

# Quantifying the impacts of past and future climate and eutrophication on the dynamics of dissolved oxygen in the shallow waters of Chesapeake Bay

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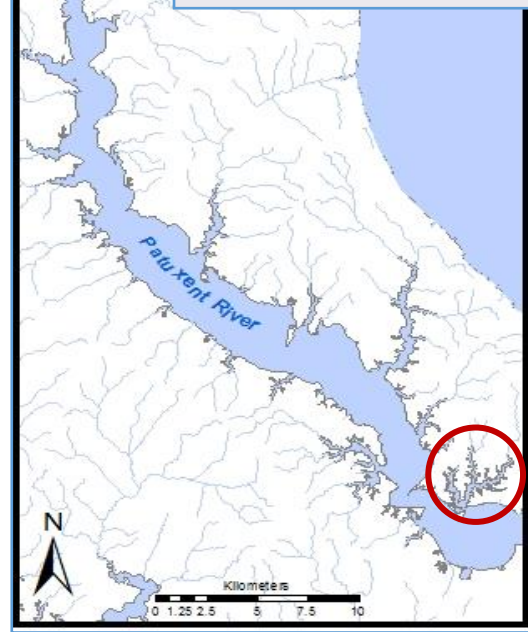
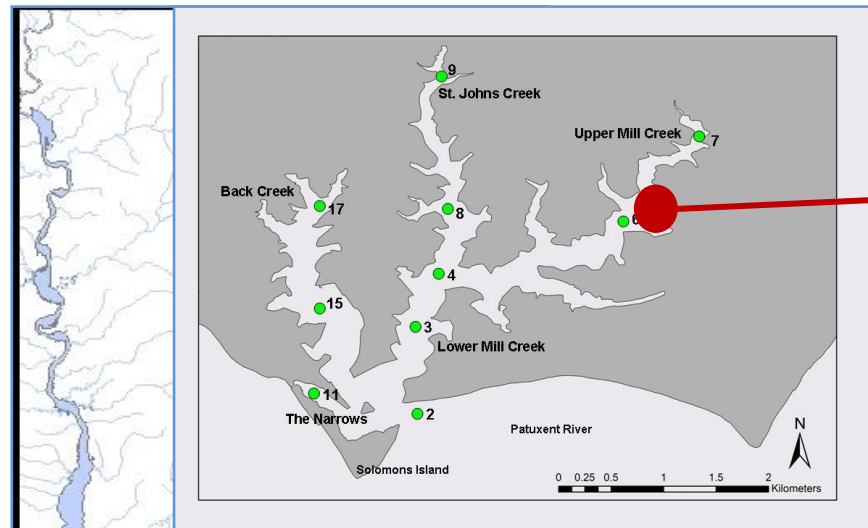
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<sup>4</sup>*Virginia Institute of Marine Sciences*

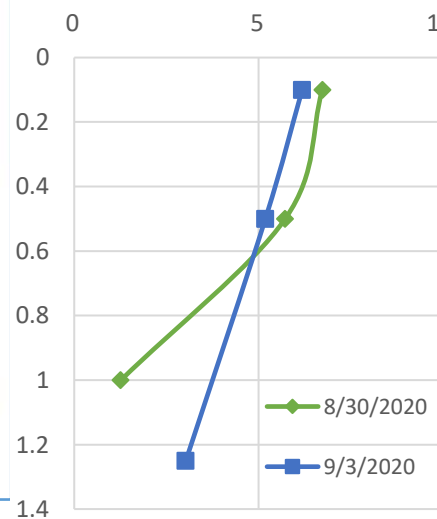
<sup>5</sup>*USGS*

<sup>6</sup>*UMCES/Chesapeake Bay Program*

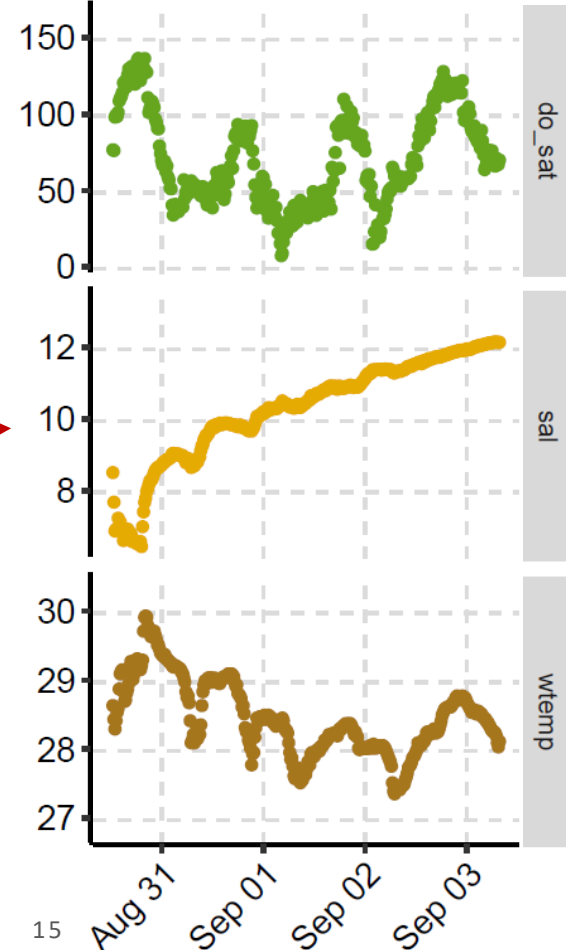
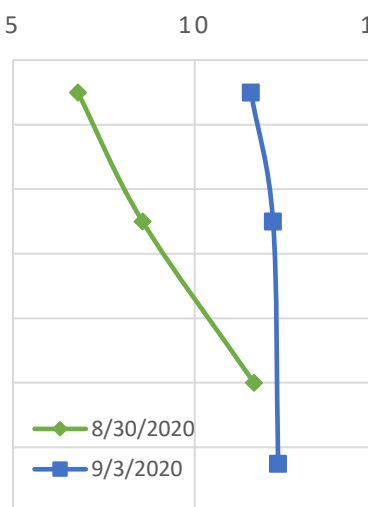
# Shallow Water Oxygen is a Challenging Problem



Dissolved O<sub>2</sub> (mg/L)

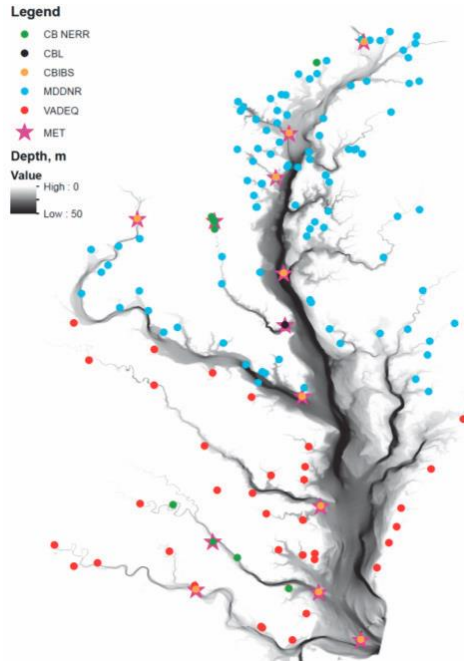


Salinity



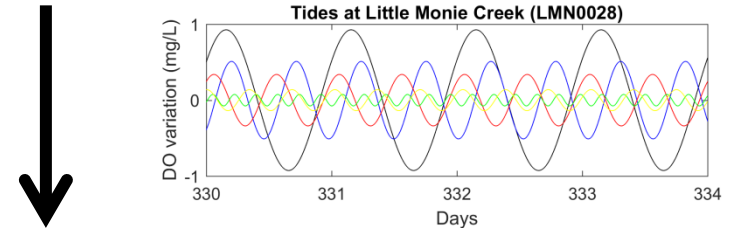
# Schematic of Analysis Design

## High-Frequency Oxygen Observations

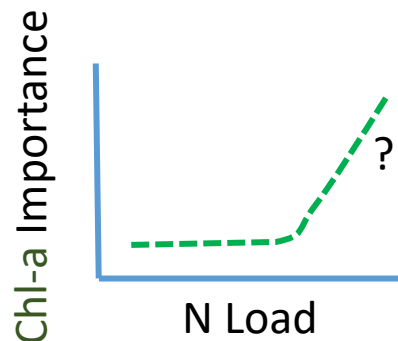


Decompose tidal contribution, isolate residual (biological?) contribution

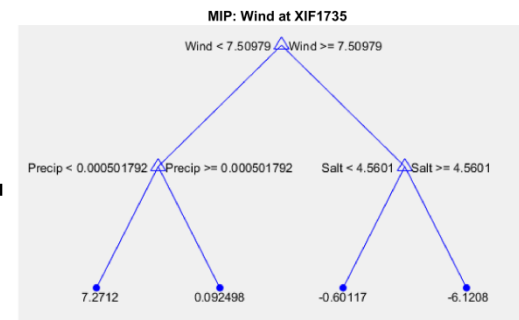
$$C_{DO}(t) = \overline{C_{DO}} + \sum_{n=1}^N A_n \cos(\omega_n t - \theta_n) + R(t)$$



Associate local conditions ( load, metabolism, physical setting) with DO variation



