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# Extracting Signal from the Noisy Environment of an Ecosystem

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CENTER FOR  
ENVIRONMENT, ENERGY  
AND ECONOMY

**SRBC**

SUSQUEHANNA RIVER  
BASIN COMMISSION

NY ■ PA ■ MD ■ USA

# Extracting Signal from the Noisy Environment of an Ecosystem

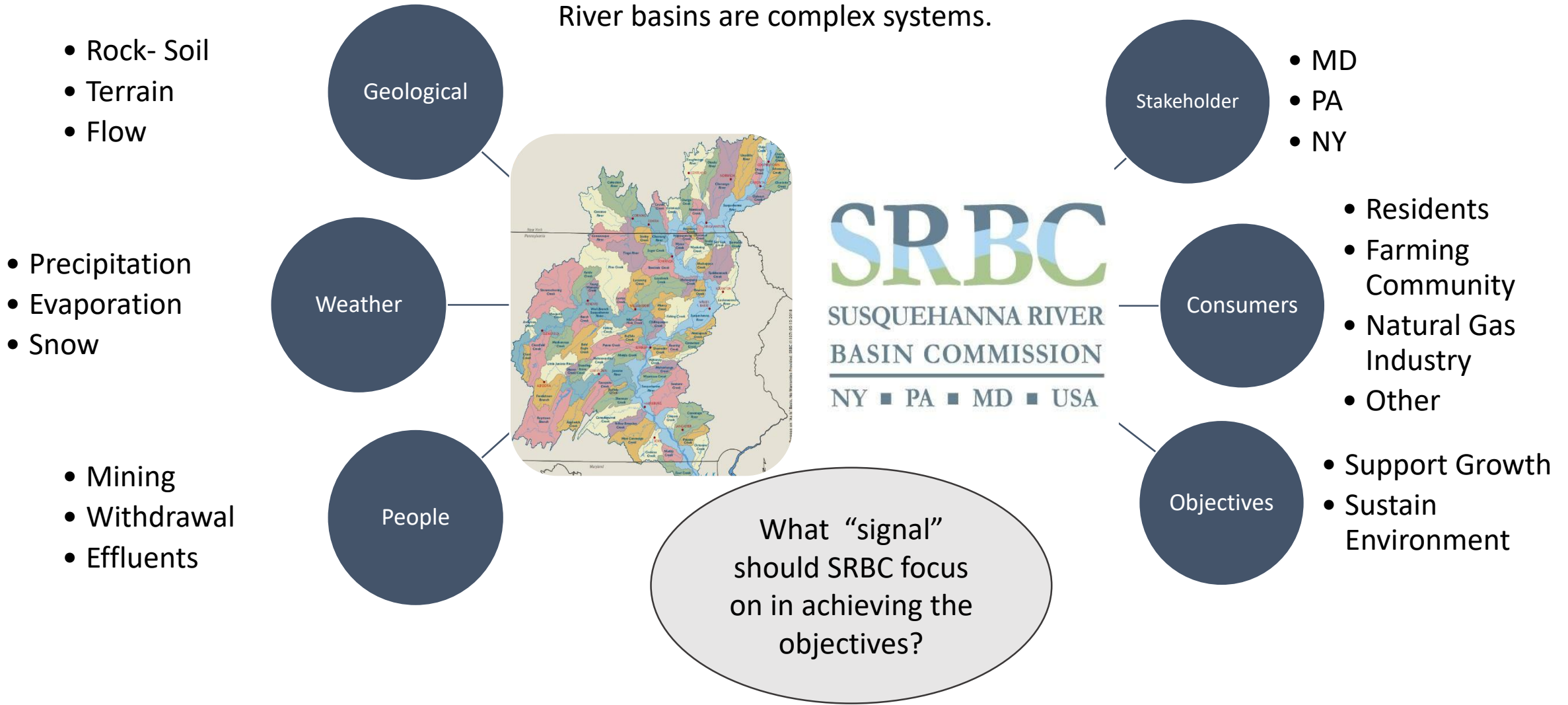
Emily Wefelmeyer, Pranita Patil, Sridhar Ravula,  
Kevin Purcell, Ziyuan Huang, & Igor Pilja

**Harrisburg University of Science & Technology**

# River Basin Management: Signal Vs Noise



River basins are complex systems.



