

Statistical Modeling and Trends in Freshwater Derived Nutrient Loads, Lavaca Bay, Texas

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Michael Schramm - michael.schramm@ag.tamu.edu

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Dr. Mike Wetz - Harte Institute, TAMU-CC

Bill Balboa - Matagorda Bay Foundation

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Graduate Students:

Shubham Jain - Ph.D Candidate TAMU Biological and Agricultural Engineering



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Background



Lavaca Bay Watershed

- 3,146 miles²
- 50% Pasture and rangeland
- 20% Cultivated cropland (cotton, soy, corn, sorghum)
- 17% Forested
- 5% Suburban, urban

Background

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Water quality trends in Texas estuaries

Kalman Bugica^a, Blair Sterba-Boatwright^b, Michael S. Wetz^{a,*}

^aHarte Research Institute for Gulf of Mexico Studies, Texas A&M University-Corpus Christi, 6300 Ocean Dr., Corpus Christi, TX 78412, USA
^bDepartment of Mathematics and Statistics, Texas A&M University-Corpus Christi, 6300 Ocean Dr., Corpus Christi, TX 78412, USA

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ABSTRACT

Coastal watersheds in Texas have experienced significant human population growth over the past several decades, yet there have been no comprehensive assessments of water quality trends in Texas estuaries. Here, analysis of historical estuarine water quality data indicates regional “hot spots” of change. Galveston Bay and Oso Bay, which have highly urbanized watersheds, currently exhibit symptoms of eutrophication. Symptoms of eutrophication were also found in the Baffin Bay-Upper Laguna Madre complex, which has a sparsely populated but agriculturally-intensive watershed. Increasing salinity was observed in estuaries of the central Texas coast and are attributed to long-term decreases in freshwater inflow. Another artifact of decreasing freshwater inflow is a reduction in the delivery of carbonate minerals to estuaries, which manifests as decreases in pH. With findings from this study, targeted studies can now be directed at the estuaries that are experiencing water quality degradation in order to guide future management efforts.

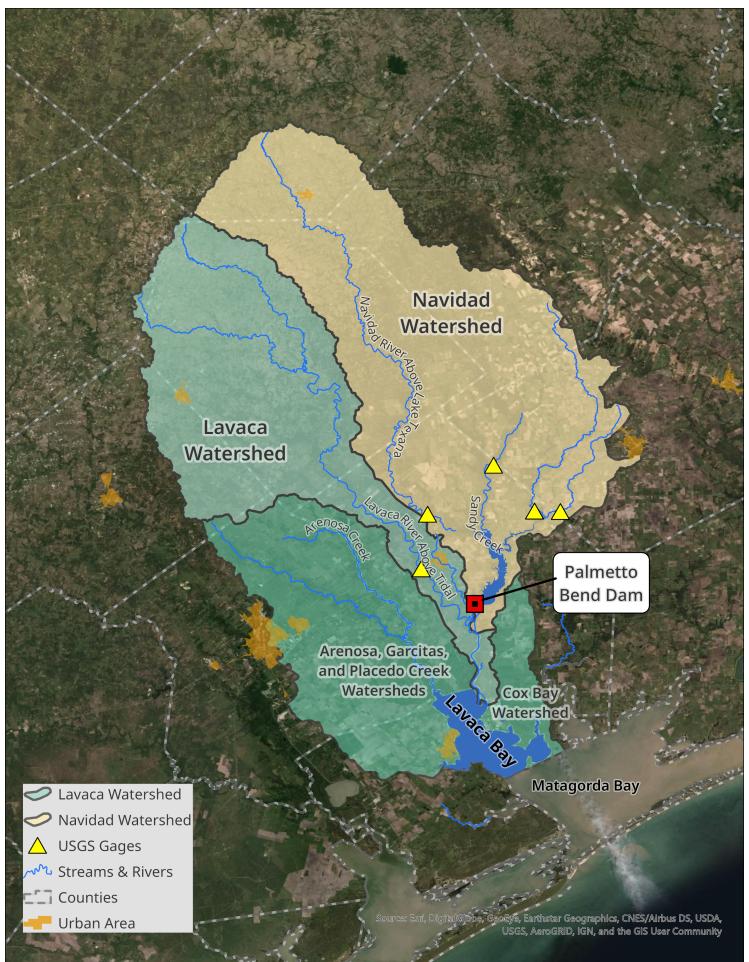
- Possible concerns for Total Phosphorus at some FW sites.
- Texas does not currently have nutrient standards for streams.
- Long-term quarterly monitoring, no historical storm or flow-biased data to this point.

- Bugica, Sterba-Boatwright, and Wetz (2020) identify eutrophication risk in Lavaca Bay due to ↑ TP and ↑ Chlorophyll-a concentrations.

