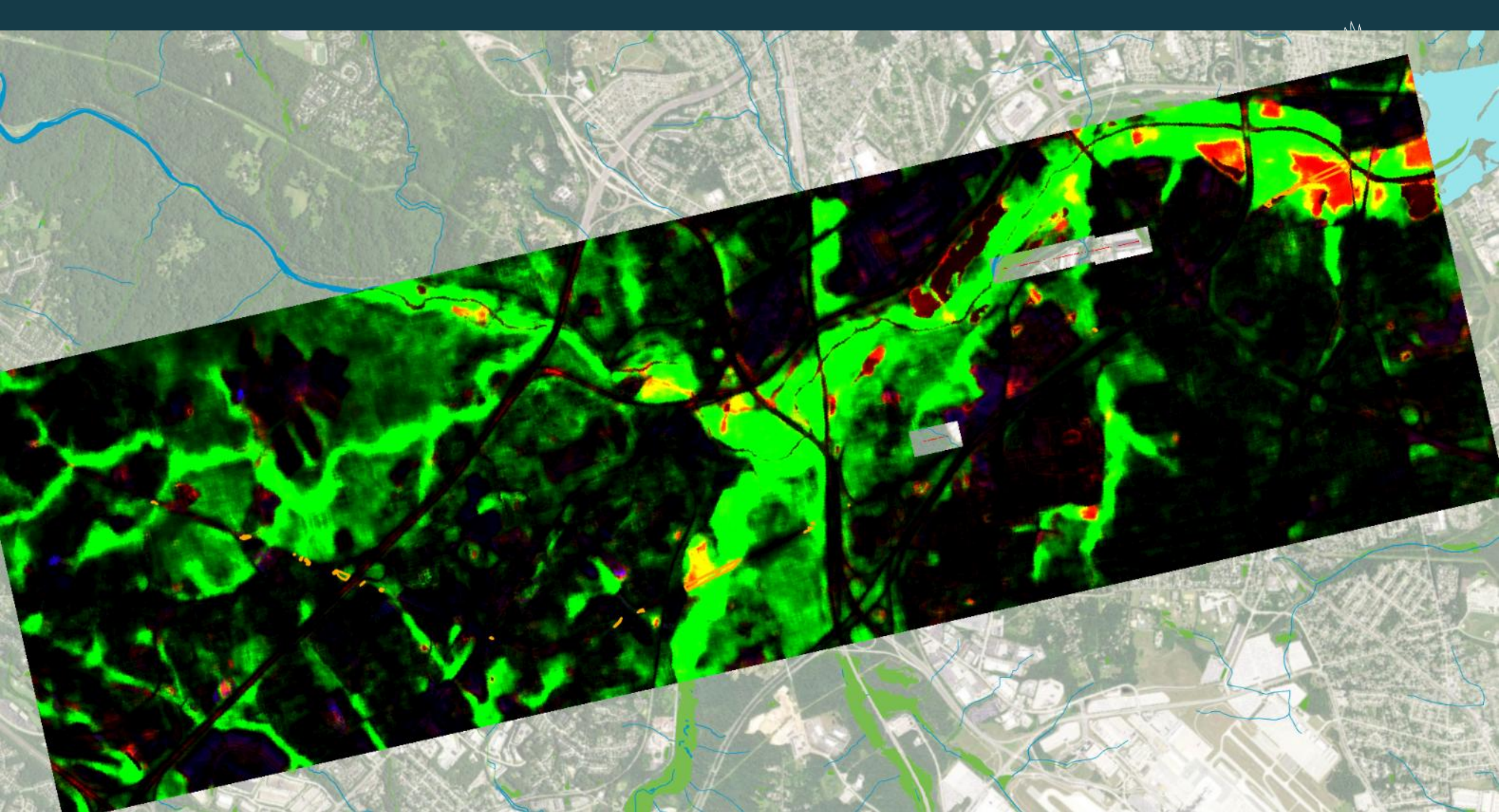
Deep Learning: Convolution

learn the shape and context of objects in images

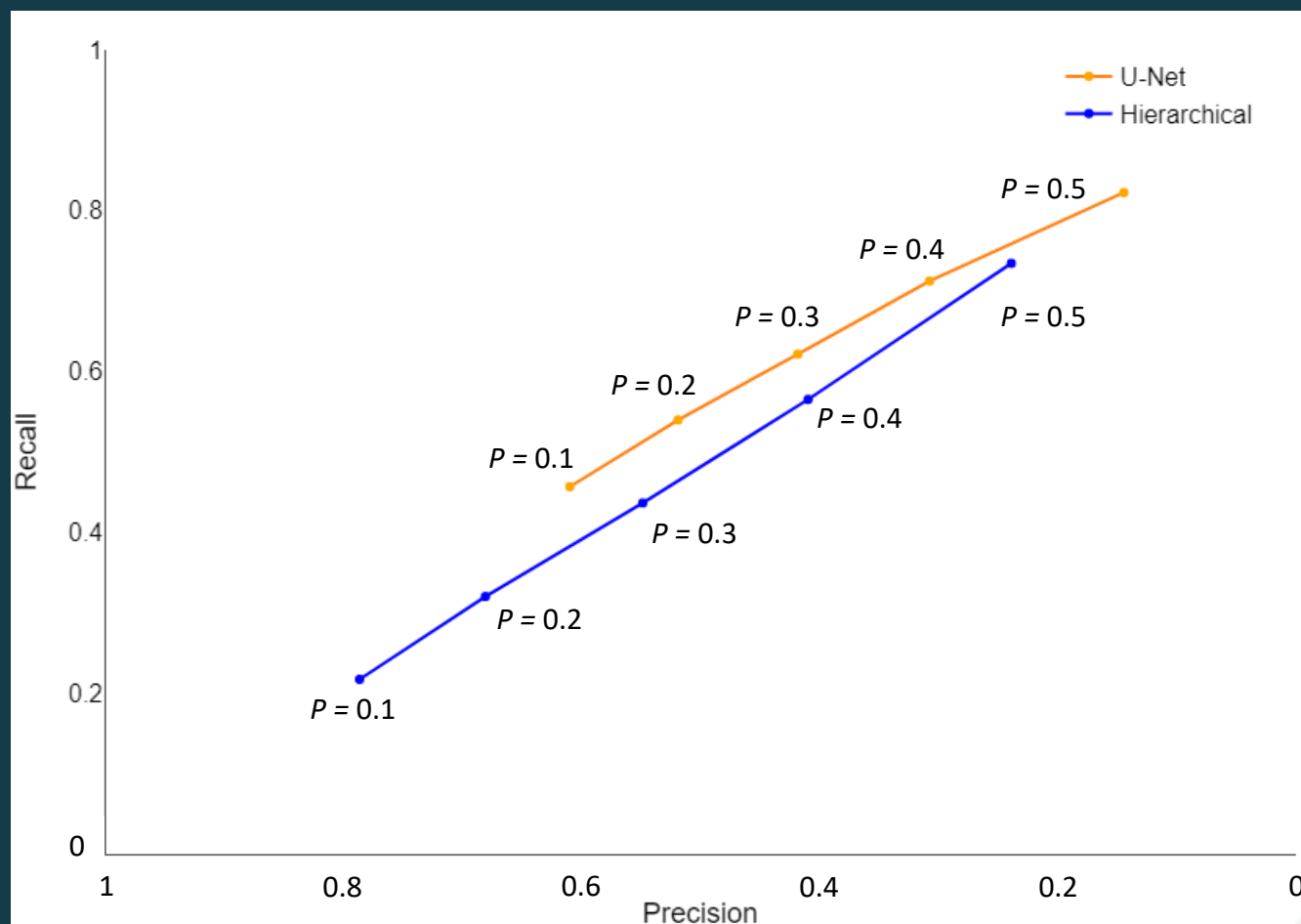


Greenbury Point, MD. Sep, 2022



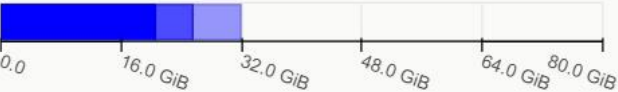




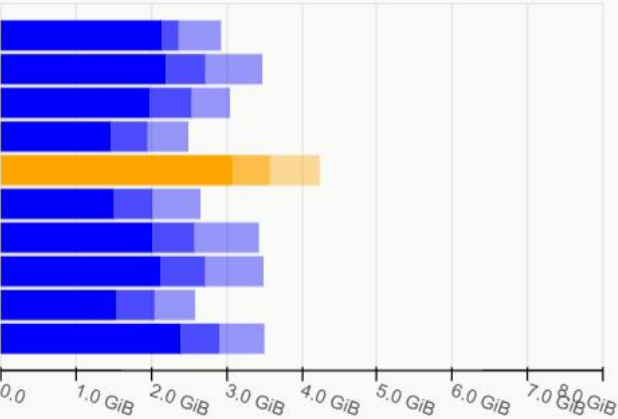




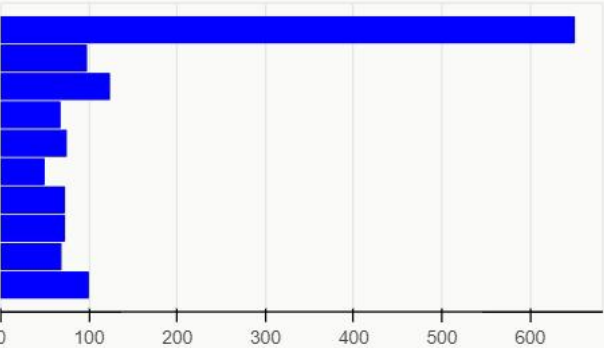
Bytes stored: 31.92 GiB



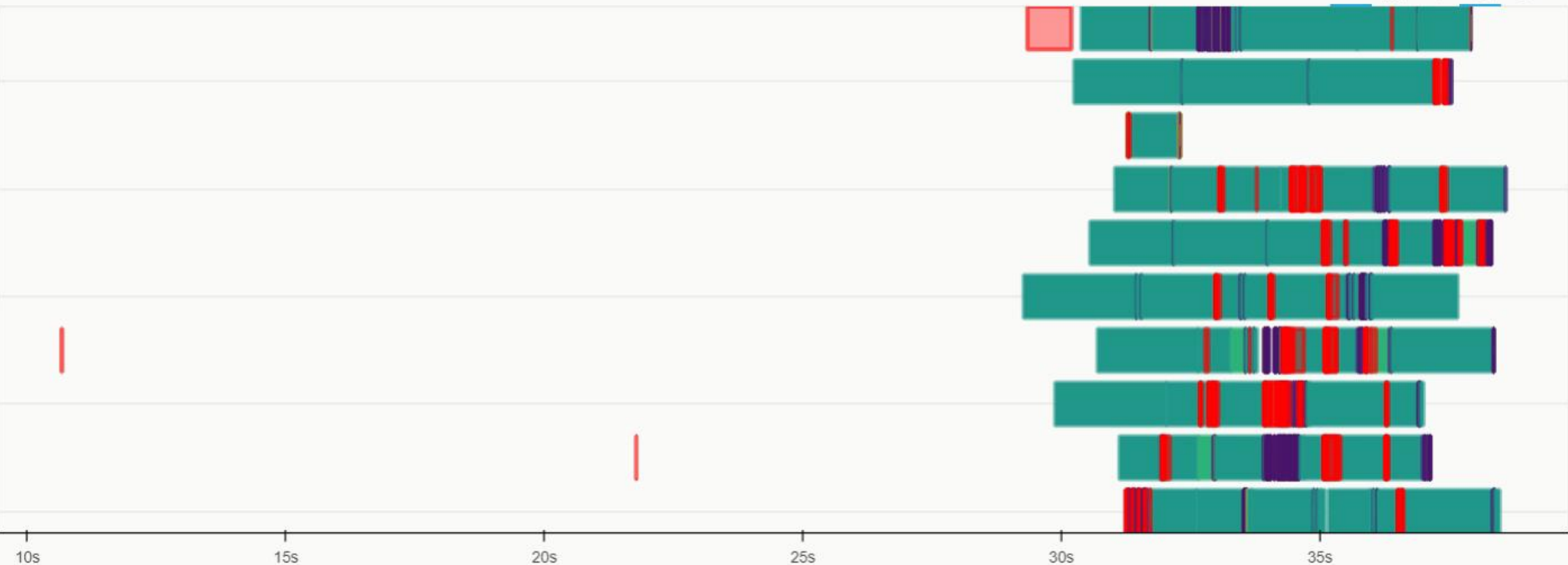
Bytes stored per worker



Tasks Processing



Task Stream



Progress -- total: 19324, in-memory: 1122, processing: 1371, waiting: 15266, erred: 0

rechunk-split	1524 / 6254	rechunk-merg...	30 / 624	moment_combi...	8 / 208
getitem	107 / 4606	mean_chunk	29 / 624	mean_combine...	8 / 208
where	324 / 792	concatenate	48 / 428	asset-table	132 / 132
rechunk-merge	130 / 676	_trim	0 / 350	asset-table	132 / 132
truediv-getitem	24 / 624	block-info	32 / 350	add	4 / 104
moment_chunk	30 / 624	predict_chunk	0 / 350	rechunk-spli...	30 / 104
where-getitem	30 / 624	getitem-overlap	0 / 350	mean_agg-agg...	4 / 104
nanmedian	30 / 624	rechunk-merg...	4 / 259	rechunk-merg...	2 / 91



