Perceptions of greenness: Wetland assessments through field measurements, remote sensing, and citizen engagement

University of Louisville Department of Geographic and Environmental Sciences Andrea E. Gaughan

David Brown Kassidy Haynes Sam Hutchins Madison Cicha

University of Louisville Department of Biology Andrew Mehring Mark Tierney

FY2022, FY2023 USGS 104b grant





#### It's Not Easy Being Green



Dept of Geographic and Environmental Sciences







Kassidy Haynes



Madison

Cicha



David Brown



#### Dept of Biology







McKenzie Goodwyn Mark Tierney

Jacob Webb





Cowles



Dr. Andrew Mehring

#### It's Not Easy Being Green



#### **Ecological Component**

**Remote Sensing Application** 

#### Citizen/Community Engagement

K Rese	ucky Water arch Institute	By Developing and Communicating Water-Related Solutions and Tools					
About Us 🗸	Research 🗸 Opportunities 🗸	Watersheds 🗸	Water Utilities 🗸	Resources 🛩	join Our Listserv		
	he Kentucky Water Research Institu ationwide association of 54 federall esource institutes and centers locat niversities throughout the United S erritories. KWRI has over 50 years o esources isources and has established nk between water-related personne atitutions, government agencies, a WRI manages a wide variety of rese rovides technical support to its stal ducates and trains the next generat operts.	y authorized water ed at land grant tates and its of involvement in wat itself as an important ef at academic ad the private sector, arch projects, acholders, and					

## Wetlands

Areas of land that are inundated for at least part of the year, leading to physico-chemical and biological conditions characteristic of shallowly flooded systems (IPCC, 1996b)



#### Wetlands

Areas of land that are inundated for at least part of the year, leading to physico-chemical and biological conditions characteristic of shallowly flooded systems (IPCC, 1996b)

- Wetlands are one of the most ecologically productive types of ecosystems on the planet, producing high amounts of plant material and storing energy.
- Wetlands are also among the Earth's most biodiverse ecosystems, with ~40% of all species living or breeding in wetland landscapes.
- They protect human welfare physically and economically through disaster mitigation.

AL CONTRACT P

#### Nature-based solutions

#### 0.3°C of cooling by 2100

Girardin et al 2021 – Nature

wetland restoration and protection



reforestation and avoided forest conversion

over 1/3 of the cost-effective mitigation needed to limit warming below 2°C

Griscom et al 2017 - PNAS

#### Nature-based solutions

Wetlands can store more carbon than forests

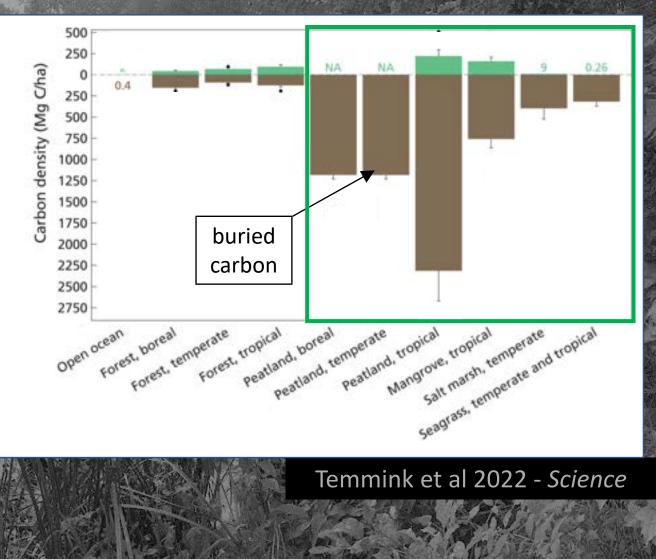
#### FRONTIERS IN ECOLOGY and the ENVIRONMENT

Research Communication 🛛 🖻 Free Access

High carbon burial rates by small ponds in the landscape

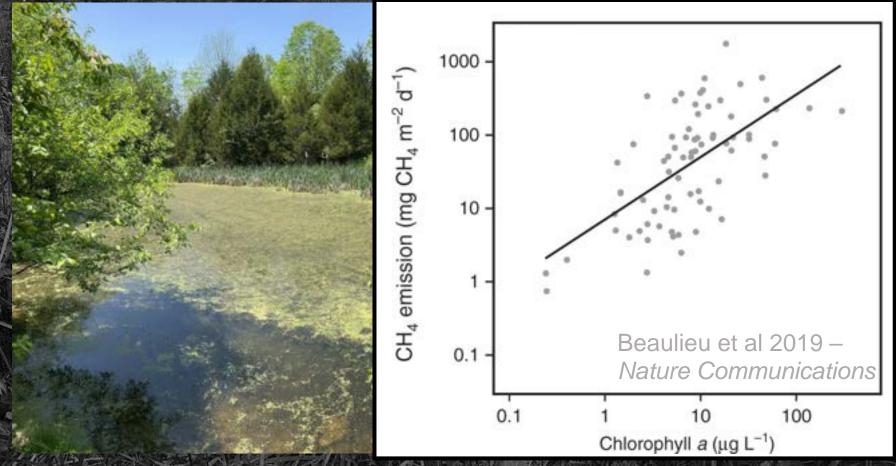
Scott Taylor, Peter J Gilbert, David A Cooke, Michael E Deary, Michael J Jeffries 🔀

