



Characterizing spatial and temporal patterns of conductivity in freshwater tributaries within the Chesapeake Bay watershed

Rosemary Fanelli
USGS South Atlantic Water Science Center

Integrated Trends Analysis Team (ITAT)
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Term definitions

Salinity: concentration of salt ions dissolved in water

Freshwater salinization: increased concentrations of salt ions in naturally low saline waters

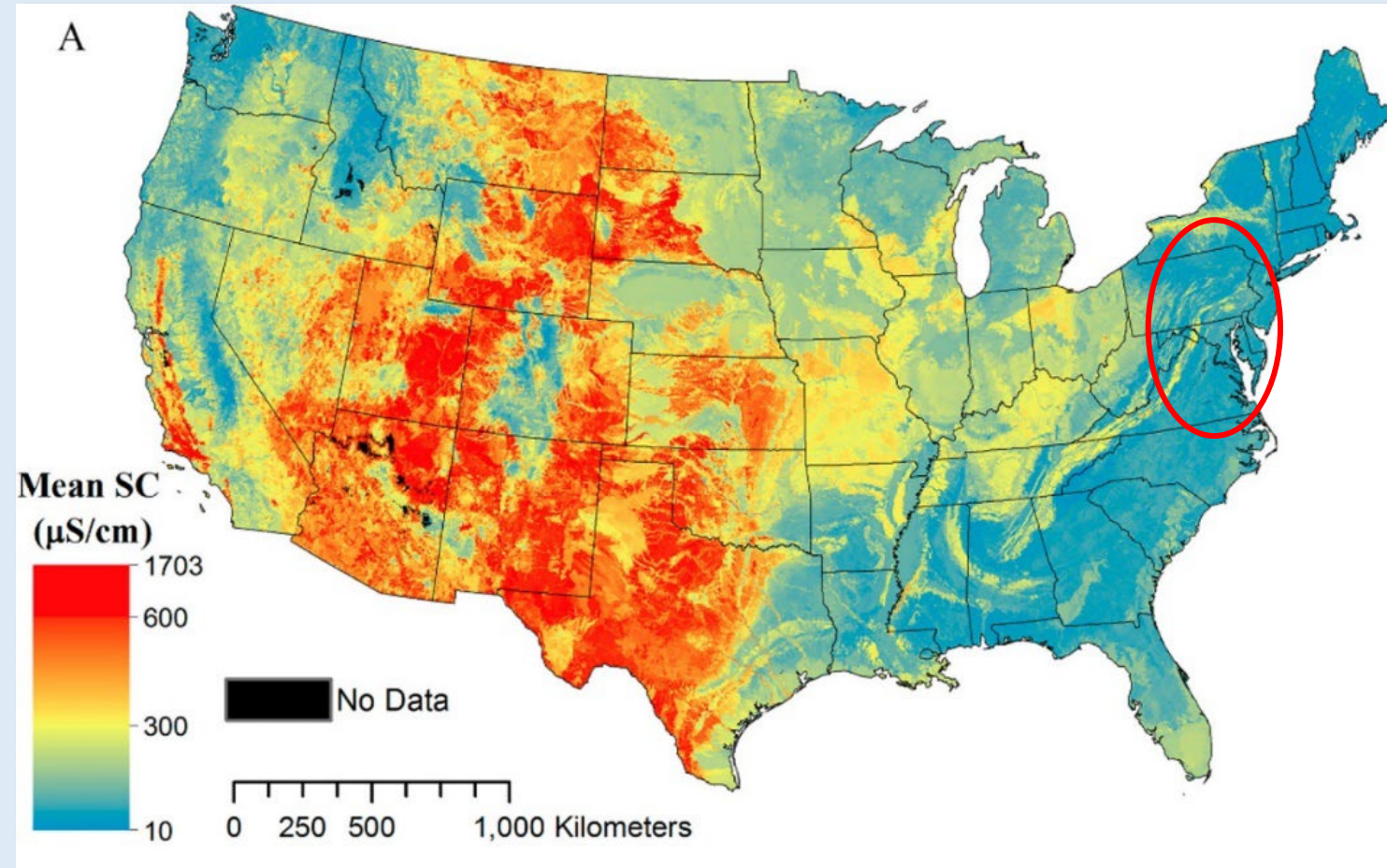
Conductivity: electrical conductance of water (increases with salinity)

Specific conductance (SC): conductivity of 1 cm³ water at 25°C



SC in the CB watershed

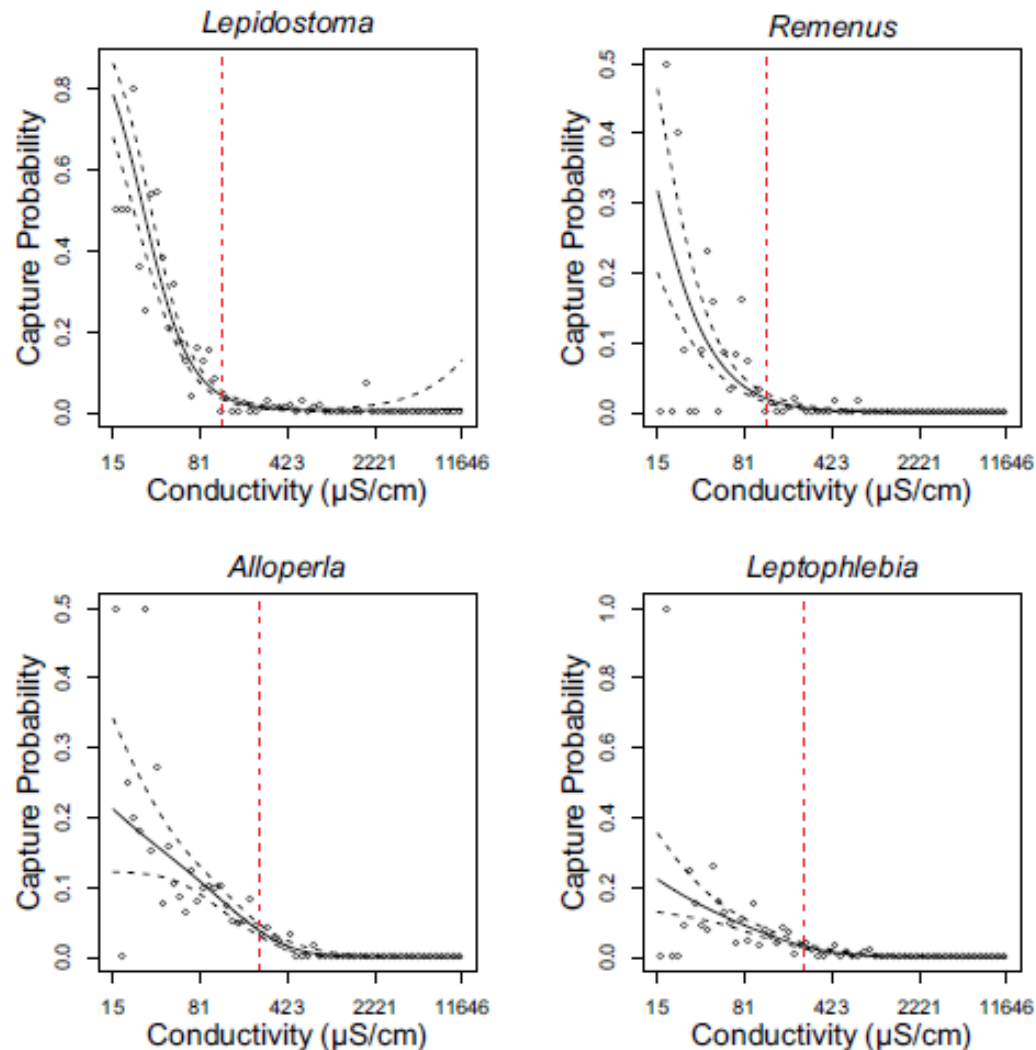
- Usually 20-400 $\mu\text{S}/\text{cm}$
- Natural sources
 - Carbonate lithology
 - Seawater aerosol
 - Precipitation/deposition
- Anthropogenic sources
 - Deicer applications
 - Material weathering
 - Agricultural land use
 - Resource extraction
 - Point source discharge



Olson, J. and S. Cormier. 2019. Modeling Spatial and Temporal Variation in Natural Background Specific Conductivity. *Environ. Science and Technology*. 53, 4316–4325.
<https://pubs.acs.org/doi/10.1021/acs.est.8b06777>

Effects of freshwater salinization

E-2



- Elevated ions can be an ecological stressor to aquatic organisms
- Excess ions can mobilize metals
- Influences biogeochemical cycling
- Can make other contaminants more toxic (6PPD, a synthetic tire compound)
- Human health implications
 - Corrosivity of water infrastructure
 - Water supplies meeting sodium-restricted diet thresholds

Clements and Kotalik, 2016. Effects of major ions on natural benthic communities: an experimental assessment of the US Environmental Protection Agency aquatic life benchmark for conductivity. *Freshwater Sci.* 2016, 35, 126–138. <https://www.journals.uchicago.edu/doi/epdf/10.1086/685085>

Regional stakeholder needs

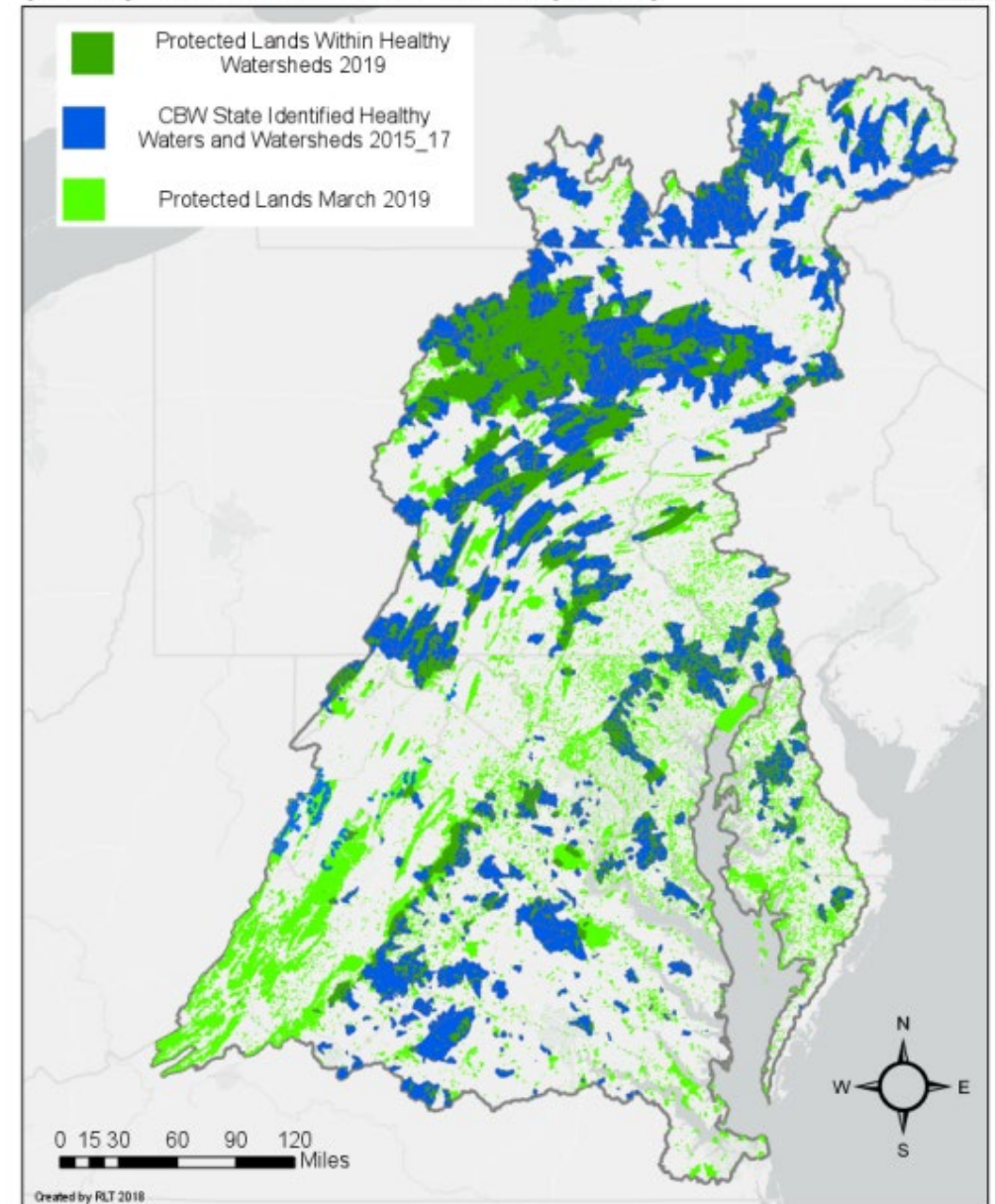
Healthy Watersheds Goal Implementation Team

- Goal is to “is to maintain local watersheds at optimal health across a range of landscape contexts”
- More information on in-stream conditions is needed to assess healthy watershed status and determine vulnerability