Project Objectives

- 1. Develop estimates of NO₃-N and TP loading from Lavaca and Navidad Rivers
- 2. Link nutrient loads and river discharge to changes in nutrient concentration in Lavaca Bay

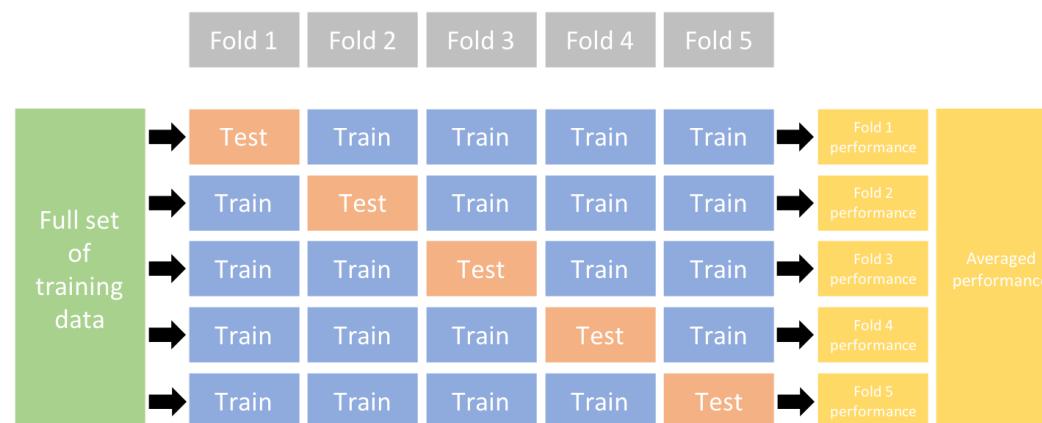
Lavaca Bay Watersheds



- 1.3 million acre-feet per year
- 65% from Lavaca/Navidad watershed
 - 61% from Navidad at Palmetto Bend Dam
 - 32% from Lavaca near City of Edna
 - 7% ungaged downstream runoff

Methods

- Loading estimates:
 - **Specify Concentration Regression Model** - Generalized Additive Models ([Kuhnert et al. 2012](#); [Robson and Dourdet 2015](#); [McDowell et al. 2021](#))
 - model error structure and specify link function
 - predictor variables can be smooth functions allowing non-linear responses.
 - **Model performance** - Repeated 5-fold cross-validation
 - **Predict daily loads** - Point estimates with uncertainty

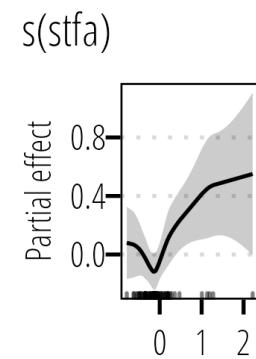
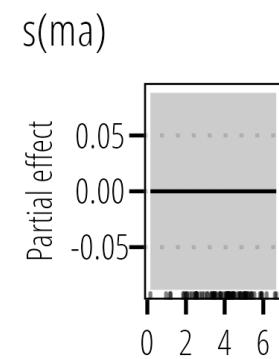
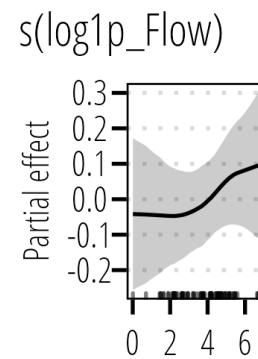
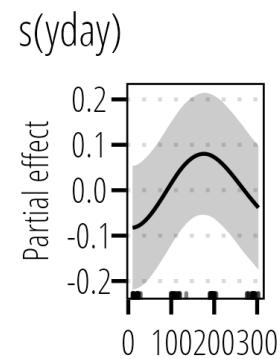
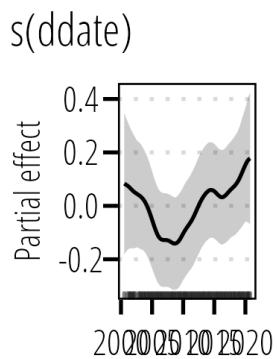


5-fold CV procedure. Image from Boehmke & Greenwell 2020 (<https://bradleyboehmke.github.io/HOML/>)

Methods

$$Y = s(ddate) + s(yday) + s(\log1p(Flow)) + s(ma) + s(fa)$$

- Y = NO_3 or TP concentration;
- s = smoothing function;
- $ddate$ = decimal date;
- $yday$ = numeric day of year;
- $Flow$ = mean daily discharge;
- ma = exponential moving average ([Kuhnert et al. 2012](#); [Zhang and Ball 2017](#)).
- fa = short- or long-term flow anomaly (stfa, ltfa) ([Vecchia et al. 2009](#); [Zhang and Ball 2017](#));
- Gamma family with log-link



ddate
Basis: TPRS

yday
Basis: Cyclic CRS

$\log1p_Flow$
Basis: TPRS

ma
Basis: TPRS

stfa
Basis: TPRS

Methods

Loading estimates:

- Prediction of **daily loads** from GAM models at each site
 - predicted concentrations × mean daily streamflow
 - aggregated to monthly and annual totals
- Report **model uncertainty**
 - **95% credible intervals** developed from 1000 draws of parameter estimates from the multivariate normal posterior distribution of model parameters provided by `mgcv::gam` function in R.
- Account for variance in mean daily discharge
 - **Flow-normalized estimates** calculated similar to WRTDS, assume daily flow variables are random occurrence from all possible values on that day of year.

Methods

Do variations in Flow and Load explain Bay nutrient concentration?

