



#### Part 2: Water Quality of Larger Inland Water Bodies

Instructors: Sherry L. Palacios, PhD & Amita Mehta, PhD Guest Speaker: Daniela Gurlin, PhD, Wisconsin Department of Natural Resources

## **Training Objectives**

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Learn to:

- Understand which data products are used for water quality monitoring
- Follow rigorous practices for obtaining and processing aquatic remote sensing data
- Build skills in image processing for water quality monitoring for coastal and inland water bodies using NASA's SeaDAS image processing software



#### **Training Outline**

June 5 Water Quality in the Coastal Zone

#### June 12

Water Quality of Larger Inland Water Bodies

#### June 19

Aquatic Remote Sensing Skill Development and Best Practices





### **Outline for Part 2**

- Review Part 1
- Water Quality Monitoring in Freshwater Systems
- Sensors & Data Products for Freshwater Systems
- Examples of Freshwater WQ Monitoring Programs
- Guest Speaker:

Integrating Remote Sensing into a Water Quality Monitoring Program Dr. Daniela Gurlin, Wisconsin Department of Natural Resources

 Demonstration of Landsat 8 OLI Image Data Access, Atmospheric Correction, Processing to Level 2 Products





# Review of Part 1

# NASA's Applied Remote Sensing Training Program (ARSET)

#### http://arset.gsfc.nasa.gov/

- Empowering the global community through remote sensing training
- Seeks to increase the use of Earth science in decision-making through training for:
  - policy makers
  - environmental managers
  - other professionals in the public and private sector
- Training topics focus on:
  - air quality land
  - disasters water

Helping Professionals Solve Problems Including...





#### Part 1 Review

## How In Situ and Satellite Observations Roughly Correspond



In Situ	Satellite
Water Temperature	Sea Surface Temperature (SST)
Colored Dissolved Organic Matter (CDOM)	Absorption by CDOM (adg_443_giop)
Suspended Solids – Turbidity	Diffuse attenuation of light at 490 nm (Kd_490)
Water Clarity	Chlorophyll-a, Normalized Fluorescence Line Height (nFLH)
Cyanobacteria	Cyanobacteria Index (CI)
	Euphotic Zone Depth (Z <sub>eu</sub> )
Algal Pigments	Experimental Phytoplankton Functional Type Algorithms



#### Part 1 Review

# **Current Satellite Missions for Water Quality Monitoring**

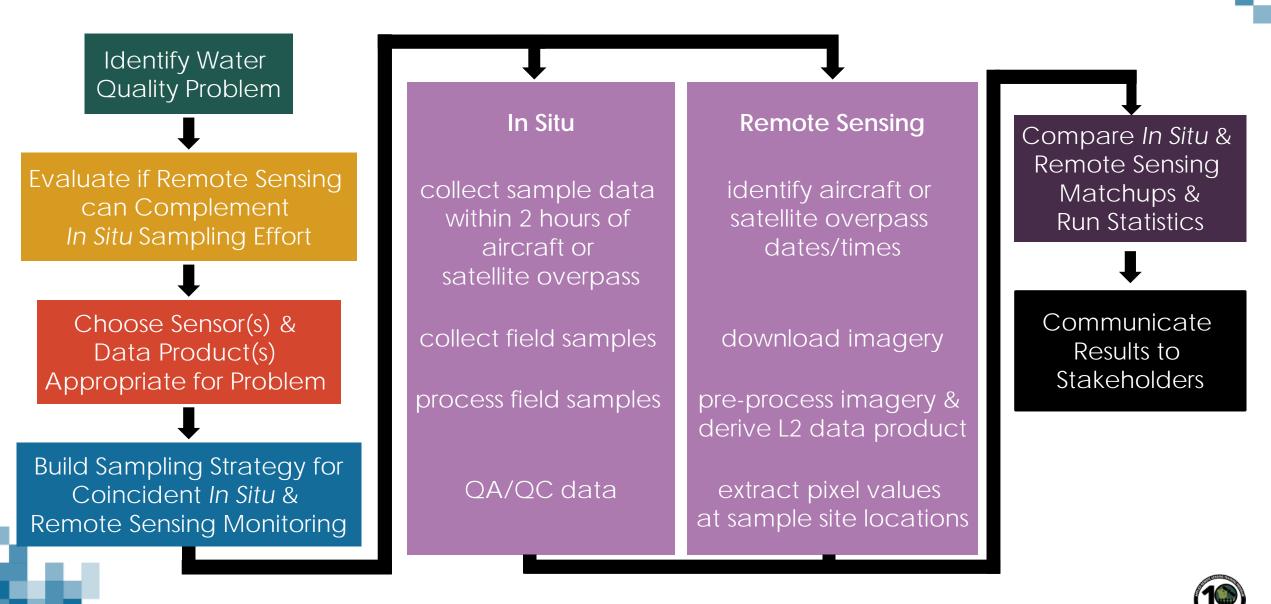
- Landsat 7 (4/15/1999 present)
- Landsat 8 (2/1/2013 present)
- Terra (12/18/1999 present)
- Aqua (5/4/2002 present)
- Suomi National Polar Partnership (SNPP) (11/21/2011 – present)
- Sentinel-2A (6/23/2015 present)
- Sentinel-2B (3/7/2017 present)
- Sentinel-3A (2/16/2016 present)