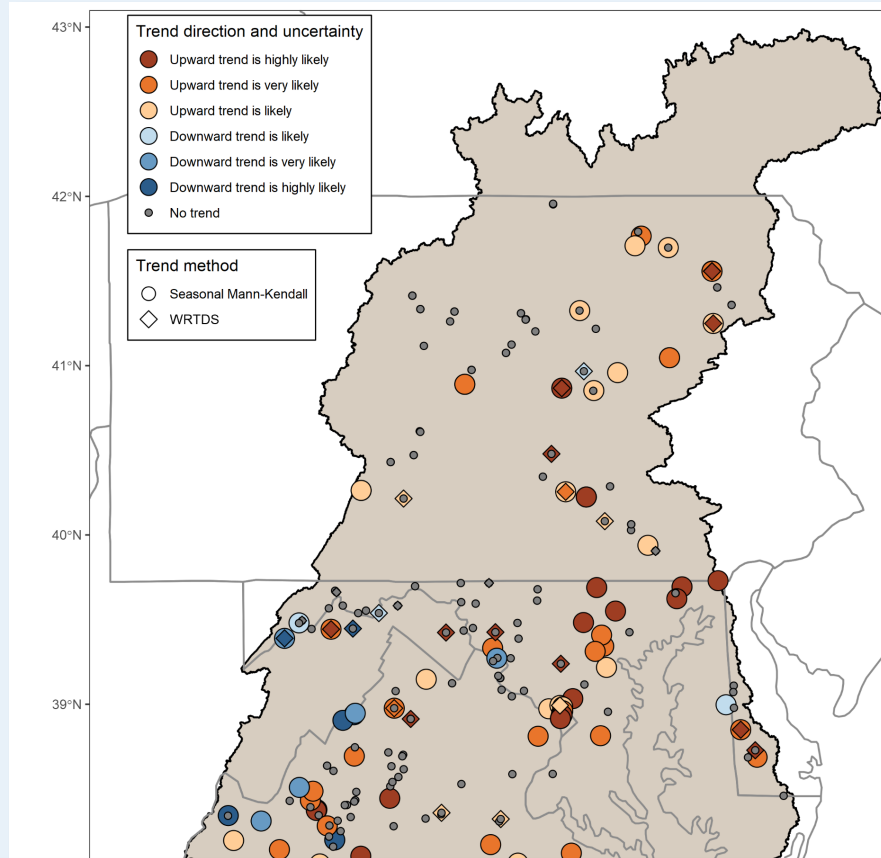
Stream Health Workgroup

- Supporting CBP goal to improve biological conditions in 10% of stream reaches across the watershed
- Need information on what in-stream stressors may be impairing stream health

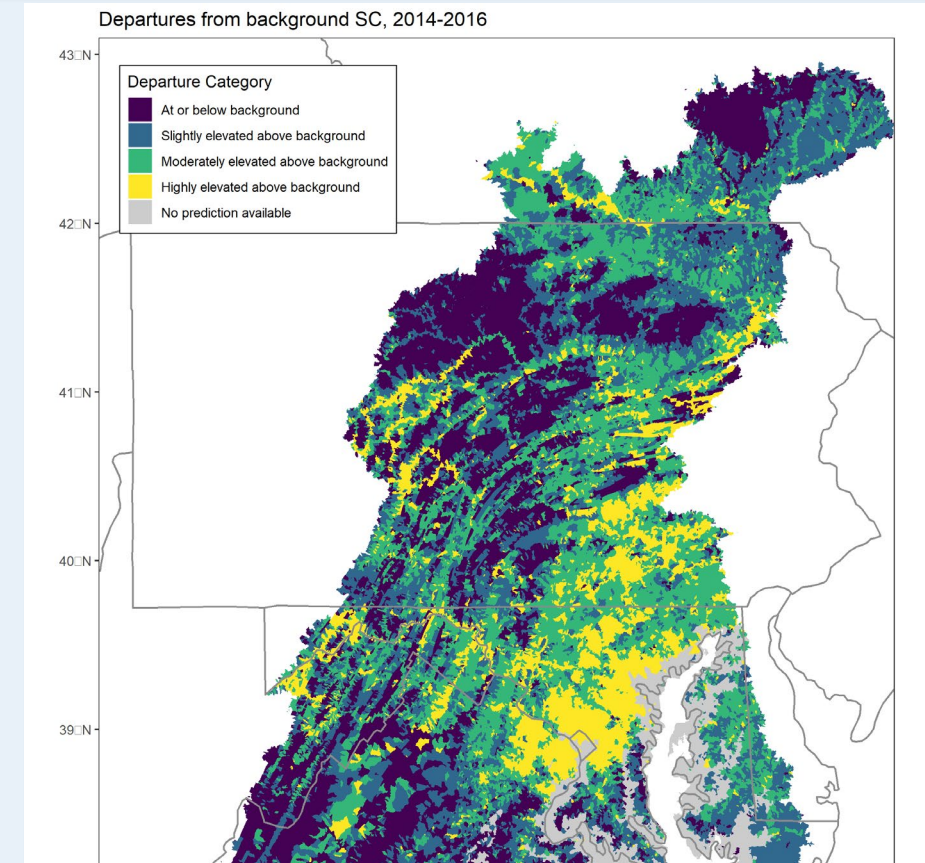
State Identified Healthy Waters and Watersheds (2017) and Protected Lands (2019)



Ongoing regional USGS projects on freshwater salinization



Project 1: Quantify status and trends in SC using discrete observations in freshwater tributaries

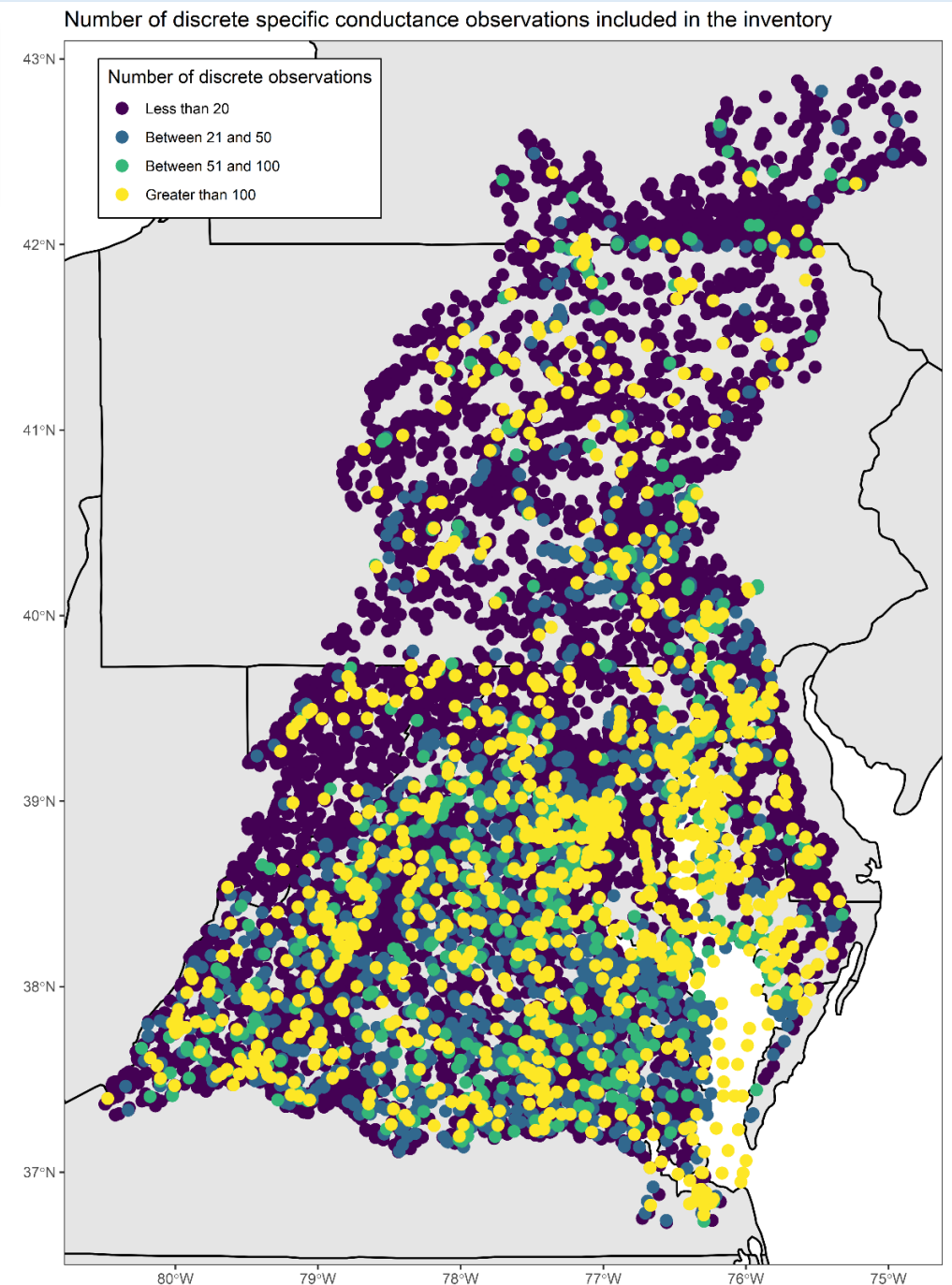


Project 2: Predict SC and departures from background SC across all freshwater tributaries

SC inventory for CB watershed

- Retrieved SC data National Water Quality Portal and USGS NWIS
- 1.2+ million discrete obs at 16,900+ sites
- 89 sites with continuous SC
- Latest data pull: July 2022
- Dataset clean up and unit harmonization
 - QA/QC samples removed
 - Surface water samples retained
 - Units harmonized to $\mu\text{S}/\text{cm}$
 - Database screened for duplicate entries

Fanelli, R.M., Sekellick, A.J., and Hamilton, W.B., 2023, Compilation of multi-agency specific conductance observations for streams within the Chesapeake Bay watershed, U.S. Geological Survey data release, <https://doi.org/10.5066/P98O2HQJ>.





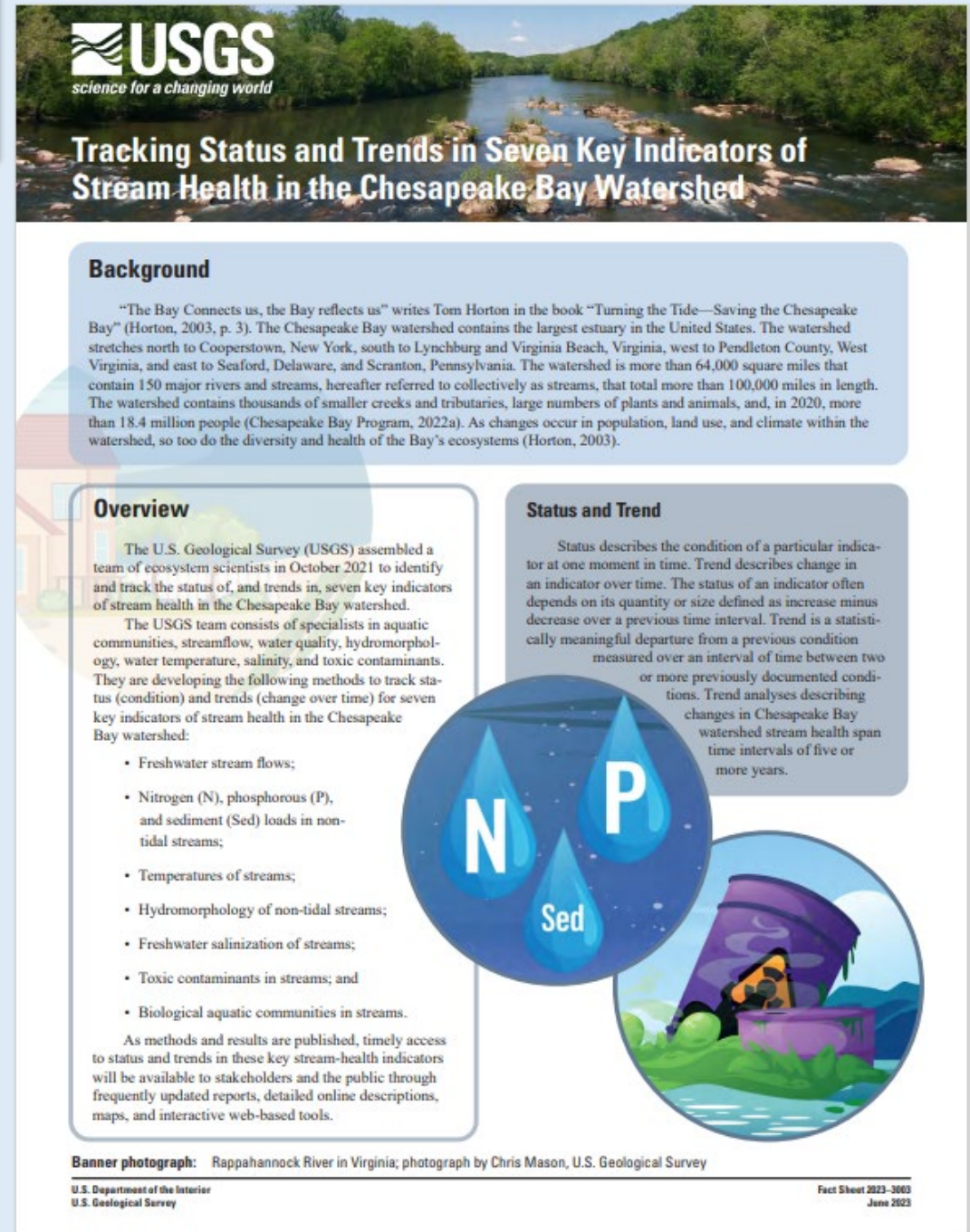
Project 1: Quantify status
and trends in SC using
discrete observations in
freshwater tributaries

Rosemary Fanelli, SAWSC, U.S. Geological Survey
Kaitlyn Elliott, SAWSC, U.S. Geological Survey
(plus many others from the USGS)

CB status and trends

- Multi-year effort to quantify status and trends for seven indicators of **stream health**
 - Biological endpoints (Lindsey Boyle)
 - Physical habitat (Matthew Cashman)
 - Stream temperature (John Clune)
 - Nutrients and sediment (Chris Mason)
 - **Salinity**
 - Flow (Sam Austin)
 - Toxic contaminants (Trevor Needham)
- Initial product: USGS SIR outlining methods and results for status and trends