Temporal Model

$$Y = s(\text{Day}) + s(\text{Date}) + ti(\text{Day}, \text{Date})$$

Flow Model

$$Y = s(\text{Day}) + s(\text{Date}) + ti(\text{Day}, \text{Date}) + s(\text{Flow})$$

Full Model

$$Y = s(\text{Day}) + s(\text{Date}) + ti(\text{Day}, \text{Date}) + s(\text{Flow}) + s(\text{Load})$$

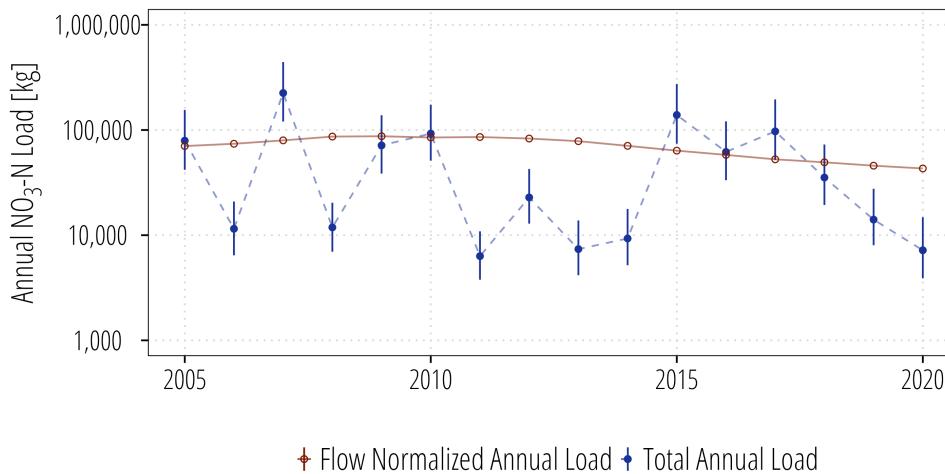
- *Flow* is seasonally adjusted
- *Load* is flow adjusted
- Simplified methodology following Murphy et al. (2019) and Murphy et al. (2022).
- Compare AIC and other model metrics

Results

Lavaca River

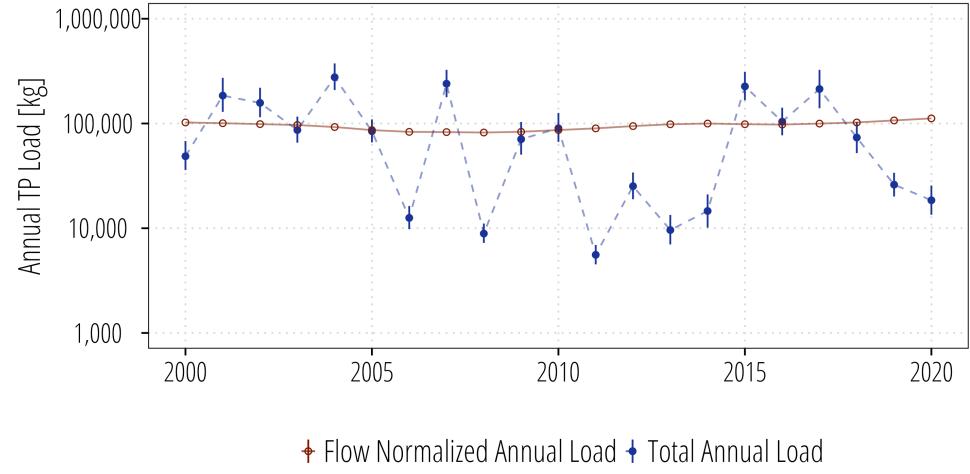
NO₃-N

Metric	Median (IQR)
NSE	0.758 (0.714, 0.765)
R ²	0.761 (0.728, 0.771)
Percent Bias	-7.80 (-9.02, -4.15)



TP

Metric	Median (IQR)
NSE	0.77 (0.71, 0.81)
R ²	0.77 (0.72, 0.82)
Percent Bias	-7.45 (-9.10, -6.35)