Use CART to link control variables to non-tidal DO variations



Temp  
PAR  
CHL-a  
Turbidity  
Salinity  
Precip.  
Wind

# Hourglass Approach

“Blind” Analysis  
of All Data

Verify Nature of  
Relationships,  
Identify Mechanisms

Explain Why  
Mechanism Important  
Across Space

Time-Series Analysis  
of Mechanisms  
and Relationships  
(e.g., GAM)

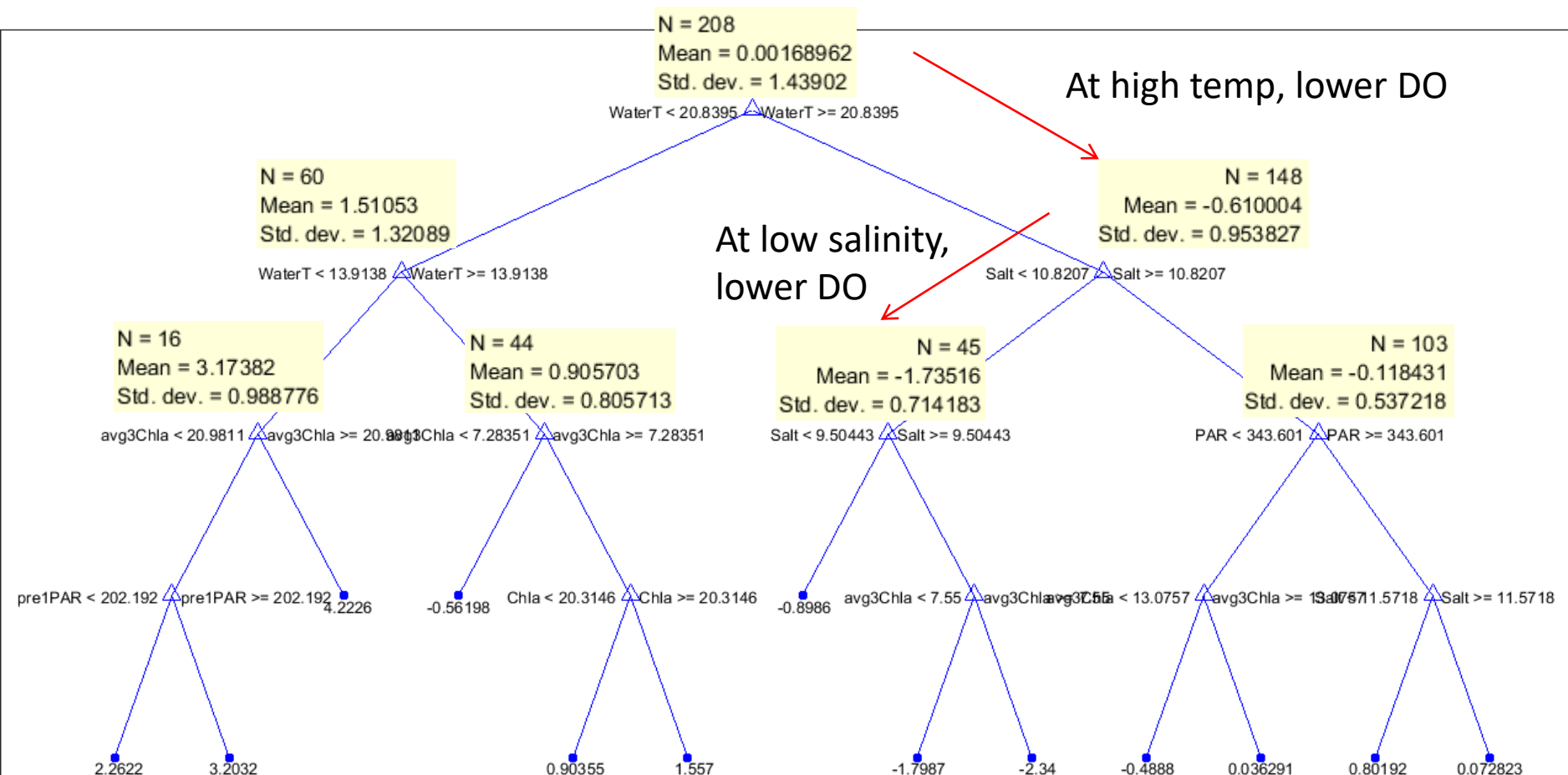
CART (machine learning)  
All Stations and Time Periods  
Key controlling Variables Identified

Spatial Clustering Analysis  
All Stations

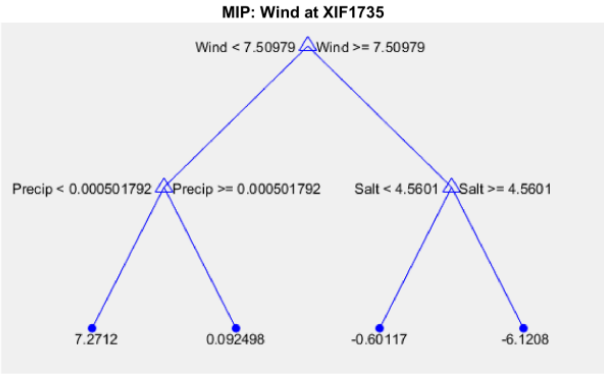
Load, Nutrient Relationships

# CART to Discern Key Variables Driving Residual DO

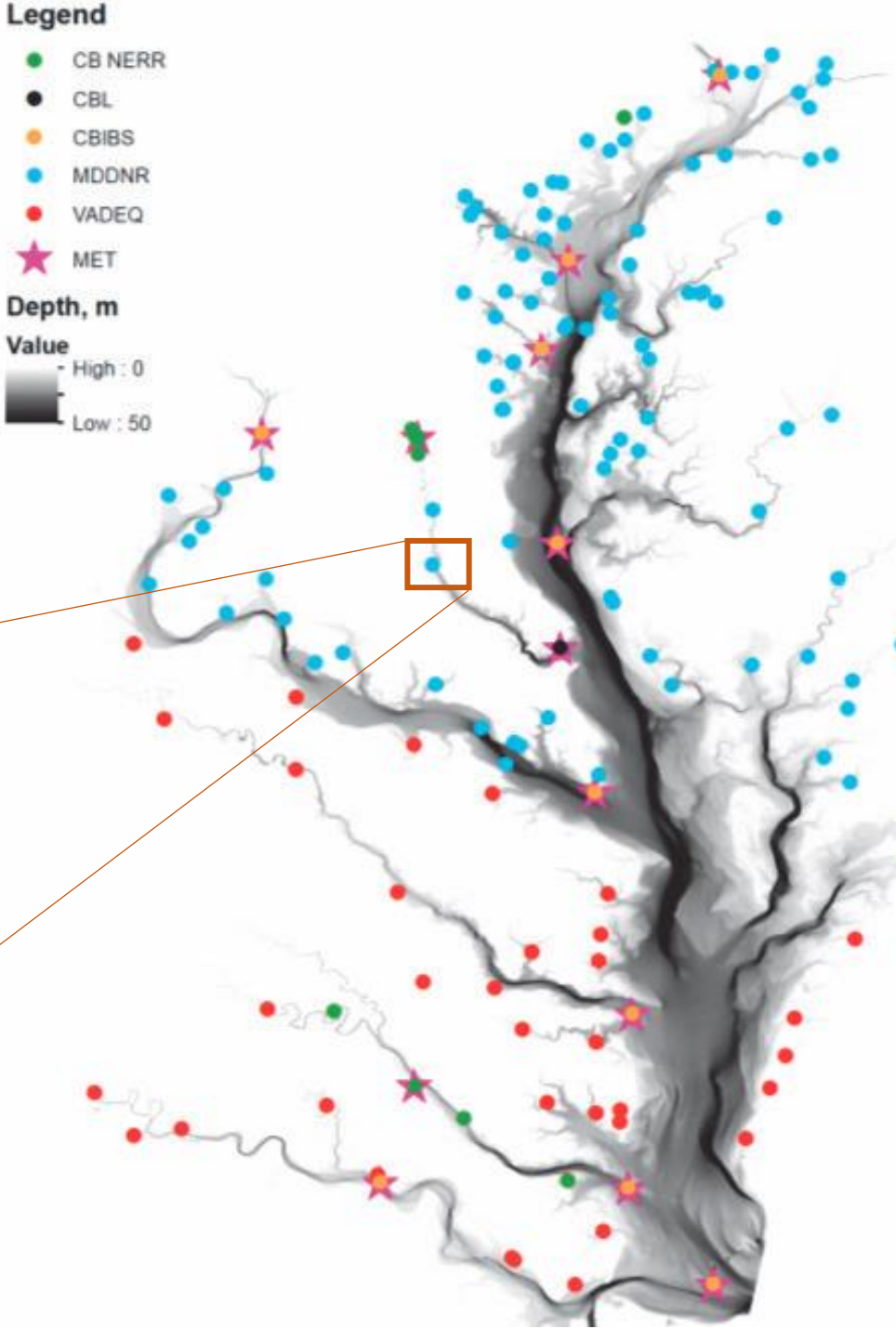
Example: Regression Tree for **DO residual** (mg/L) at Little Monie Creek