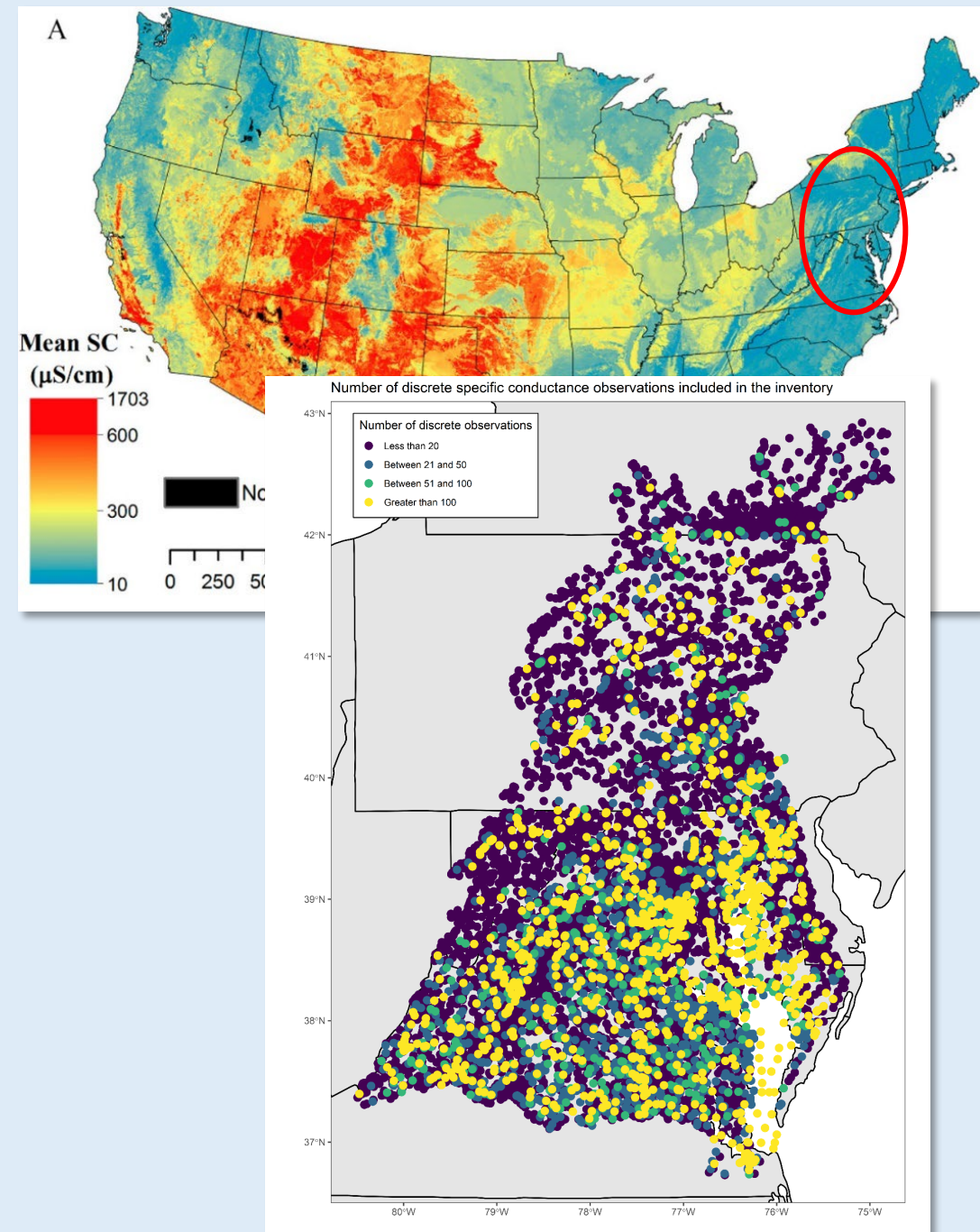
Austin, S.H., Cashman, M.J., Clune, J., Colgin, J.E., Fanelli, R.M., Krause, K.P., Majcher, E.H., Maloney, K.O., Mason, C.A., Moyer, D.L., and Zimmerman, T.M., 2023, Tracking status and trends in seven key indicators of stream health in the Chesapeake Bay watershed: U.S. Geological Survey Fact Sheet 2023–3003, 6 p., <https://doi.org/10.3133/fs20233003>



Computing SC status

- The “status” of all seven indicators was characterized using the most recent data available (usually 2016-2020)
- SC status = 3-year median SC for years 2015-2017
- Sites with at least one sample per quarter for the 10-year trend window (2008-2017)
- Compared to SC background dataset
- Departures from background SC
 - At or below background SC
 - 1-2 times the background SC
 - 2-3 times the background SC
 - Greater than 3 times the background SC

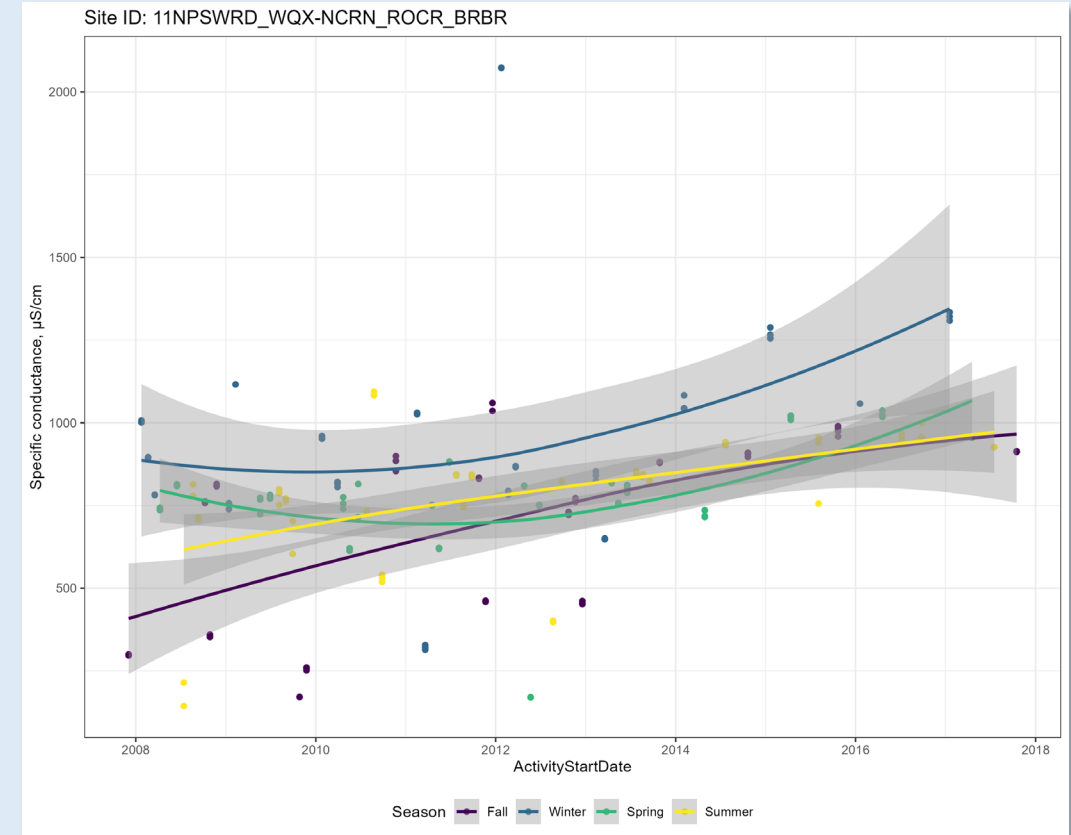
Olson, J. and S. Cormier. 2019. Modeling Spatial and Temporal Variation in Natural Background Specific Conductivity. *Environ. Science and Technology*. 53, 4316–4325.
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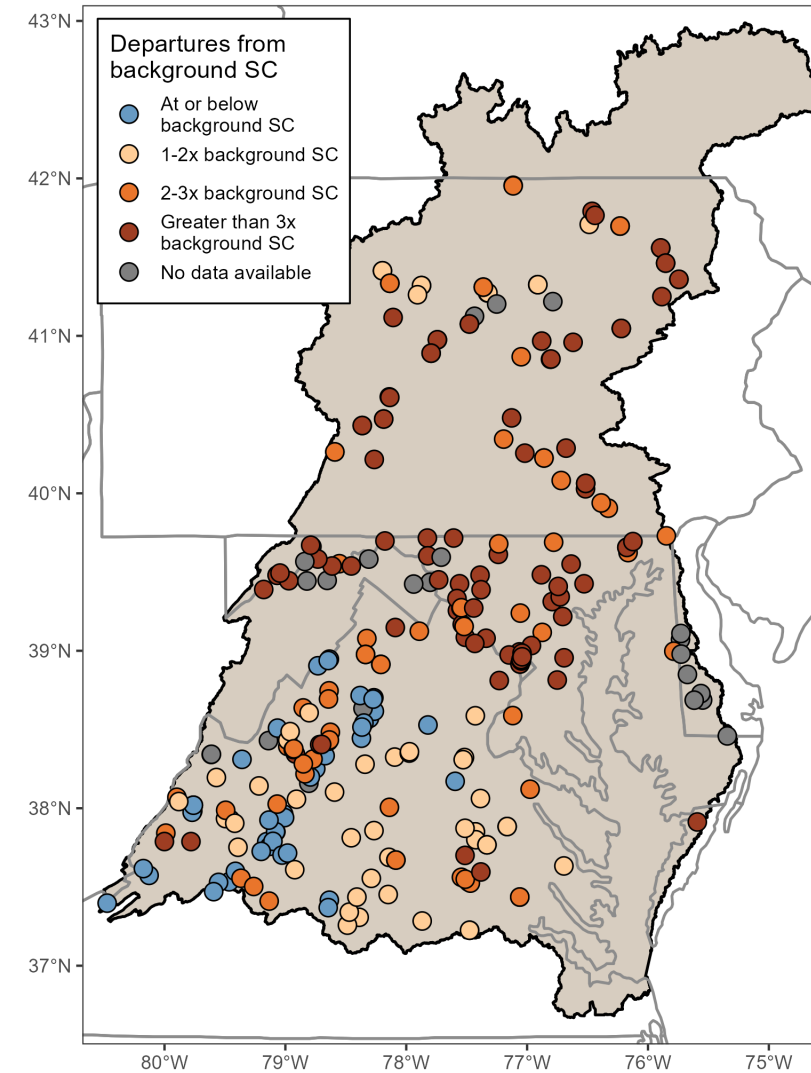
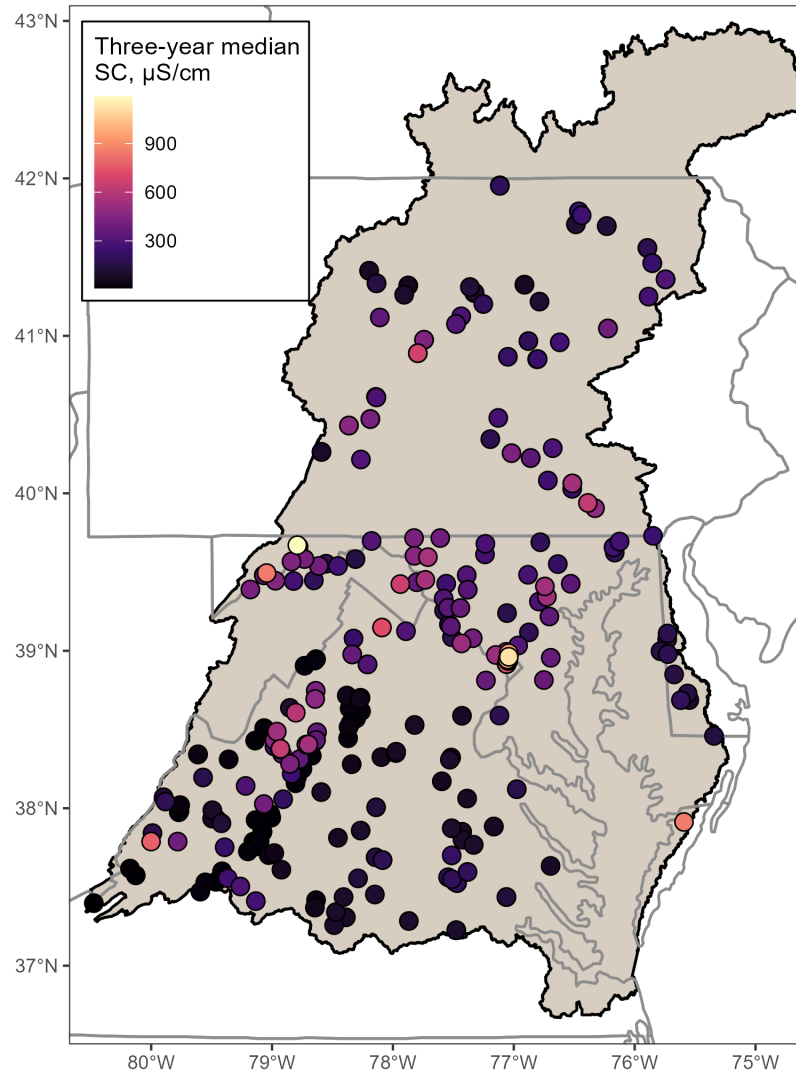
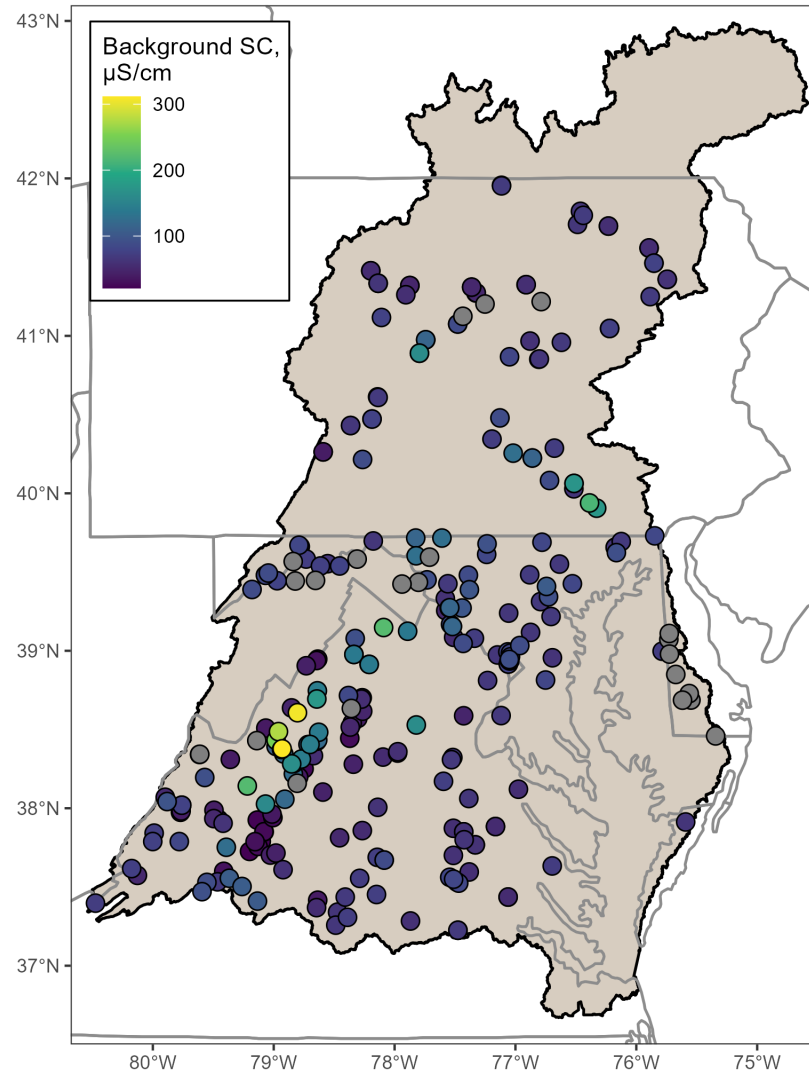
Computing SC trends