GEOMARVEL



Conservation
Innovation Center

Remote Sensing Data

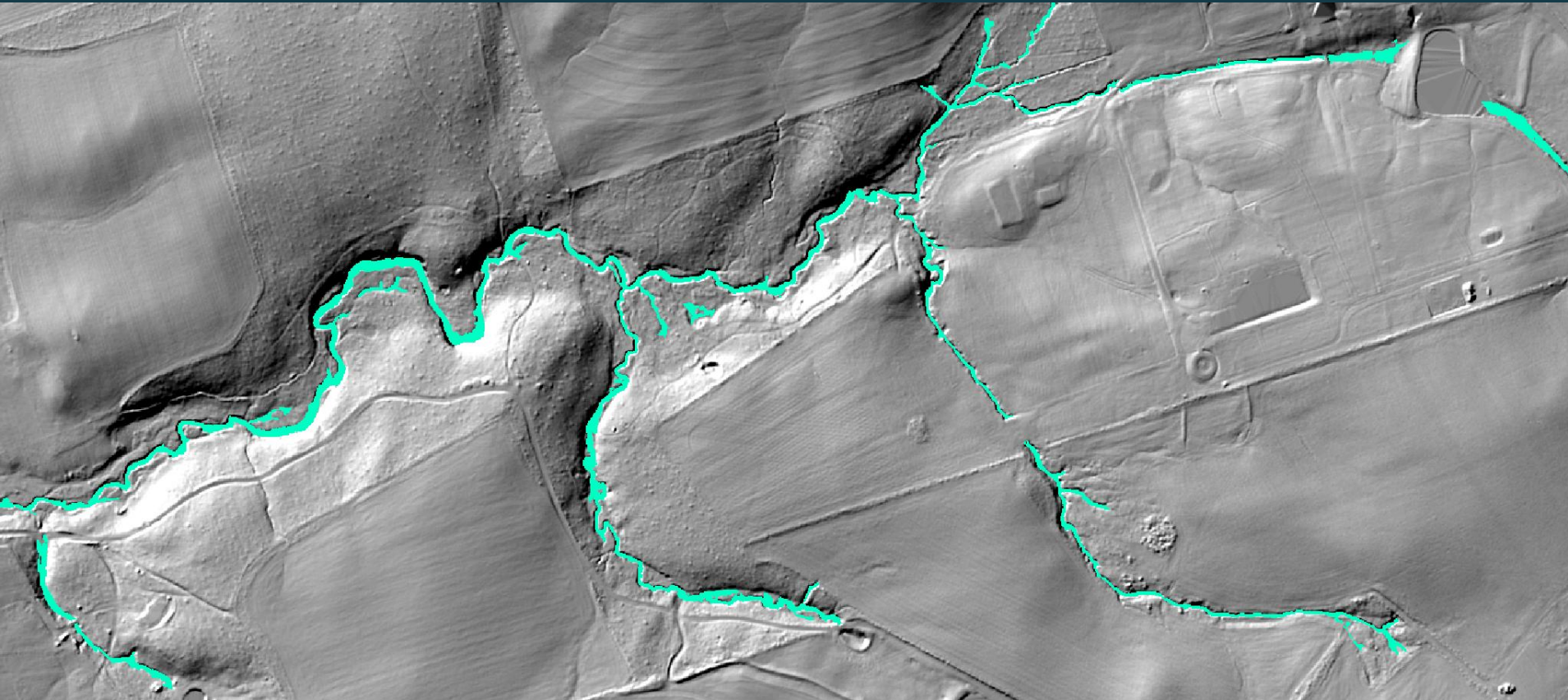


1. Data about Earth's surface
2. Collected by satellite or plane
3. Can have multiple 'bands'
4. Many types of data (radar, lidar, etc.)





Hyper-resolution hydrography



Partnership between CC,
UMBC, and EPA's
Chesapeake Bay Program

Maps headwaters and
lateral positioning of
stream channels more
precisely than NHD



High Resolution Hydrography Viewer



- ▷ Boundaries ...
- ▽ NHDFlowline ...
 - NHDFlowline ⓘ
- ▽ Draft High Resolution Hydrography ...
 - Draft_High_Resolution_Hydrography ⓘ
- ▷ World Imagery ...