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#### Part 1 Review

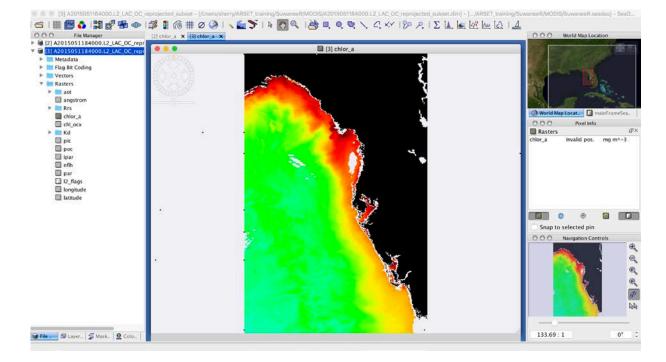
# Water Quality Monitoring Program Workflow



### Part 1 Homework Reminder

#### https://forms.gle/Uw9dTtfktm2iy45Q9

- Complete the exercise from Part 1
- Some of the data saved in the Part 1 exercise will be used for Part 3, be sure to process data for chlor\_a, sst, and adg\_443\_giop and to crop all images to the same geographic coordinates
- Your answers to Part 1 Homework are due on June 21<sup>st</sup>





# Water Quality in Freshwater Systems

# What are Some Goals of Monitoring in Freshwater Systems?

To Monitor for...

- cyanobacteria
- pathogens
- man-made pollutants
- nutrient inputs
- water clarity

Why? Impacts to...

- drinking water
- domestic animals
- wildlife
- ecosystems

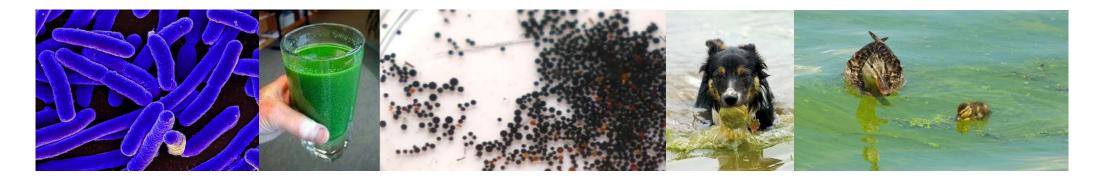


Image Credits: E. coli, cyanobacteria, plastic microbeads, domestic animals, wildlife

# What Are Typical *In Situ* Observations for Monitoring in Freshwater Systems?

- chlorophyll concentration
- temperature
- water clarity
- nutrients
- metals
- pH & alkalinity
- dissolved organic matter

- phytoplankton taxonomy
- cyanobacteria
- condition of indicator species
- suspended sediments
- E. coli
- plastics

# Which Remote Sensing Data Products Are Relevant for Freshwater Systems?

- chlorophyll
- water surface temperature
- absorption of light by CDOM
- diffuse attenuation coefficient
- water clarity
- cyanobacterial index
- light absorption and scattering at wavelengths diagnostic of particular algal taxa
- user-defined, custom algorithms developed for specific use and particular region



# Which Sensors are Best for Monitoring in Freshwater Systems?

- Considerations depend on the problem
- Sensor choice is based on:
  - spatial resolution
  - temporal resolution
  - spectral resolution and data products





# Why Are Cyanobacteria Such a Problem?

- Toxin producers
- Excessive biomass
- Thrive in warm waters, so blooms will likely increase with warming climate

Genus	Colony or Filament	Surface Scum?	Toxin(s)	
Microcystis	Colony	yes	Microcystin, Anatoxins	
Aphanizomenon	Filament	yes	Cylindrospermopsin, Anatoxins, Saxitoxins	
Anabaena	Filament	yes	Microcystin, Cylindrospermopsin, Anatoxins, Saxitoxins	
Planktothrix (Oscillatoria)	Filament	no	Microcystin, Anatoxins, Saxitoxins	
Cylindrospermopsis	Filament	no	Cylindrospermopsin, Anatoxins	
Lyngbya	Filament	no	Cylindrospermopsin, Saxitoxins	

Credit: <u>R. Stumpf</u>, <u>EPA</u>



## Cyanobacteria and Satellite Remote Sensing

- Satellites cannot detect toxins
- Taxa that form surface scums are readily detected in most satellite imagery
- Spectral features at particular wavelengths can be diagnostic of cyanobacteria
  - phycocyanin absorption at 620 nm
  - chlorophyll-a absorption at 667 nm
  - backscattering at 709 and 779 nm
- Variations in spectral signature can be used to distinguish cyanobacteria
  - Cyanobacteria Index (CI) (Wynne et al. 2008)
  - Maximum Chlorophyll Index (MCI) (Gower et al. 2008)
  - Maximum Peak Height (MPH) (Matthews & Odermatt 2016)

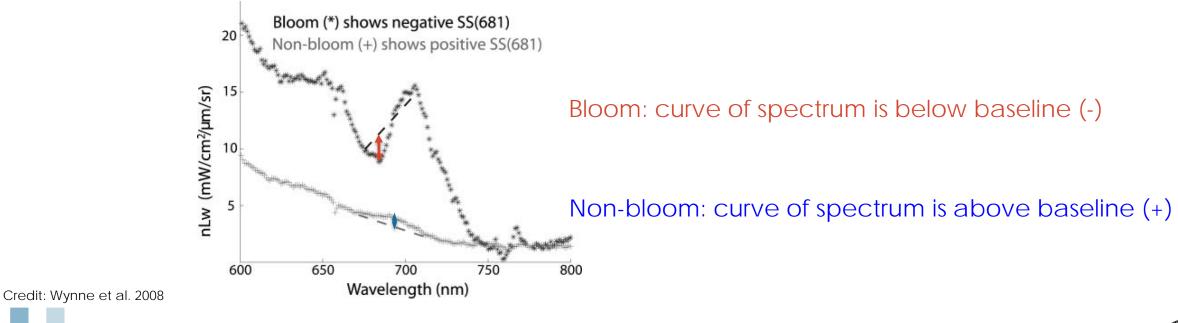
Work fine without atmospheric correction



