

Model Correlations and Forecasting Using Virtual Beach

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Agenda

- Modeling
- Virtual Beach
- Case study Ottawa WTP



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Problem: Delayed notification

Today's concentrations of E. coli are not available until tomorrow



- Notify the public based on the previous day's concentrations
- Sanitary conditions can change overnight



Site-specific multiple linear regression models

- Contain two or more variables related to target concentrations
- Results available within an hour to make timely management decisions
- Multiple linear regression models have been shown to work well to predict recreational water quality at Great Lakes beaches
 - Francy and others 2013

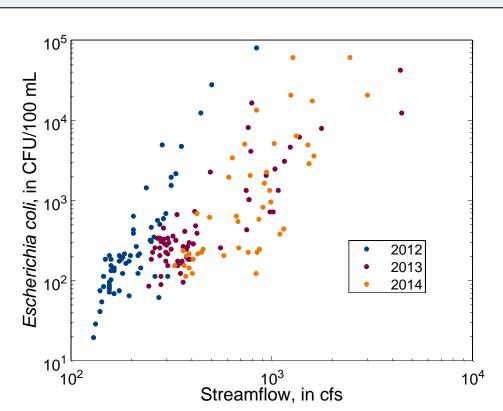




Model development

Multiple years of data are necessary prior to model development to try to capture the annual variability

Relations between E. coli and environmental and(or) water-quality variables can vary from year to year based on weather patterns





Virtual Beach

- Free software developed by the USEPA
- Used for studying relations between water-quality indicators or threats and ambient environmental conditions
- Has typically been used to estimate E. coli and enterococci concentrations

https://www.epa.gov/ceam/virtual-beach-vb

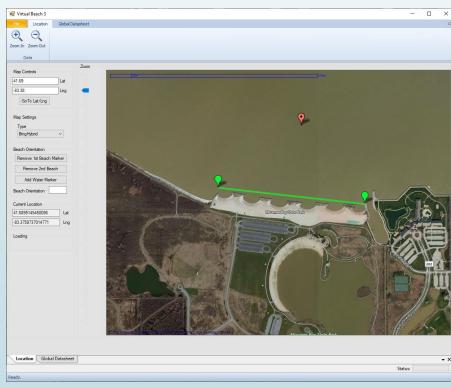






Virtual Beach components

- Map to define site orientation and calculate wind, wave, and current variables
- Spreadsheet processing and analysis of imported data
 - Key ability to optimally transform the response and independent variables



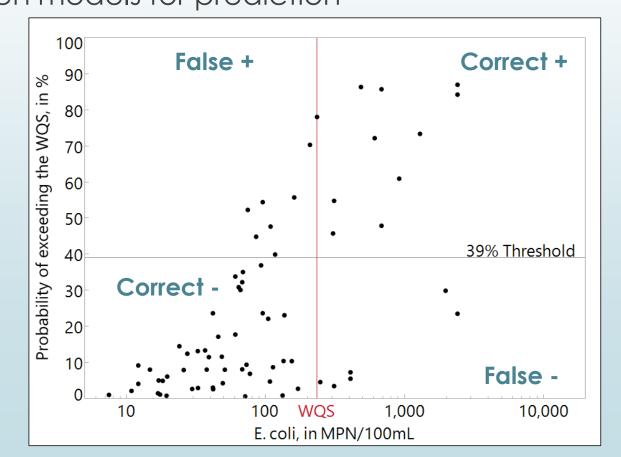


Virtual Beach components

Linear regression analysis with model selection tools
Use of chosen models for prediction

Two types of output: 1. Predicted

concentration
Probability of exceeding a concentration





Two types of cyanoHAB toxin models

- <u>Real-time models</u> include factors that are easily or continuous measured
- <u>Comprehensive models</u> use factors from samples collected and analyzed in a laboratory