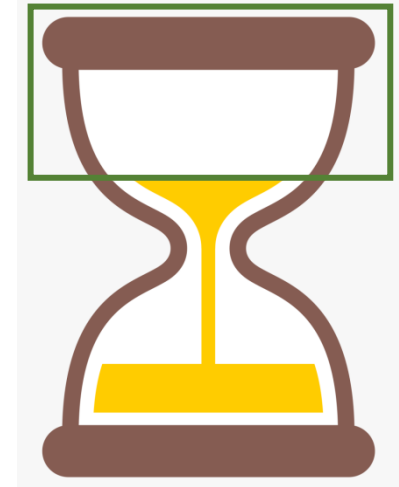
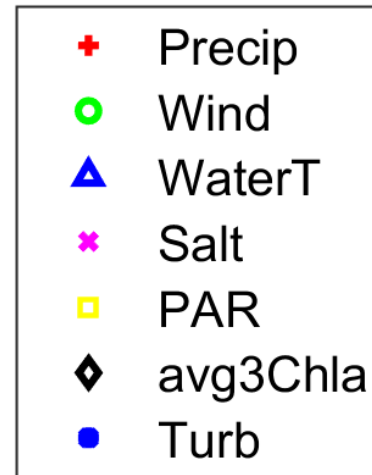
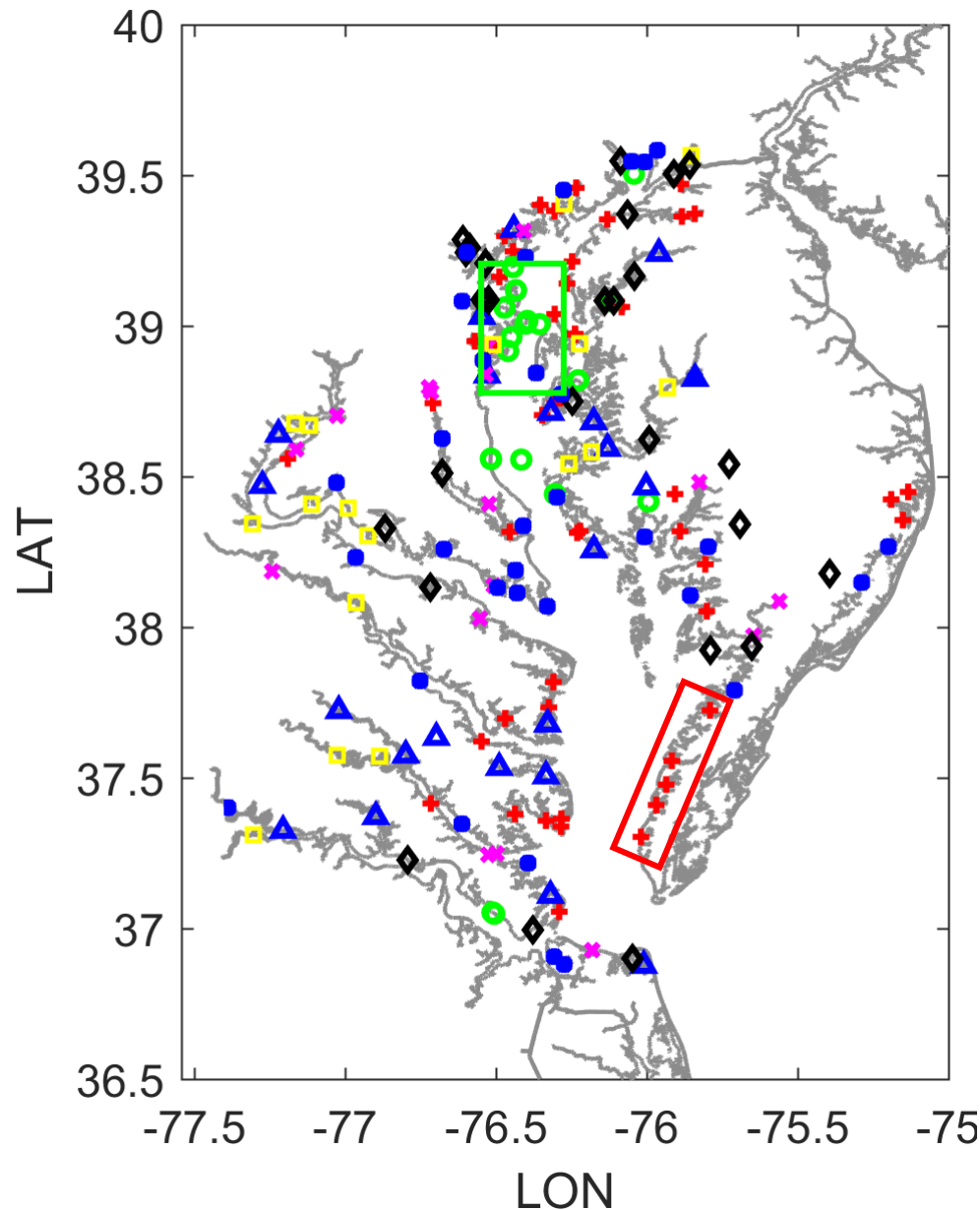
# Application of CART Across Stations