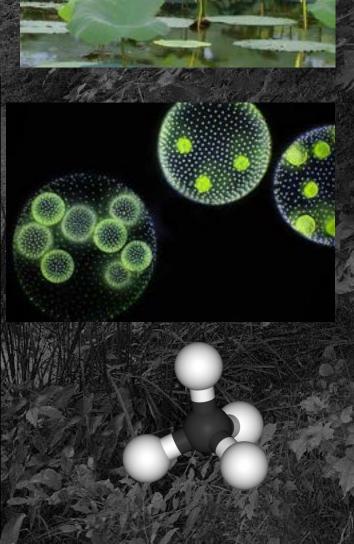




#### Source or sink?

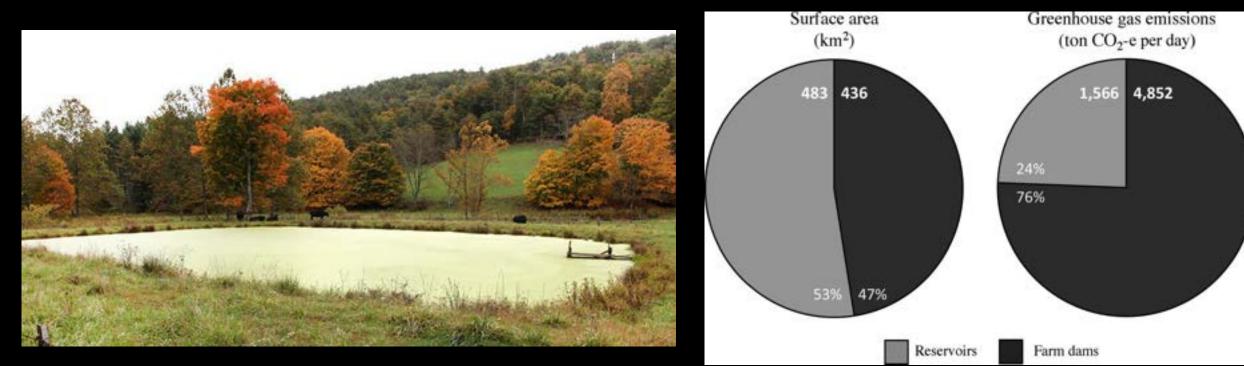
- Most vegetated wetlands are CO<sub>2</sub> sinks
- Plants grow faster than they decompose, bury C in sediments
- Wetlands largest natural emitters of methane (CH<sub>4</sub>) (Kirschke et al 2013 – *Nat. Geoscience*)





### "Little things mean a lot"

- Small ponds (area <1,000 m<sup>2</sup>) make up ~9% of global standing water
- Wetlands impacted by fertilizer runoff emit more greenhouse gas
- May account for 15% and 41% of global standing water CO<sub>2</sub> and CH<sub>4</sub> emissions



Ollivier et al 2019 – Global Change Biology

(Holgerson 2016 *Nature Geoscience*; Smith et al. 2002 *STOTEN*; Renwick et al. 2005 *Geomorphology*; Downing 2010 *Limnetica*)

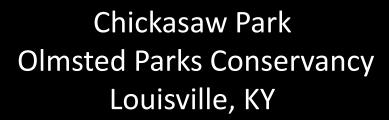
Wetlands - small ponds, disproportionate impacts

Can we improve the ability to rapidly detect healthy ponds and wetlands vs GHG-emissions hotspots?

One Component: How can we leverage remote sensing data to assist us in measuring and monitoring pond dynamics over space and time?

Another Component: How to public perceptions of wetlands match against expert opinion?









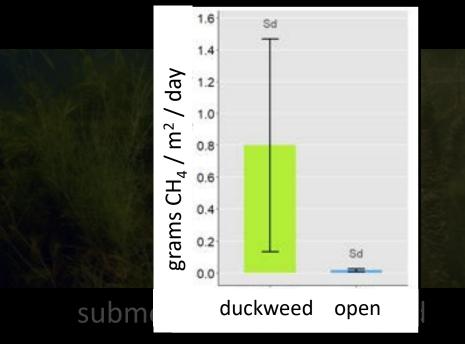


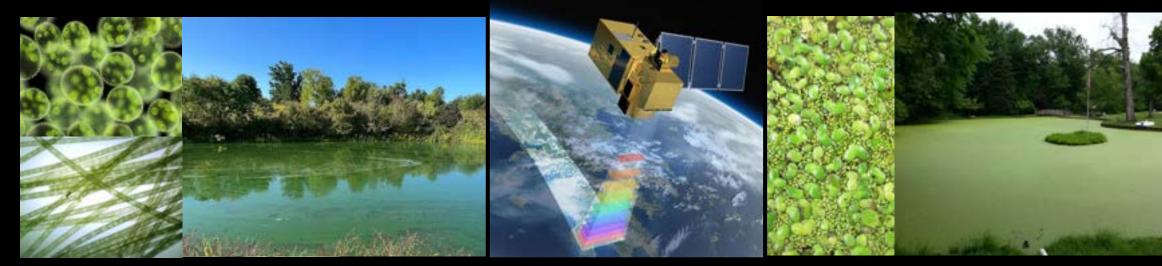
#### Alternate states of "greenness"