Regulators need to identify "Signal" from noisy data

# Signal Extraction and Definition



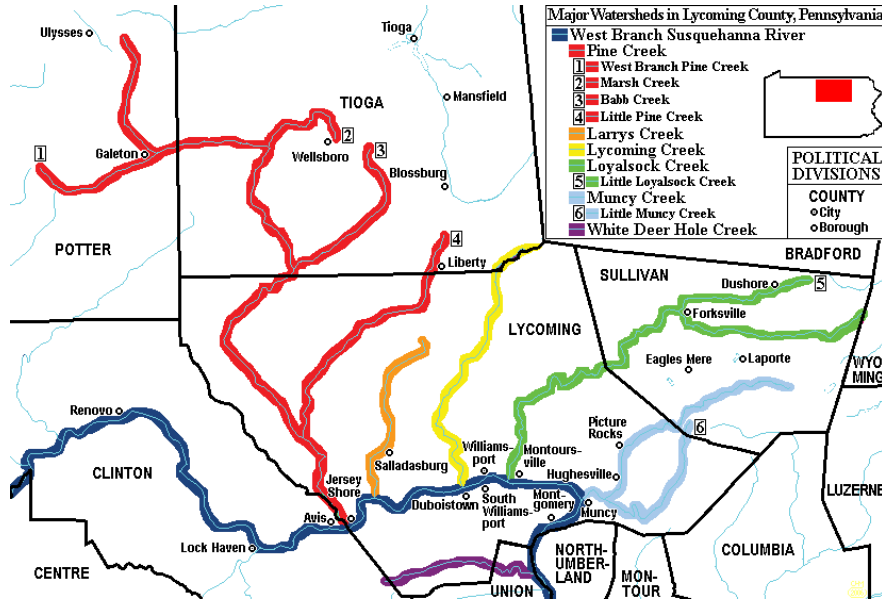
- Volume and variety of data
- Data scientist was sought
- How to systematically define and extract signal from noise
  - Lack of guiding framework
  - Limited tool set
- Explored available tools and literature
- Signal desired attributes
  - Clearly definable
  - Must be able to detect
  - Ability to explain movement in response variable with help of explanatory variables.
  - Has an effect on environment or is an indicator of change in environment
- Rational approach to variable selection
  - Maximize the information obtained from the analysis of variable
    - Linked to a large set of spatial, temporal environmental variables.

# Available Data



Initial scope limited to “Pine Creek”

- Largest tributary of the West Branch Susquehanna River
- 87.2 miles (140.3 km) long
- Largest watershed of all the West Branch’s tributaries



File	<i>n</i>	Parameters	Start	End	Missing Data	Frequency/Periodicity
Water Quality	195,142	6	6/23/2011 12:00	12/31/2017 23:45	<0.15%	4 hours until 10/9/2014; 15 minutes since then
Chemistry	754	51	8/10/1983 00:00	3/5/2018 08:45	~70%	No noticeable pattern; 1983, 1994, 2002, & 2008 – 2018
Fish	52	59	9/09/2008 12:00	8/03/2017 09:45	~0%	No noticeable pattern
Macroinvertebrates	134	223	8/10/1983 00:00	10/17/2017 14:00	<0.001%	No noticeable pattern; 1983 – 2002 & 2008 – 2017
Biotic integrity community data	59	95	No dates given	Seems to be averages	~22%	n/a

# Community Metrics



- Community metrics (Diversity) from observed species count
  - For fish and macroinvertebrates
  - Available for numerous sites
- Community (Diversity) metrics advantages
  - Good overall metrics of environmental quality
  - Environmental quality is a core standard by which SRBC has to accomplish its mission
- Multiple response variables computed for diversity

Metric	Formula	Remark
Margalef's species Richness	$S_{\text{Marg},y} = \frac{S_y - 1}{\log F_y}$	Where $S_y$ is species count and $F_y$ is the total count of all individual fish caught
Pielou evenness	$J_y = \frac{-\sum_{s=1}^{S_y} N_{y,s}/N_y \log(N_{y,s}/N_y)}{\log S_y}$	Where $N_y$ is abundance and $N_{y,s}$ is average density of species 's' (individuals km <sup>-2</sup> )
Hill's N1 Diversity	$N1_y = \exp\left(-\sum_{s=1}^{S_y} \frac{N_{y,s}}{N_y} \log \frac{N_{y,s}}{N_y}\right)$	$N_y$ and $N_{y,s}$ defined as above
Hill's N2 dominance	$N2_y = \frac{1}{\sum_{s=1}^{S_y} (N_{y,s}/N_y)}$	$S_y, N_y$ and $N_{y,s}$ defined as above