Temp  
PAR  
CHL-a  
Turbidity  
Salinity  
Precip.  
Wind

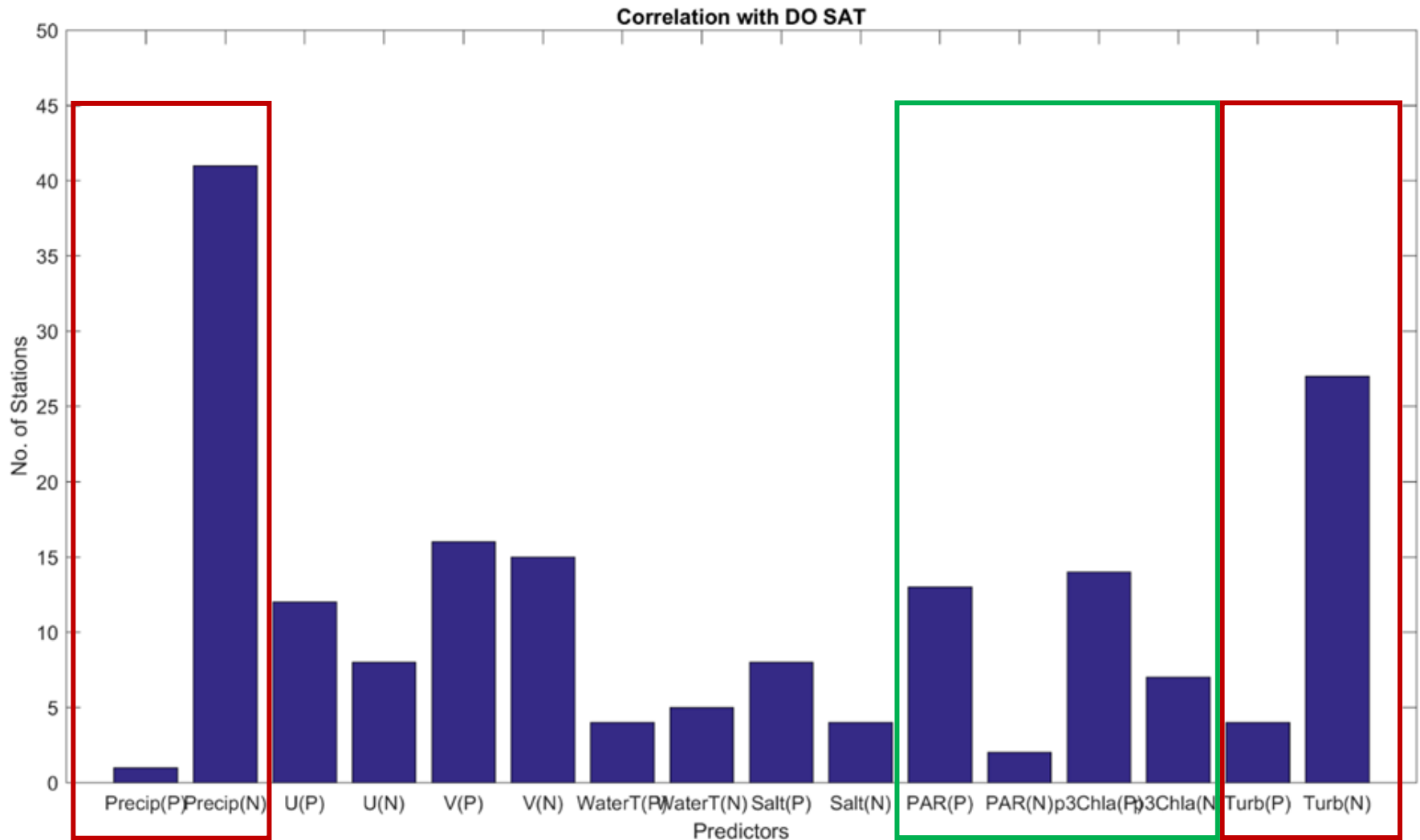


# Most important predictor, by station



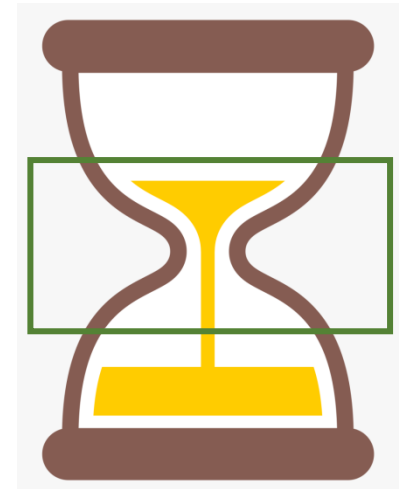
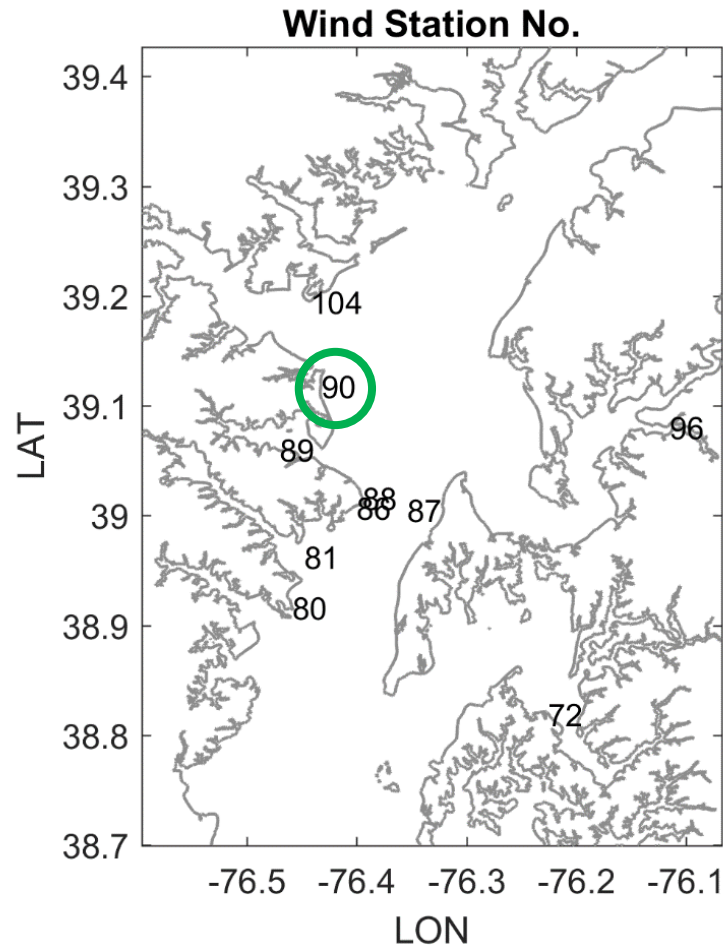
- Diversity of Important variables
- Specific clusters
  - Wind in upper mainstem
  - Precipitation in lower Bay ES
  - Chl-a upstream Baywide

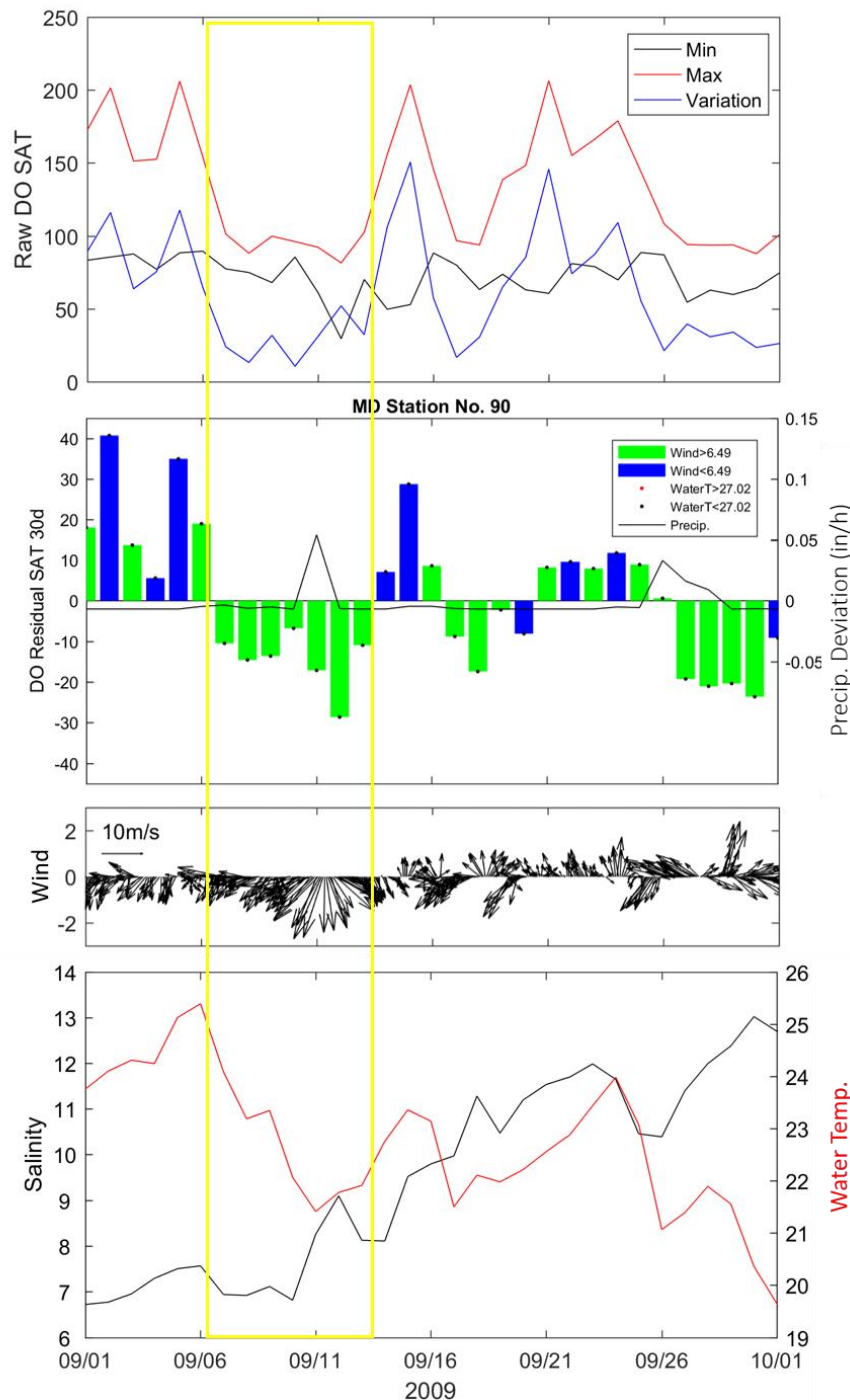
# Frequency of most important predictor



# Time Series of sites with wind as MIP

*How does wind impact DO?*

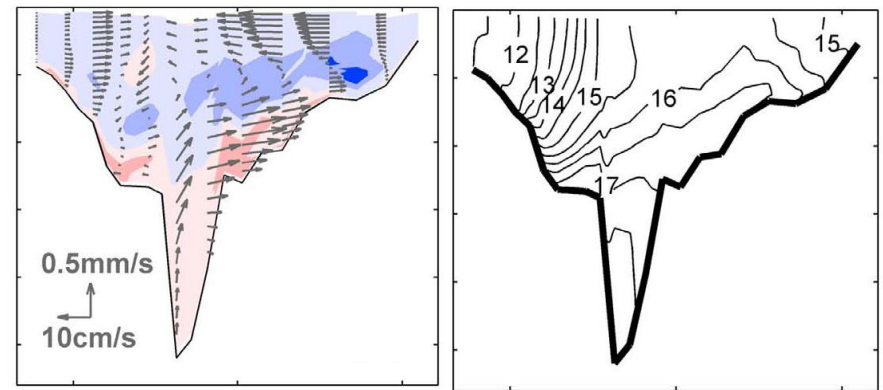




**Primary Hypothesis:** Wind coming from the north generates Ekman transport/counter-clock lateral circulation (looking upstream), which brings relatively salty water from the eastern shore to the western shore sites