Boundaries

Chesapeake Bay Watershed
Counties



NHDFlowline

NHDFlowline

Draft High Resolution Hydrography

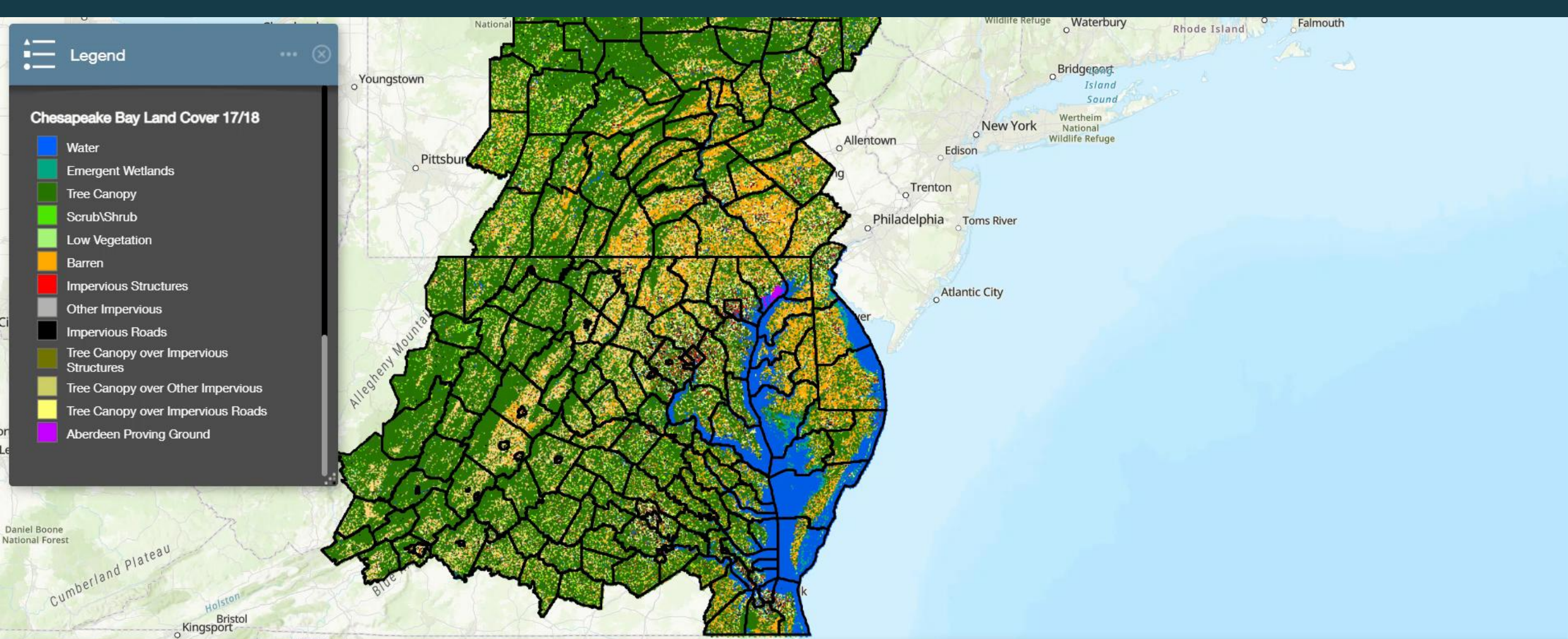
Draft_High_Resolution_Hydrography

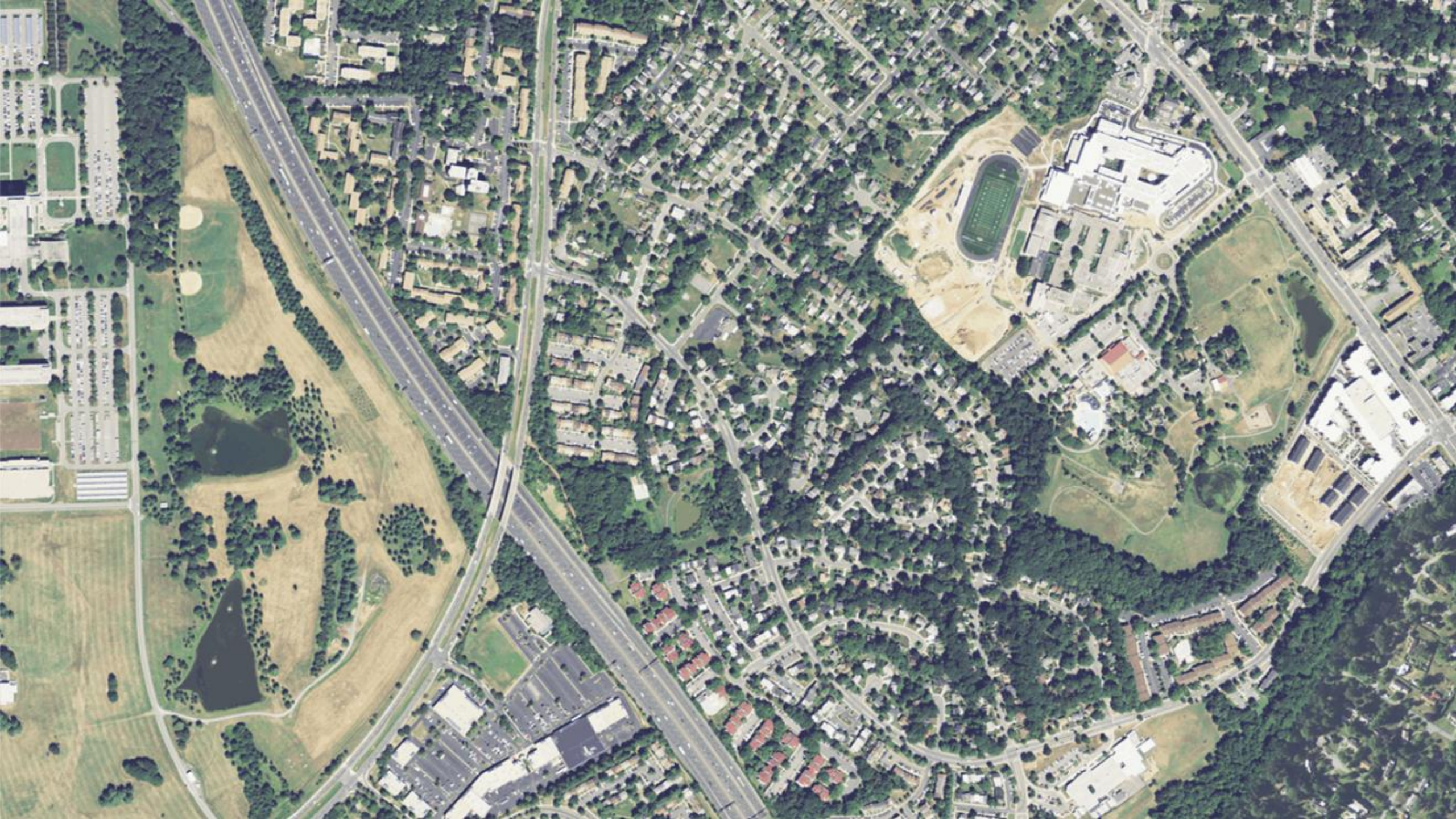
Line Type

channel

connector

<https://cicgis.org/portal/apps/experiencebuilder/experience/?id=8b64a2aa6a8544b88e3dd599ca132210>