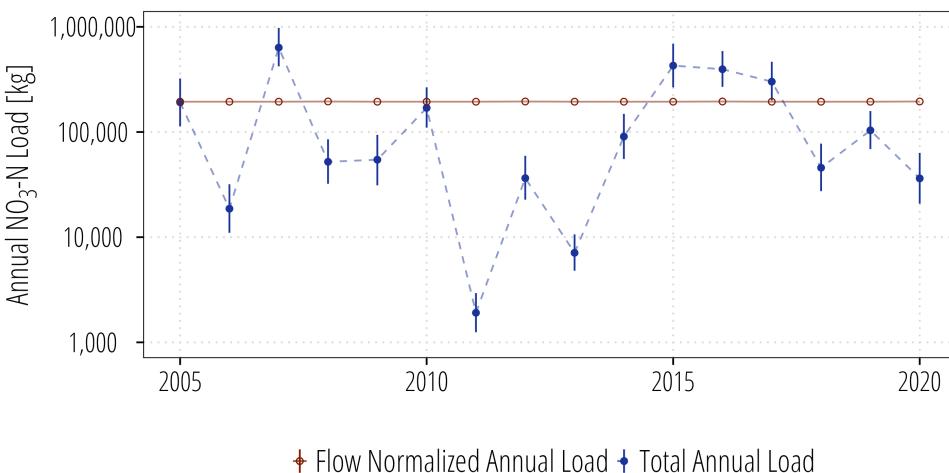


Results

Navidad River/Palmetto Bend Dam

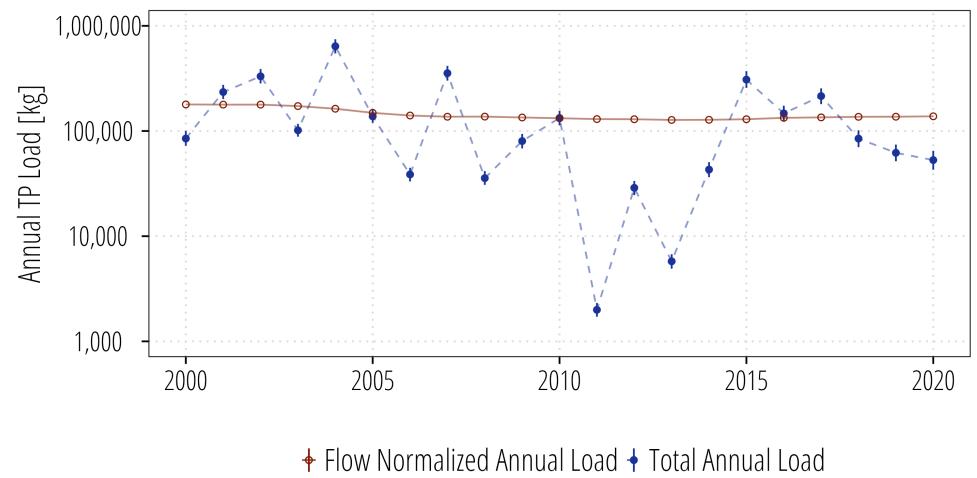
NO₃-N

Metric	Median (IQR)
NSE	0.42 (0.34, 0.46)
R ²	0.60 (0.52, 0.66)
Percent Bias	-43 (-47, -38)