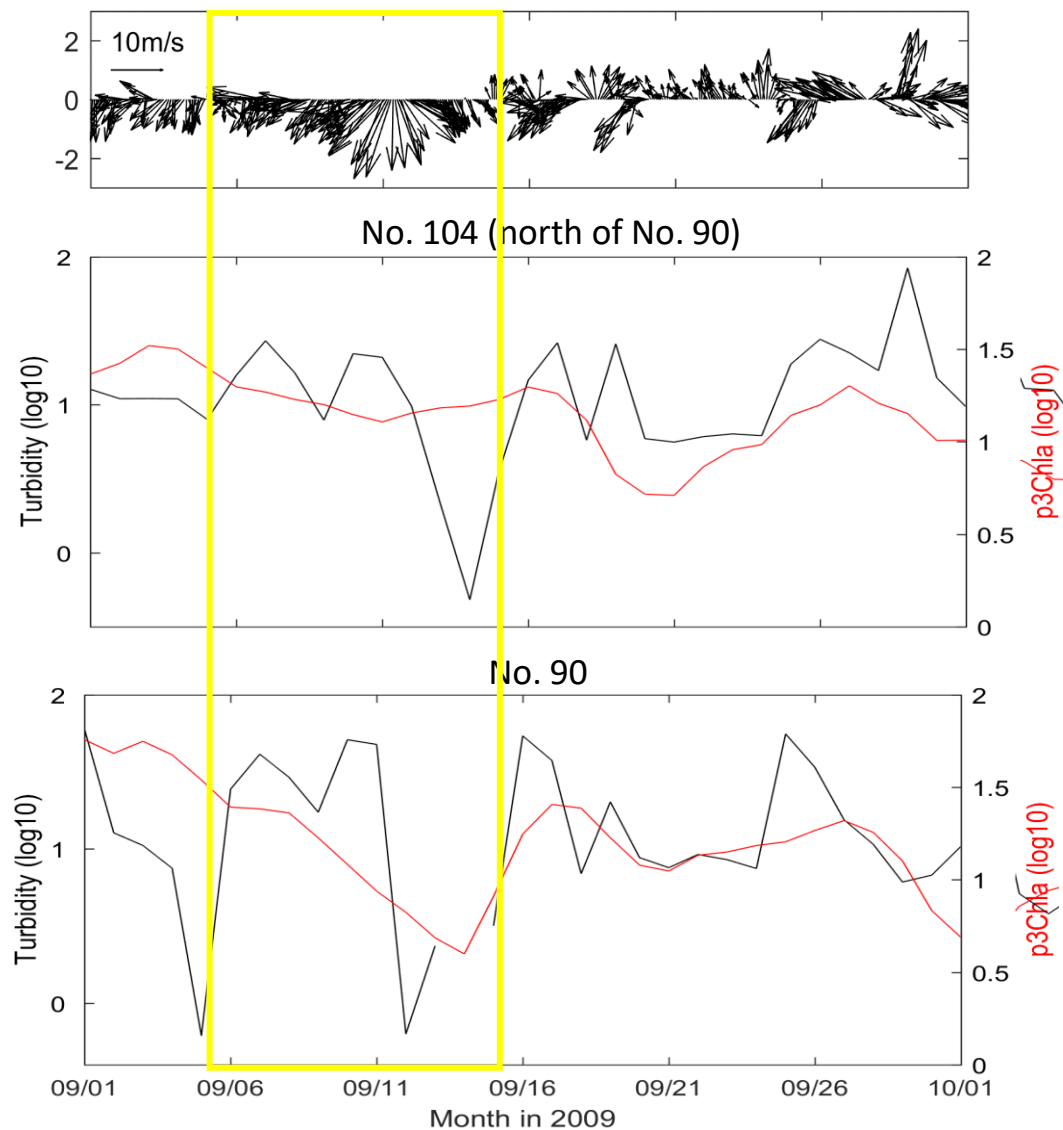
Alternatives for lower do residual: (a) Wind transports upstream organic matter downstream. (b) Air-sea exchange (min and max ~100% saturation)

**Wind coming from the north**



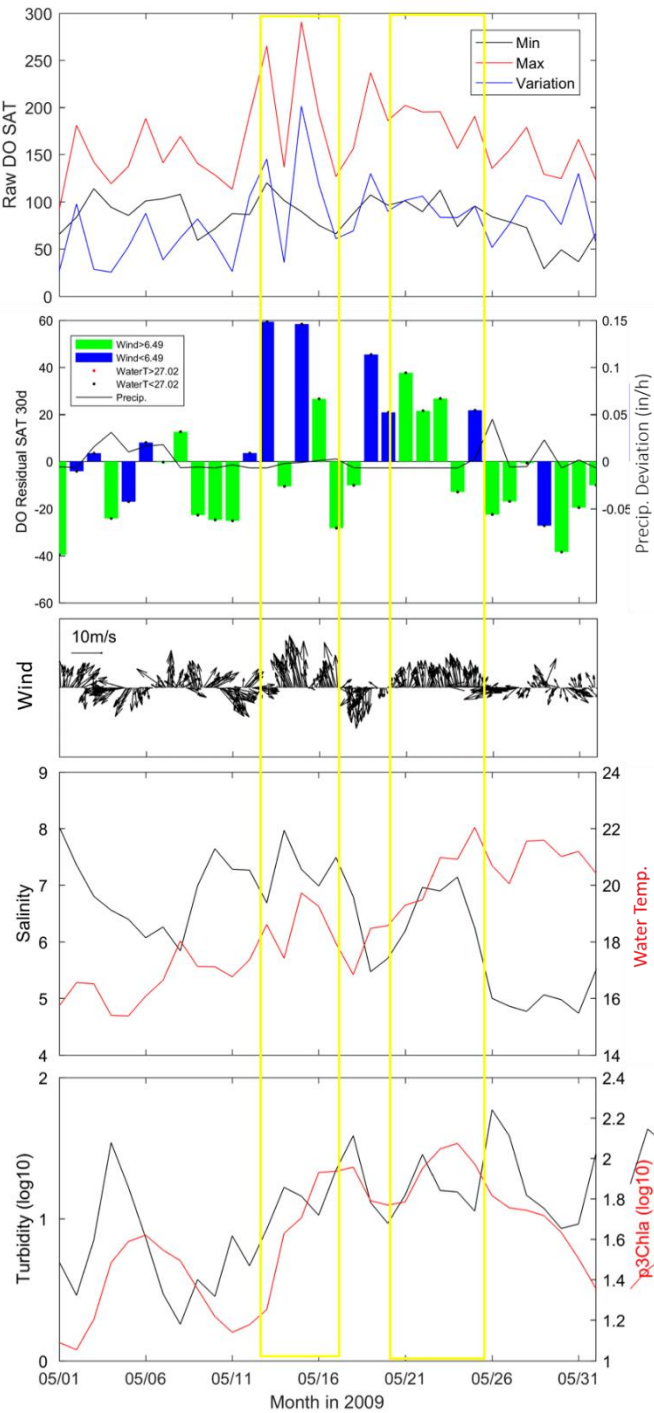
(Looking upstream)

Under strong North Wind, Chlorophyll-a declines and turbidity increases = lower DO

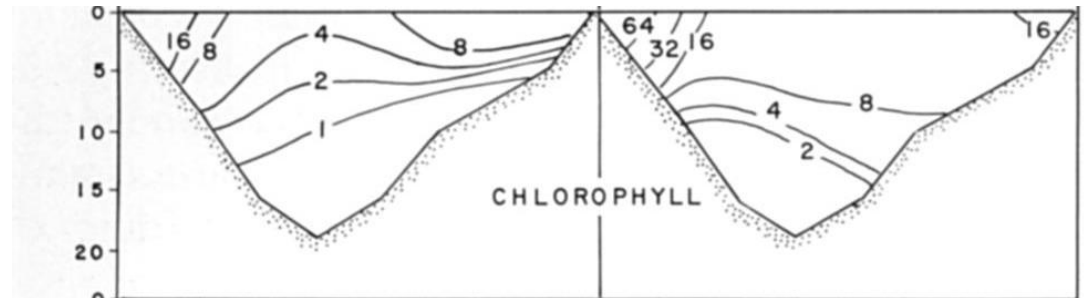


Primary Hypothesis: Wind from the south leads to the opposite effect: clockwise lateral circulation and upwelling in the western shore, which leads to salinity increase and higher DO SAT, due to mixing of nutrients to support blooms.