Credit: R. Stumpf, EPA

# Cyanobacteria Index (CI)

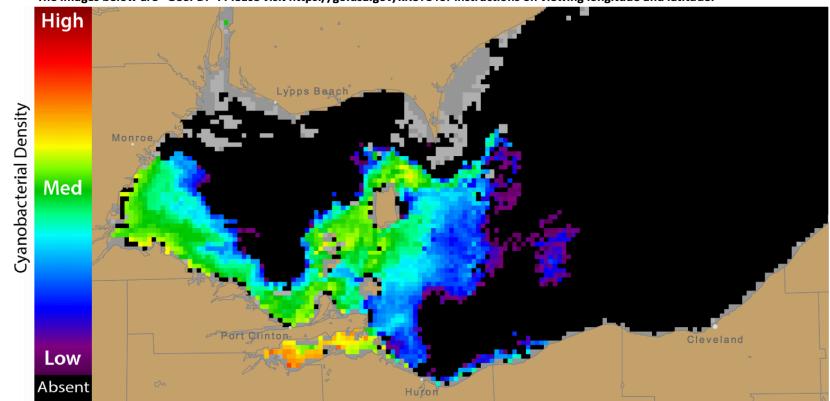
- Does not require surface scum, but also works with scum
- Less sensitive to sediment and water vapor in atmosphere
- CI equates to cell concentration
- CI is equivalent to the spectral shape at 681 nm





### CI Gives an Estimate of Cyanobacteria Cell Concentration

https://tidesandcurrents.noaa.gov/hab/lakeerie\_bulletins/HAB20180813\_2018016\_LE.pdf



The images below are "GeoPDF". Please visit https://go.usa.gov/xReTC for instructions on viewing longitude and latitude.

Figure 1. Cyanobacterial Index from NASA MODIS-Aqua data collected 11 August, 2018 at 14:21 EST. Grey indicates clouds or missing data. The estimated threshold for cyanobacteria detection is 20,000 cells/mL.

Credit: NOAA Lake Erie Harmful Algal Bloom Bulletin, 13 Aug 2018

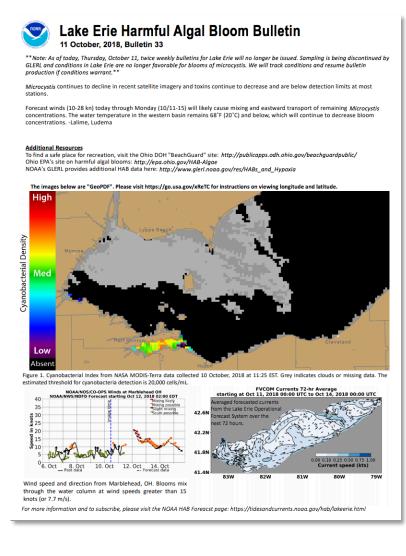


Cyanobacterial Harmful Algal Bloom Monitoring Programs

### **NOAA Operational Great Lakes HAB Bulletin**

#### https://www.glerl.noaa.gov/res/HABs\_and\_Hypoxia/bulletin.html

- Operational Harmful Algal Bloom
   Bulletin
- Issued twice weekly when conditions warrant – typically in the summer and early autumn
- Reports Cyanobacterial Index (CI)





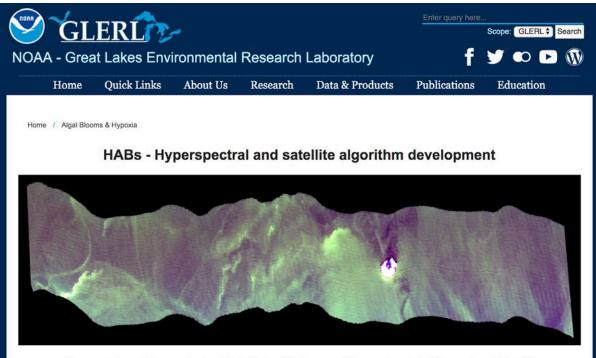
#### NOAA Experimental Lake Erie HAB Tracker

https://www.glerl.noaa.gov/res/HABs\_and\_Hypoxia/habTracker.html



# **NOAA Great Lakes Hyperspectral Monitoring**

#### https://www.glerl.noaa.gov/res/HABs\_and\_Hypoxia/airSatelliteMon.html

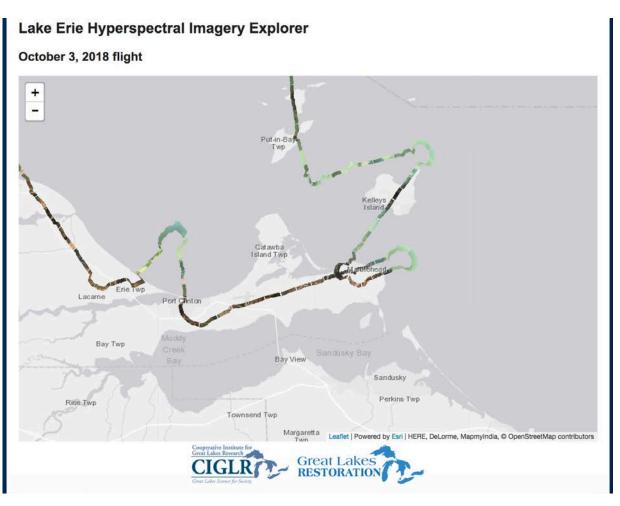


Hyperpectral scan taken nearby the Toledo Harbor Lighthouse and Maumee Bay, Lake Erie, on August 31, 2015

#### View true-color hyperspectral imagery

The links below, organized by year, will take you to a map interface to view georeferenced, true-color hyperspectral image data from the respective date. Given the total size of the data can range up to and beyond 1 GB of data, we recommend closing other browser tabs and windows to better guarantee viewing of the data.