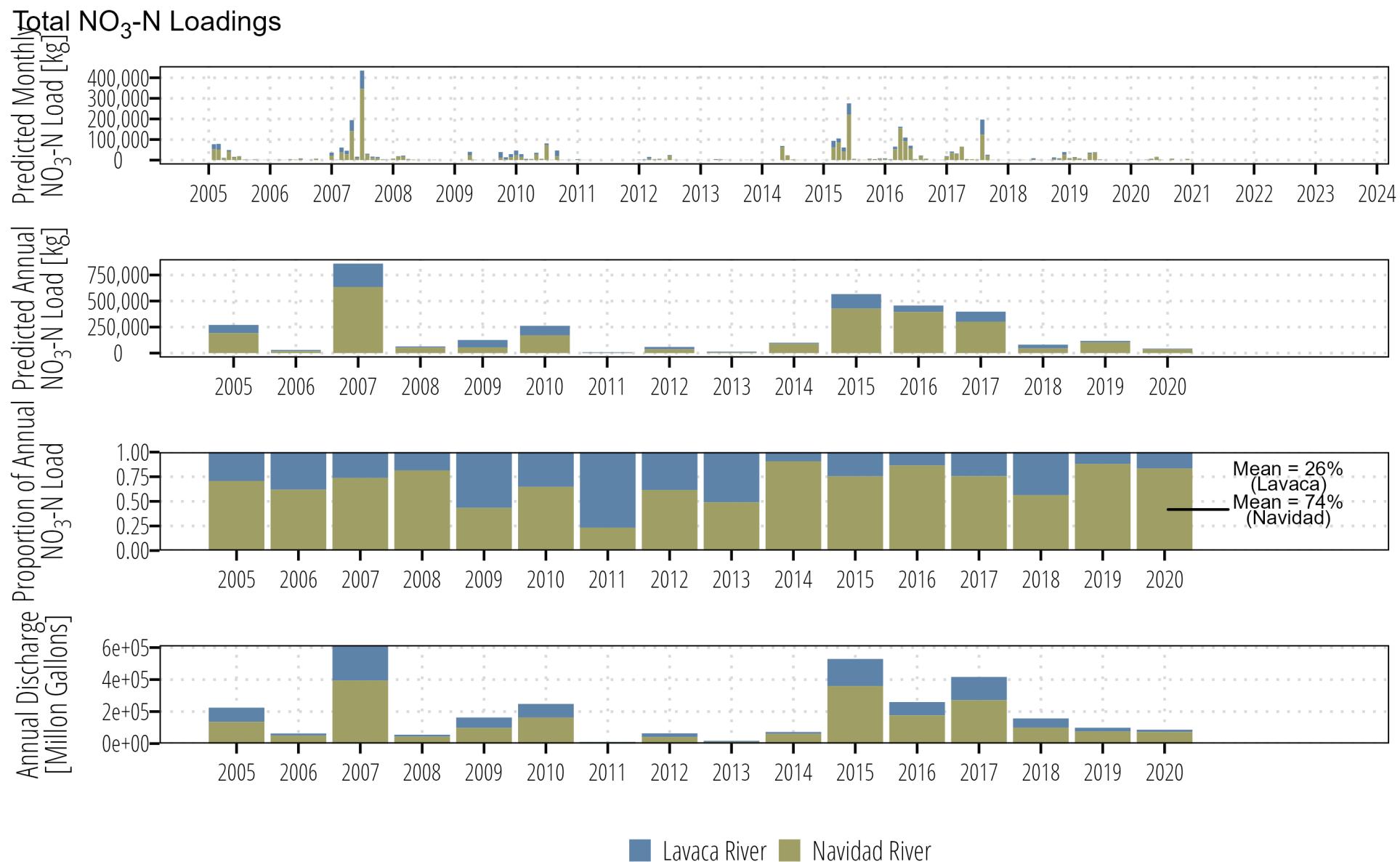


TP

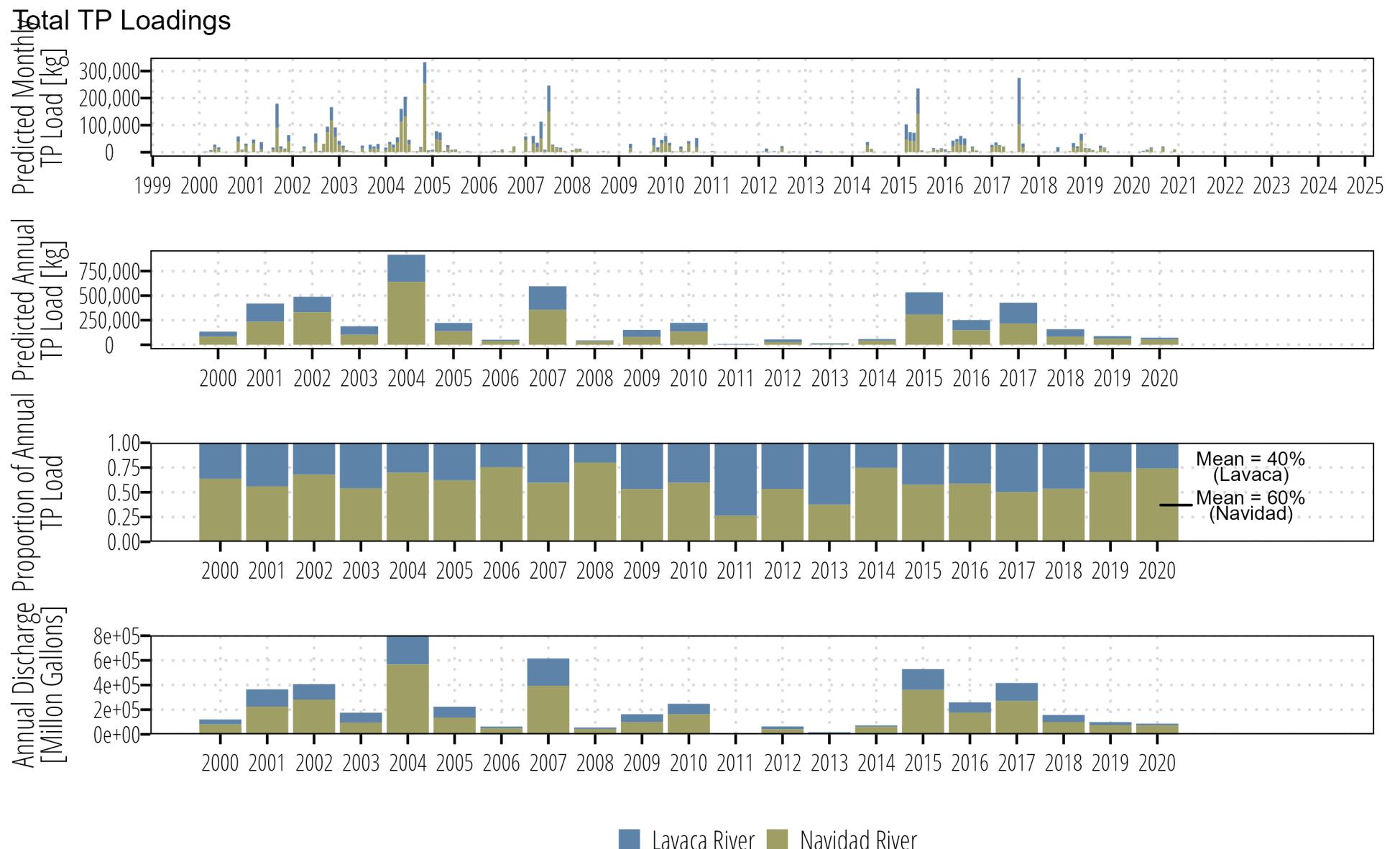
Metric	Median (IQR)
NSE	0.877 (0.862, 0.911)
R ²	0.961 (0.956, 0.975)
Percent Bias	-17.6 (-21.1, -12.7)



Results



Results



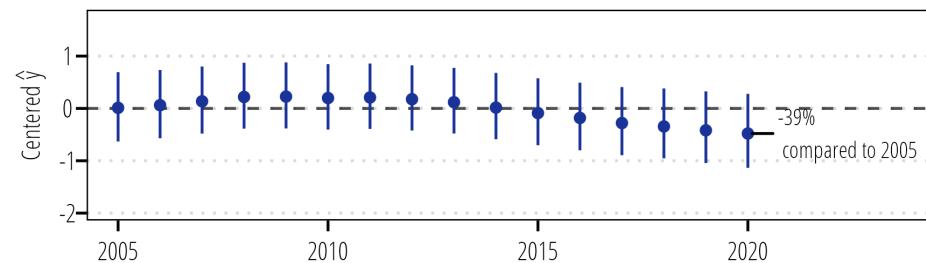
Regional Study Comparison

Parameter	Reported Yield (kg/km ² /yr)	Approach	Time Period	Reference
TP	42.9 (34.4, 54.0)	GAM	2000-2020	-
TP	45.2	SPARROW	2012	Wise, Anning, and Miller (2019)
TP	42	SWAT	1977-2005	Omani, Srinivasan, and Lee (2014)
TP	20.81-91.58	SPARROW	2002	Rebich et al. (2011)
TP	28.9	LOADEST	1972-1993	Dunn (1996)

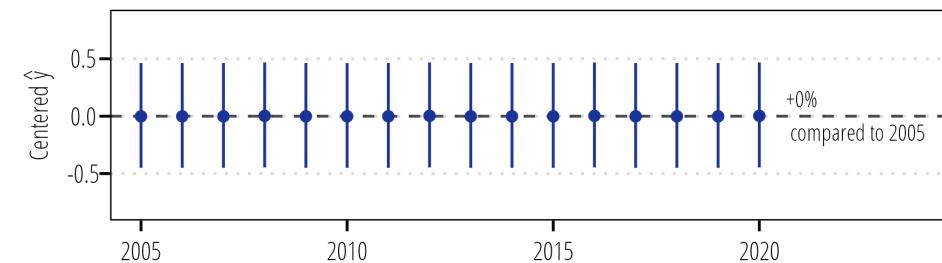
What About Trends?