Previously reported in mainstem CB



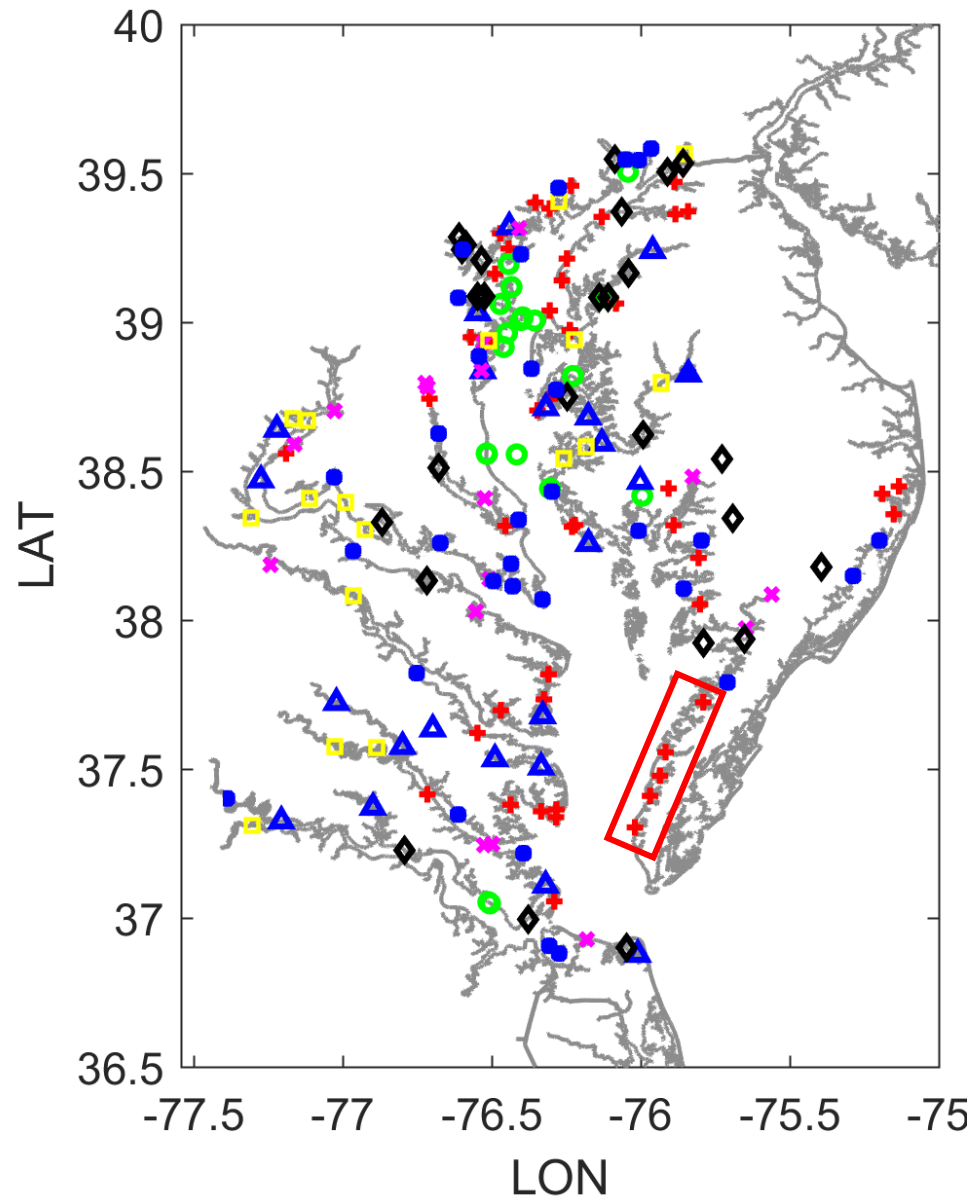
South Wind Event



Malone et al. 1986

# How does precipitation effect DO?

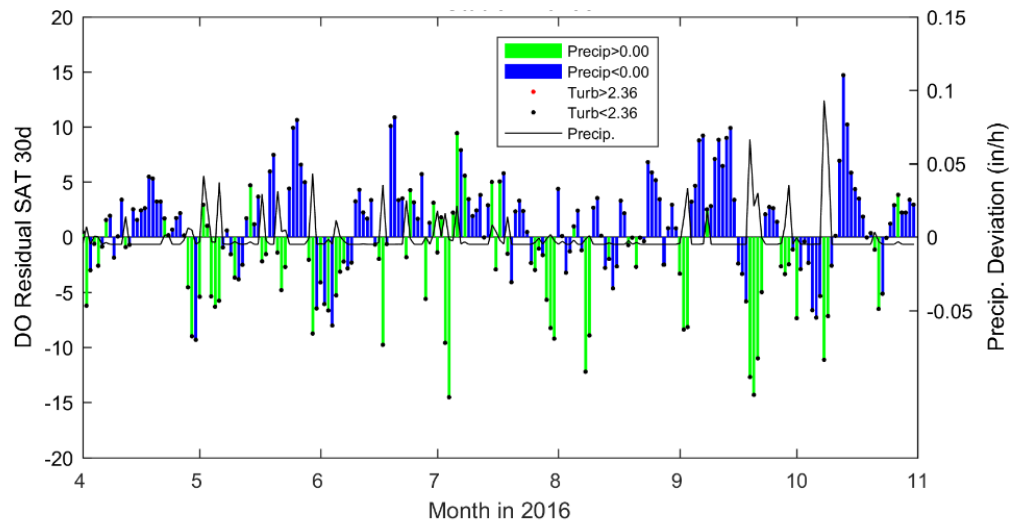
*Eastern shore cluster with precipitation as MIP*

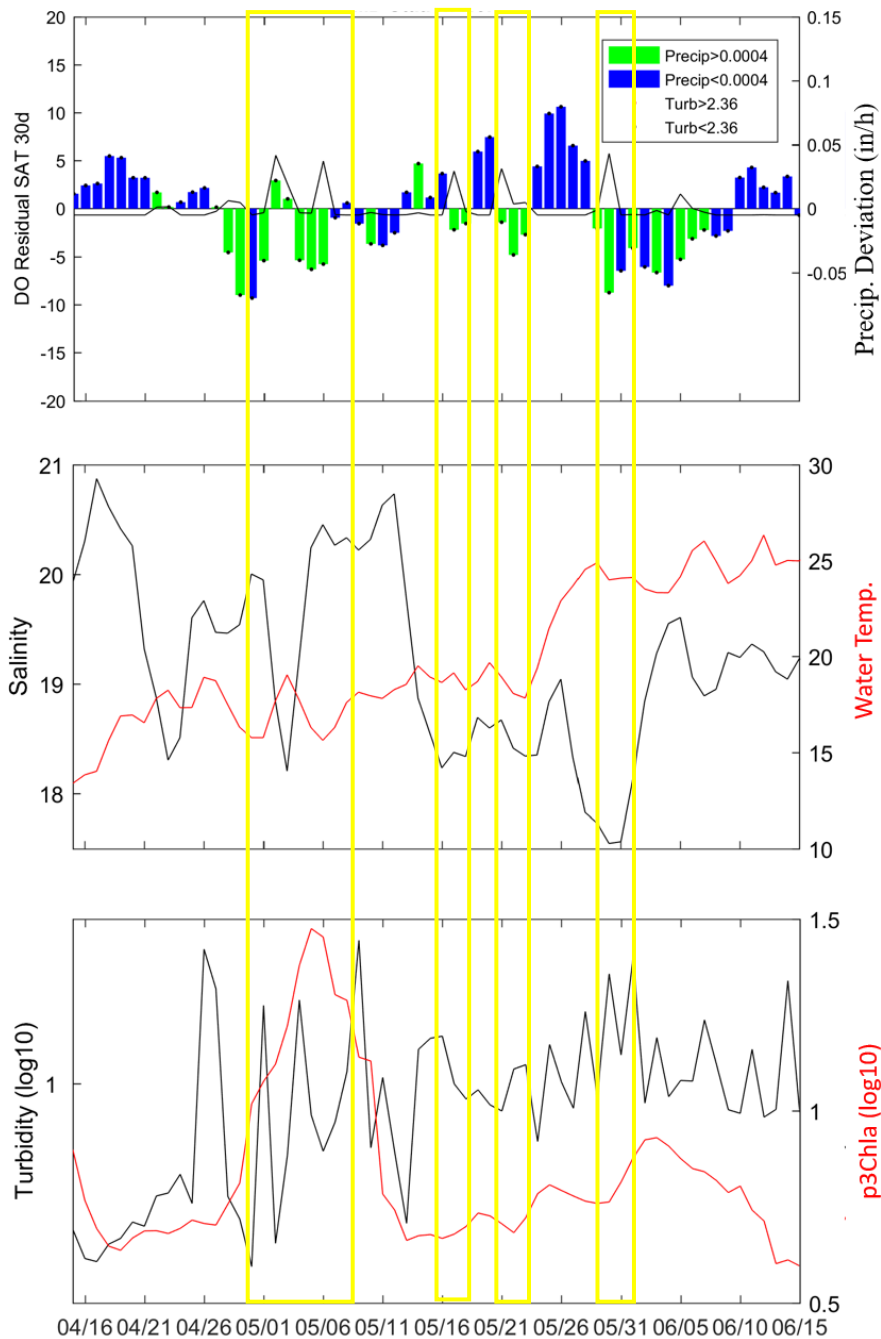


DOSAT is sensitive to precipitation change.

Precipitation divides positive/negative DO SAT.

Precip. Cluster Station No.	Split value of Precip. (cm/day)	Mean DOSAT above	Mean DOSAT below
VA 7	0.0366	-4.9282	1.4560
VA 12	0.1097	-3.8185	1.4318
VA 32	0.0244	-3.2516	1.7603
VA 34	2.0320e-05	-2.3140	1.8397
VA 35	0.1219	-3.6525	1.3514



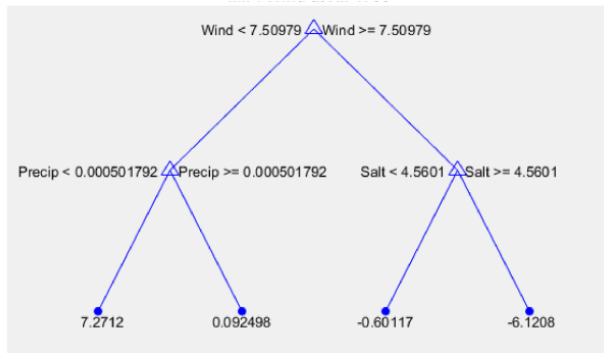


*Primary Hypothesis:* High precipitation events deliver nutrients into the rivers, leads to phytoplankton growth with increased Chla, then more respiration consumes DO, which eventually lead to low DO SAT.

*Alternative:* (a) Increasing water Temp. leads to the decreasing trend of DO SAT, (b) elevated flow leads to organic matter input (to increase respiration) and higher turbidity (to slow photosynthetic oxygen production)