emergent veg-dominated

#### Janssen et al 2020 – Freshwater Biology methane emissions

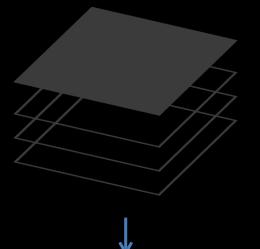


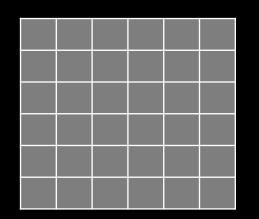


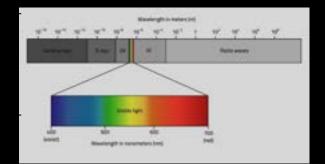
#### algae-dominated

#### floating veg-dominated

#### Remote Sensing Data





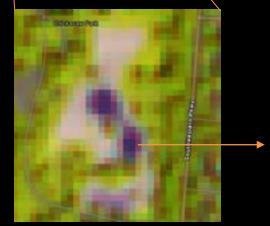


Electromagnetic Spectrum

Columns and Rows

#### **Biophysical/Remote Sensing Protocol**





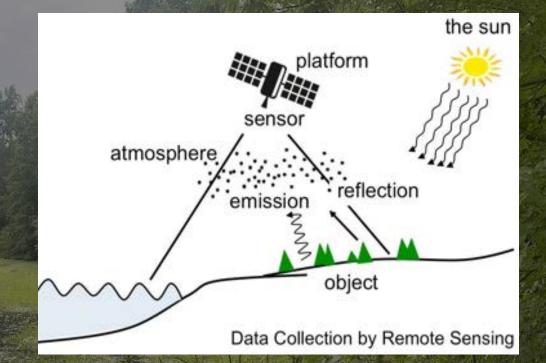
Sentinel 2 January 2023 Louisville, KY

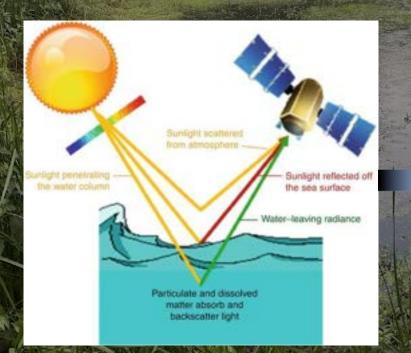
Pixel

#### Water - From Above

#### Reflected Solar Radiation (~color of water)

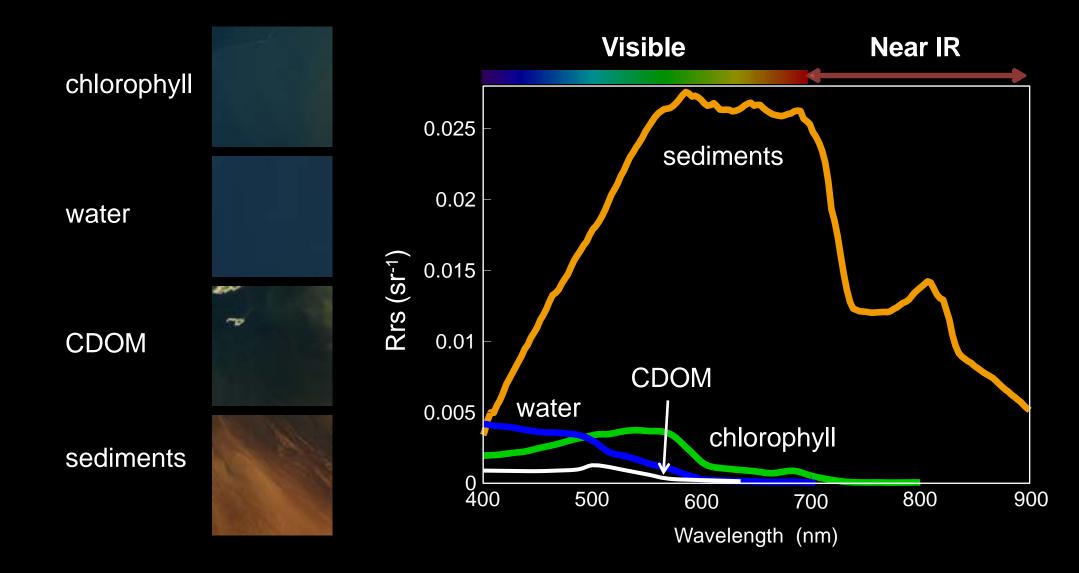
- Measured by satellite sensors
- Used to derive the properties of opticallyactive water constituents

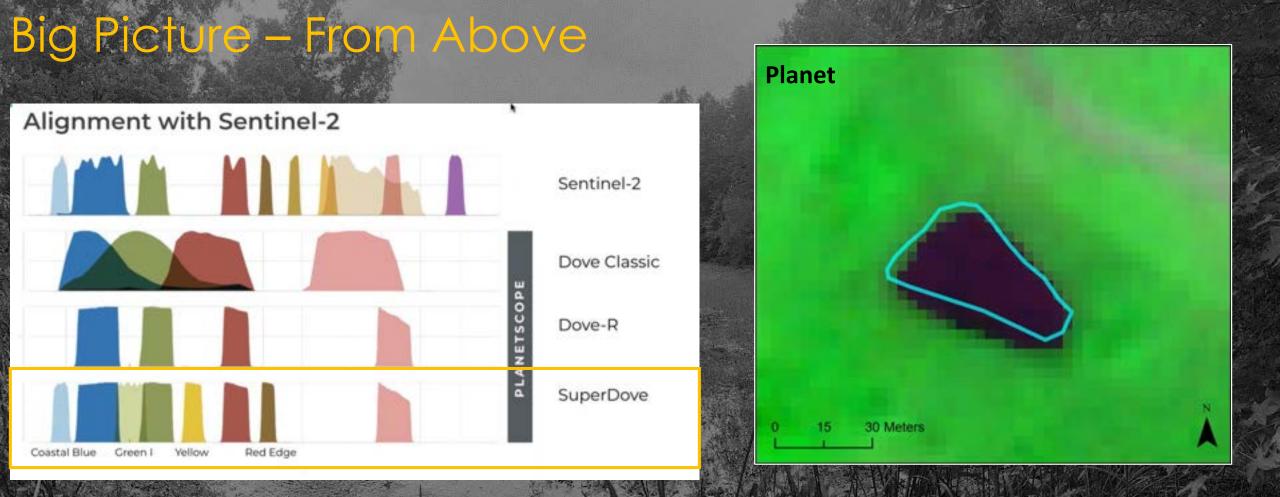




- Suspended Sediments
- Algae
- Colored Dissolved Organic Matter
- Detrital Organic Matter
- Submerged or floating vegetation

# Inherent Optical Properties (IOPs) and the 'Color' of Water





		· · · · · · · · · · · · · · · · · · ·	
Satellite (Sensor)	Spatial Resolution	Temporal Resolution	Spectral Bands
Sentinel 2	10 m, 20 m, 60 m	5 day revisit	13 bands
PlanetScope	3 m	2 times per day	8 bands

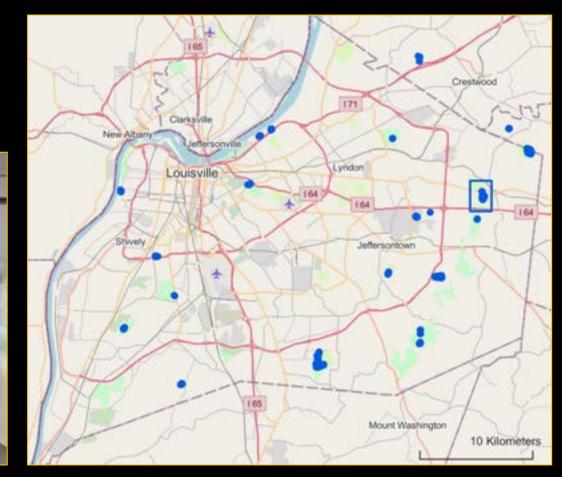
1. 1. 1. S. 1.5.

Pond sample total: n=35

Field Data: Water sample Reference data Turbidity estimate



Created At	Open Water % Cover	Emergent Vegetation 16 Cover	Floating Algae % Cover	General Notes	Photo 1
30th Aug. 2023	0%	25-50%	0%	Load of Typha / cattalis No wat.	had
3081 Aug. 2023	0%	0-25%	0%	Bottle is FP-1, nothing on cap	1
30h Aug. 2023	0%	0%	0%	Duckweed city Bottle is P3-02	-
2011 Aug. 2023	75-100%	0-25%	0%	Emergent veg is a super name.	-
30m Aug. 2023	75-100%	0-25%	0-25%	Duckweed is here but barely	and the second second
30h Aug. 2023	75-100%	0-25%	0-25%	Very little floating algae just re	
30m Aug. 2023	75-100%	0-25%	0-25%	Small Amt of floating algae. Su	
30th Aug. 2023	75-100%	0-25%	0%	Creeping primose willow pres	
<b>e</b> pi <b>c</b>	collect5	0%	0-25%	Algae floating with the wind ne	-