Real-time variables

- Phycocyanin
- ► pH
- Temperature
- Daily mean streamflow
- Satellite data

Comprehensive variables

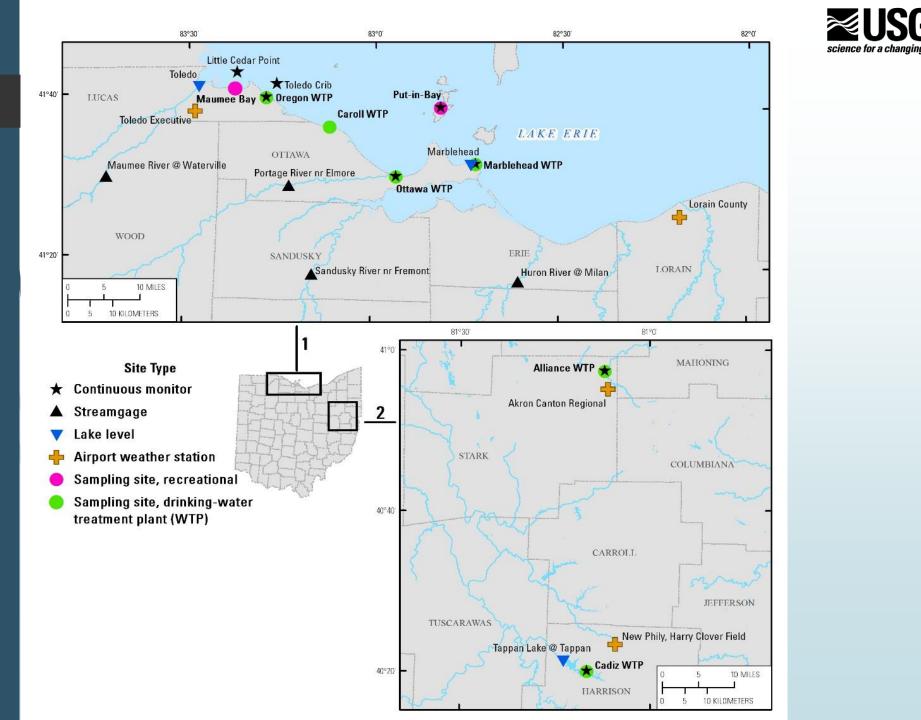
- Cyanobacterial genes
- Nutrient concentrations



Case Study Model Development – Ottawa WTP

- Lake Erie intake, influenced by Portage River
- Collect samples from their wet well before addition of permanganate
- Data 2016 2018
- Samples collected semiweekly (n = 105)



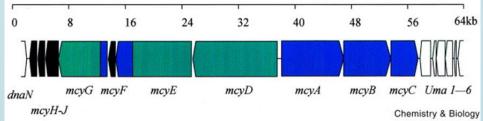




Data collection and compilation

- Microcystin and nutrients
 - Ohio State University, Stone Laboratory, 2016-2017
 - USGS National Water Quality Laboratory, nutrients 2018
 - Oregon WTP, microcystin 2018
- Cyanbacterial genes Ohio EPA
 - General cyanobacteria
 - Microcystin gene
 - Saxitoxin gene
 - Cylindrospermopsin gene







Data collection and compilation

Physical parameters measured at the plant

- Turbidity
- Water temperature
- Environmental data
 - NOAA weather data
 - NOAA water levels
 - USGS streamflow
 - Continuous monitor data
 - Ottawa WTP
 - Oregon WTP
- Satellite data
 - NASA data