Imagery in the maps linked below that appear a pink-like color are a product of sun glint.





### Harmful Algal Blooms Analysis Tool

#### https://cchab.sfei.org/

- Screening level analysis tool to prompt field verification and sampling to confirm suspected cyanoHAB
- Browser provides map view of blooms to show spatial extent of bloom and time series at particular pixel location
- Region includes US States California and parts of Oregon and Nevada

# Harmful Algal Blooms Analysis Tool Purpose Purpose Disclaimer Image: SFEI AQUATIC CENTER

This project is part of My Water Quality Portal

The satellite imagery analysis tool provides a screening level analysis to prompt field verification and sampling to confirm the status of a suspected cyanobacteria harmful algal bloom and presence of toxic species. This map displays estimated levels of cyanobacteria in large water bodies, calculated from satellite imagery in order to better understand potential risks to public health. Data is displayed in map form to show the spatial extent of blooms and is also viewable in long and short timelines to show how concentrations vary over time. Additionally, field data can be displayed providing users a combination of data and tools to better understand the status and trends of cyanobacteria harmful algal blooms and the potential risks to public health.

The tool features a mix of data from discrete samples, which are very precise for a specific location, and satellite imagery data, which can provide a broad-scale understanding of cyanobacteria density but should be regarded as provisional in nature. The tool offers water resource managers and the public an opportunity to review and compare satellite data with sampling data (where available) in an exploratory interface. As the satellite imagery improves over time, the data quality will likely also increase, as will its value for decision-making.

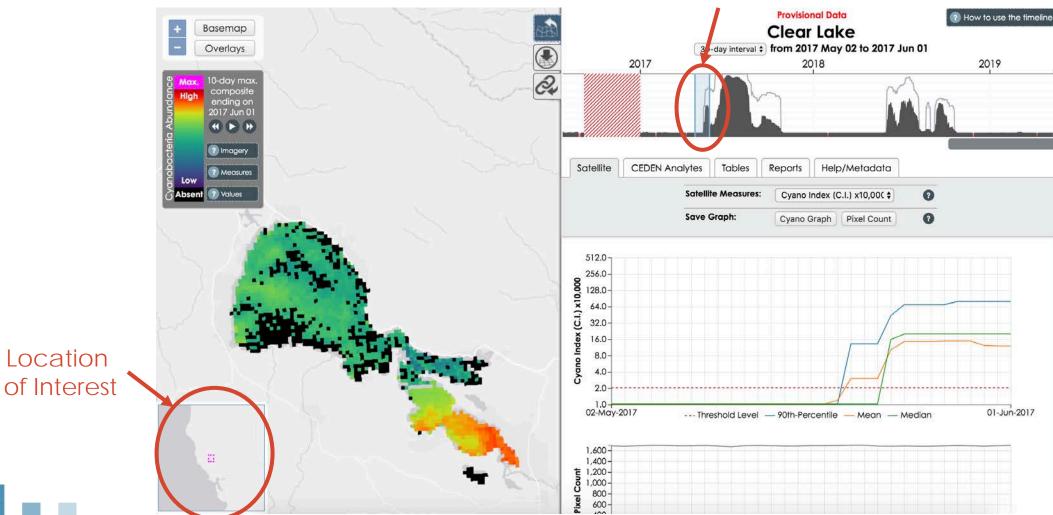


© San Francisco Estuary Institute



## Harmful Algal Blooms Analysis Tool

#### https://cchab.sfei.org/



#### Time Period of Interest



Considerations When Using Remote Sensing for Freshwater Water Quality Monitoring

#### Advantages of Remote Sensing for Freshwater Systems

- Longtime imagery record for time series analysis
- Ongoing commitment from space agencies to continue data collection
- Reliable data for operational early warning and forecasting systems
- Some sensors have spatial resolution appropriate for lakes
- Imagery is typically freely available and of high quality

#### **Disadvantages of Remote Sensing for Freshwater Systems**

- Shallow water interference from the bottom
- Water bodies too small for the spatial resolution of sensors
- Limited number of standard algorithms for these optically complex waters
- Atmospheric correction
- Highly variable systems
- Ground truthing is costly



# Dr. Daniela Gurlin Wisconsin Department of Natural Resources





# Integrating Remote Sensing into a Water Quality Monitoring Program

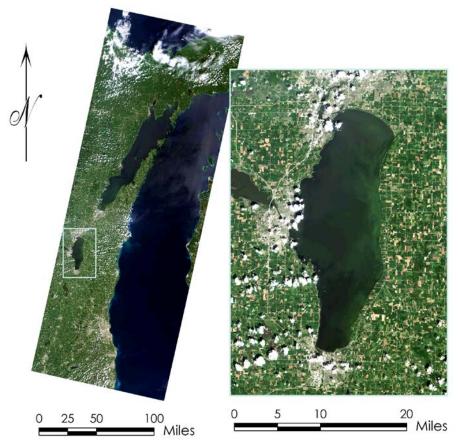
Daniela Gurlin (Wisconsin Department of Natural Resources)

12 June 2019

#### Outline

- Remote Sensing Challenges
- Earth Observation Sensors
- Remote Sensing Activities
- Satellite Water Clarity Monitoring
- Data Use for Integrated Reporting
- Data Dissemination
- Remote Sensing Research Projects

#### Algal bloom in Lake Winnebago



Algal bloom in eastern Lake Winnebago as seen by Landsat 8 OLI on 07/26/2016 (Source of Landsat 8 OLI data: USGS).