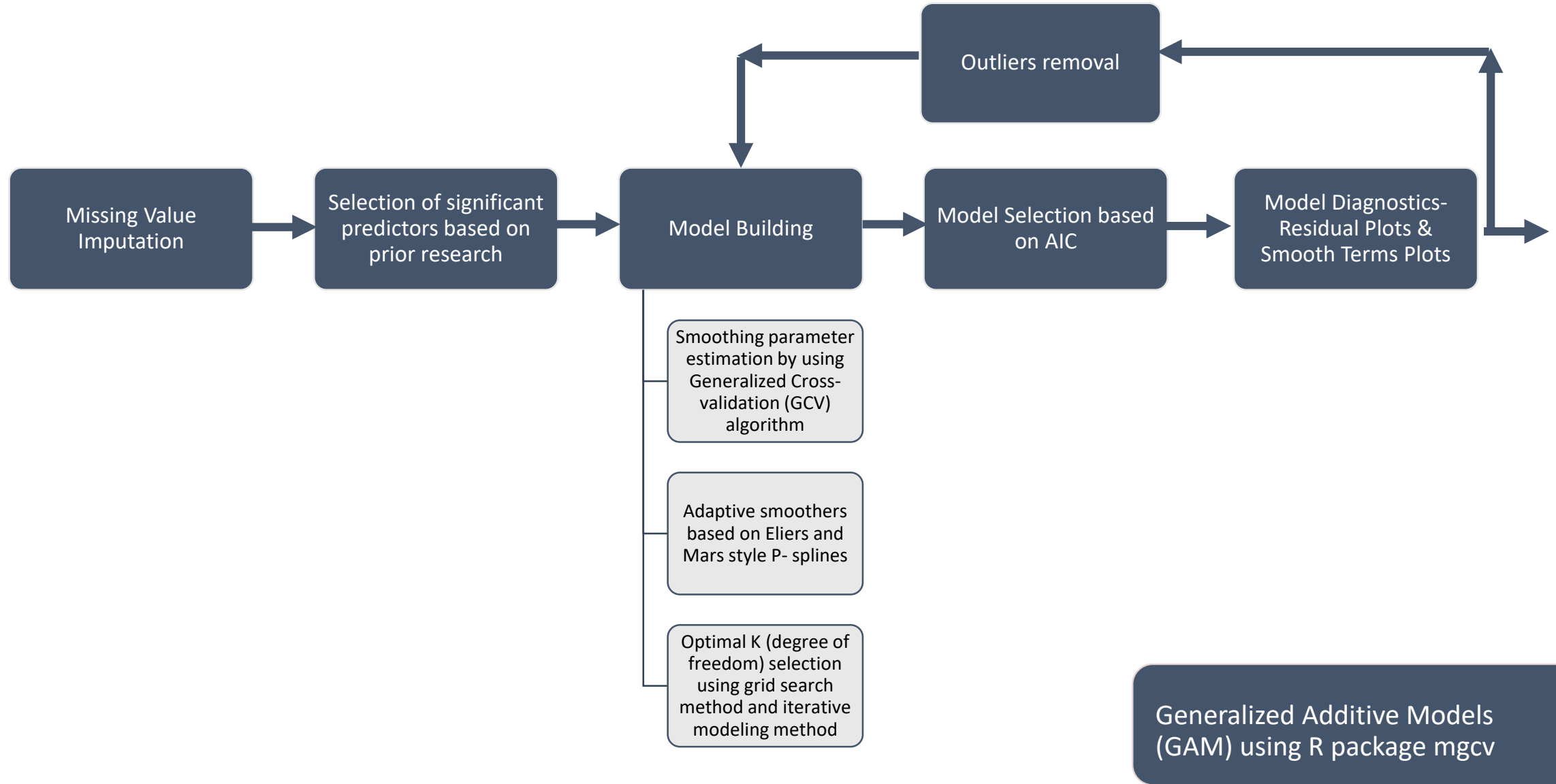
# Model Selection: Generalized Additive Model (GAM)

- More powerful than linear model due to inclusion of non-linear smoothers
- Parametric and non-parametric functions to explore linear and non-linear patterns
- Addition of smooth functions of covariates
- Smooth/basis function estimated from the data
- Mostly used when
  - Non-linear relationships
  - Distribution other than normal (response)
  - Need regularization(to avoid overfitting)
- Applications: air quality, ecology, medicine, genetics, molecular biology, etc.