- High variability in actual loads that reflect total discharge
- Flow-normalized loads:

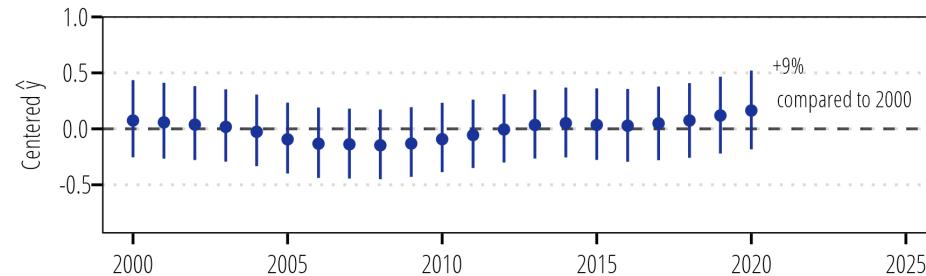
Flow-Normalized NO₃-N Load, Lavaca River



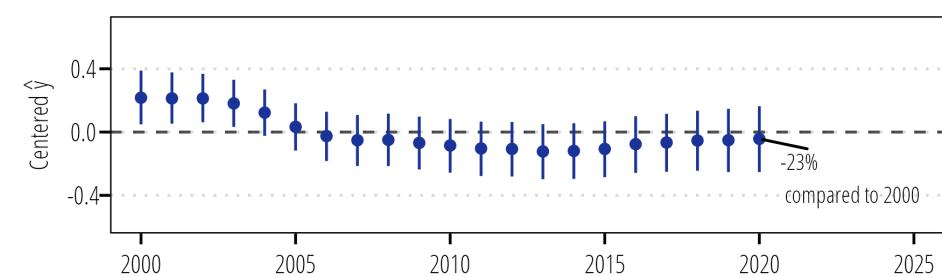
Flow-Normalized NO₃-N Load, Navidad River