Lemma/Wolffia Cover 75-100%



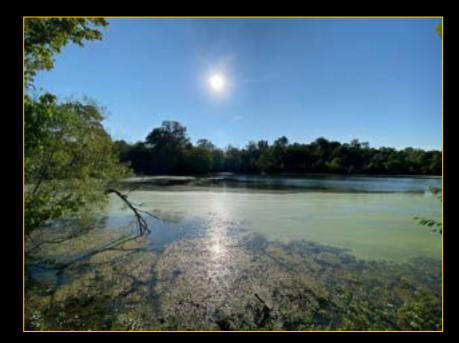






Lemma/Wolffia Cover 25-75%

FP-4 30 Aug 2023



#### FP-3, 30 Aug 2023



Lemma/Wolffia Cover 0-25%

MC-1 30 Aug 2023



CP=1, 30 Aug 2023



Algae

TH-1 30 Aug 2023



IP-1, 30 Aug 2023



Turbidity Estimates

Turbidity might block light so we want to try and control for that in the modeling

Low tech <u>turbidity tube</u>!







- Pond sample total: n=35
- Team divides and conquers
- Field Data: Water sample Turbidity estimate Reference data
- January 2023: n=24
- April 2023: n=34
- August 2023: n= ~35









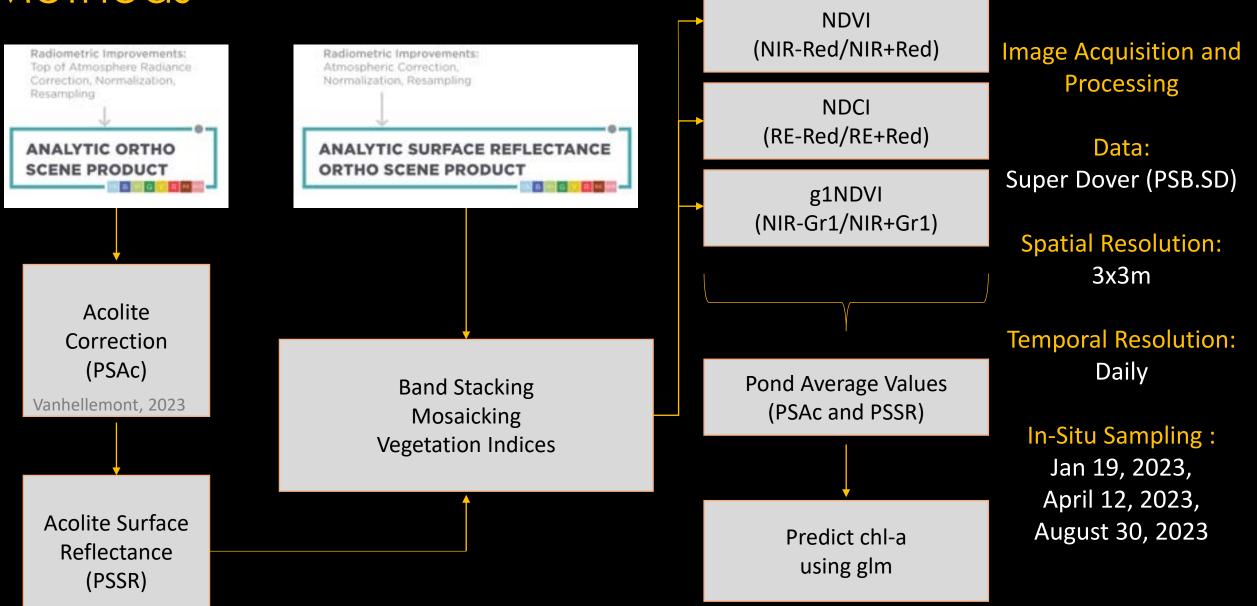








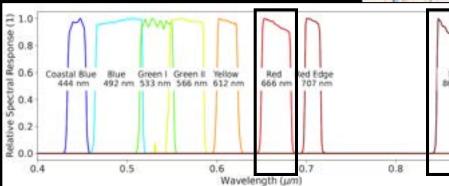




### Study Area

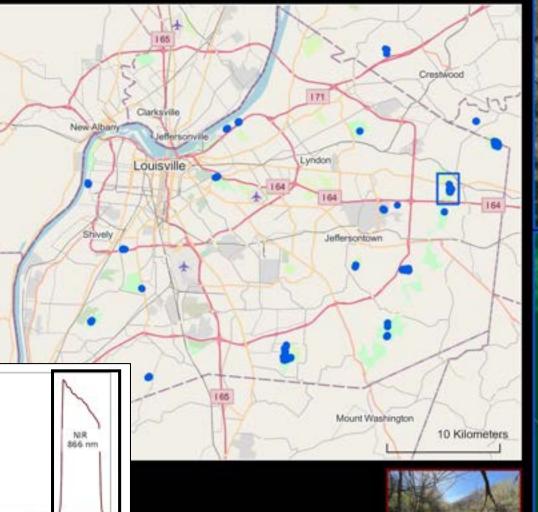
Table 1. SuperDove band names, centre wavelengths and full width at half maximum in nm as derived from the RSR in Fig. 3.

Band Name	Wavelength (nm)	FWHM (nm)		
Coastal Blue	444	20		
Blue	492	52		
Green I	533	37		
Green II	566	38		
Yellow	612	23		
Red	666	32		
Red Edge	707	16		
NIR	866	41		



Louisville, KY, Jefferson County

0.9



The Parklands Beckley Creek Park







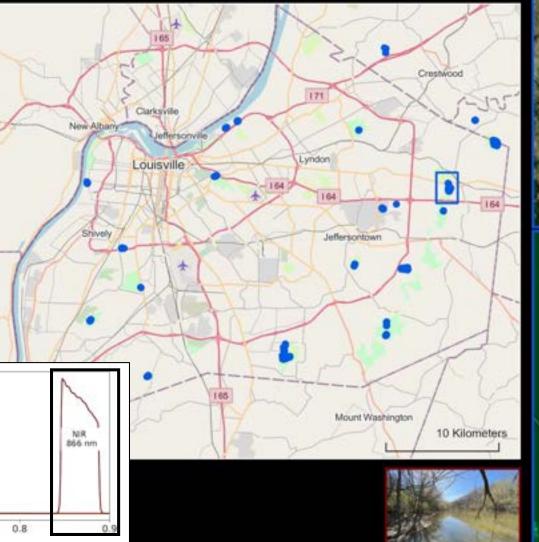
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#### sponse (1) 0.8 2 0.6 Coastal Blue Blue Green I Sreen Il Yellow 444 nm 492 nm 533 nm 566 nm 612 nm Red Red Edge 666 nm 707 nm spectral 1 \$ 0.2 2 0.0 0.6 0.8 0.4 0.7 Wavelength (um)





