

# Cover crop mapping in Maumee River watershed

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## INTRODUCTION

- Excessive nutrient loadings from agricultural lands are the major contributing factors to **water quality problems** at both regional and national scales.
- Increasing occurrences of **harmful algal blooms** pose serious threats to aquatic and human health.
- Cover cropping** is one of the best management practices (BMPs) to improve water quality and promote soil conservation [1].
- Cover crop benefits** include reduced soil erosion, compaction and nutrient loss; higher organic matter; and improved soil health.
- Despite substantial efforts to promote BMPs through conservation programs (e.g., conservation reserve program (CRP) and environmental quality incentive program (EQIP)), **widespread adoption of these practices has not followed**.

- There is **knowledge gap** from micro (soil microbiology) to macro (local to national policy) scales.
- Current estimates of cover crop adoption rate and benefits are based on **extensive field-based surveys**, which are costly, time consuming, and are not representative of the larger landscape [2].
- Satellite remote sensing technology** provides a cost-effective solution for rapid estimation of cover crop presence, their growth and nutrient uptake at a watershed scale.

## OBJECTIVES

- To leverage historical satellite time-series data to identify spatial and temporal trends in cover crop adoption practices in Maumee River Watershed, Ohio.
- To assess the impacts of cover crops on water quality.

## STUDY AREA

- About 80% of the Maumee River watershed (Fig1) is mainly on corn and soybean production, which is estimated to contribute **>85% of the nutrient loads** to the Maumee River.
- The nutrients from Maumee River is the major contributor to **algal blooms** in the Lake Erie.

## Data

- Sentinel 2A and Sentinel 2B satellites' imagery
- Raw data on cover cropping status by field for 2019 from Farm Service Agency
- USDA crop land data layer
- In-situ water quality data from 2015 for four winters

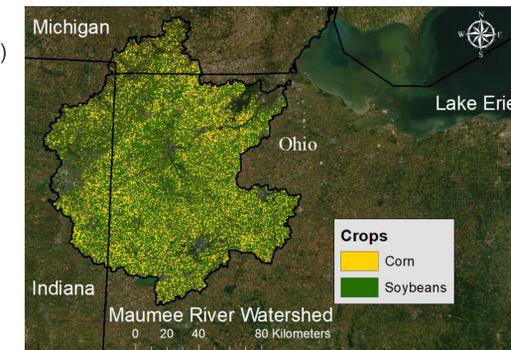


Figure 1: Maumee River Watershed (17326 sq.km), OH

## METHODS

### Data Processing

- Compiled satellite imagery.
- Preprocessed to avoid snow cover and cloud cover.
- Estimated vegetation indices using various combinations of spectral bands of satellite imagery.
- Data processing and analyses in Google Earth Engine.

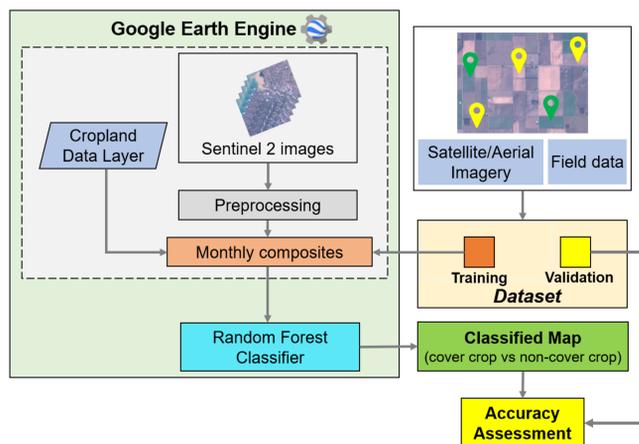


Figure 2: Data processing and classification framework

Table 1: Features of the images used in the classification

SWIR 1	Shortwave Infrared 1
SWIR 2	Shortwave Infrared 2
NDVI	Normalized Difference Vegetation Index
NGRDI	Normalized Green Red Difference Index
DVI	Difference Vegetation Index
RVI	Ratio Vegetation Index

### Machine Learning Algorithm:

- Random Forest for supervised classification of satellite images.
- Overall classification accuracy > 80%.