#### **Remote Sensing Challenges**

# 27

#### Advantages

- Data with a high spatial and temporal resolution
- Evaluation of environmental problems and potential health risks
- Historical data for studies of trends in water quality
- Data for integration into early warning systems

#### Disadvantages

- Optically complex conditions in lakes and reservoirs
- Interference from the lake bottom
- Dynamic water quality changes
- Limited number of water quality parameters
- Collection of ground-truth data required



	Landsat 7	Landsat 8	Sentinel-2A	Sentinel-2B	Sentinel-3A	Sentinel-3B	
Satellite Sensor System	ETM+	OLI/TIRS	MSI	MSI	OLCI	OLCI	
Spatial Resolution (m)	15, 30, 60	15, 30, 100	10, 20, 60	10, 20, 60	300	300	
Spectral Bands	8	11	13	13	21	21	
Revisit Cycle (days)	16	16	5	5	2	2	
Swath Width (km)	183	183	290	290	1270	1270	
Launch Date	April 1999	February 2013	June 2015	March 2017	February 2016	April 2018	
Years in orbit/minimum design life (years)	20/5	6/5	3/7	2/7	3/7	1/7	
Example Data Sources	U.S. Geological Survey EarthExplorer ( <u>https://earthexplorer.usgs.gov/</u> )			Copernicus Open Access Hub ( <u>https://scihub.copernicus.eu/</u> )			

#### Satellite Sensor Abbreviations

Landsat 7: Enhanced Thematic Mapper Plus (ETM+)

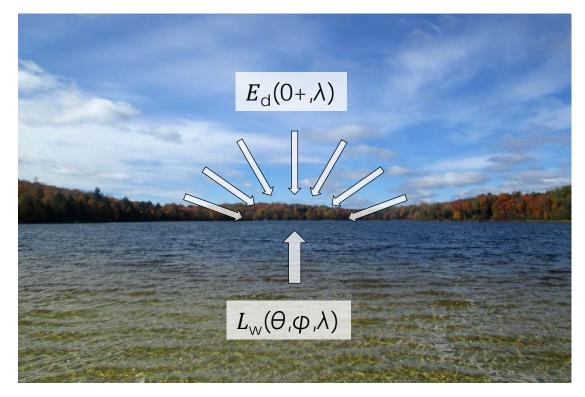
Landsat 8: Operational Land Imager (OLI) and Thermal Infrared Sensor (TIRS)

Sentinel-2: Multispectral Instrument (MSI)

Sentinel-3: Ocean and Land Color Instrument (OLCI)



#### What is remote sensing reflectance?



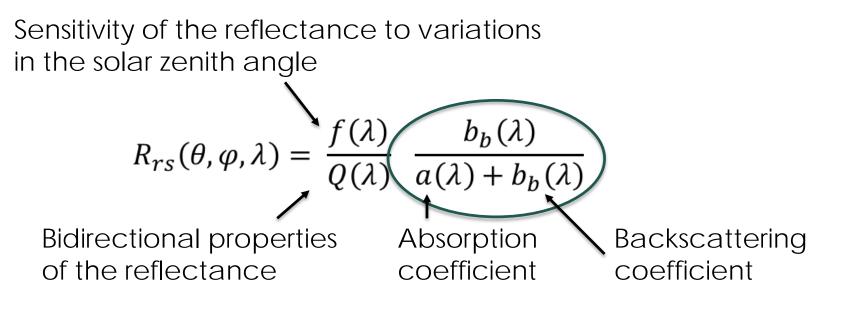
$$R_{\rm rs}(\Theta, \varphi, \lambda) = \frac{L_{\rm W}(\Theta, \varphi, \lambda)}{E_{\rm d}(0^+, \lambda)}$$

- $R_{rs}$  ( $\theta$ , $\phi$ , $\lambda$ ): remote sensing reflectance
- $L_w(\Theta, \phi, \lambda)$ : water leaving radiance
- $E_d(0+,\lambda)$ : downwelling irradiance
- $\theta$ : solar zenith angle
- φ: solar azimuth angle
- $\lambda$ : wavelength

Calculation of the remote sensing reflectance of waterbodies. This equation relates the ratio of the water leaving radiance and the downwelling irradiance ( $L_w(\Theta, \varphi, \lambda)$ ) and  $E_d(0^+, \lambda)$ ) to the remote sensing reflectance ( $R_{rs}(\Theta, \varphi, \lambda)$ ).