- SC trend time period: 2008-2017
- Approach 1: WRTDS
 - Accounts for interannual flow variability
 - Requires discharge record
 - SC data criteria: three samples per season for each of the 10 years in trend period
 - 35 sites qualified for WRTDS
- Approach 2: Seasonal Mann-Kendall
 - Does not require streamflow records
 - SC data criteria: one sample per season for each of the 10 years in trend period
 - 278 sites qualified for SMK
 - Defined criteria for characterizing trend significance and uncertainty

P-value from Seasonal Mann-Kendall test	Theil-sen Slope estimate	Trend category
> 0.2	NA	No trend
< 0.2 and ≥ 0.1	Greater than zero	Upward trend is likely
< 0.1 and ≥ 0.05	Greater than zero	Upward trend is very likely
< 0.05	Greater than zero	Upward trend is highly likely
< 0.2 and ≥ 0.1	Less than zero	Downward trend is likely
< 0.1 and ≥ 0.05	Less than zero	Downward trend is very likely
< 0.05	Less than zero	Downward trend is highly likely



Results: SC status



Results: SC trends

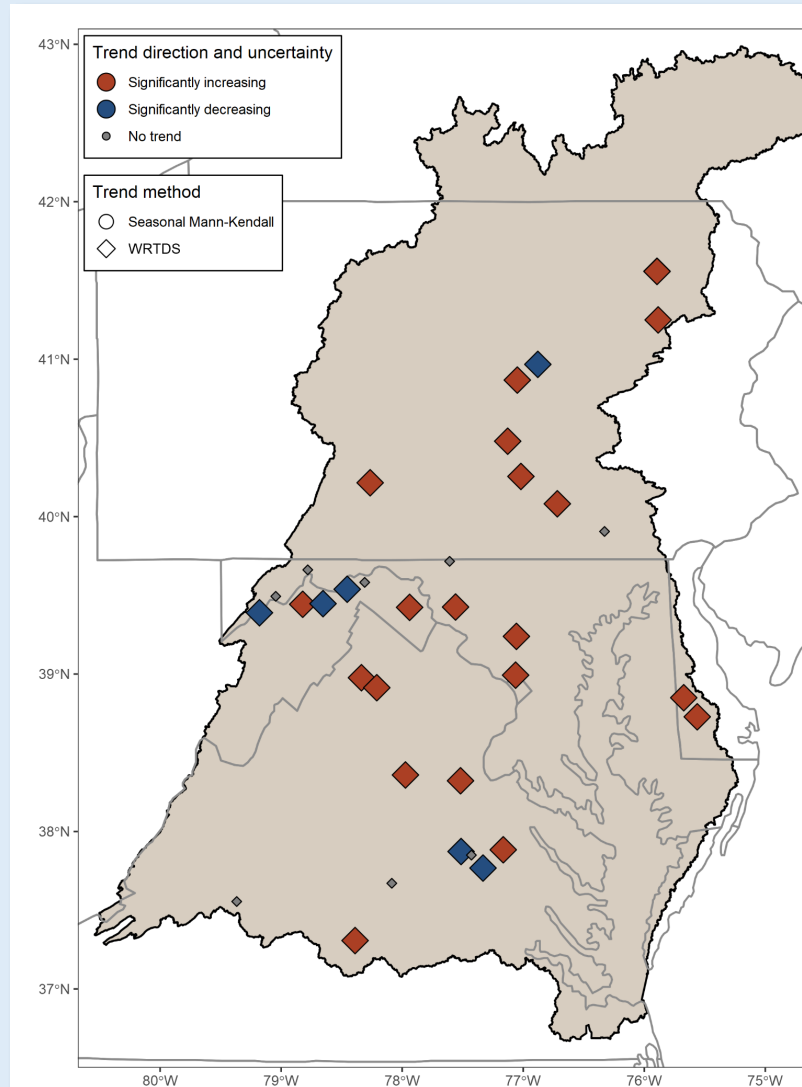
WRTDS (35 sites)

- 60% (21 sites) had inc. trends
- 23% (8 sites) had no trend
- 18% (6 sites) had dec. trends

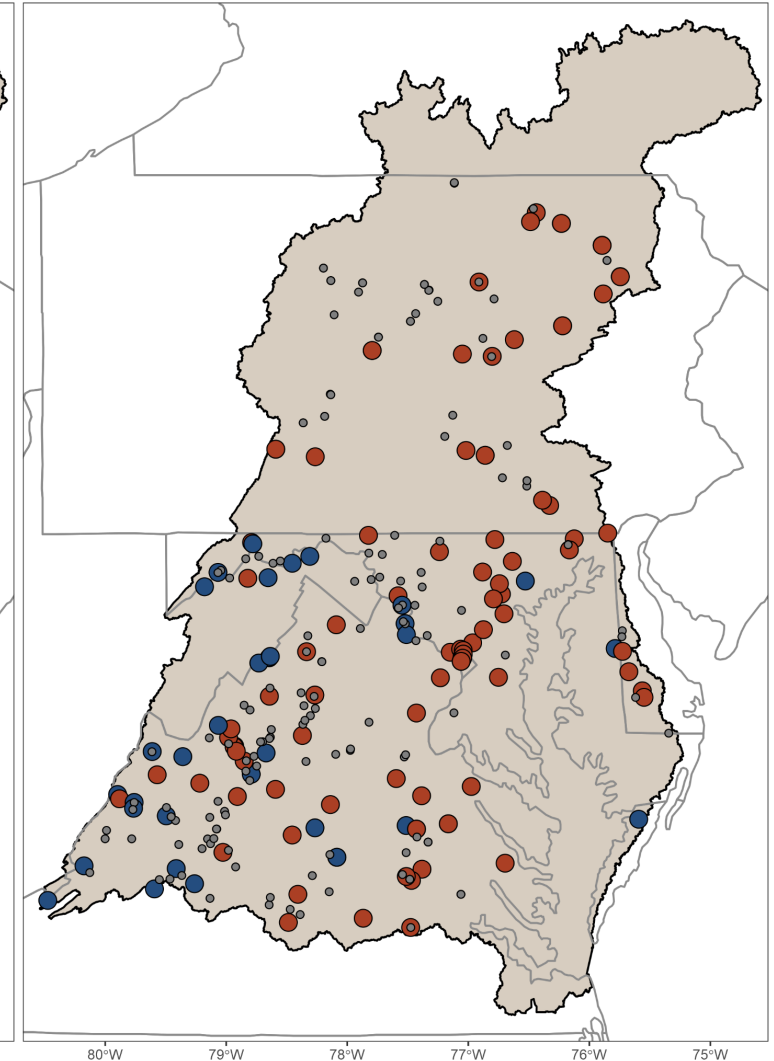
SMK (278 sites)

- 55% (152 sites) had no trend
- 33% (92 sites) had inc. trends
- 12% (34 sites) had dec. trends

WRTDS



Seasonal Mann-Kendall



Results: SC trends

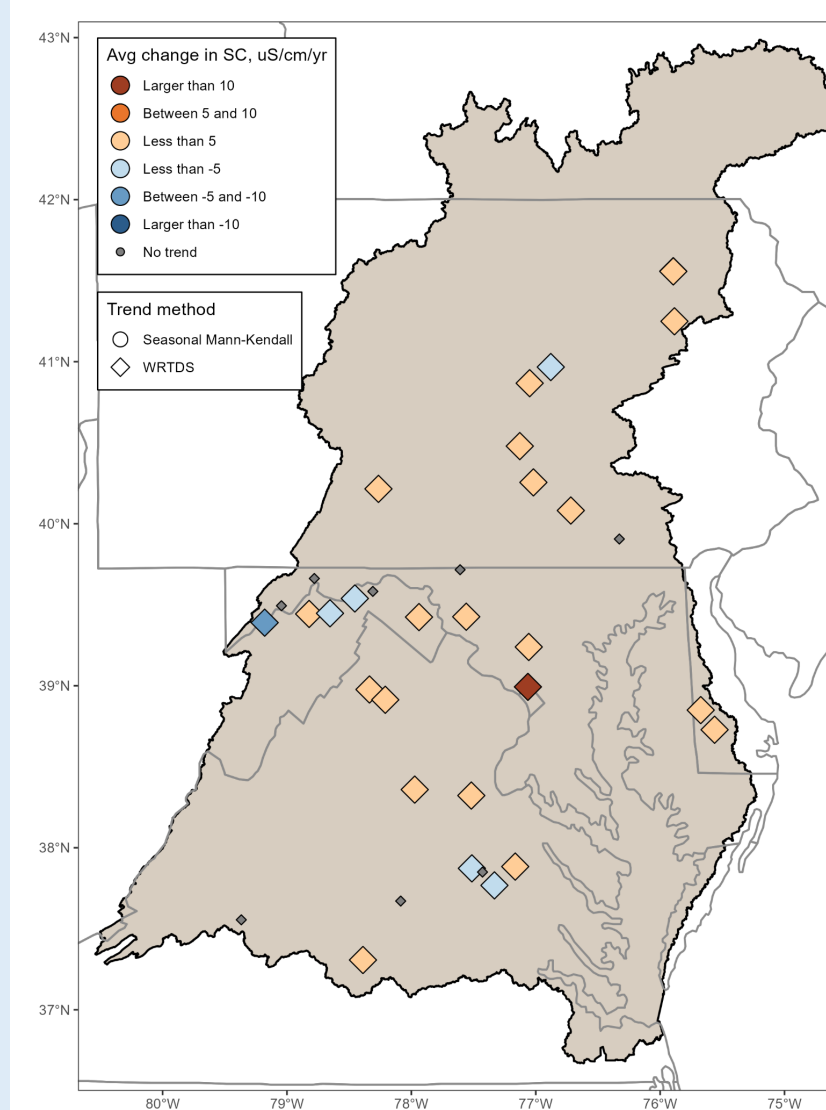
WRTDS

- 95% of upward trending sites (20 sites) had small rates of change ($< +5 \mu\text{S/cm/yr}$)
- 83% of the downward trending sites (5) had small rates of change ($< -5 \mu\text{S/cm/yr}$)