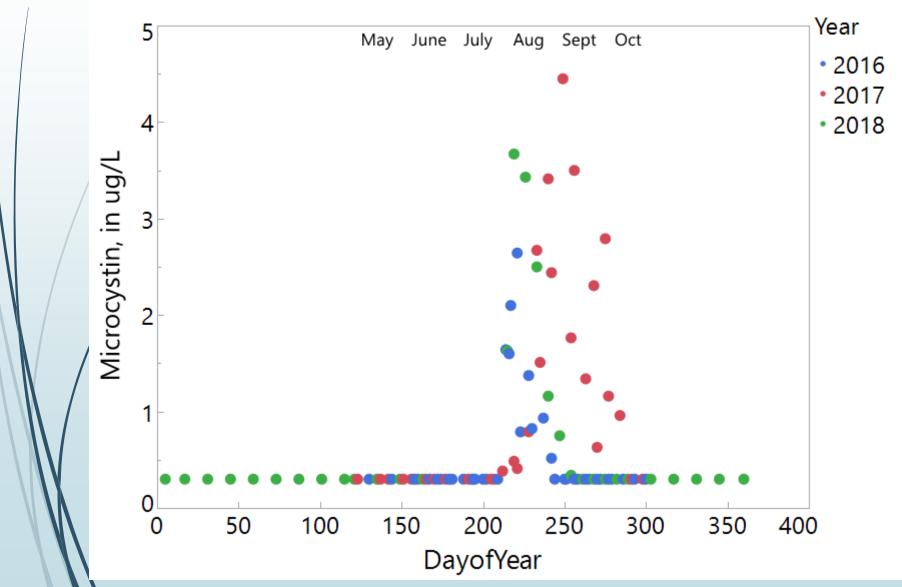








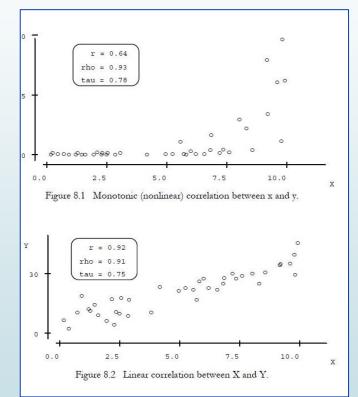
Microcystin at Ottawa intake





Correlations

- Strength of the association: -1 to 1
- Spearman's rho (monotonic) and Pearson's r (linear)
- Significance: p value (<0.05)
- Does not imply cause and effect
 - Helsel and Hirsch, Statistical Methods in Water Resources, 2002





Spearman's correlations (rho)

Independent Variables	Ottawa WTP
Nutrients	
Nitrite plus nitrate, lagged	-0.34
Dissolved reactive phosphorus, lagged	-0.66
N to P ratio; lagged	-0.24
Ammonia, lagged	-0.32
Cyanobacterial genes	
General cyanobacteria, lagged	0.67
Microcystin gene, lagged	0.77

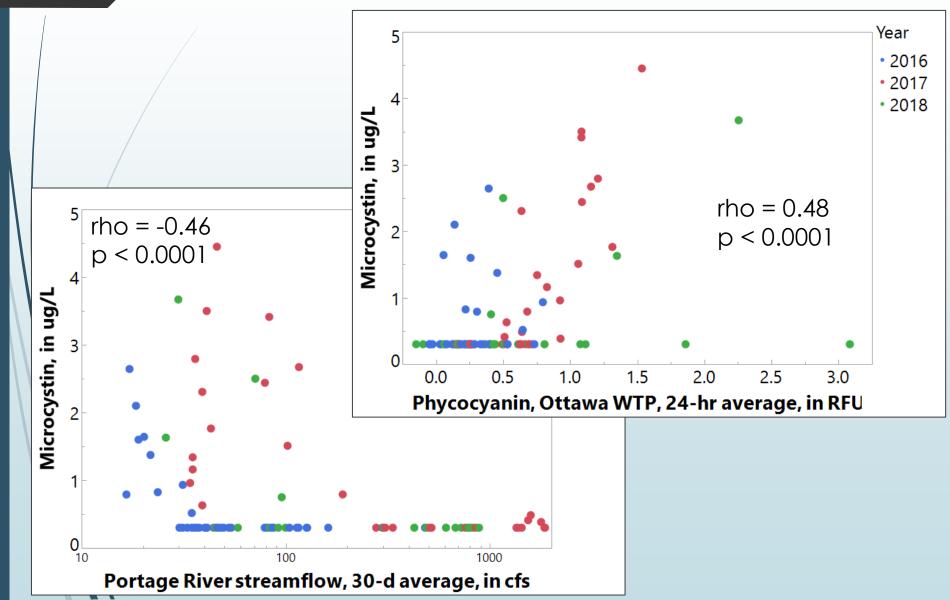


Spearman's correlations (rho)