Next Steps

1. Produce public web maps
2. Create data for entire CBW
3. Tool development
4. Evaluate application to NWI

12/16/2025



18S 348056 4342676



NWI

 Wetland

 Water

AI Wetland
Probability

Emergent



Wooded



Farmed



18S 349331 4341842



 Wetland

 Water

AI Wetland
Probability

Emergent



Wooded



Farmed



18S 352494 4341268



AI Wetland
Probability
Emergent



Wooded



Farmed



18S 356004 4342228



Wetland

Water

AI Wetland
Probability

Emergent



Wooded



Farmed



18S 358074 4343035



 Wetland

 Water

AI Wetland
Probability

Emergent



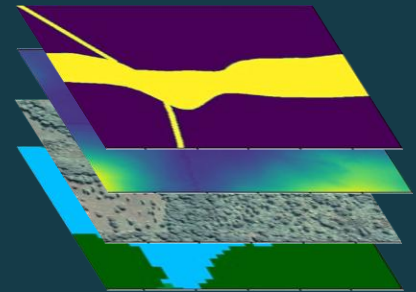
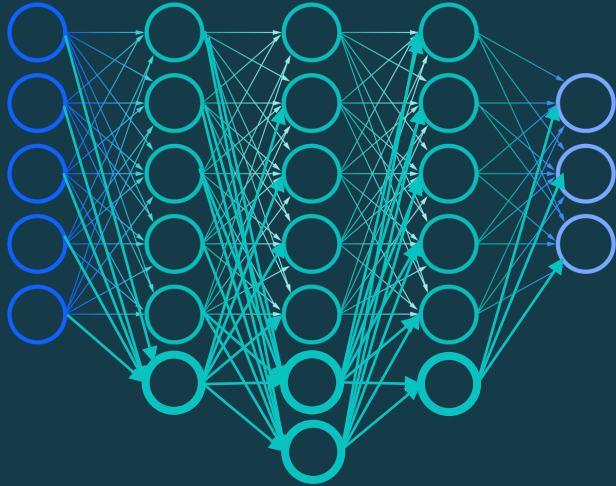
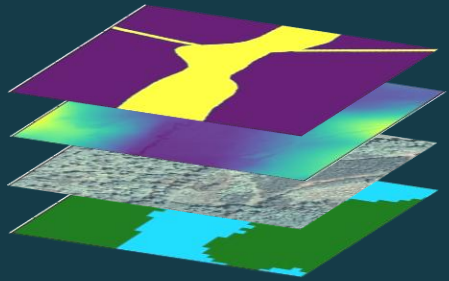
Wooded