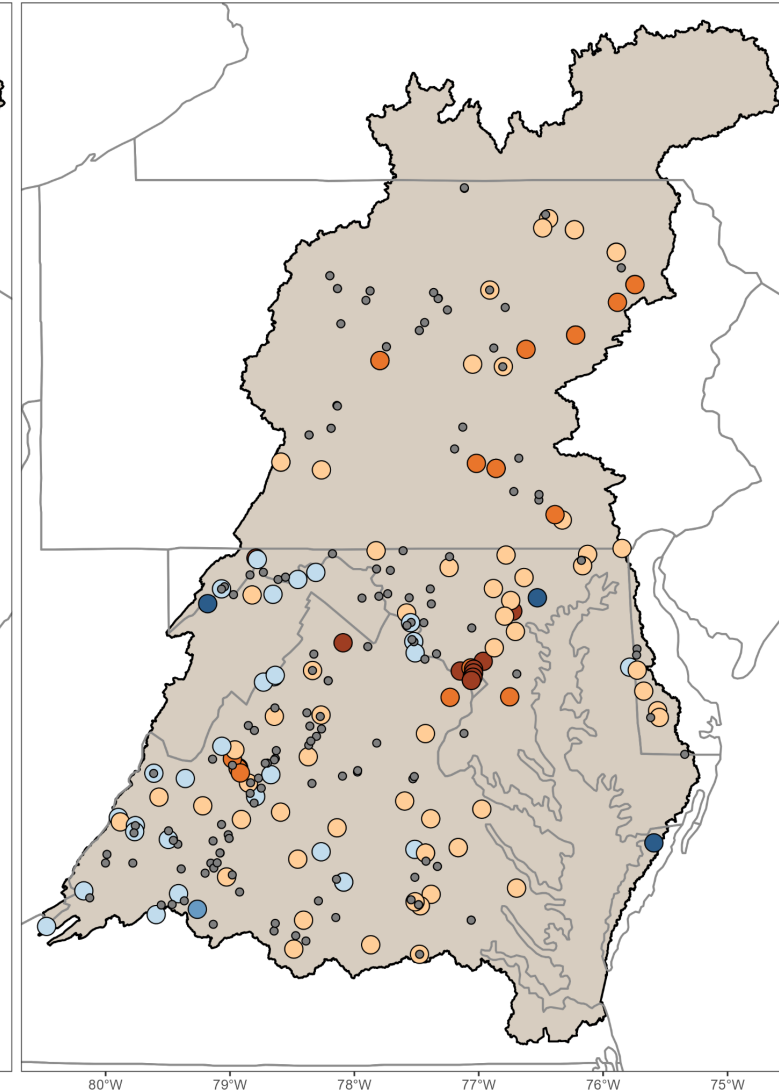
Seasonal Mann-Kendall

- 66% of upward trending sites (51 sites) had small rates of change ($< +5 \mu\text{S/cm/yr}$)
- 13% of upward trending sites (12) had large rates of change ($> +10 \mu\text{S/cm/yr}$)
- 85% of downward trending sites (29) had small rates of change ($< -5 \mu\text{S/cm/yr}$)

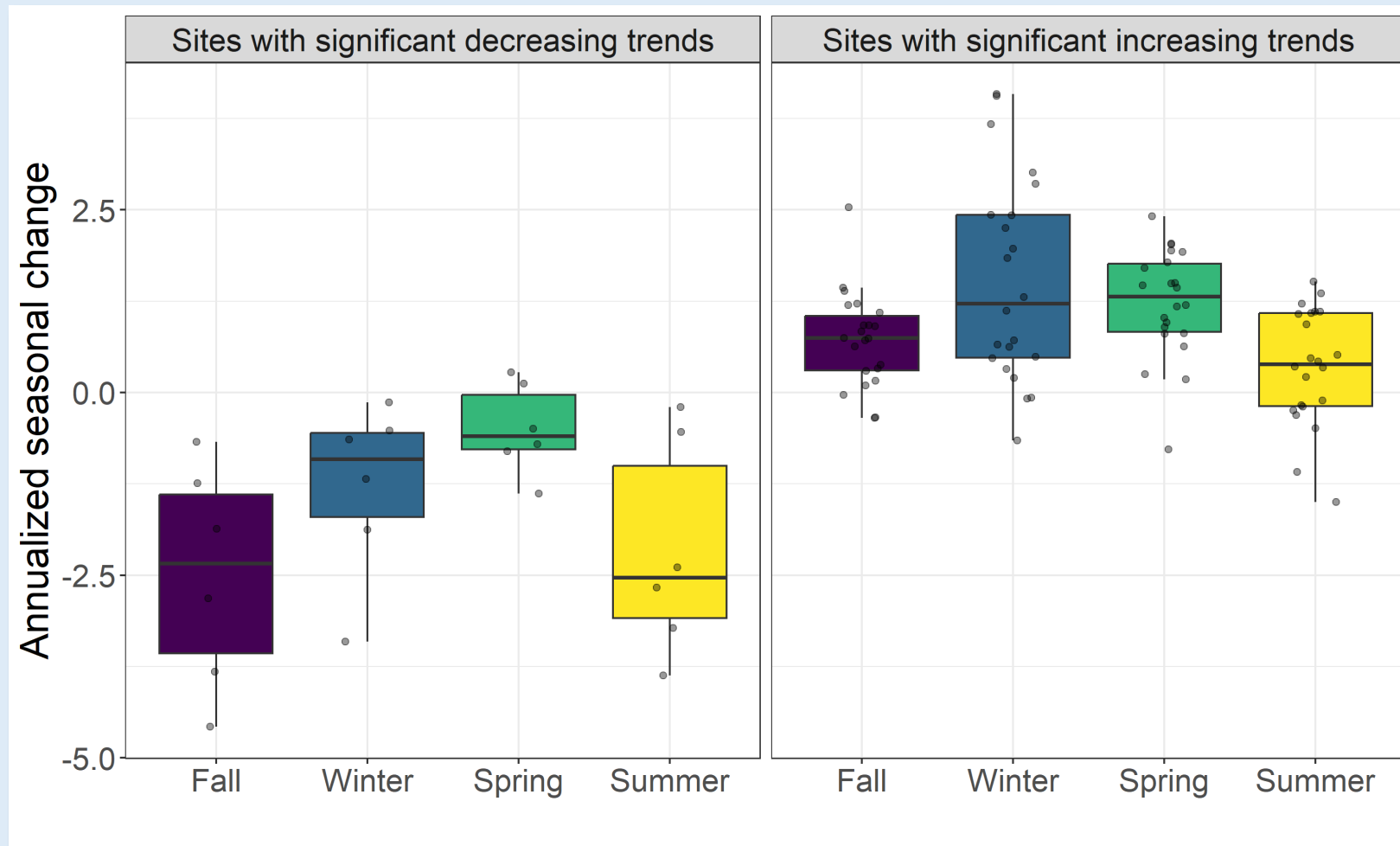
WRTDS



Seasonal Mann-Kendall



Results: Seasonal changes for WRTDS trend sites

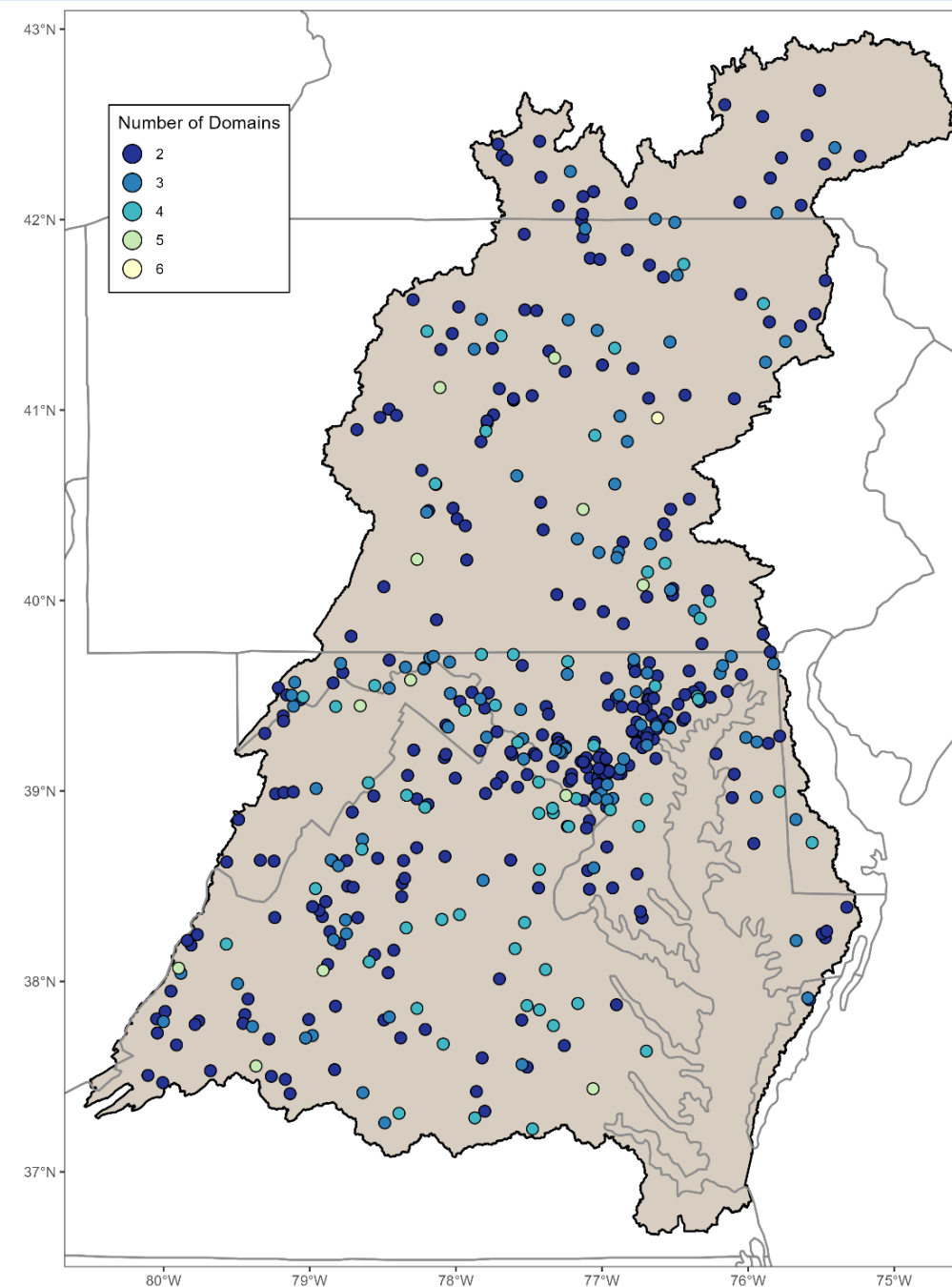
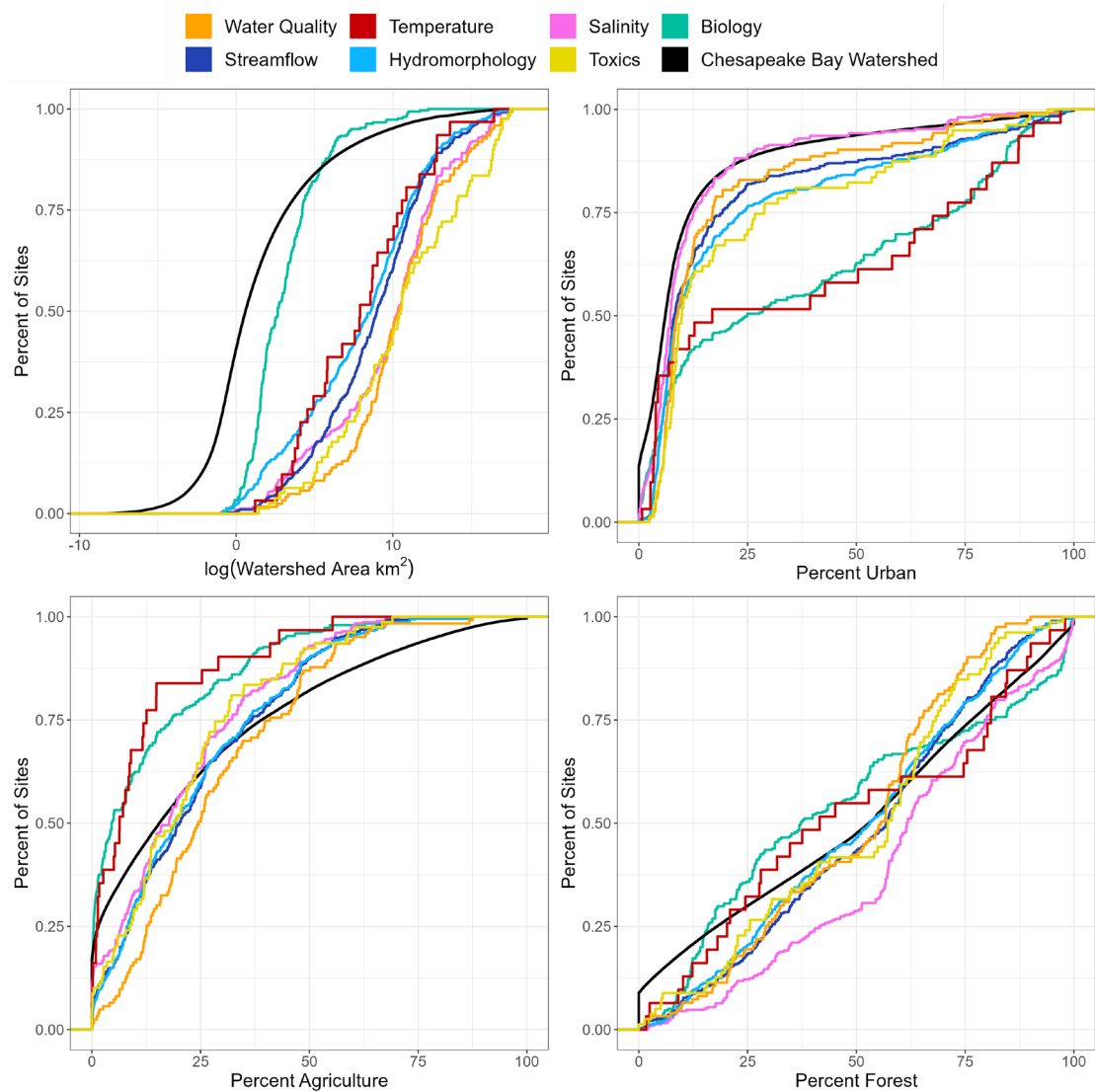


Annualized seasonal change:

- Used predicted daily SC values from WRTDS
- Computed seasonal mean SC for each year
- Computed change in seasonal means between start and end year
- Divided it by the seasonal mean in initial year
- Units are % per year

Rumsey, C. A., Hammond, J. C., Murphy, J., Shoda, M., & Soroka, A., 2023. Science of The Total Environment, 858, 159691.
<https://doi.org/10.1016/j.scitotenv.2022.159691>

Indicator site representation



This information is preliminary and is subject to revision. Not for Citation or Distribution.