**These factors made us chose GAM over other models**

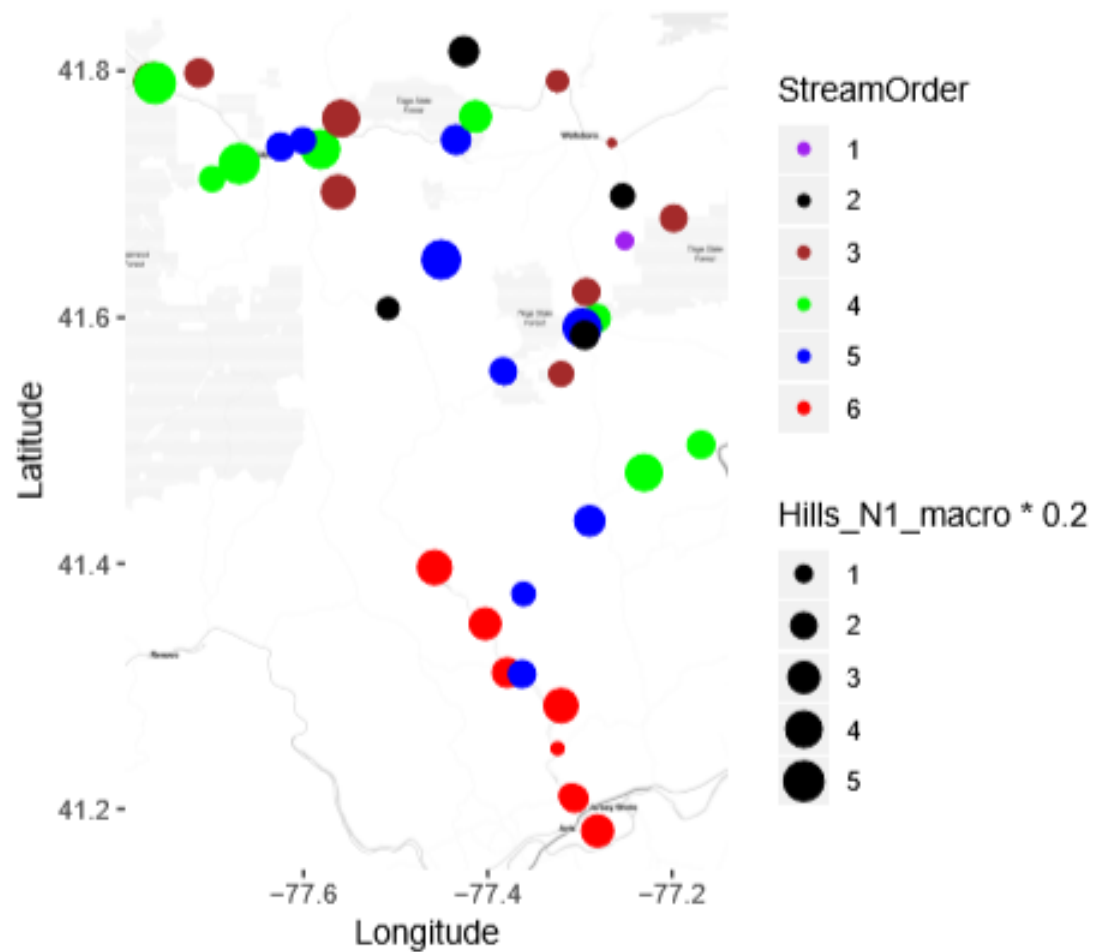
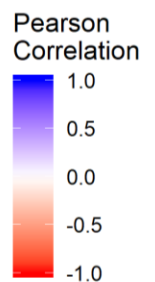
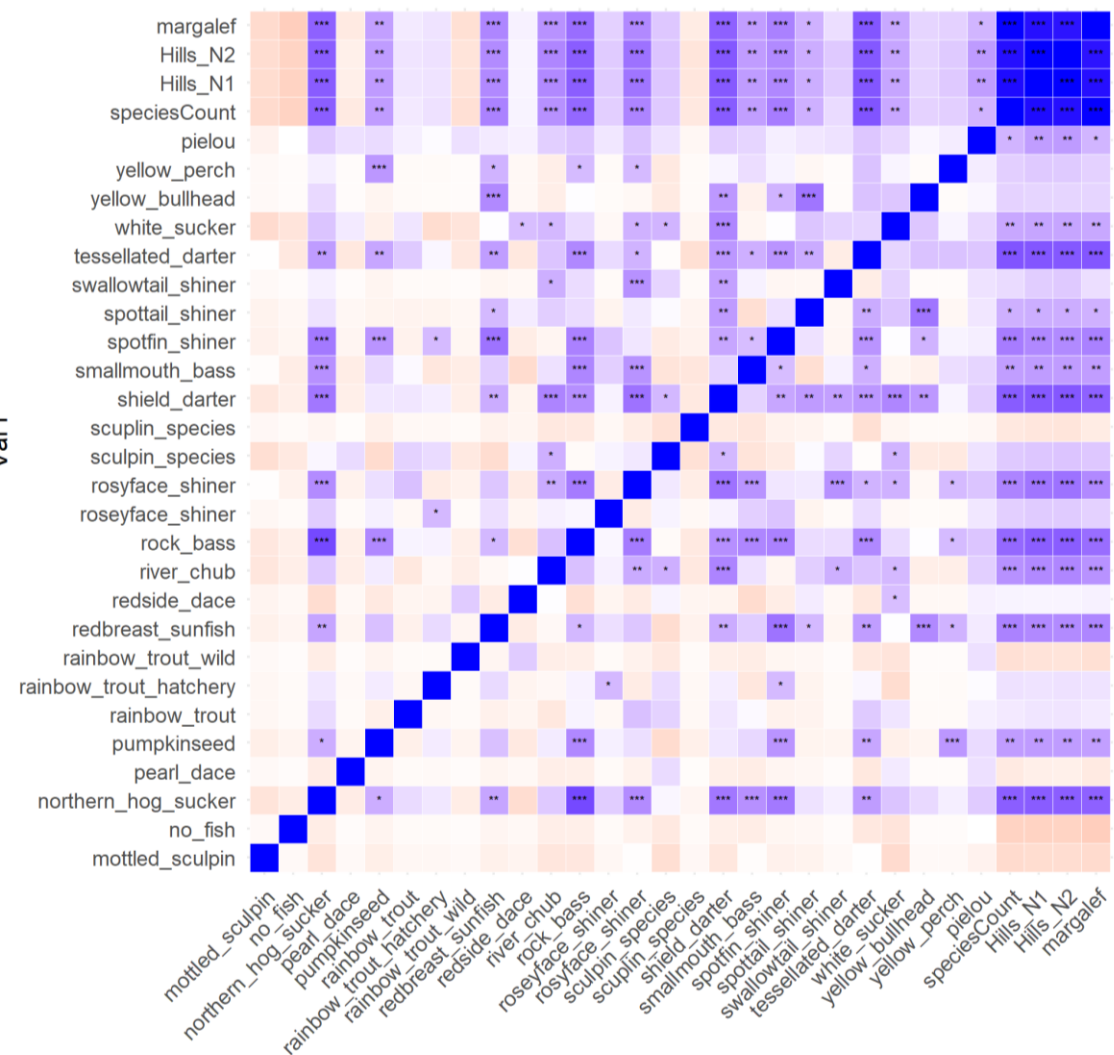