Farmed



Model Training: Active Learning



Not Wetland = **0.83**

Emergent Wetland = **0.12**

Forested Wetland = **0.05**

Not Wetland = **0.33**

Emergent Wetland = **0.41**

Forested Wetland = **0.26**

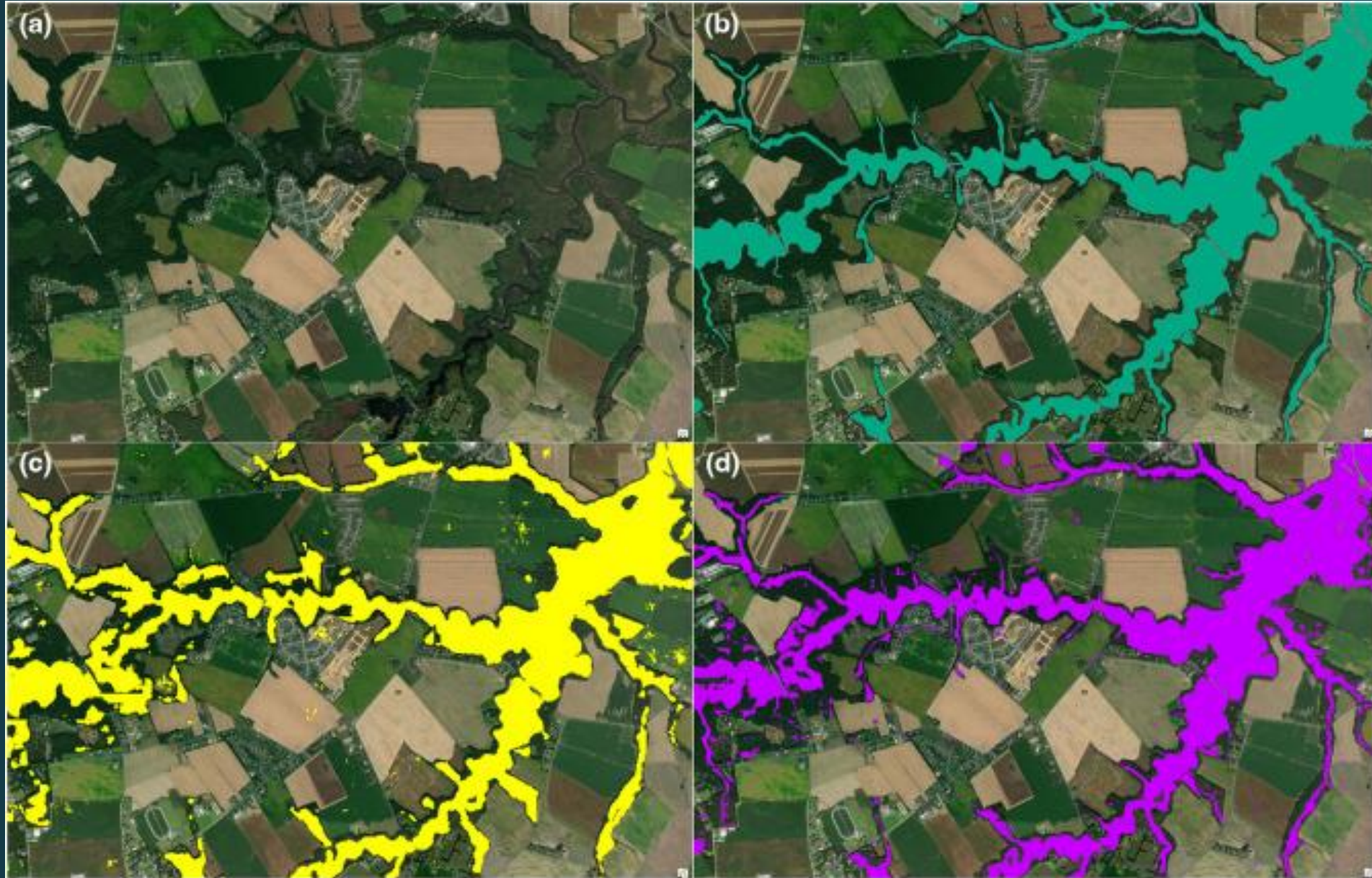


Model Outputs

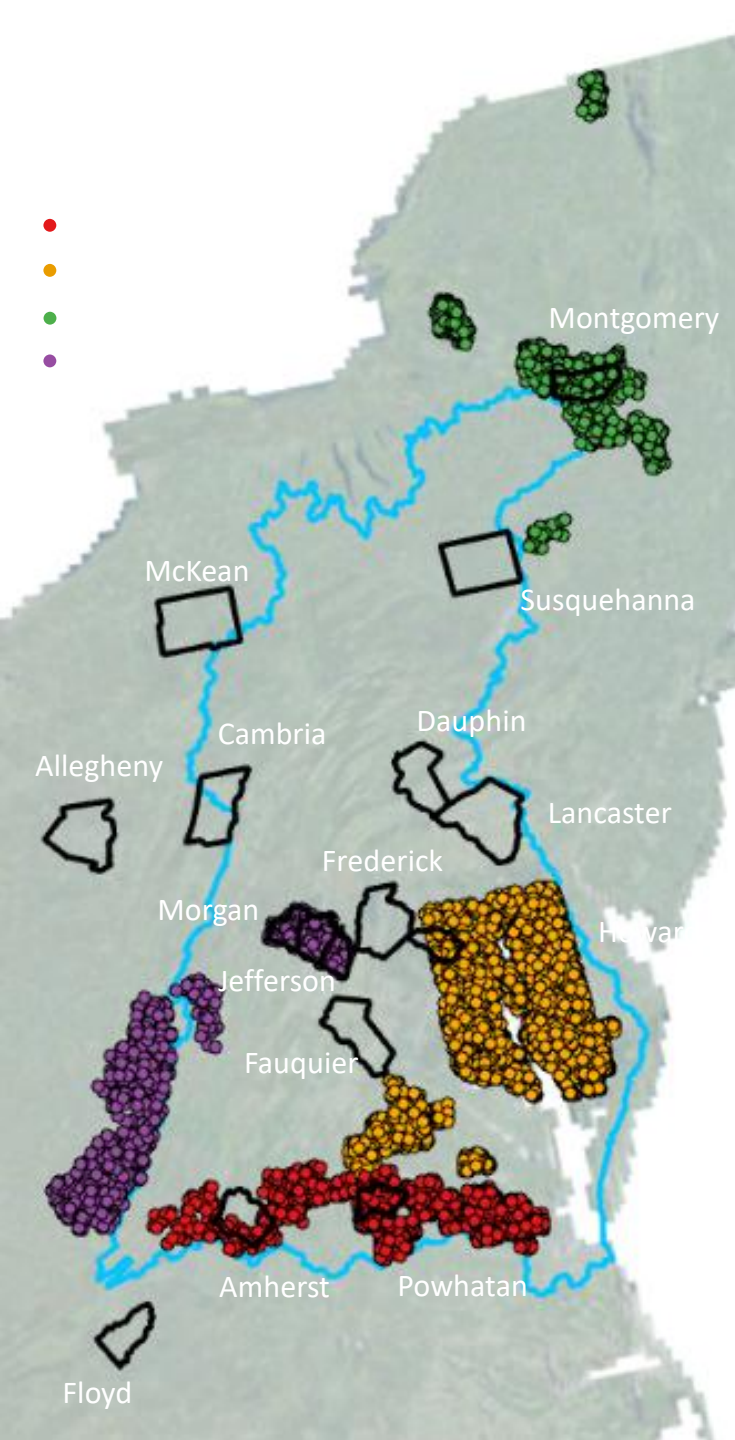


NWI label

Full model prediction



Basic model prediction