Takeaways and next steps

- Only 15% of sites (43) had 3-year median SC values at or below background SC
- SC is increasing at the majority of WRTDS trend sites (60%)
- Rates of increase were modest ($< 5 \mu\text{S}/\text{cm}/\text{yr}$), except in urban regions where SC is increasing at a faster rate
- Rate of change are greatest in winter and spring months at sites with increasing SC trends

Next steps (FY24 and beyond)

Publish SIR (FY24)

Trend explanation based on stakeholder needs

Incorporate major ion information (Cl^- , SO_4^{2-})

Re-run trends analysis using refreshed data





Project 2: Predict SC and departures from background SC across all freshwater tributaries

Rosemary Fanelli, SAWSC, U.S. Geological Survey

Joel Moore, Towson University

Charlie Stillwell, SAWSC, U.S. Geological Survey

Andrew Sekellick, MD-DE-SC WSC, U.S. Geological Survey

Rich Walker, Upper Iowa University



Conductivity assessment – Why is this important

- More information on **where streams are most impacted by freshwater salinization** is needed
- Companion assessment to three other regional assessments to predict basin-wide conditions:
 - Benthic macroinvertebrates (Maloney et al. 2022; <https://doi.org/10.1016/j.jenvman.2022.116068>)
 - Fish populations (Maloney et al., 2021; <https://doi.org/10.1016/j.ecolind.2021.108488>)
 - Physical stream habitat metrics (Cashman et al., *in prep*)
- All assessments use NHDplus v2.1 (1:100K scale) network as geospatial framework
- Assessments will be used in a synthesis in FY25



Data and methods

- **Input datasets**

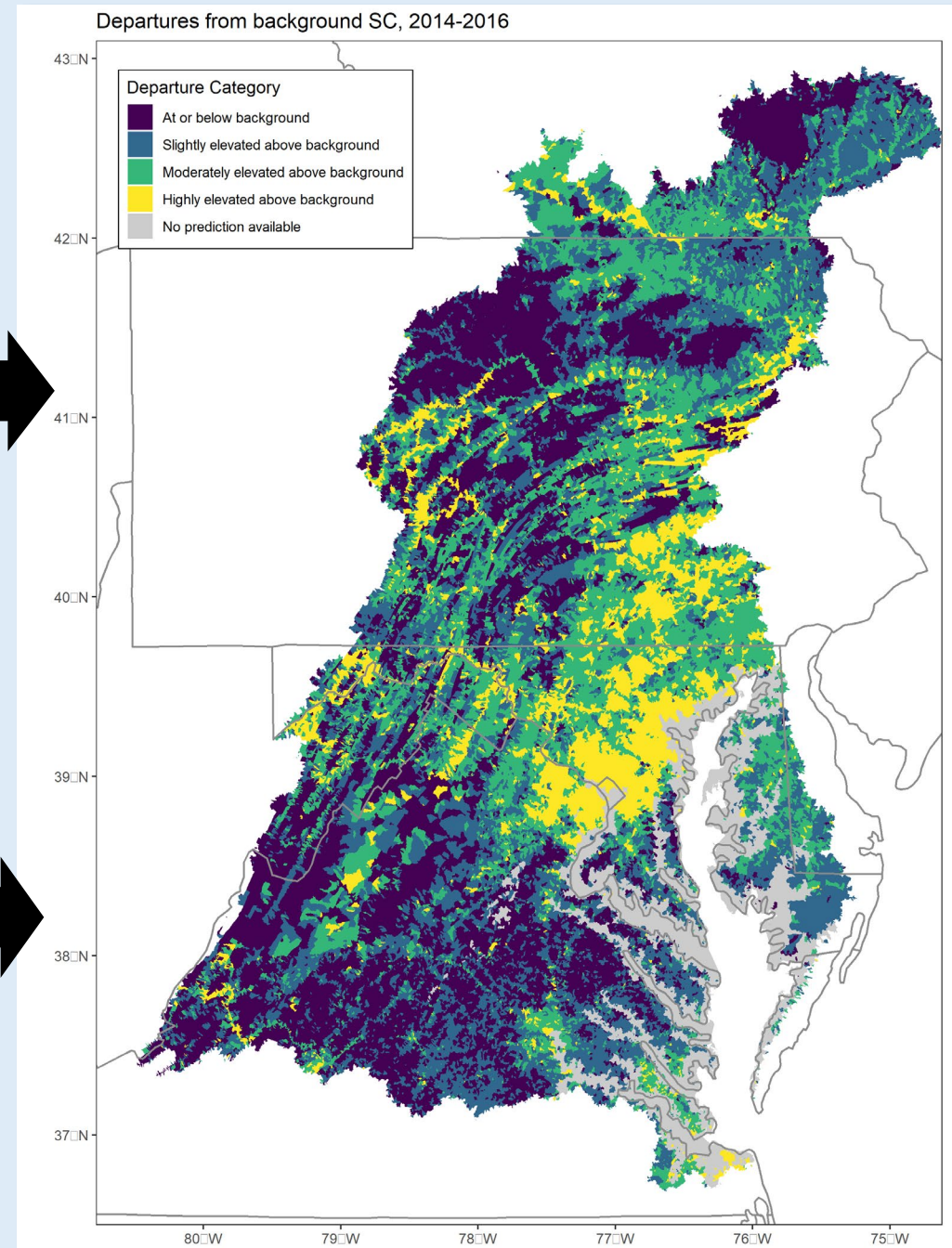
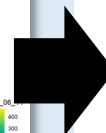
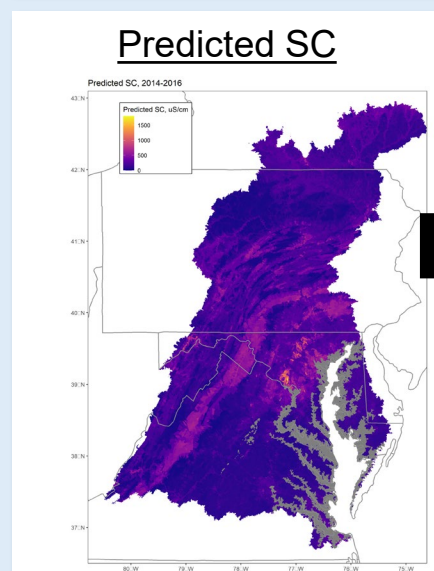
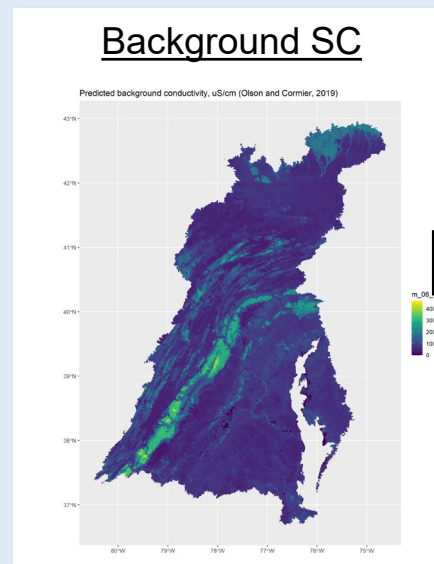
- Observed SC (*Fanelli and others, 2023*)
- Watershed characteristics: land use, climate, geology, point sources
- Modeled background SC (*Olson and Cormier 2019*)

- **Predicting SC**

- Used 1:100K NHD v 2.1
- Grouped observations into four time periods (2001, 2006, 2011, 2016)
- Used random forests regression models to predict median annual SC for four time periods

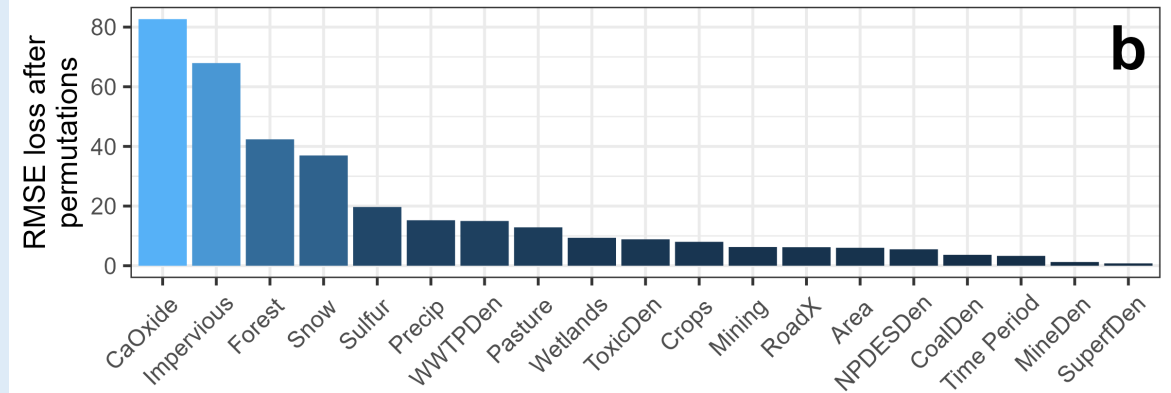
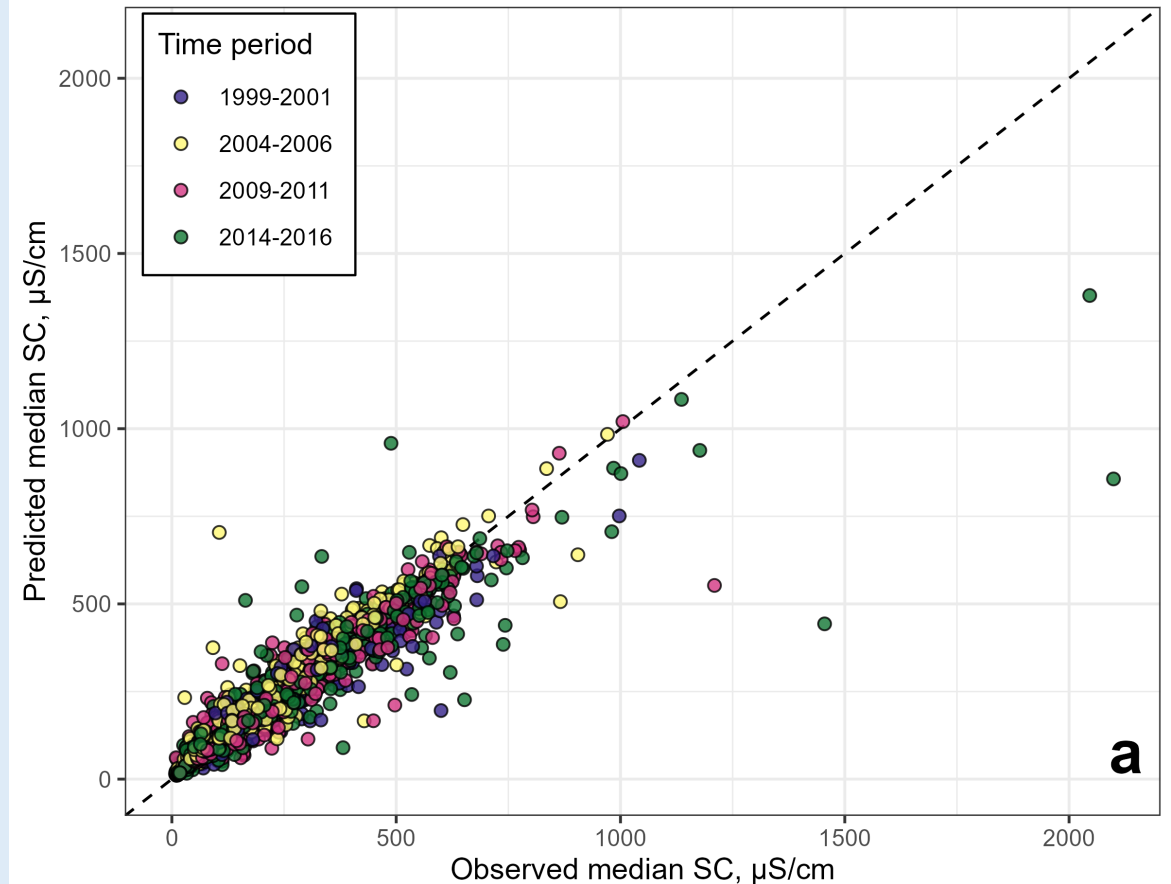
- **Quantifying departures**