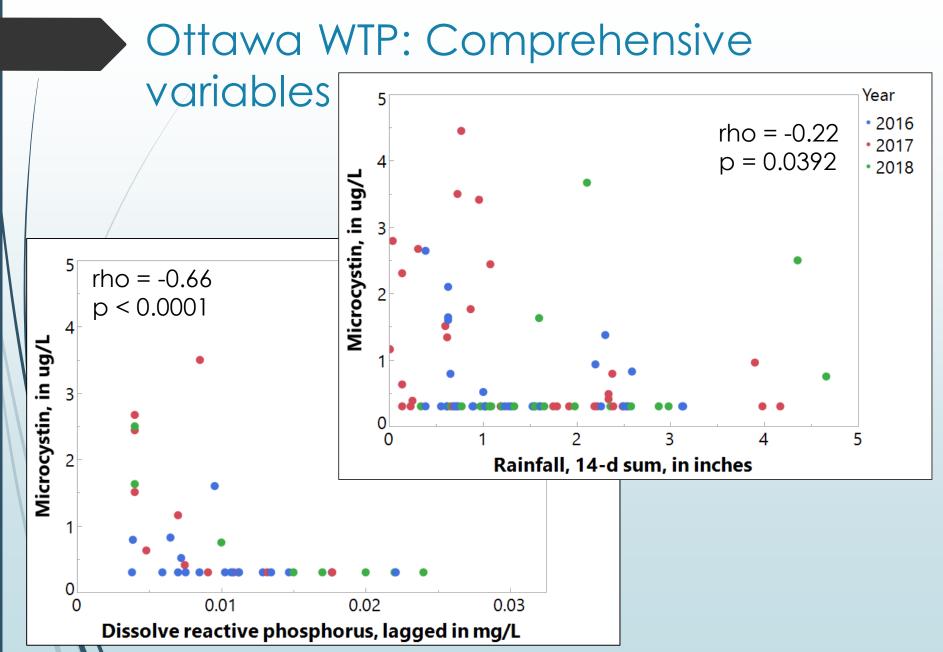
Independent Variables	Ottawa WTP
NOAA and USGS derived data	
Rainfall, 14d sum	-0.22
Wind speed, instantaneous	-0.45
Lake level, 7d ave	0.19
Portage Rv discharge, 30d ave	-0.46
Continuous monitor data	
Phycocyanin, 24hr ave, Ottawa	0.48
Phycocyanin, 14d ave, Oregon	0.70
Turbidity, 14d ave, Oregon	-0.13
Specific conductance, 14d ave, Ottawa	-0.56
pH, 14d ave, Oregon	0.72



Ottawa WTP: Real-time variables



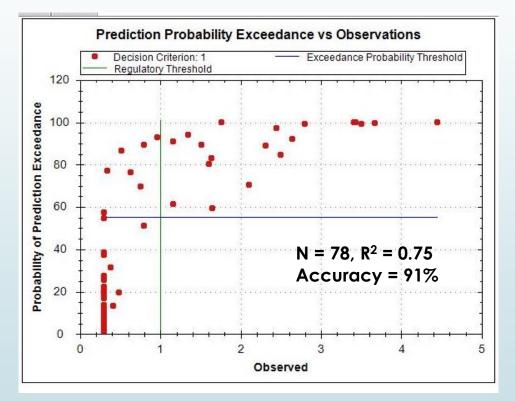






Ottawa WTP real-time model: 2016–18

- Cosine, day of year
- Phycocyanin, 24 h, Ottawa sonde
 - /pH, 14d, Oregon sonde
 - Discharge, Portage River, 30d average



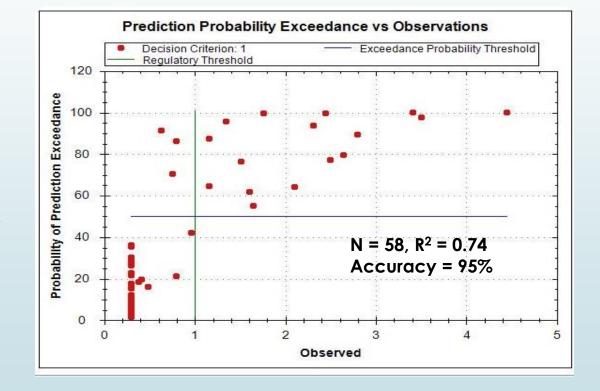
Microcystin = -11.32 - 1.019*(SQUARE(cos_DOY)) + 1.456*(Leoc_BGA_RFU_Ave24hr) + 1.446*(Leorgn_PH_Ave14d) - 0.5159*(LOG10(Portage_Dis_30dAve))

Preliminary information-subject to revision. Not for citation or distribution.



Ottawa WTP comprehensive model: 2016–18

- pH, 14d, Oregon sonde
- Orthophosphate, lagged
- Phycocyanin, 14d, Oregon sonde
- Rain, 14d sum, LPR



Microcystin = -9.896 + 1.313*(Leorgn_PH_Ave14d) - 20.81*(OrthoP_mgL_LAG_PLUS) + 0.2508*(Leorgn_BGA_RFU_Ave14d) - 0.1887*(LPR_Rain_sum14d)

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Benefits of modeling

Real-time predictions

- Provide information to trigger sample collection
- Provide data to optimize water treatment and intake options for current conditions

Comprehensive predictions

- Provide advanced warning of the potential for a toxic cyanoHAB
- Provide an understanding of what factors are related to toxin production



Acknowledgements





Thank you!

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