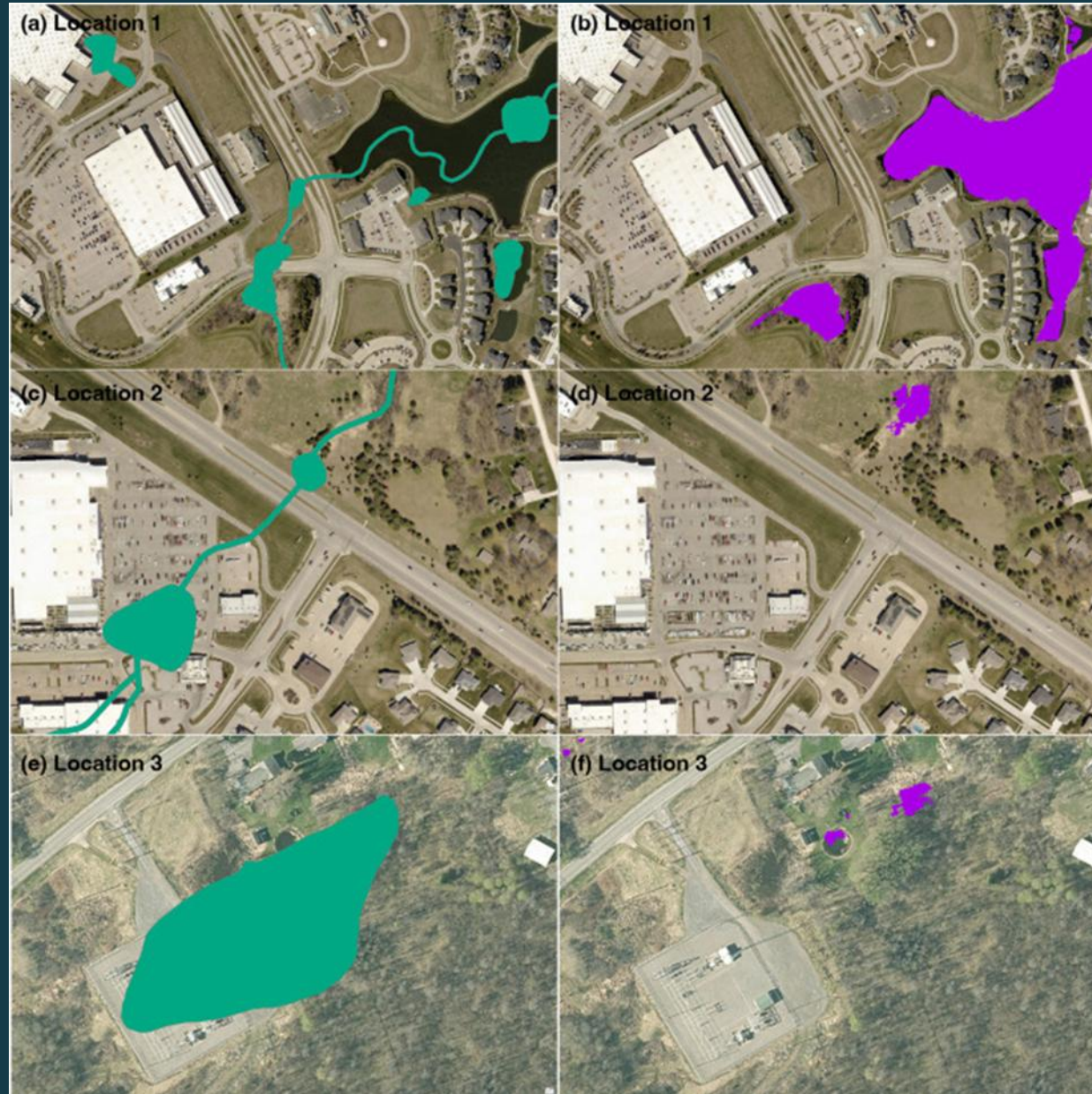


Model Outputs

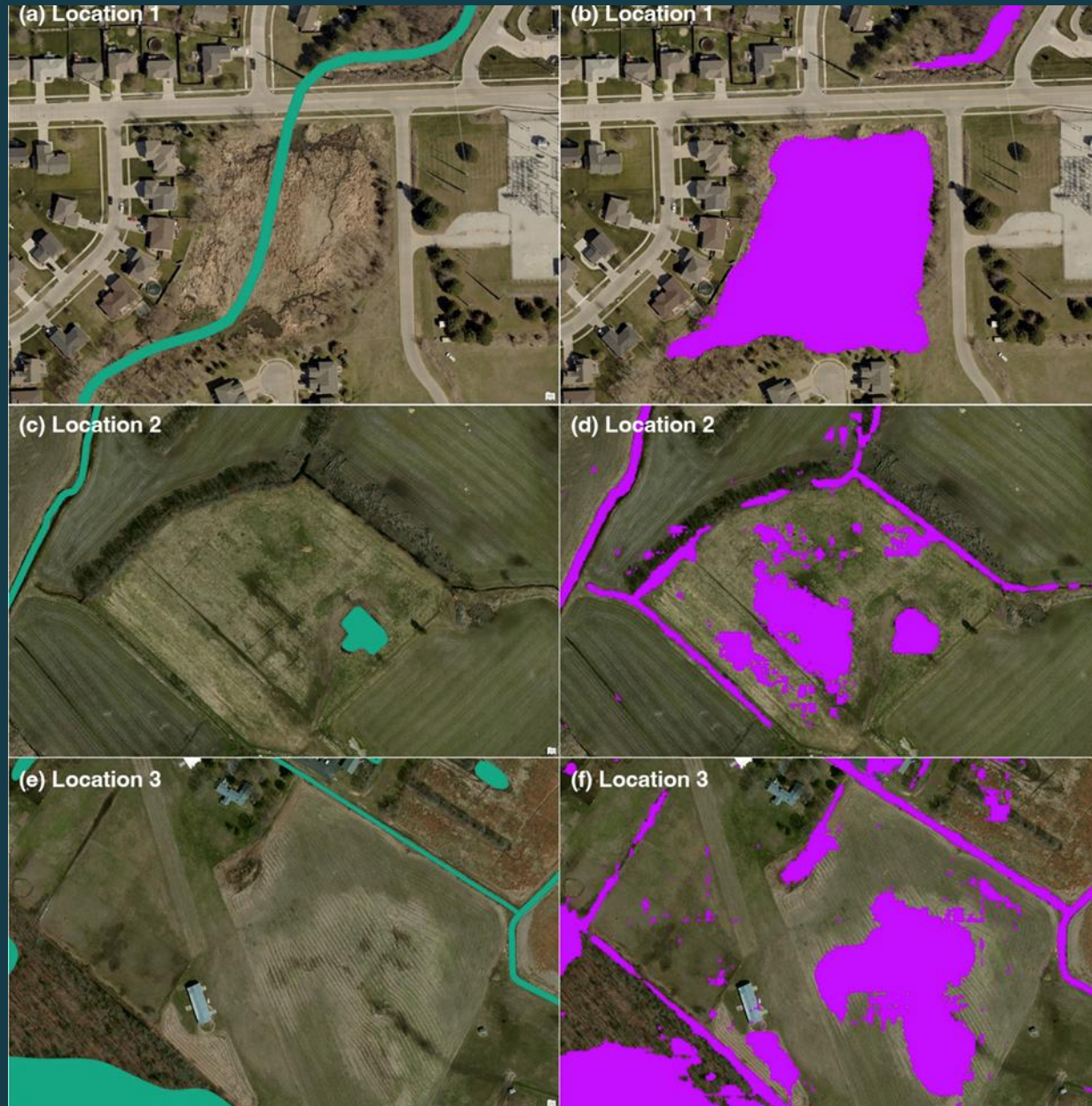


 NWI label

 Full model prediction



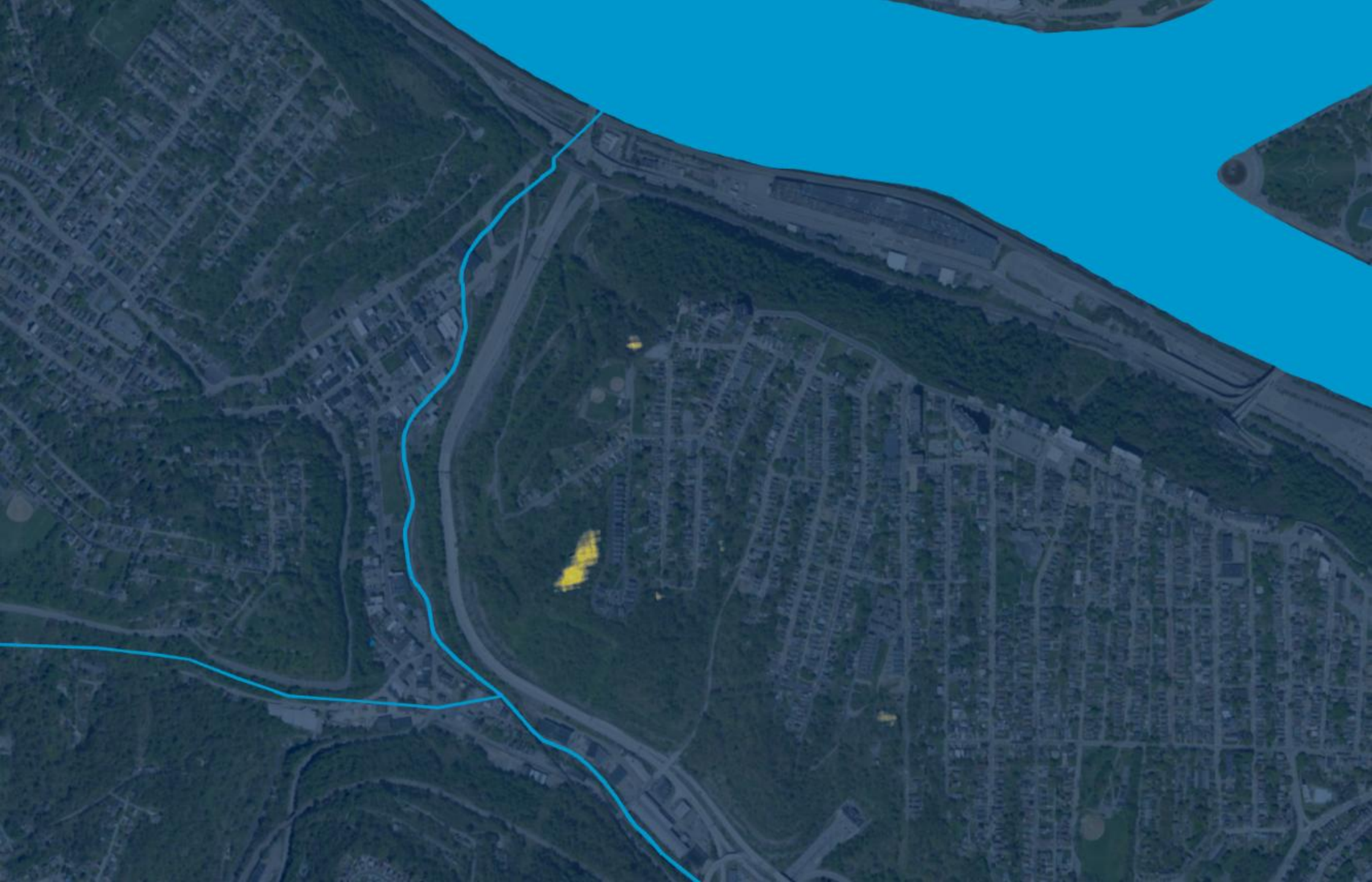
Model Outputs



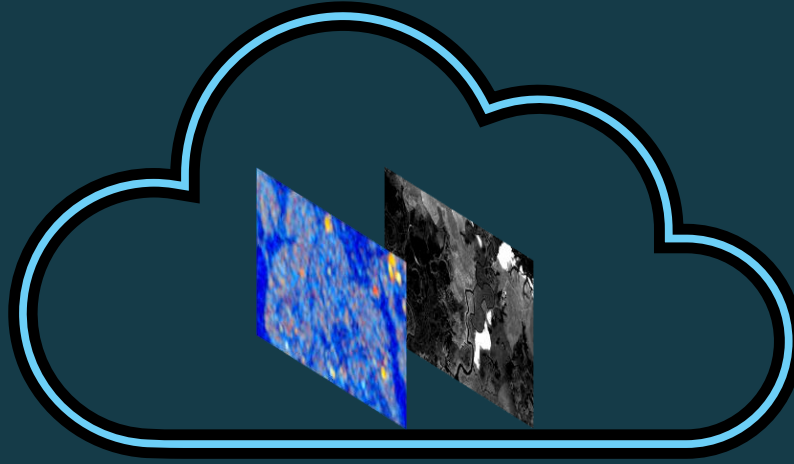
 NWl label

 Full model prediction

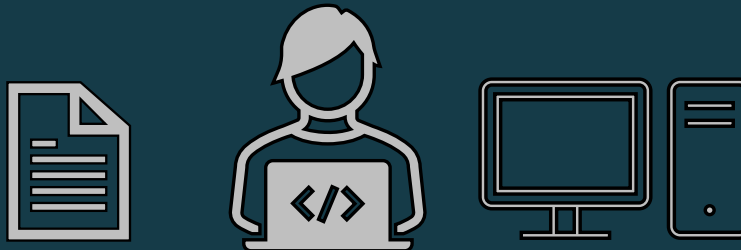




Cloud Computing



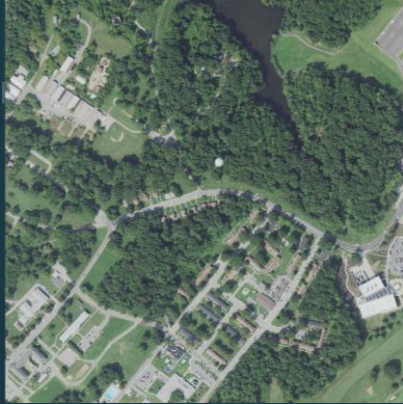