- Accounted for background SC to determine departures from background SC

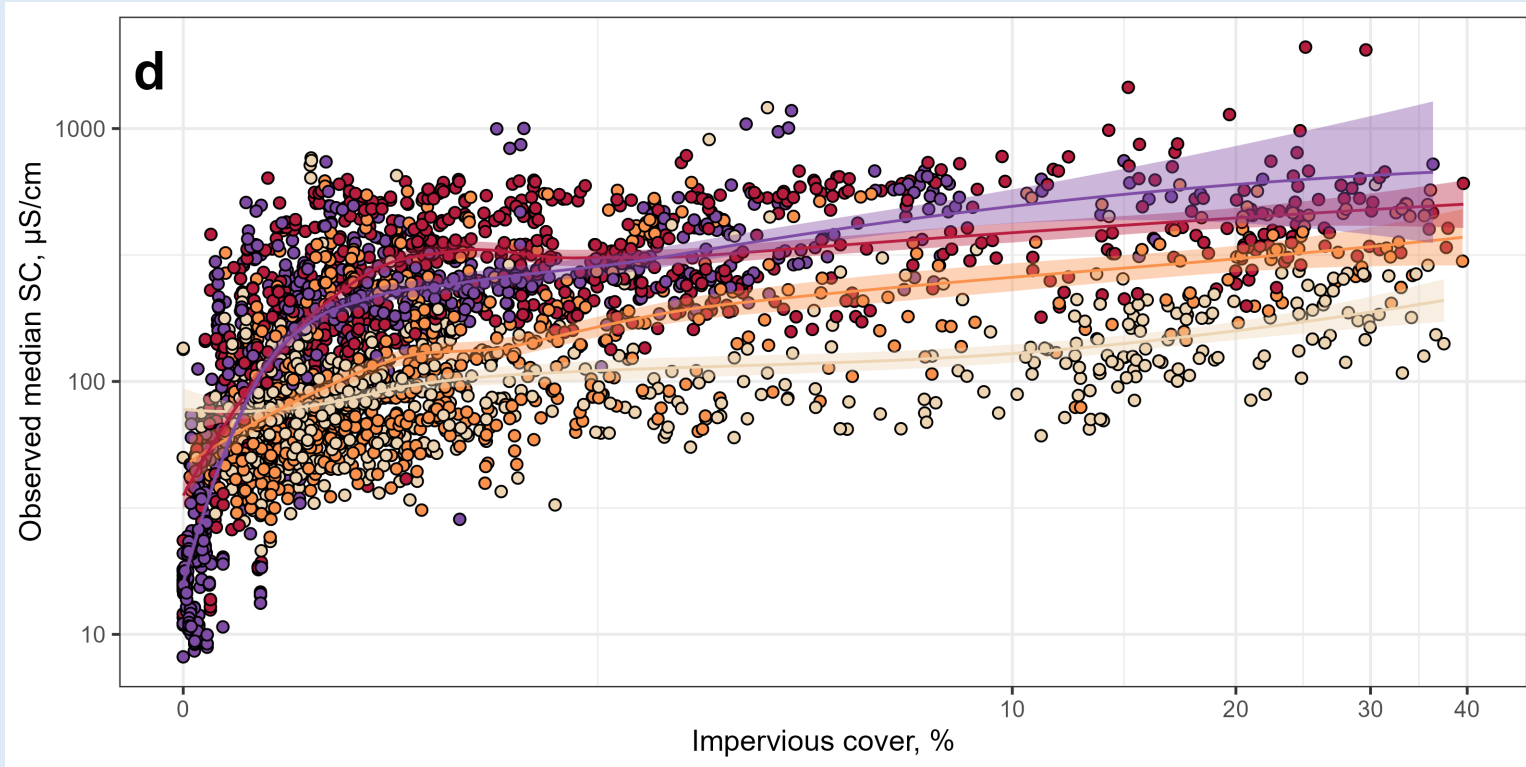
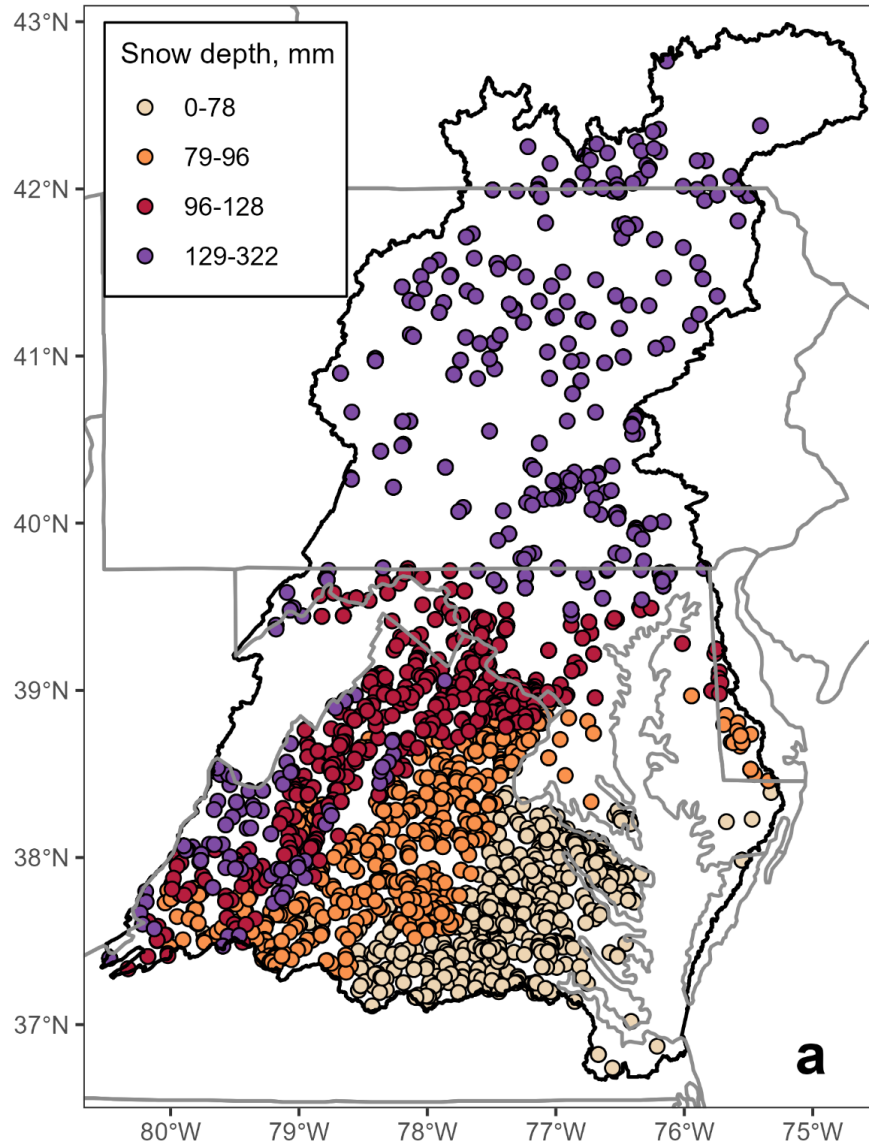


Model set up and performance

- **Random forest regression modeling**
 - Response variable = Median annual SC
 - 18 predictor variables including time period
 - Stratified sites in testing and training datasets based on carbonate geology
 - Model parameters were tuned in 10-fold cross validation procedure
- **Results**
 - MAE = 38.4 $\mu\text{S/cm}$
 - Mean $R^2 = 0.81$
 - Top three variables: calcium oxide (proxy for carbonate geology), impervious cover, and forest cover

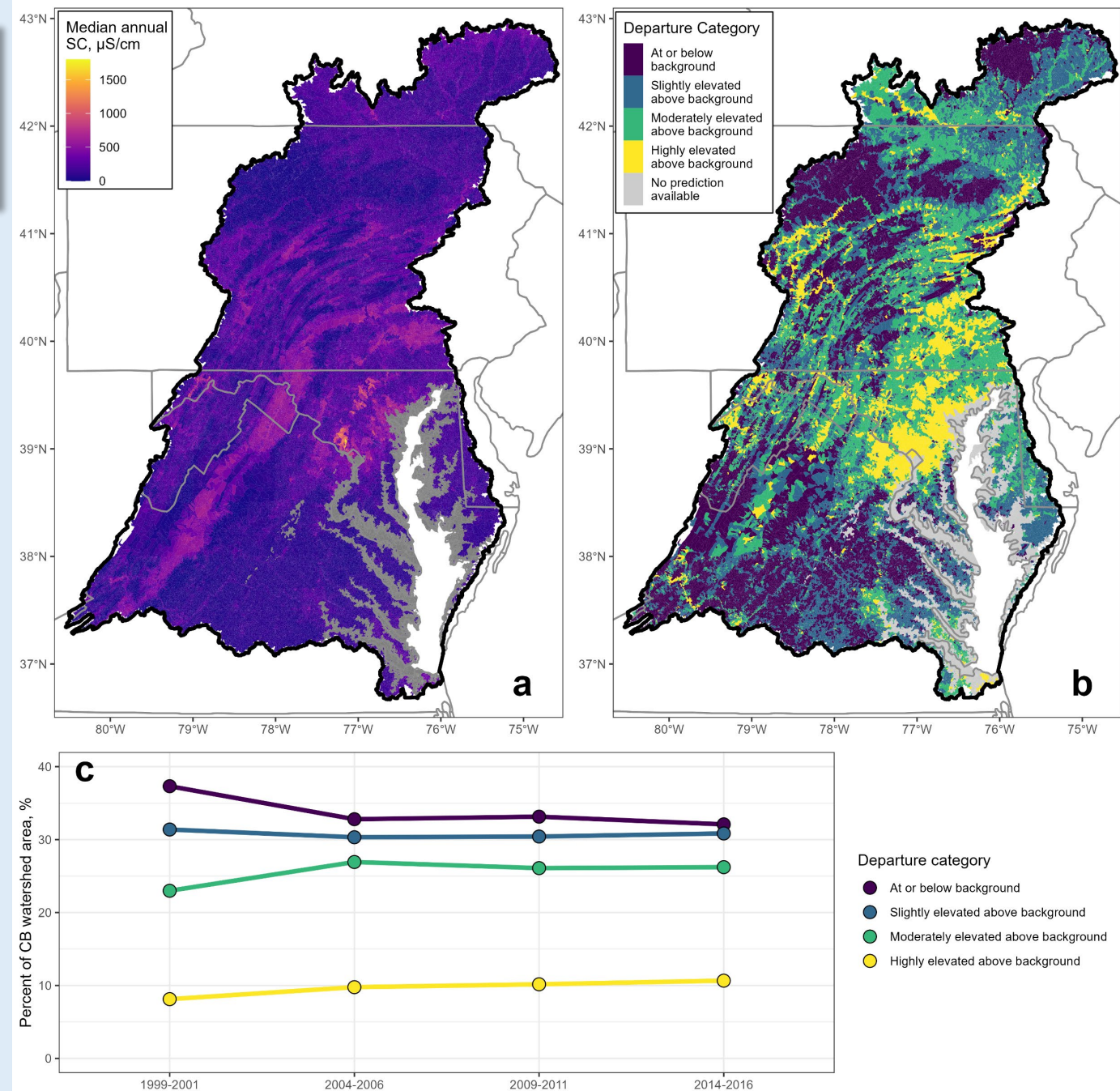


Results: Snow depth amplifies the effect of impervious cover on SC



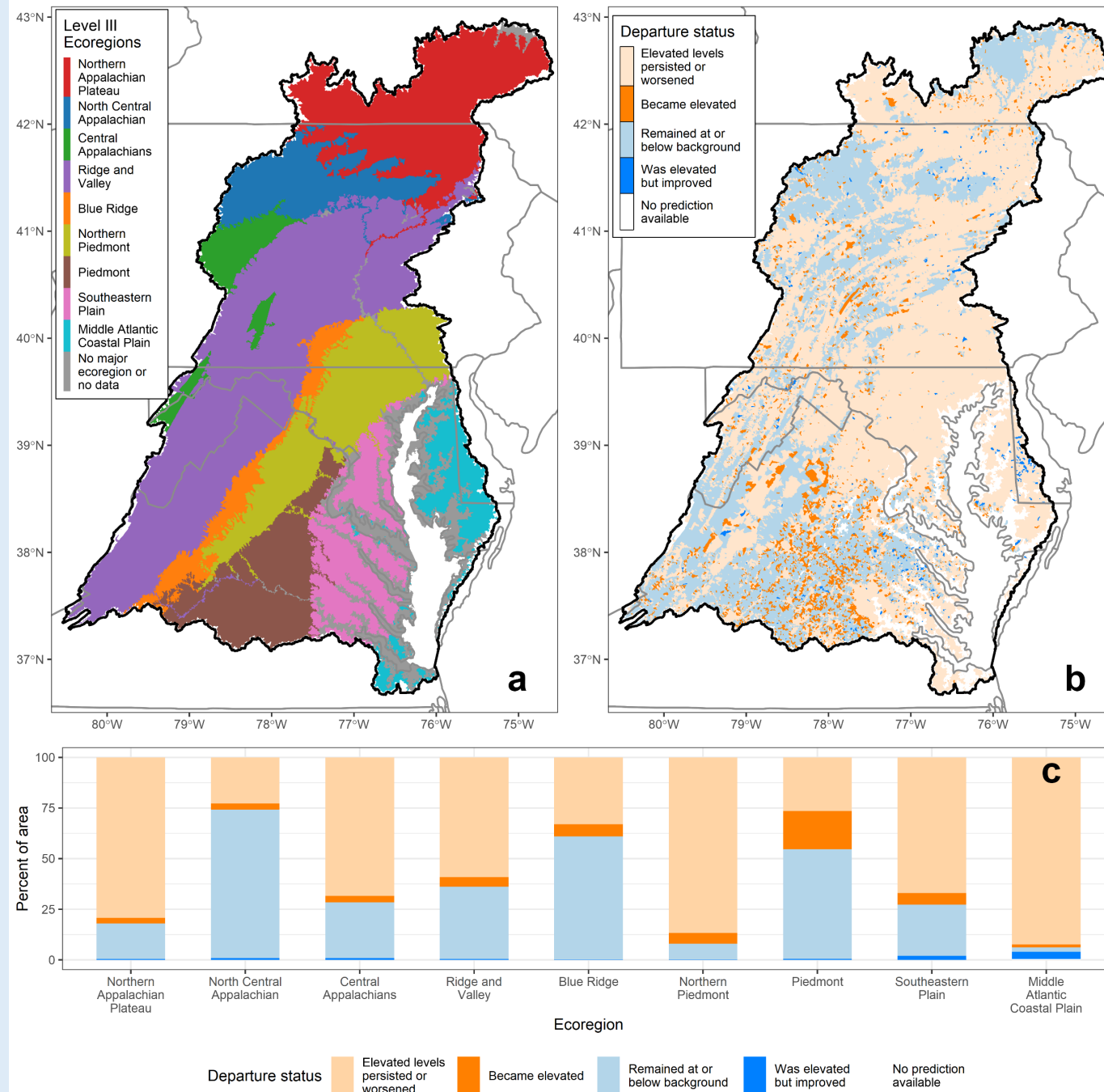
Basin-wide predictions of SC and SC departures