# Modeling Steps





# Model Inputs Selection



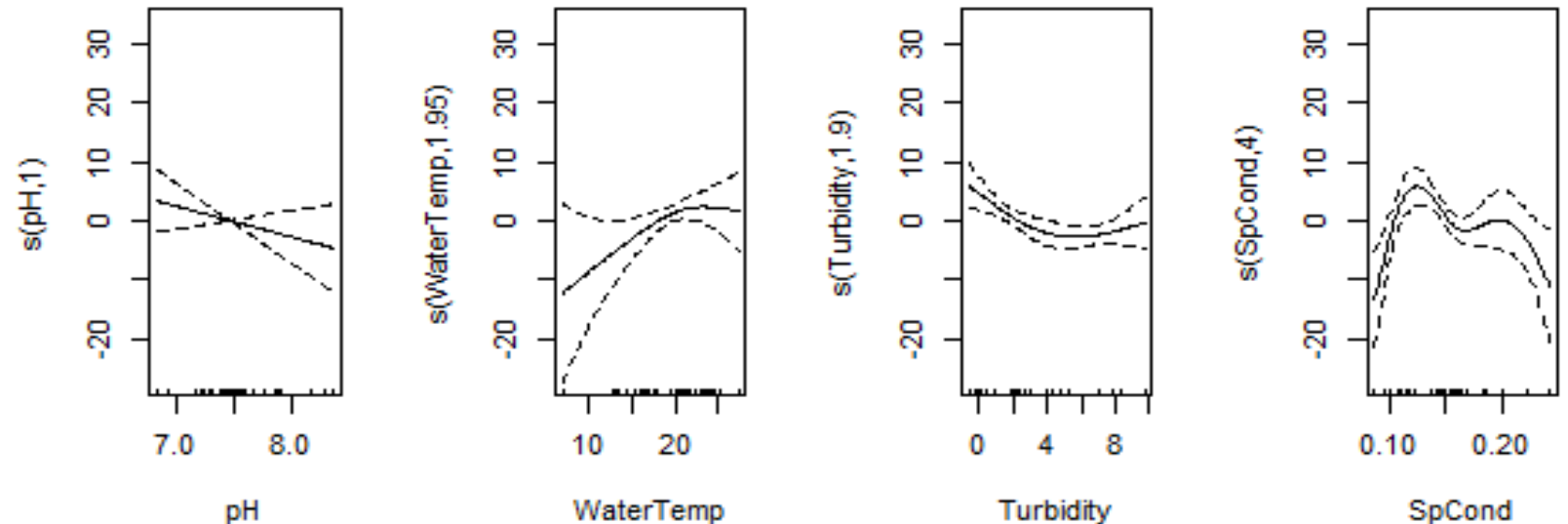
Var2

# Hill's N1 GAM Model: Fish



- Data
  - 50 observations
  - Repeated measurement at few stations
- Grid search for degrees of freedom (3 to 8)
- Best AIC: 288.94
- Deviance Explained: **74%**
- Outliers removed