Cloud Computing



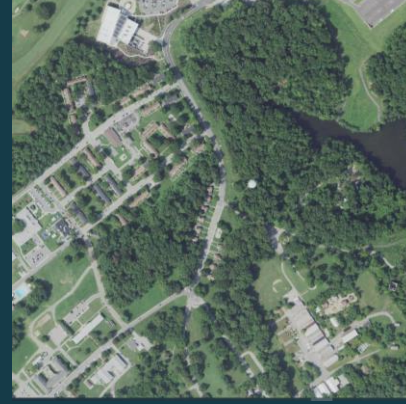
Model Training: Image Augmentation



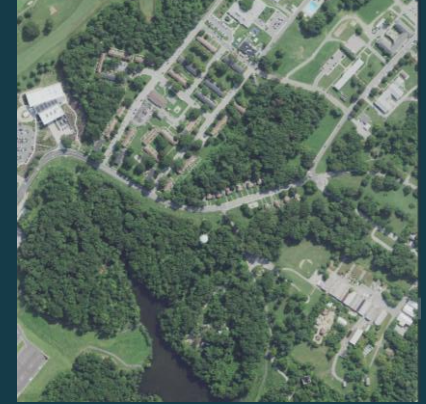
Original



Rotate 90°



Rotate 180°



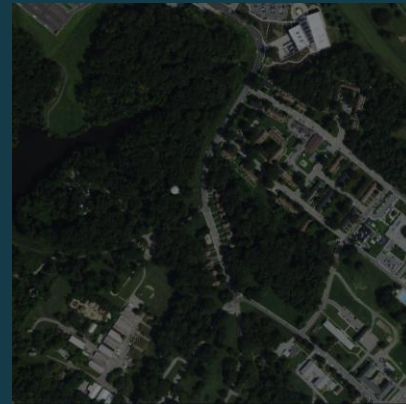
Rotate 270°



Brightness -5%
Contrast -5%



Brightness +5%
Contrast -5%



Brightness -5%
Contrast +5%



Brightness +5%
Contrast +5%