**The Parklands Beckley Creek Park** 







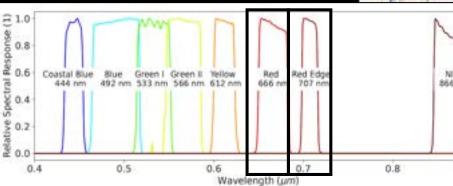


gNDVI (NIR-Gr/NIR+Gr)

### Study Area

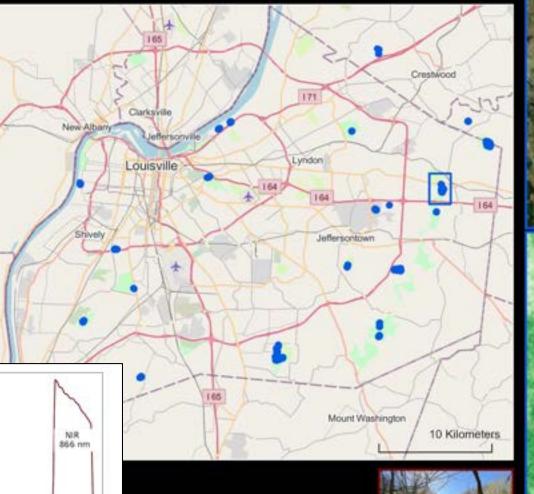
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Louisville, KY, Jefferson County

0.9





CONCEPT

DCI (RE-Red/RE+Red

The Parklands Beckley Creek Park

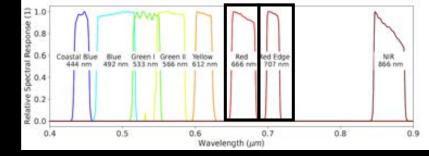


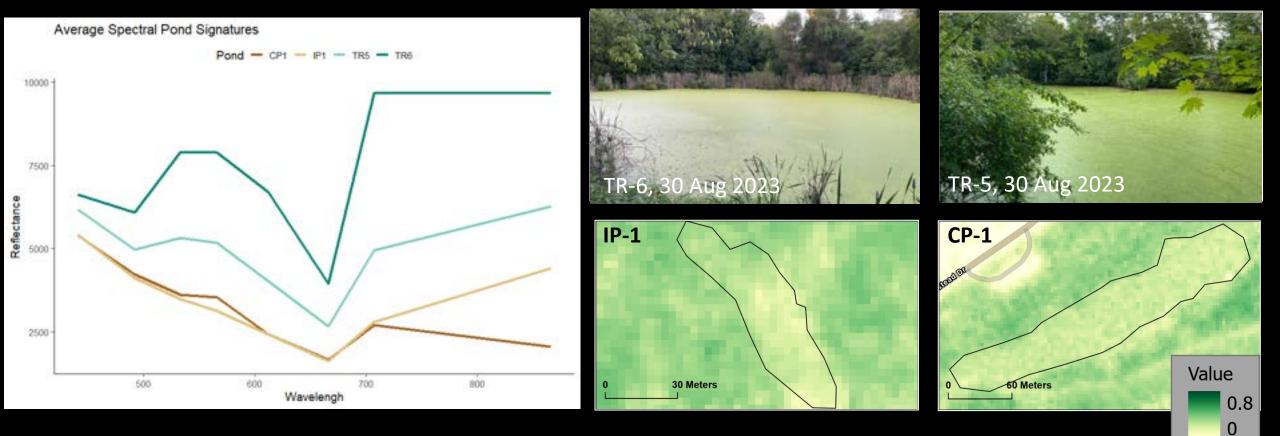




### Results

#### AUG NDCI (RE-Red/RE+Red)





#### Model Results

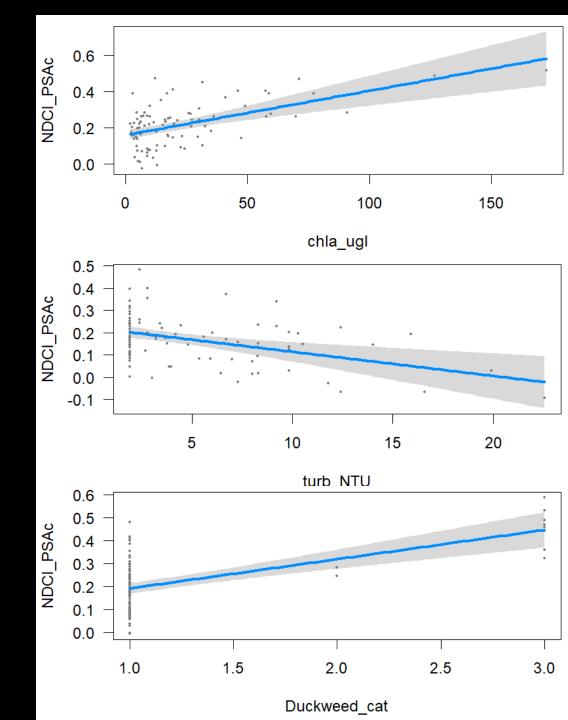
VI	model	AICc	deltaAICc R2		MSE	RMSE
NDCI_PSAc	glm(NDCI_PSAc~chla_ugl+Duckweed_cat+turb_NTU)	-156.0	0 0	0.43	0.101994	0.319364
NDCI_PSSR	glm(NDCI_PSSR~chla_ugl+Duckweed_cat+chla_ugl*turb_NTU)	-142.2	13.8	0.35	0.132209	0.363605
g1NDVI_PSAc	glm(g1NDVI_PSAc~chla_ugl+Duckweed_cat+turb_NTU)	-62.7	93.3	0.24	0.170444	0.412849
NDVI_PSAc	glm(NDVI_PSAc~chla_ugl+Duckweed_cat+turb_NTU)	-36.5	5 119.5	0.28	0.195373	0.44201
NDVI_PSSR	glm(NDVI_PSSR~Duckweed_cat+turb_NTU+Duckweed_cat*turb_NTU)	19.7	175.7	0.26	0.319727	0.565444
g1NDVI_PSSR	glm(g1NDVI_PSSR~chla_ugl+Duckweed_cat+turb_NTU)	77.7	233.7	0.08	0.36457	0.603797
VI	model	AICc	deltaAICc R2		MSE	RMSE
NDCI_PSAc	glm(NDCI_PSAc~chla_ugl+turb_NTU)	-135	5 0	0.27	0.103449	0.321635
NDCI_PSSR	glm(NDCI_PSSR~chla_ugl+turb_NTU+chla_ugl*turb_NTU)	-125.5	5 9.5	0.08	0.132502	L 0.364007
g1NDVI_PSAc	glm(g1NDVI_PSAc~chla_ugl+turb_NTU+chla_ugl*turb_NTU)	-52.6	6 82.4	0.15	0.182439	0.427129
NDVI_PSAc	glm(NDVI_PSAc~chla_ugl+turb_NTU+chla_ugl*turb_NTU)	-27.0	108.0	0.20	0.221722	0.470874
NDVI_PSSR	glm(NDVI_PSSR~chla_ugl+turb_NTU+chla_ugl*turb_NTU)	17.6	5 152.6	0.14	0.323828	0.569059
g1NDVI_PSSR	glm(g1NDVI_PSSR~chla_ugl+turb_NTU+chla_ugl*turb_NTU)	63.9	9 198.9	0.10	0.381732	L 0.617844