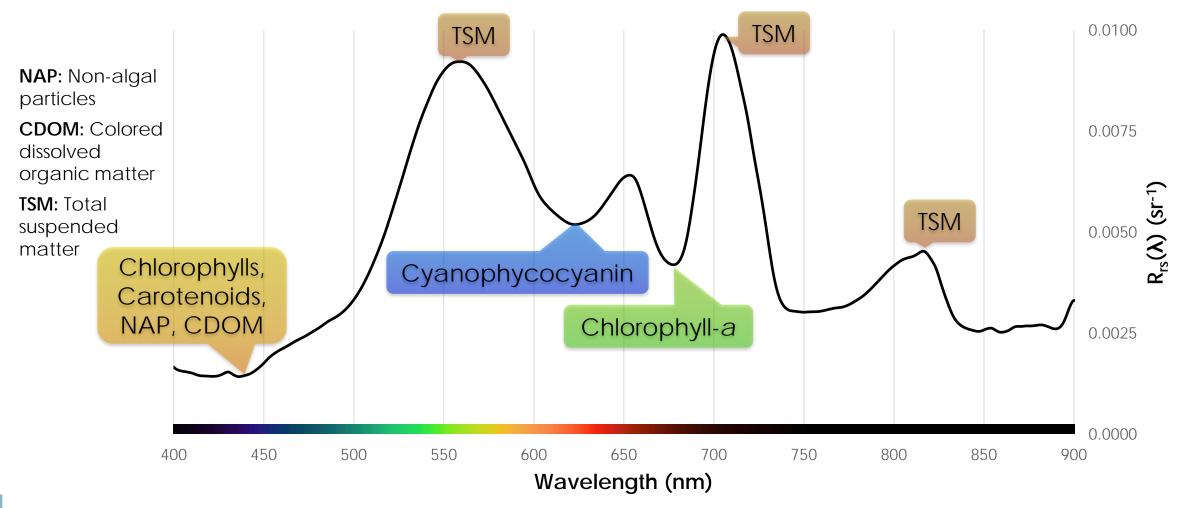
What is remote sensing reflectance?



 $a(\lambda) = a_{\varphi}(\lambda) + a_{NAP}(\lambda) + a_{CDOM}(\lambda) + a_{w}(\lambda)$ 

φ: Phytoplankton, NAP: Non-algal particles, CDOM: Colored dissolved organic matter, w: water

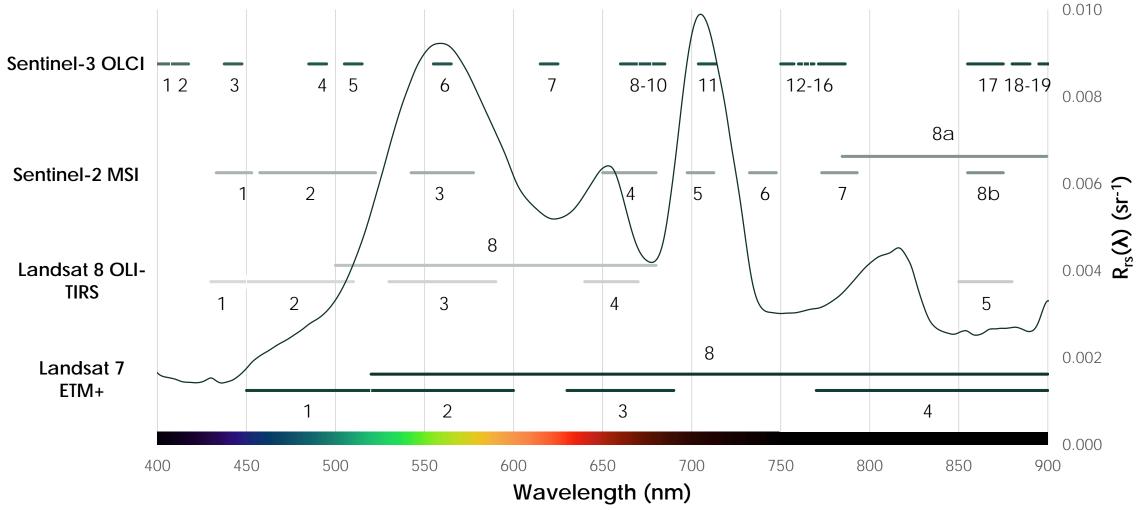




Remote Sensing Reflectance Spectrum for Lake Winnebago, Acquired 09/21/2015



### **Earth Observation Sensors**



Remote Sensing Reflectance Spectrum for Lake Winnebago, Acquired 09/21/2015

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# **Remote Sensing Activities**

- Systematic retrieval of the water clarity from Landsat 7 ETM+ and Landsat 8 OLI-TIRS data
- Studies of the major drivers of lake water clarity
- Increase in Earth observation monitoring capabilities

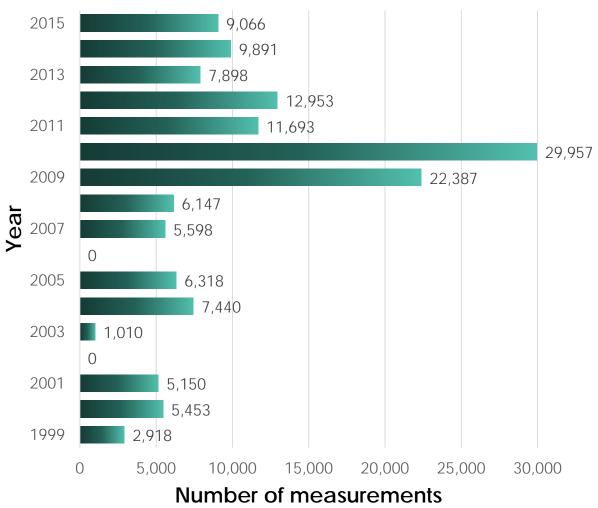
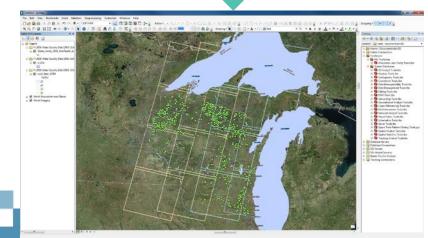
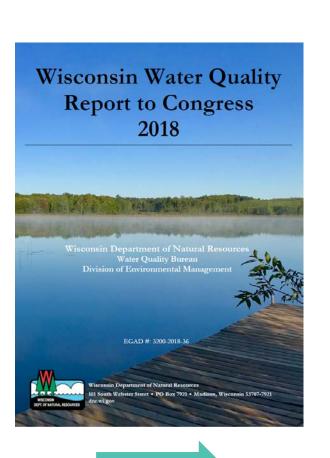