Flow-Normalized TP Load, Lavaca River



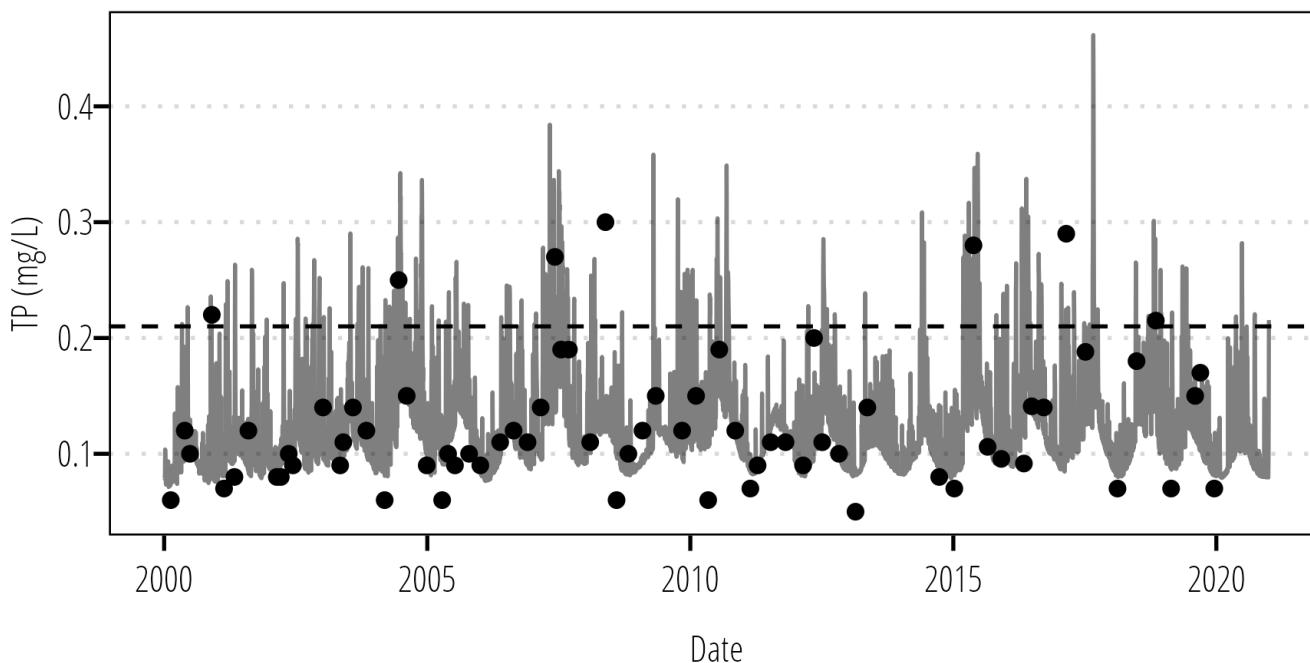
Flow-Normalized TP Load, Navidad River



Estuary Models

Example: Site 13563 TP

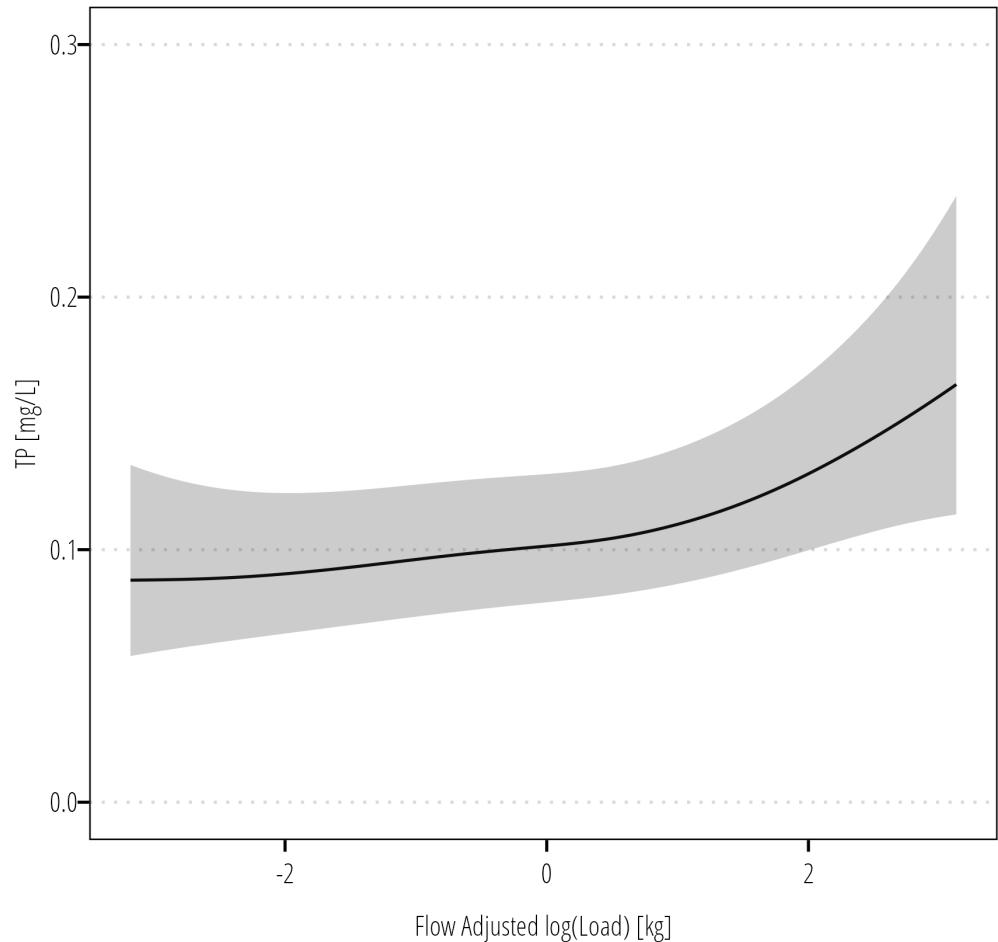
	df	AIC	adj.r.sq	dev.expl
Temporal Model	9.735954	-222.5104	0.1492214	0.2597455
Flow Model	8.479287	-241.3721	0.3383421	0.4107901
Flow and Load Model	14.272933	-252.4411	0.4755137	0.5754284