Model Smooth Functions Plots

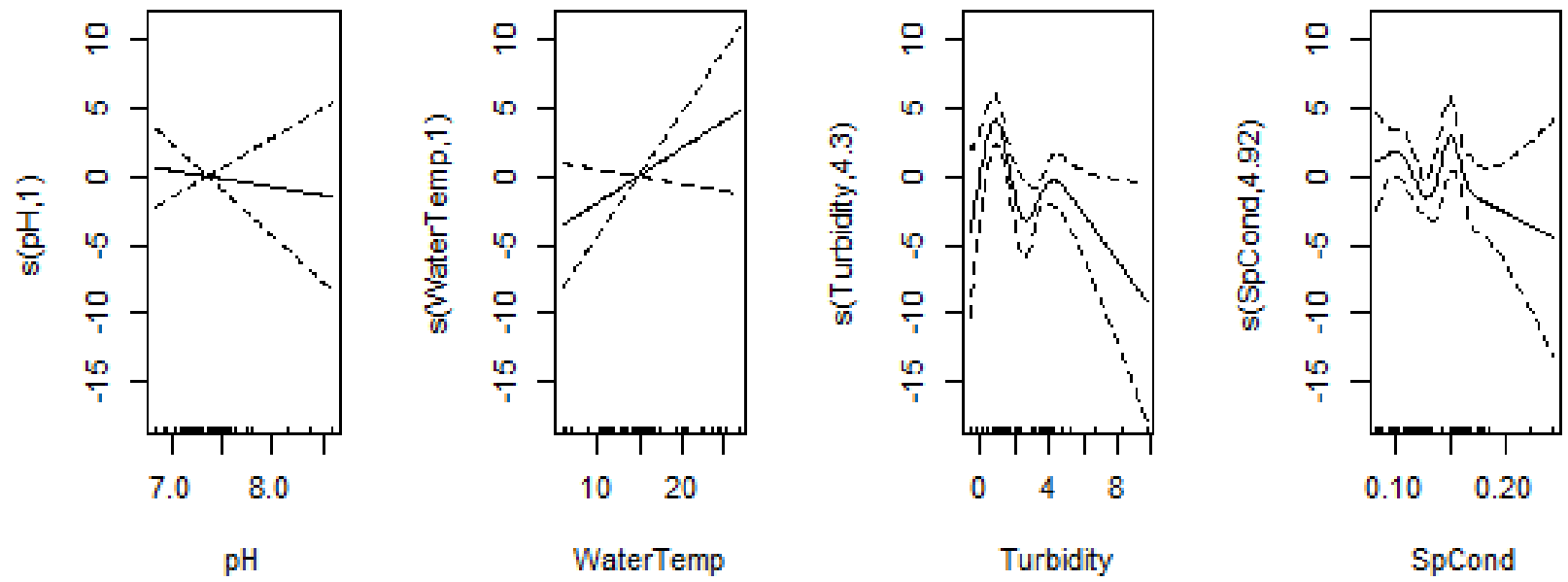




# Hill's N1 GAM Model: Macroinvertebrates

- Data
  - 115 observations
  - Repeated measurement at few stations
- Iterative modeling method for degrees of freedom
- Best AIC: 640.61
- Adaptive Smoothing Parameter
- Deviance Explained: **61.9%**
- Outliers removed

Model Smooth Function Plots



# GAM Models Summary



- Extract significant information from noisy dataset
- Explained significant % of variance in Hill's N1, Hill's N2, and Margalef diversity metrics

## Fish GAM Model

- 72% to 75% variance explained

Diversity Metric	% Variance Explained	AIC
Hill's N1	73.8	289
Hill's N2	72.4	283
Pielou	54	-156
Margalef	74.9	111

## Macroinvertebrate GAM Models

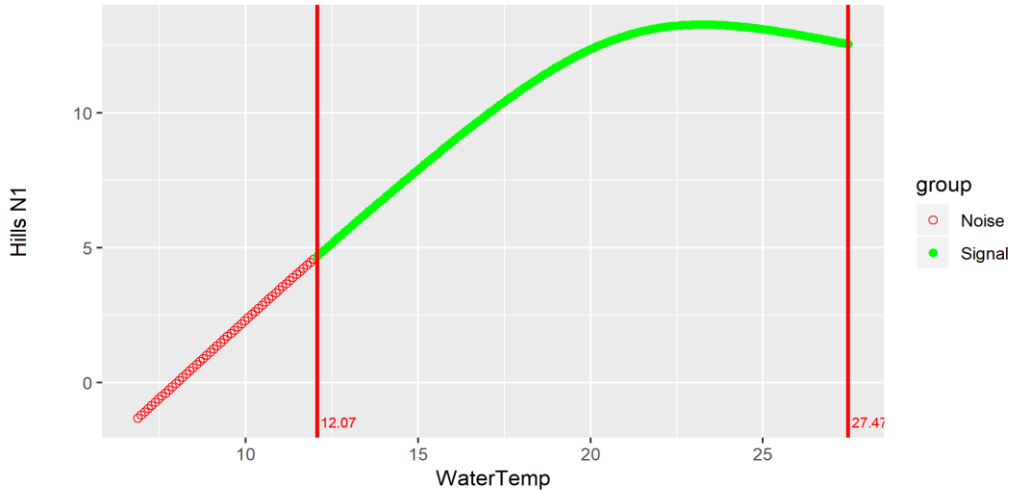
- 54% to 66% variance explained

Diversity Metric	% Variance Explained	AIC
Hill's N1	61.9	641
Hill's N2	54.2	598
Pielou	45.5	-251
Margalef	65.9	296