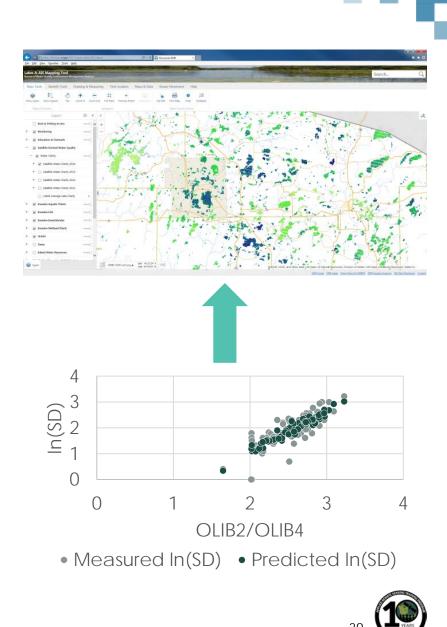




Photo credit: Amy Kowalski







2016 Satellite Retrieval of Water Clarity

- Pre-processing and mosaicking of Landsat Images
- Extraction of field data signatures from Landsat images
- Multiple linear regression of field and satellite data
- Retrieval of water clarity from Landsat images
- Software packages used include ArcGIS 10.4.1 for Desktop, ENVI 5.4.0, IDL 8.6.0, R for Windows 3.3.1, and RStudio

Download of Landsat images and ancillary data (Landsat Collection 1 Level-1 data)

Conversion of image digital numbers (DN) to top of atmosphere (TOA) reflectance

Removal of clouds and cloud shadows

Reprojection of images

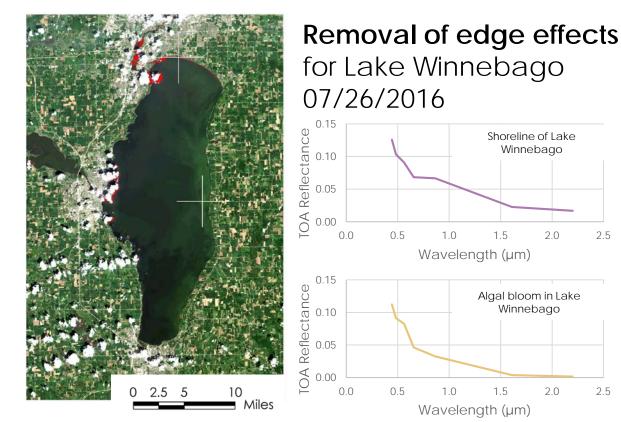
Removal of land and residual clouds

Identification and removal of shallow water, aquatic vegetation, and lake edges

Building of mosaic rasters for images from the same image acquisition dates



2016 Satellite Retrieval of Water Clarity



Differentiation of the shoreline and an algal bloom for the Landsat 8 OLI images acquired on 07/26/2016 (Source of Landsat 8 OLI data: USGS).

$$ln(SD) = a + b \times \frac{OLI_{B2}}{OLI_{B4}} + c \times OLI_{B2}$$

In(SD) – Natural logarithm of the Secchi depth

OLIB2 - Operational Land Imager Band 2

OLIB4 – Operational Land Imager Band 4

76 Landsat 7 and 8 images

2.5

2.5

20

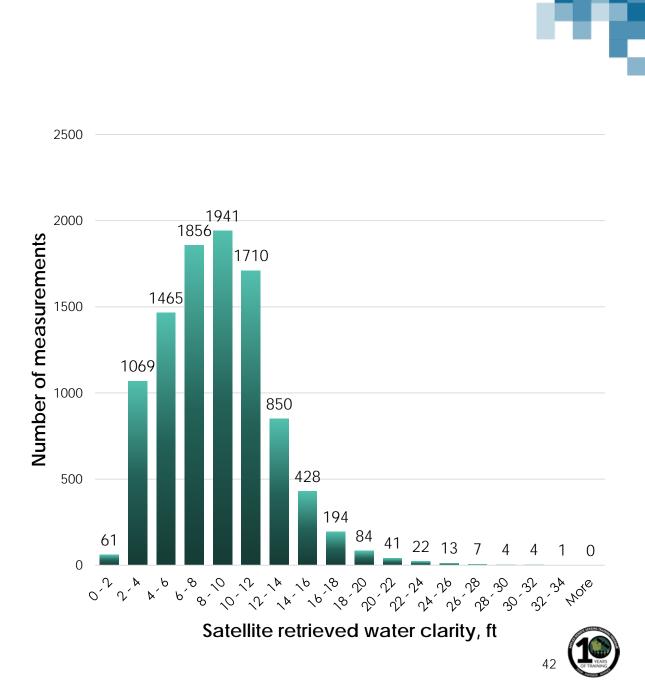
2.0

- 11 image mosaics for algorithm development
- 608 ground-truth measurements for algorithm development



2016 Satellite Retrieval of Water Clarity

- 9750 water clarity estimates
- 4500 water bodies
- Average mean normalized absolute error of 2.0 ft (0.6096 m)



# Data Use for Clean Water Act "Integrated Reporting"

### What is integrated reporting?

- Fulfill federal reporting requirements for statewide water quality conditions
- Water quality standards are used to define goals for a waterbody through use designations, use protection, and water quality protection
- Water quality monitoring data is used to assess the current status of the waterbody
- General and specific assessments
- Designated uses are classified into four categories