## RESULTS

■ Cover crop  
■ No cover crop

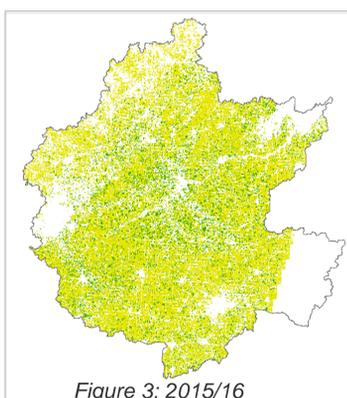


Figure 3: 2015/16

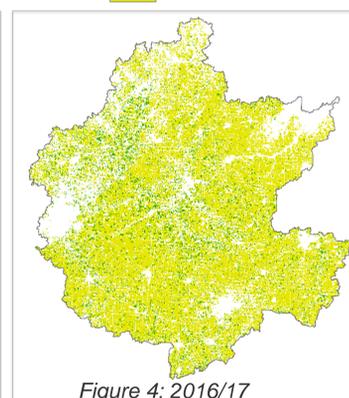


Figure 4: 2016/17

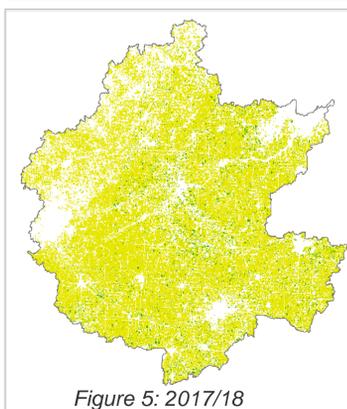


Figure 5: 2017/18

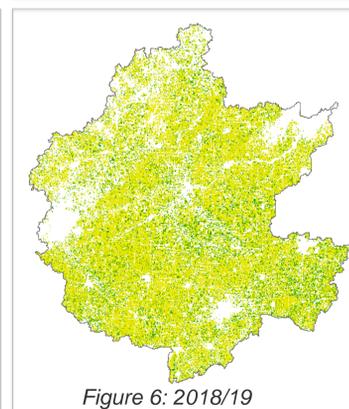


Figure 6: 2018/19

Table 2: Fraction of cover cropped area

Year	Cover crop area (%)
2015/2016	10.7%
2016/2017	8.8%
2017/2018	3.4%
2018/2019	13%

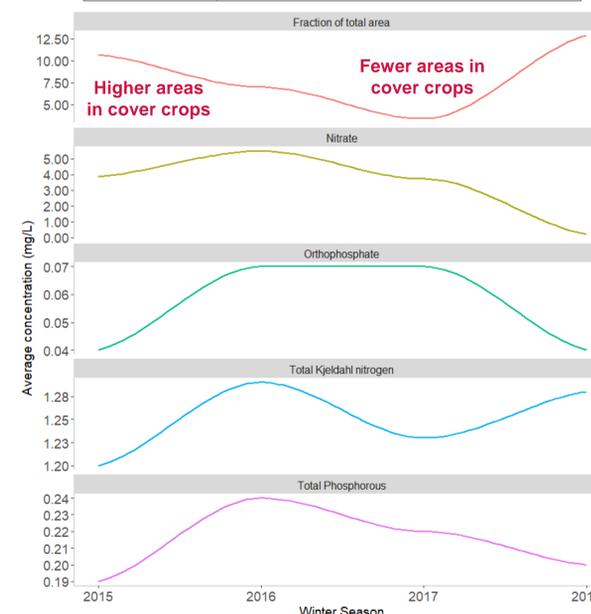


Figure 7: Relationship between cover cropping and water quality

- There is inconsistent trend in cover cropping practice in the Maumee River watershed.
- Increase in cover cropping practices seems to lower phosphorous and nitrogen concentration in the Maumee river.
- The analyses suggest the importance of cover cropping in reducing nutrient loadings from agricultural lands, thereby improving water quality.

## CONCLUSIONS AND FUTURE WORKS

- Areas in winter cover crops classified using machine learning algorithm on satellite imagery.
- The total areas of agricultural land in cover crops declined suddenly after 2015/16 and increased during 2018/19 season.
- The trend in cover cropping practice seems to have direct implication on water quality.
- Future studies will involve in-depth assessment on the relationship with water quality and crop yields.

## ACKNOWLEDGEMENTS

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## REFERENCES

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- [2] CTCIC. (2017). Report of the 2016-17 National Cover Crop Survey. Joint publication of the Conservation Technology Information Center, the North Central Region Sustainable Agriculture Research and Education Program, and the American Seed Trade Association. West Lafayette, IN.
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