# Signal Prediction Analysis: Fish Diversity

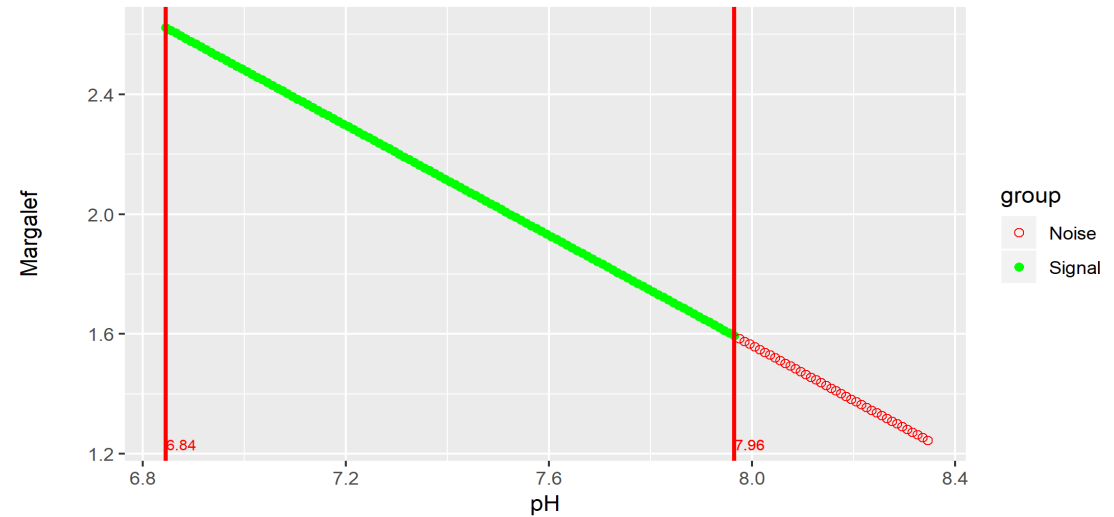


N1 vs WaterTemp- For observed range

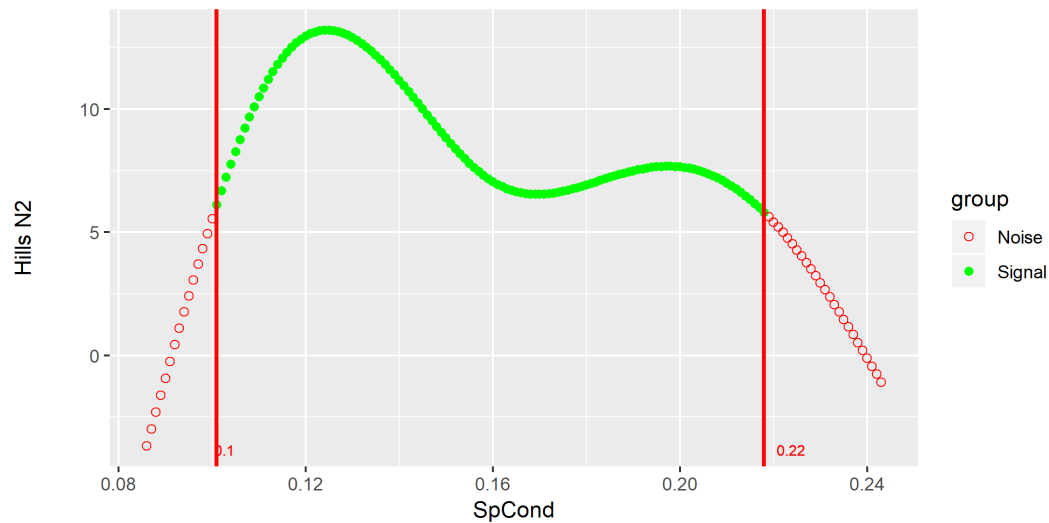


- Posterior simulations
- Change one environmental variable while keeping other variables constant
- Threshold calculations based on quantile method

Margalef vs pH- For observed range



N2 vs SpCond- For observed range

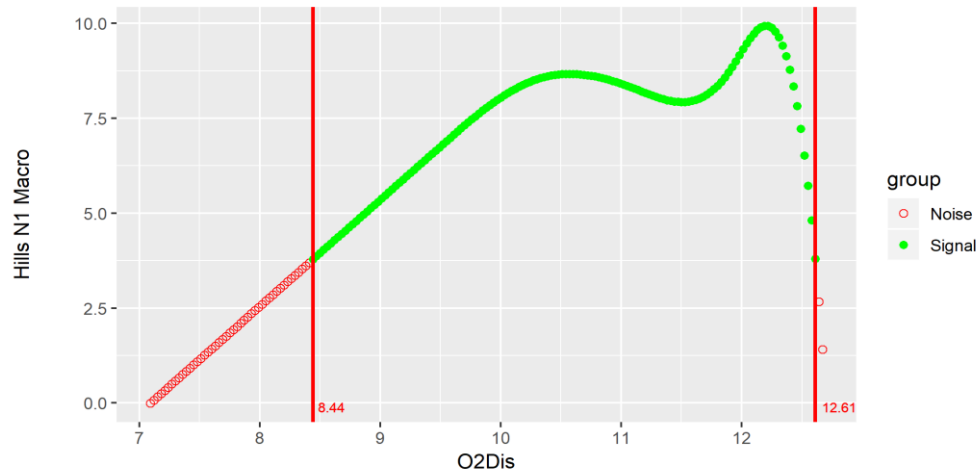


- Fish diversity deteriorates when
  - Dissolved oxygen < 8
  - Specific conductivity > 0.2