### Aquatic Life

### Recreation

Public Health & Welfare

# Data Use for Integrated Reporting

- General condition assessments include
   multiple metrics
- Carlson Trophic State Index is the most commonly used index of lake productivity
- Calculated from chlorophyll-a or Secchi depth which includes satellite inferred Secchi depth
- Calculated automatically with a programming package

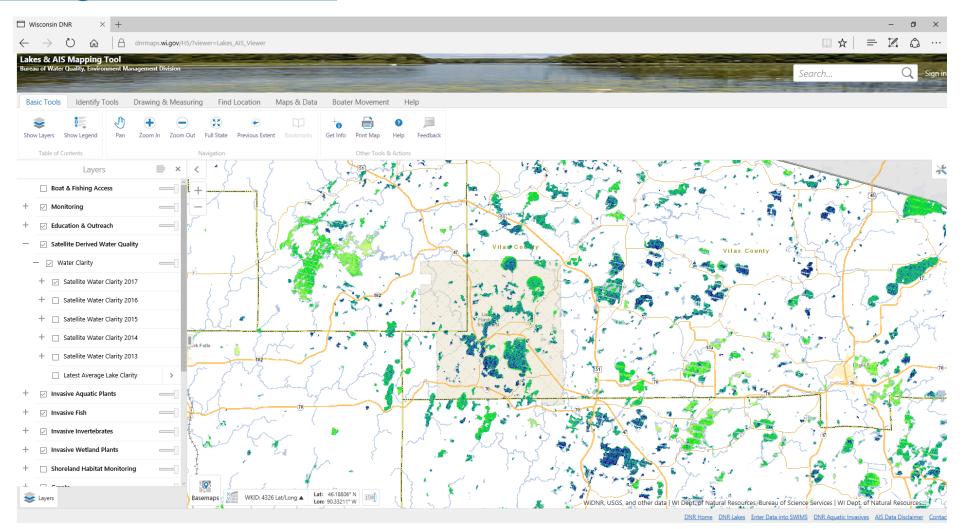
$$TSI_{SD} = 60 - 14.41 \ln(SD)$$

- TSI Trophic State Index
- SD Secchi depth (m)
- Ln Natural logarithm



### **Data Dissemination**

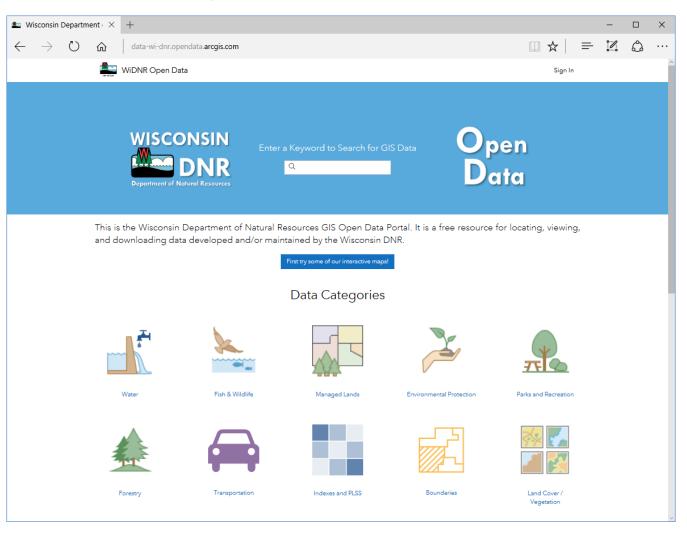
### https://dnr.wi.gov/lakes/viewer/



THE YEARS

### **Data Dissemination**

### http://data-wi-dnr.opendata.arcgis.com/







# **Remote Sensing Research Projects**

- Field data collection in summer and fall 2014 and 2015 for algorithm development
- 32 lakes in Wisconsin
- Standard water quality data
- Radiometric data
- Absorption and backscattering data
- Additional field data collection to support partnerships in summer 2016 and 2018



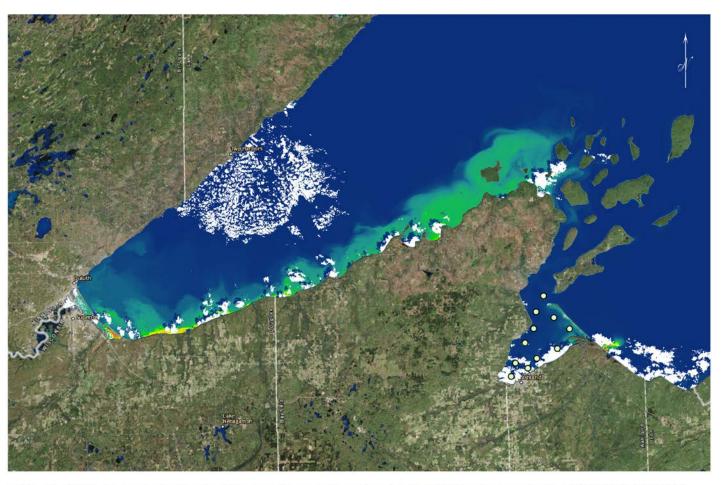
Regretting the decision to collect field data...



Field data collection completed...



### **Remote Sensing Research Projects**



Retrieval of TSM concentration for the Lake Superior nearshore from Landsat 8 OLI data acquired on 07/10/2014 (C2RCC)

O Northland College stations TSM concentration 0.10 mg L<sup>-1</sup> 158.60 mg L<sup>-1</sup>

Basemap credits: Esri, DigitalGlobe, GeoEye, Earthstar Geographics, CNES/Airbus DS, USDA, USGS, AeroGRID, IGN, and the GIS User Community





# Dr. Amita Mehta Demonstration of Landsat Download and Processing