- Predicted median annual SC varied from 10-1,720 $\mu\text{S}/\text{cm}$
- Only 37% of CB watershed had SC values at or below background SC in 2001, declined to 32% in 2016
- Percent of watershed with highly elevated SC increased from 8.1% to 10.7% from 2001-2016



Changes in departure status across ecoregions

- Change in departure status = Change in predicted SC from 2001-2016
- SC in North Central Appalachians and Blue Ridge largely remained at or below background SC
- Previously elevated SC levels persisted in Northern Piedmont and Coastal Plain ecoregions
- Piedmont had largest total area that became elevated during study period



Takeaways and next steps

- Carbonate geology important predictor of SC
- SC increases with greater impervious cover, and snow depth amplifies the effect
- Almost two-thirds of the CB watershed has SC values above background SC; pattern persists over time
- Other land uses (logging, agriculture) may contribute to elevated SC in certain ecoregions

Next steps (FY24 and beyond)

Publish journal article and data release

Re-run analysis at 1:24K scale (FY25)

Incorporate data into regional synthesis (FY25-26)



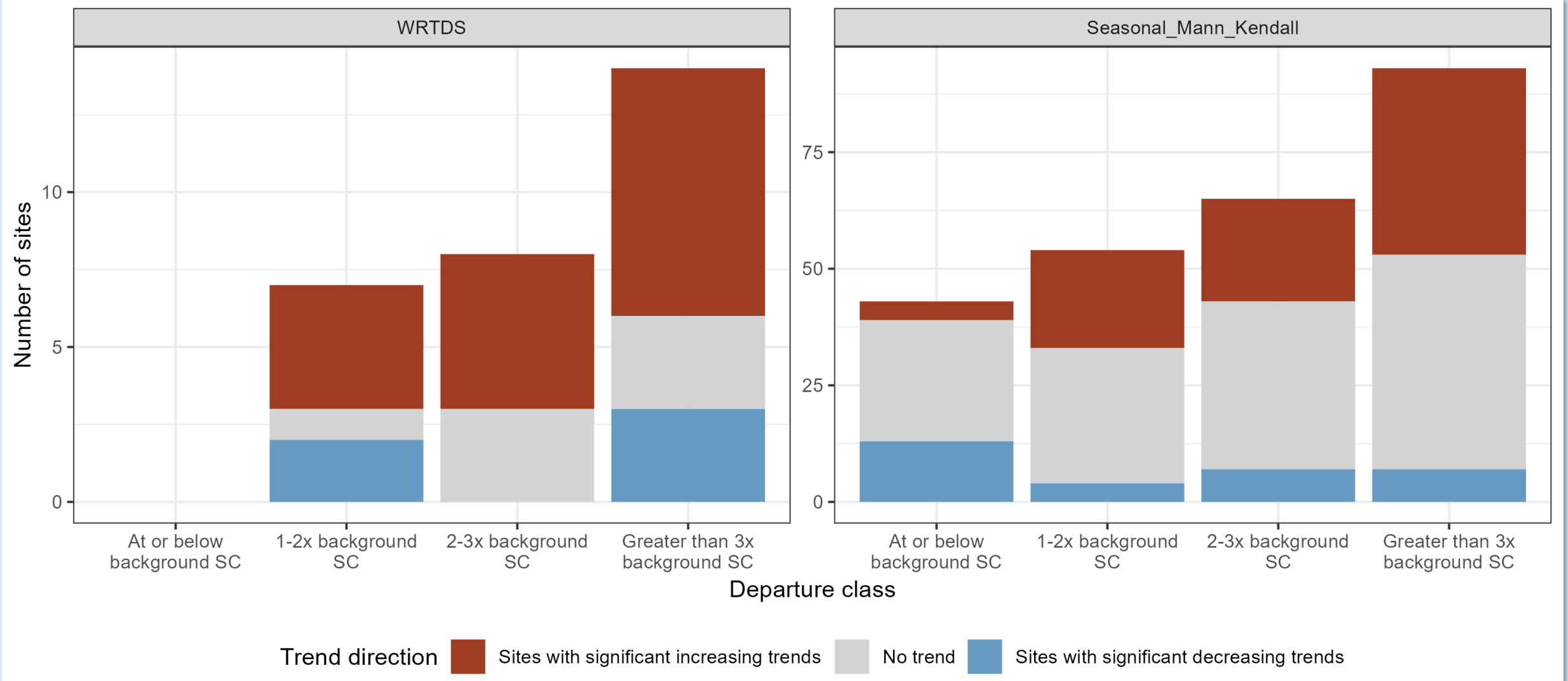


Questions or feedback?

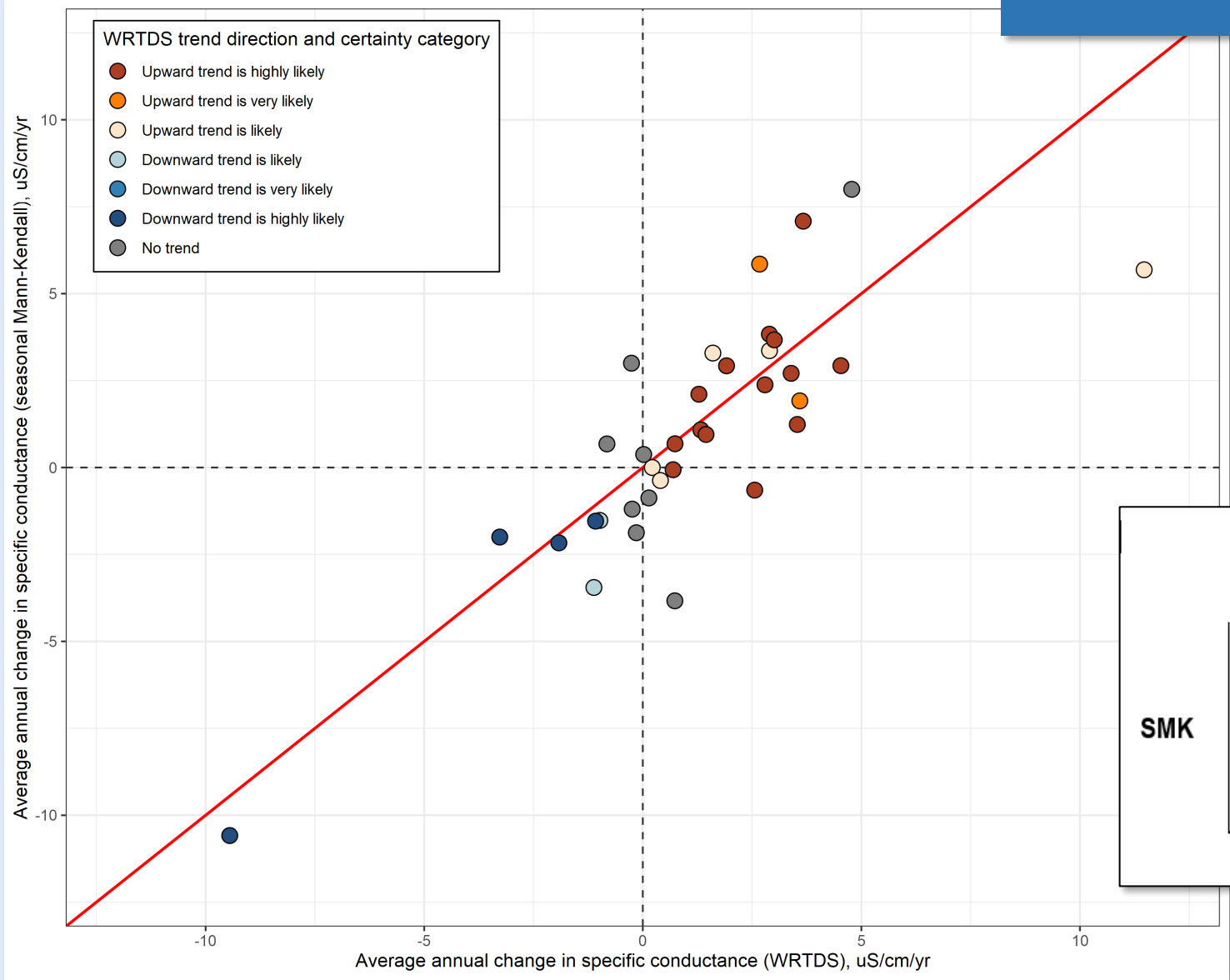
Contact info:

Rosemary Fanelli
U.S. Geological Survey
South Atlantic Water Science Center
rfanelli@usgs.gov

SC status and trends



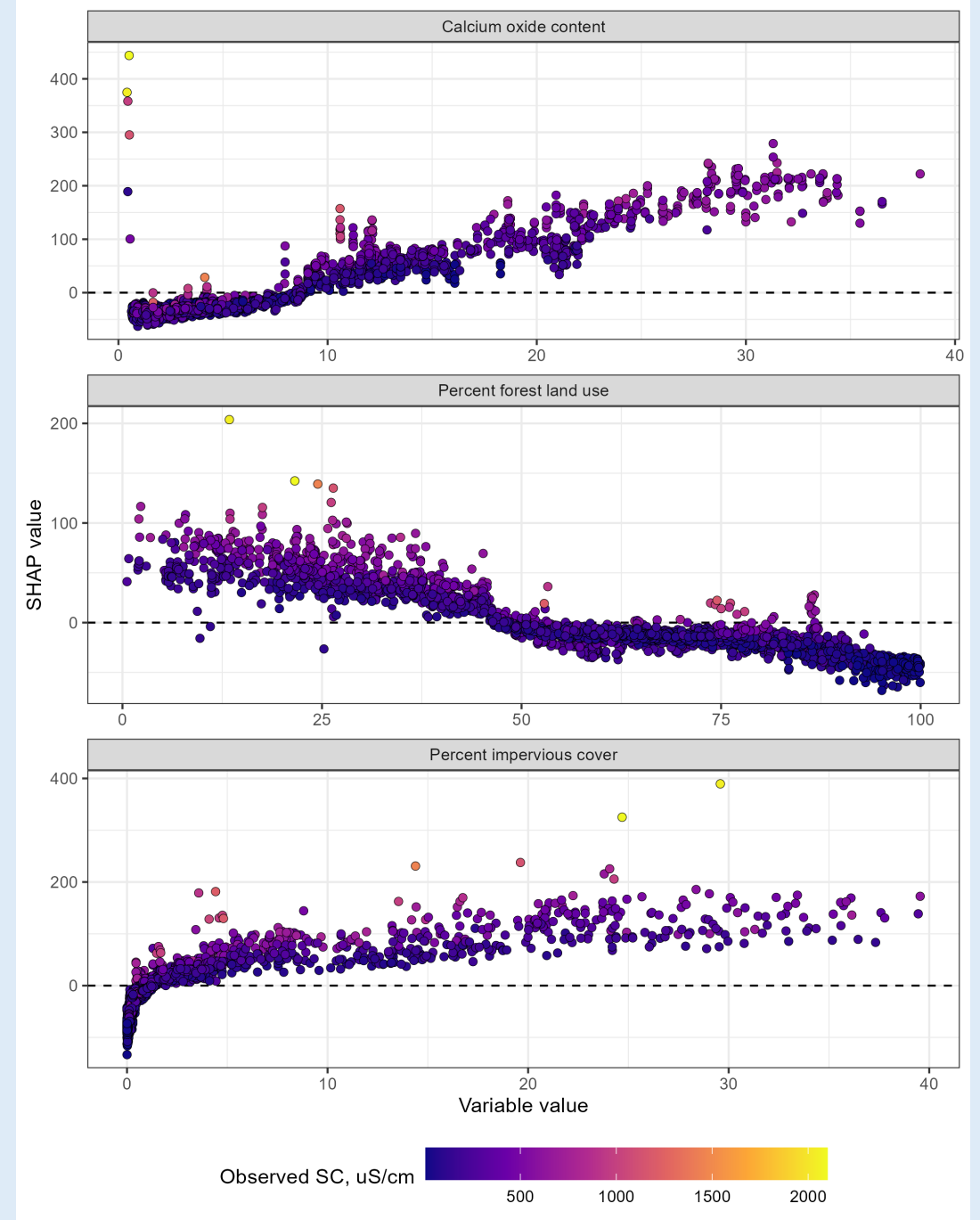
Results: SC trend comparison



		WRTDS		
		Not significant	Significantly decreasing	Significantly increasing
SMK	Not significant	9-percent	6-percent	29-percent
	Significantly decreasing	9-percent	11-percent	0-percent
	Significantly increasing	6-percent	0-percent	31-percent

Shapley values for top three predictors

- Shapley (SHAP) values = feature contributions
 - Describe the additive contributions of a single variable to a predicted SC value
- Calcium oxide SHAP values were positive when Calcium oxide $> 8\%$
- Forest cover SHAP values were negative when forest cover $> 50\%$
- Impervious cover SHAP values were positive when impervious cover $> 1\%$



Reach-scale changes in departure status