# CART to Explain Hypoxia Duration in Shallow Waters

## Duration of hypoxia



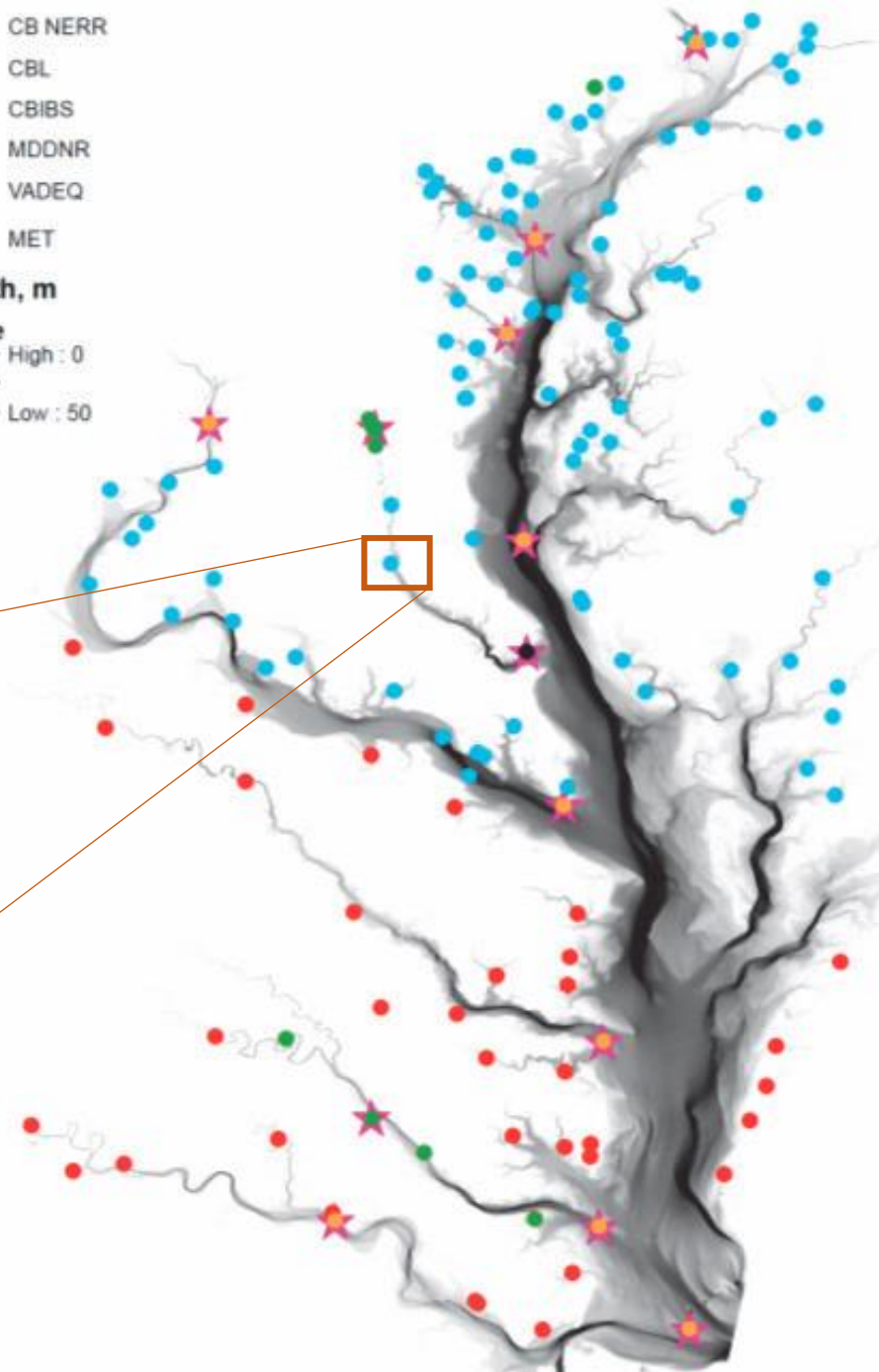
Temp  
PAR  
CHL-a  
Turbidity  
Salinity  
Precip.  
Wind

## Legend

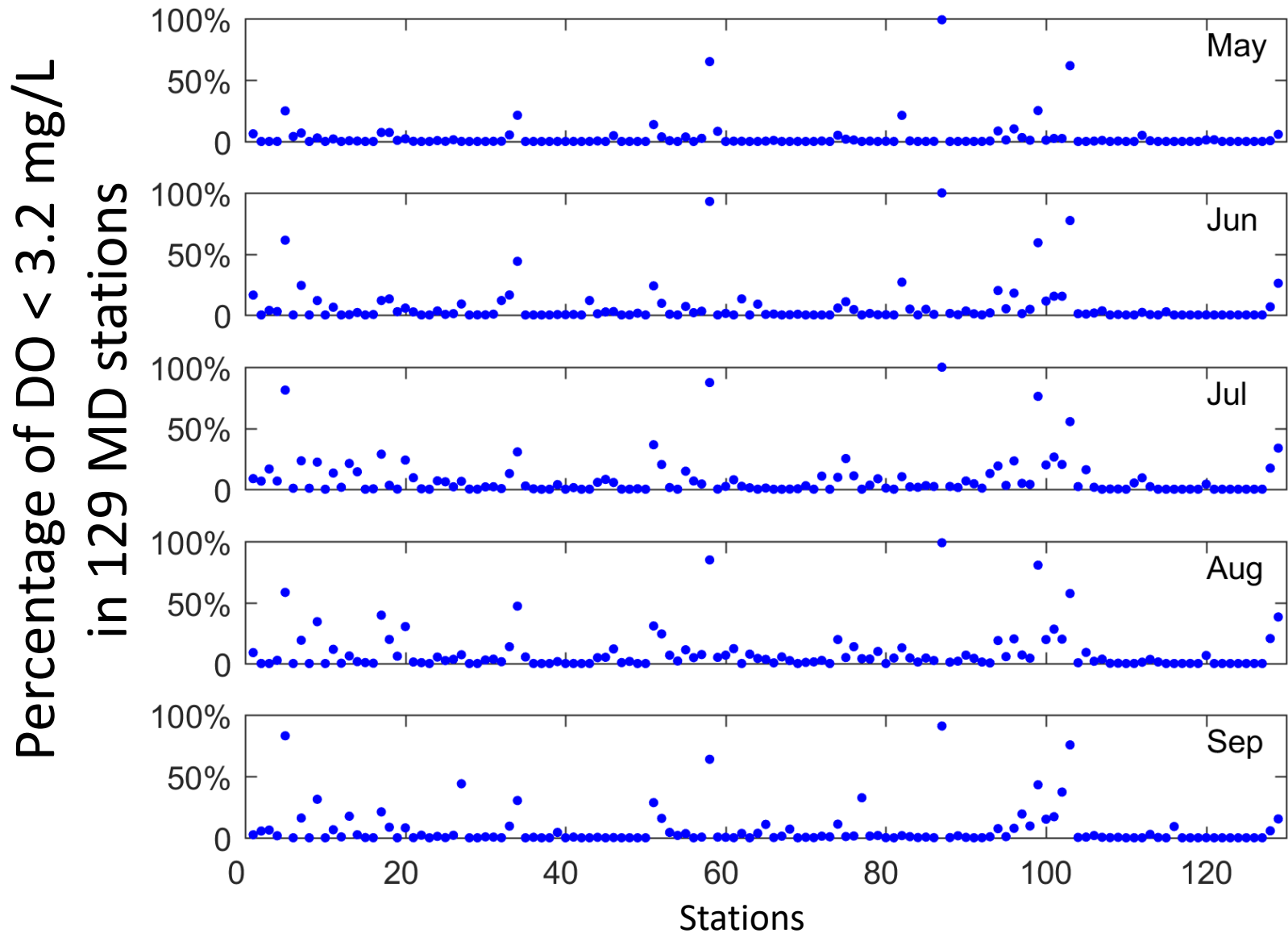
- CB NERR
- CBL
- CBIBS
- MDDNR
- VADEQ
- ★ MET