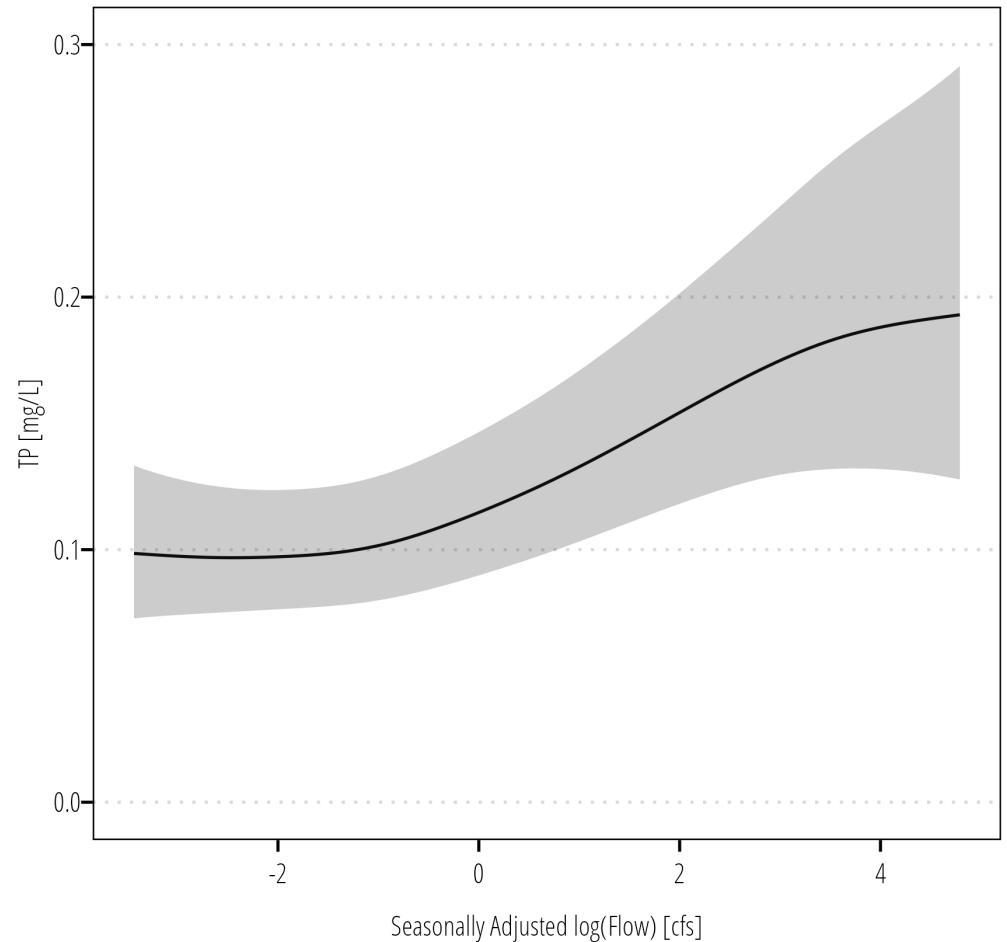
Estuary Models

Example: Site 13563 TP

Response of TP concentration to flow adjusted load



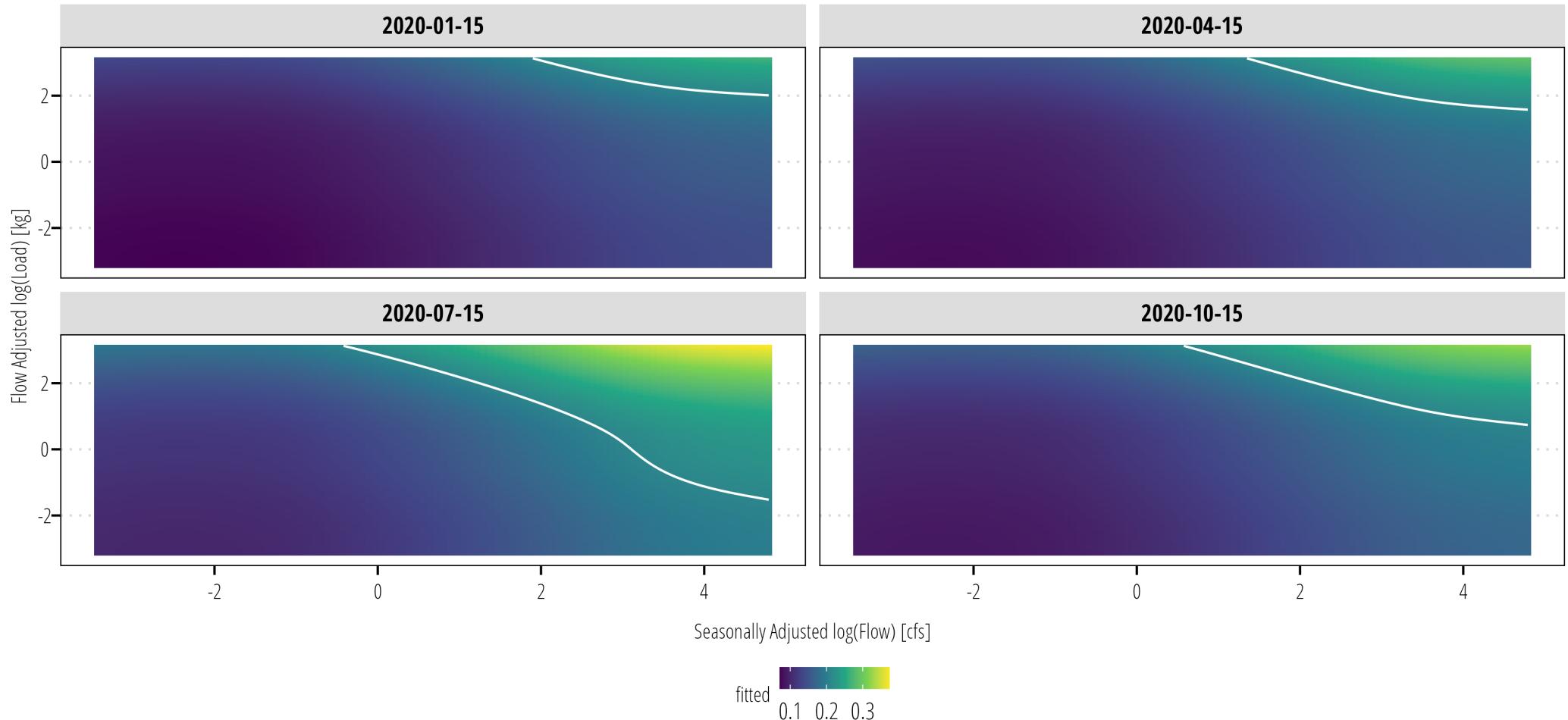
Response of TP concentration to seasonally adjusted flow



Estuary Models

Example: Site 13563 TP

Response of TP concentration to adjust flow and adjusted load



Conclusions

- GAMs
 - Useful framework for statistical load estimation and exploratory analysis of estuarine water quality.
- Nutrient Loading
 - High variance in actual loads.
 - Shifts in riverine sources during drought conditions.
 - Changes in watershed loads explains some variation in estuary nutrient concentration.

Further work

- Comparison of above lake and below lake loads;
- Develop/fund supplemental flow-biased monitoring to identify significant changes and trends;
- Continue work on estuary water quality responses...

Thank You!

Michael Schramm - michael.schramm@ag.tamu.edu

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