#### Model Results

#### Best model

glm(NDCI\_<u>PSAc</u>~chla\_ugl+Duckweed\_cat+turb\_NTU

R-sq of 0.44



### Citizen/Community Engagement

#### Main Objective:

Develop a user perceptions survey of wetland health in urban park systems

#### Goal:

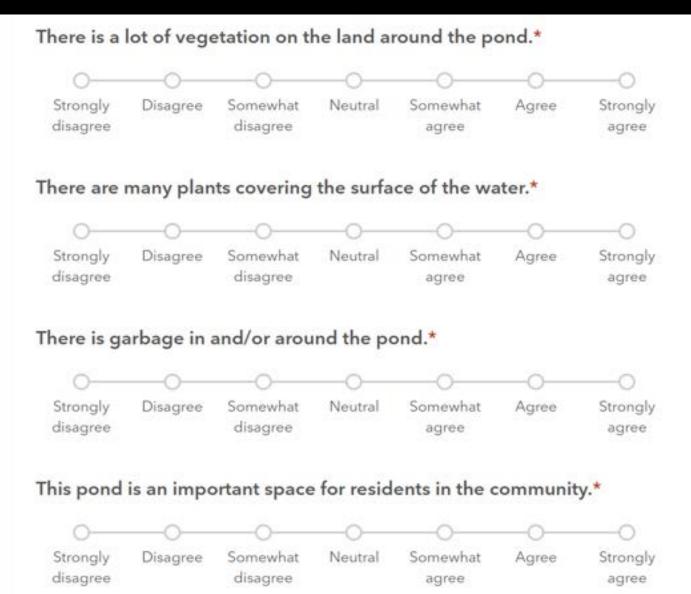
Identify similarities and differences of citizen's knowledge versus expert opinions of wetland health

#### Intended Outcome:

Enhance public involvement and provide a sense of ownership for natural resources in our public parks



### Citizen Engagement – Survey



### Citizen Engagement – Signage









Main Objective: Develop a user perceptions survey to determine similarities and differences of citizen's knowledge versus expert opinions of wetland health in urban park systems





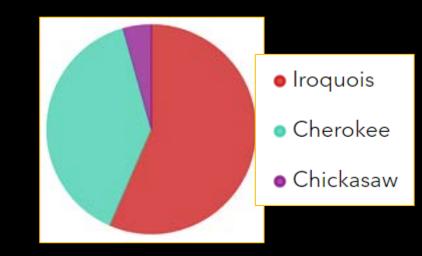
Chickasaw

Cherokee

Iroquois

Main Objective:

Develop a user perceptions survey to determine similarities and differences of citizen's knowledge versus expert opinions of wetland health in urban park systems