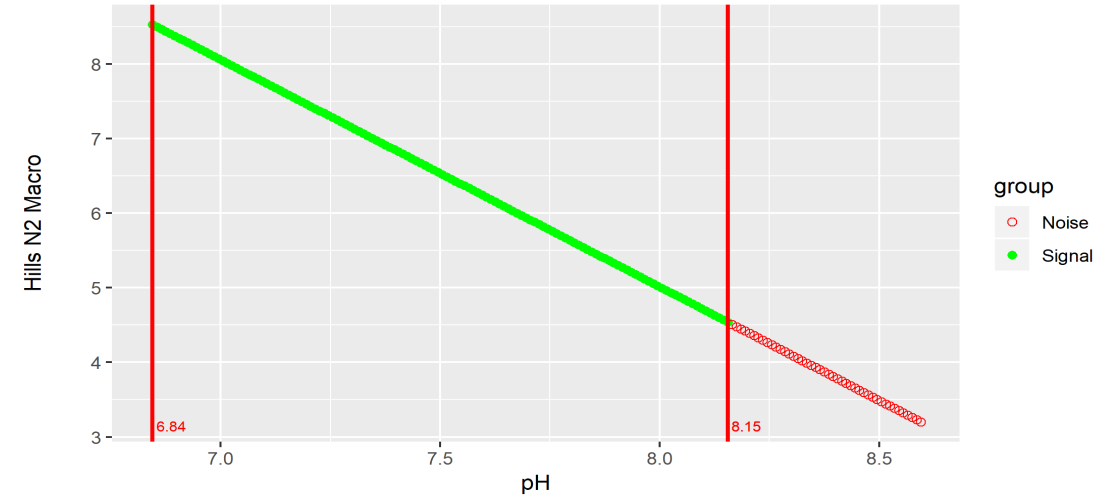
# Signal Prediction Analysis: Macroinvertebrate Diversity



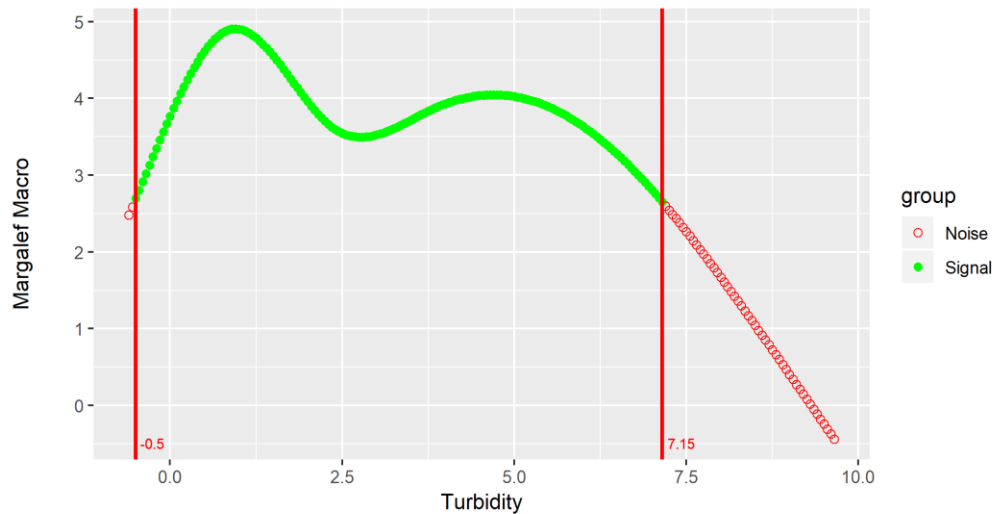
N1 vs O2Dis- For observed range



N2 vs pH- For observed range



Margalef Macro vs Turbidity- For observed range



- Macroinvertebrate diversity deteriorates when
  - Dissolved oxygen < 8.5
  - Turbidity > 7



# Our Findings

## River basin monitoring

- A framework to synthesize multiple datasets
  - Enables regulators to prioritize and communicate
  - Can help stakeholder in understanding the impact of their actions
- Flexible: diversity metric can be replaced with another metric
- Dynamic dashboard application to monitor biotic response

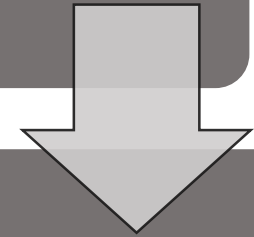




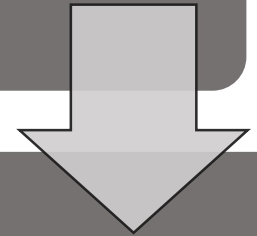
# Next Steps

Refine synthesis in collaboration with SRBC

Action Threshold  
Identification



Incorporate Variable  
Interactions



Adding More  
Information

# Questions?



## Special thanks to the following people/organizations:

John Quigley with the Center for Environment, Energy and Economy at Harrisburg University for the opportunity.

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Dr. Kevin Purcell for his help and support through the project.

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