## Depth, m

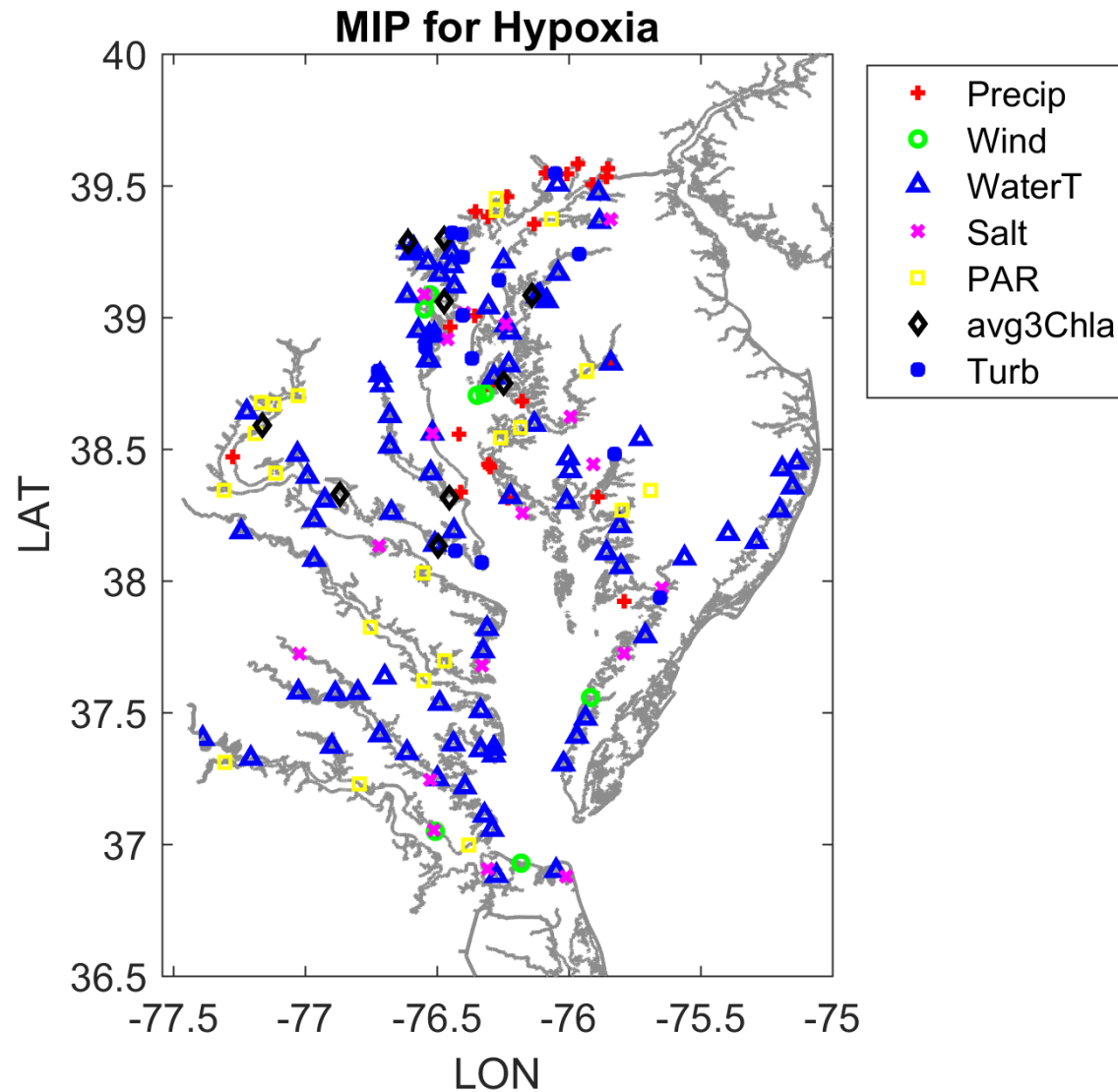
Value  
High : 0  
Low : 50



~Half of the MD stations experienced moderate hypoxia, DO < 3.2 mg/L

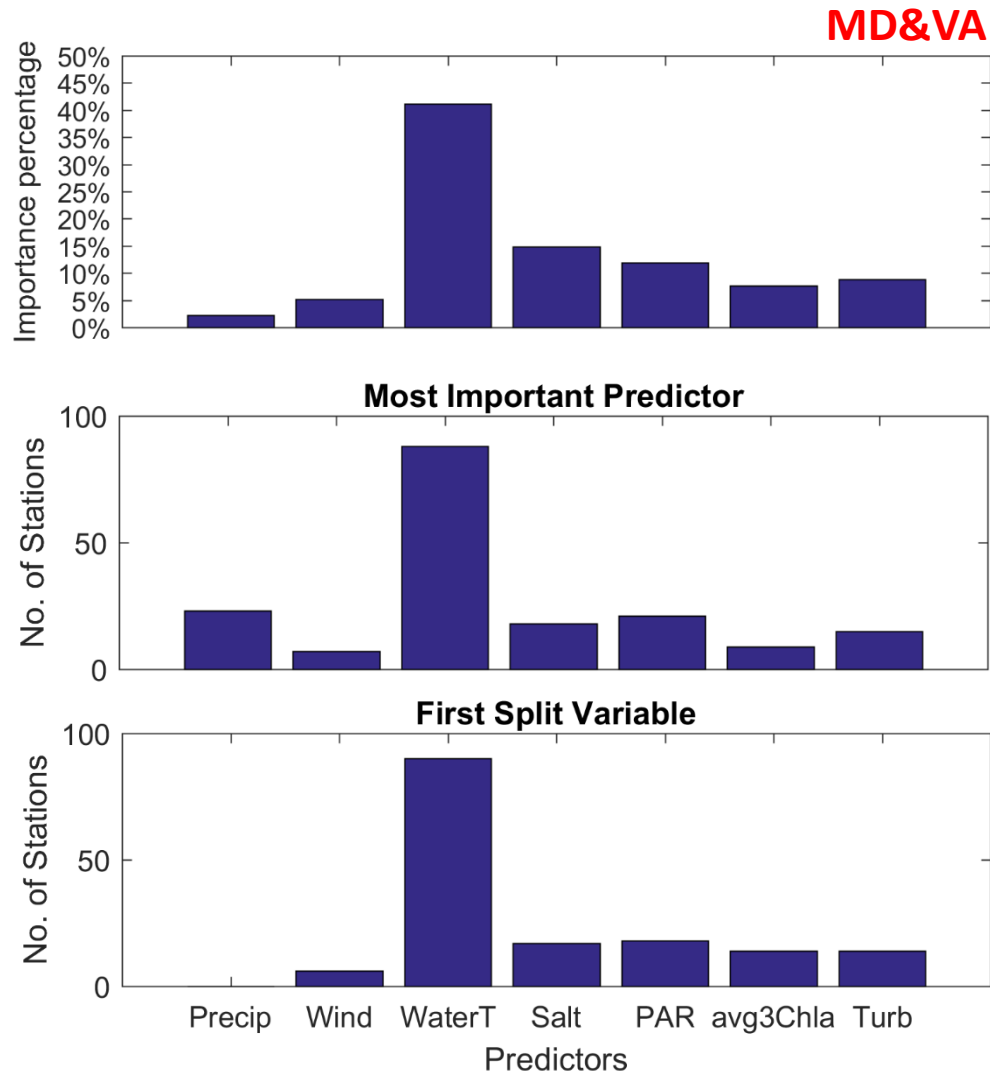


Water Temp. is dominant at most stations except  
some upper bay/river stations.

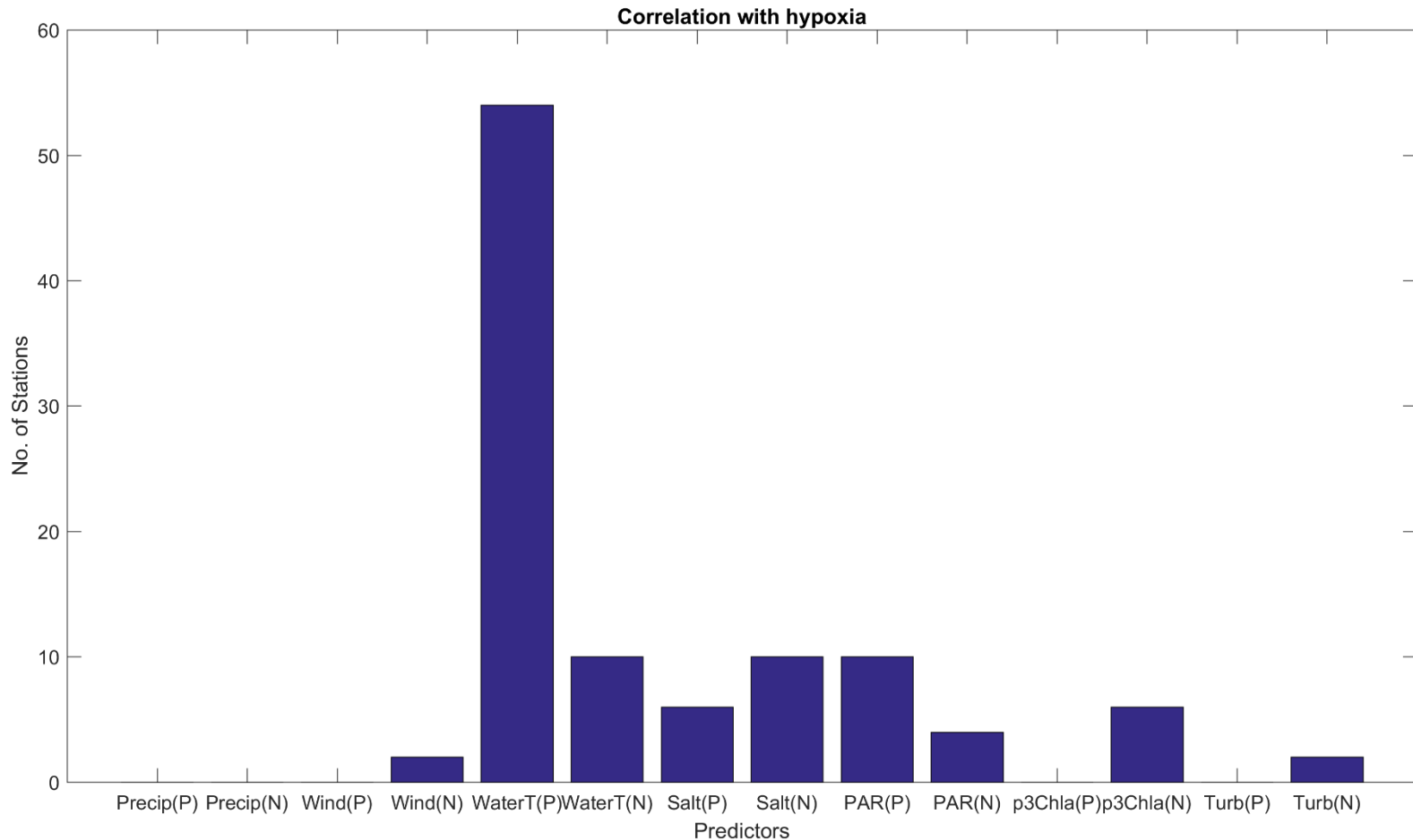


## CART for hypoxia (<4.8 mg/L)

Water Temp. is the most important controller.  
The 2<sup>nd</sup> and 3<sup>rd</sup> MIP are salinity and PAR for hypoxia.



As expected, higher water temp. leads to stronger hypoxia. Lower Chla connects with stronger hypoxia.



# Conclusions to Date

- Diversity of controls on oxygen variability, regional similarities in controls
- Climate-relevant variables are key drivers of oxygen and hypoxia (Temp, precip)
- Chlorophyll-a effect may be both positive and negative for DO

## Next Steps

- Connect CART analysis with discrete data (TSS, nutrients, loads, etc.)
- Use GAMs to Understand relationship and relative role of forcing variables (and to confirm cart results)