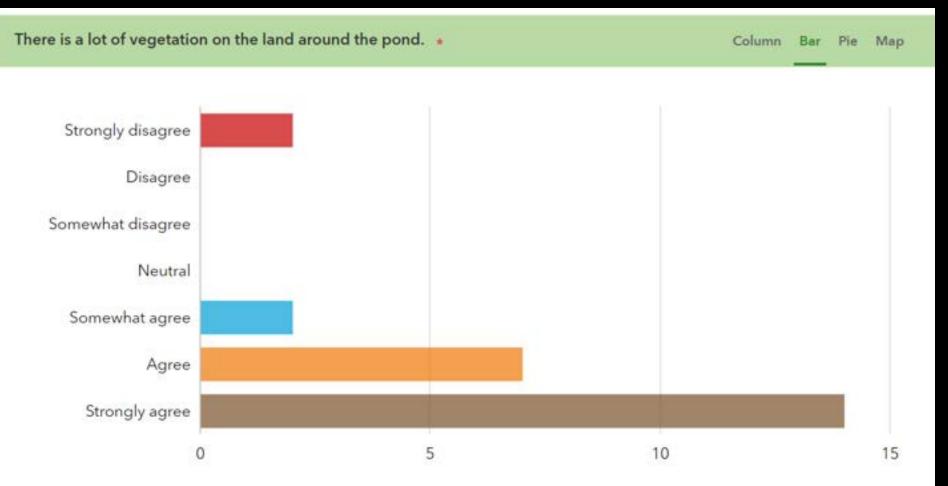


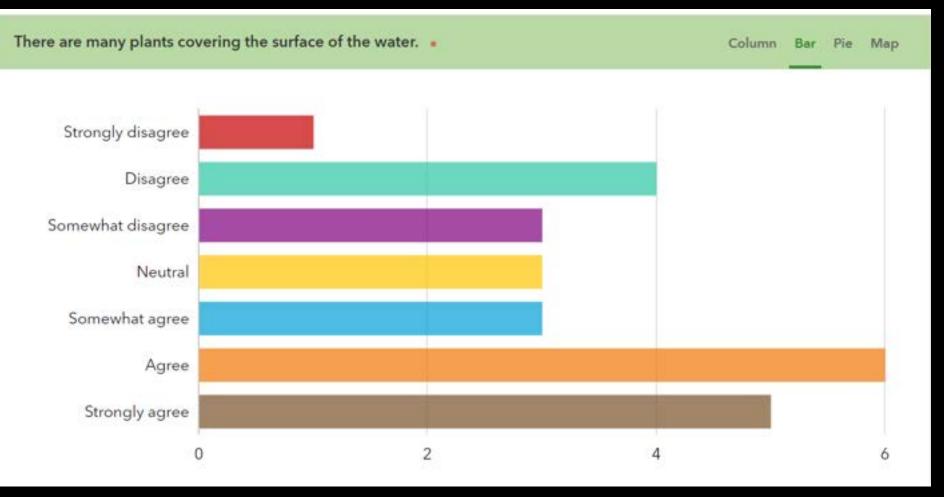


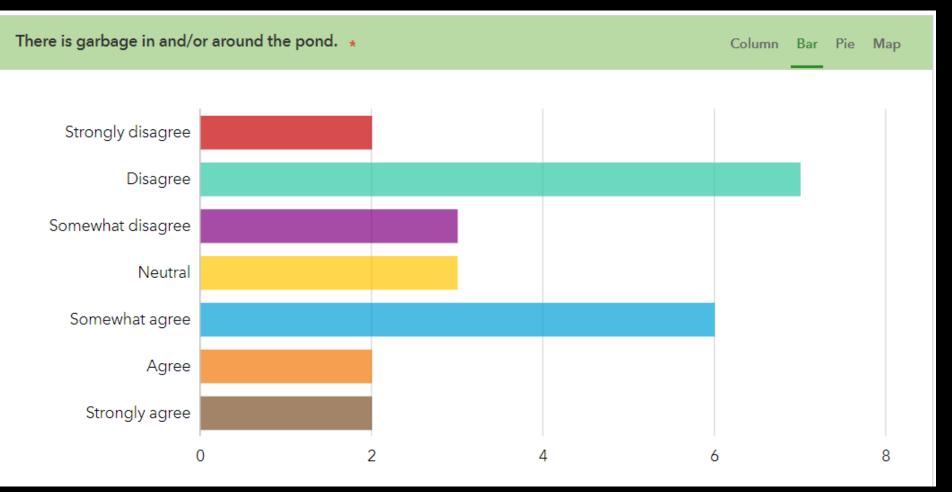
#### Chickasaw

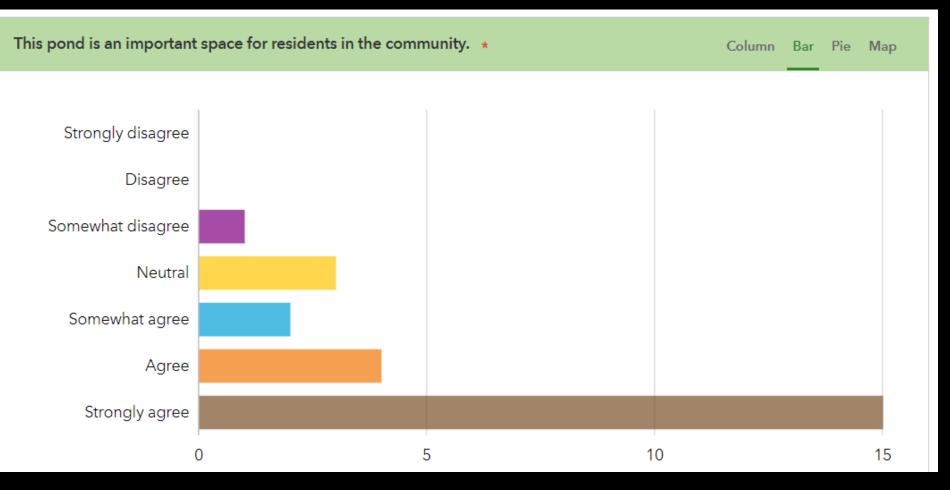
Cherokee

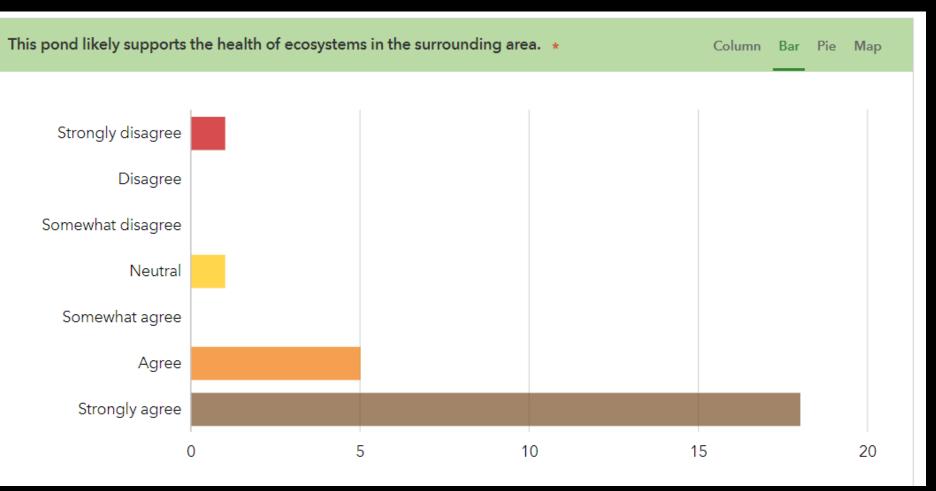
Iroquois

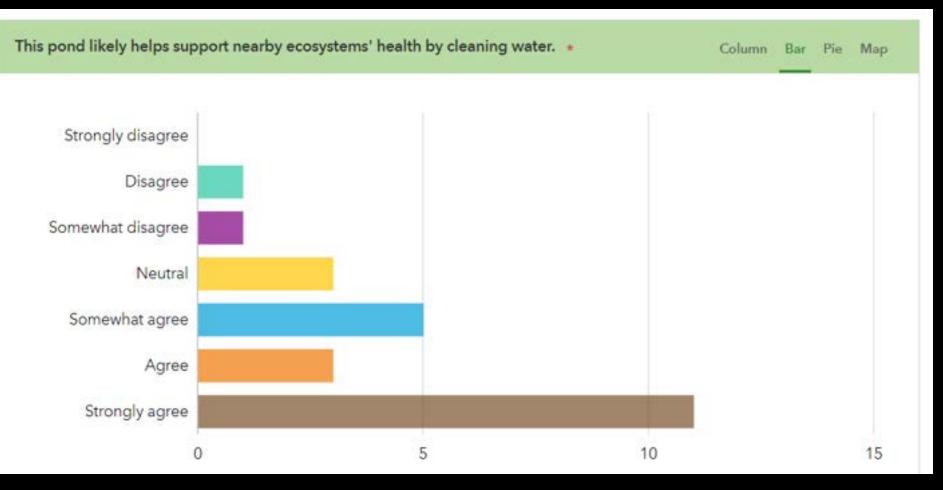


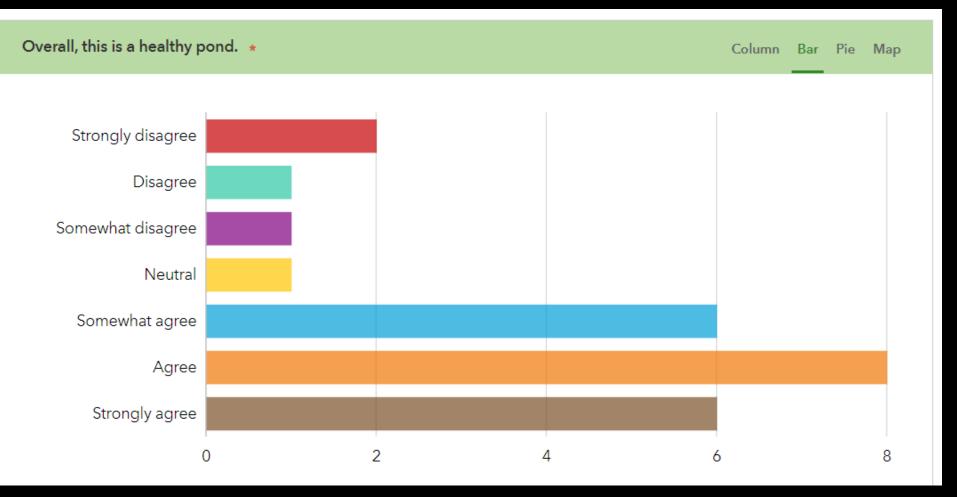


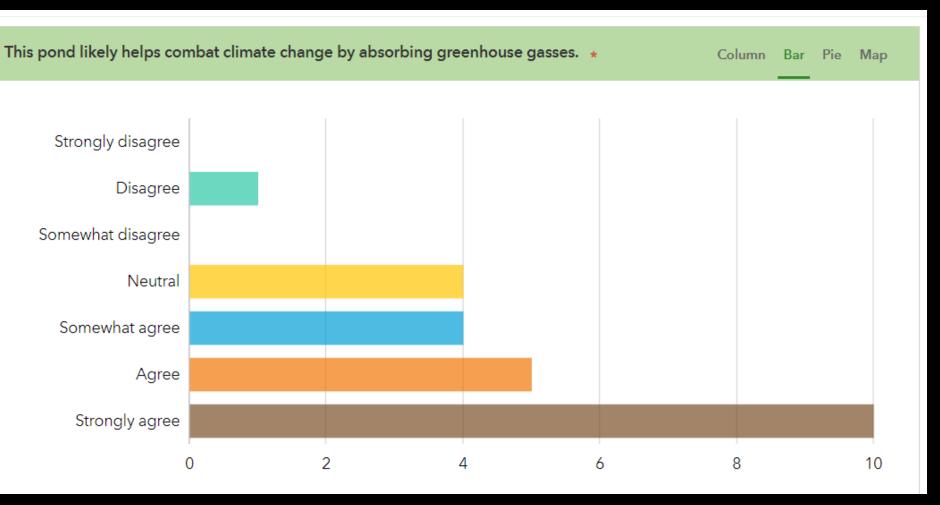


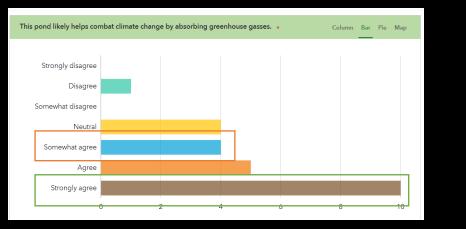














