# Cost-Effective Spatial Data Interpolation

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# **Questions to Discuss**

- 1. What is the cost-effective level of effort in conducting interpolation?
- 2. Can we adequately capture weather, time-of-day & tidal signals to inform choices for interpolation methods?
- 3. How might other data sets or dynamic models enrich the interpretation and analysis of a single day's data?
- 4. Can spatially-rich data be used to leverage long-term observations of water quality to detect trends ?
- 5. How might we improve data sampling methods to provide the most representative data for interpolation while still offering a practical approach for the field crew?

# Data and analysis options



Sparse sampling network used for routine water quality monitoring Data collected 12-20 times per year



DATAFLOW sampling (boatmounted sensors) provide more detailed spatial coverage than longterm monitoring stations Data collected 7-13 times per year

# Spatially detailed data (DATAFLOW) provide additional information for spatial interpolation of water quality



### Different Methods of Interpolation of DATAFLOW

Inverse Distance Weighting of the River (all points included)



Whole River Kriged

Kriged Pieces of the River (all points included)

### Detrending vs. No Detrending

Kriging DO with Detrending DO = *f* (distance from mouth)



Kriging DO with No Detrending (ordinary kriging)



### Difference between Detrended and Ordinary Kriging Results



### Detrending = Challenges



- Goal of developing uniform techniques for detrending multiple data sets seem limited

- Trends were highly inconsistent between months

- Detrending did not improve interpolation in our test

- Can we use other sources of variables in detrending (e.g., model data)?

- Yet residuals are highly useful for evaluating excursions from expected values

### Comparison of Kriging with & without Barriers



#### **COMPARISON OF DATAFLOW AND REMOTE SENSING RESULTS**

#### Lower Potomac River – 20 April 2007 Estimation of Chlorophyll a







SAV Habitat Analysis of Potomac River

# Goal: Understand SAV Habitat Quality & Restoration Potential

#### SAV Habitat Criteria

Potomac River	Water column light requirement (%) = g(turbidity, salinity, chla)	Chlorophyll a (µg/L)	Dissolved Inorganic Phosphorus (mg/L)	Dissolved Inorganic Nitrogen (mg/L)
Tidal Fresh	>13	<15	None*	None
Mesohaline	>22	<15	<0.01	<0.15

#### 2006 Potomac River (Mesohaline and Tidal Fresh) DataFlow Cruises





SAV Habitat Hotspots – Mesohaline Potomac 2006 Spatially Intensive Shallow Water Quality Monitoring

# **Cumulative Frequency Diagrams**

Spatially Intensive Shallow Water Quality Monitoring of the Potomac River



### Patuxent River % Observations Meeting SAV Habitat Criteria

(nutrient criteria excluded)



### **Conclusions and Recommendations**

- Interpolation techniques that go beyond basic IDW, create significant time costs, even when tasks are automated with scripts.
- More complex spatial interpolation in some cases, but not others, appears to provide substantially improved information over simpler interpolation techniques.
- Expected use of interpolation results as well as time and budget constraints can inform the level of effort that is warranted or feasible.
- The use of barriers, detrending and other techniques, raises the risk of imposing conditions on data that may not represent physical reality.
- On the other hand, failing to account for barriers introduces localized misrepresentations of data (e.g., near peninsulas).
- Failure to detrend dissolved oxygen may create spatial gradients that represent time effects rather than spatial variability.
- Alternative sampling regimes could help to minimize time-of-day effects on data
- Model data output could be a useful addition to detrending functions to better establish "expected values" given time of year and antecedent rainfall conditions if concerns about error can be addressed.

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