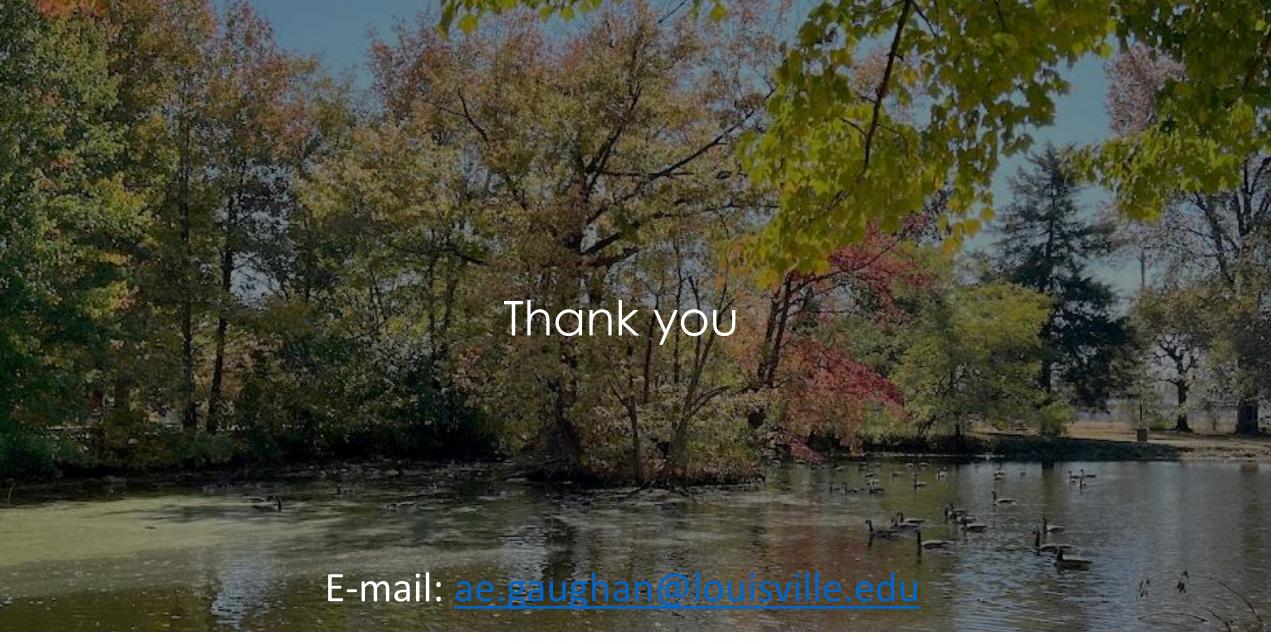
**Expert Opinion** This pond likely helps combat climate change by absorbing greenhouse gases

More likely to be a net sink for carbon



#### Project – Next Steps

- Ecological Monitoring
- Ramp up the number of ponds with signage
- Remote sensing data analysis
- Model assessment and variable inclusion
- Student Research projects



It's not east being green Story Map

# UNIVERSITY OF

FY2022 USGS 104b grant