

# UNRBA Monitoring Program Development and Implementation

PFC Meeting  
March 4, 2014





## Monitoring objectives

- Falls Lake EFDC Model improvement
- Demonstrate response of Falls Lake chlorophyll a and TOC to changes in nitrogen and phosphorus concentrations and loading
- Determine Jurisdictional Loading
- Regulatory compliance
- Link water quality and designated uses
- Prioritize BMP implementation
- Support regulatory options



## Updates

- Main focus of current analyses is to use and create models that can predict water quality or flow
  - Develop adaptive monitoring plan
  - Reduce future monitoring effort where possible
  - Optimize the level of monitoring effort and cost

# Statistical Model Development



## Statistical Model Development Objectives

- ***Inform monitoring design*** (locations and frequency of sampling)
- Assess degree to which models could be used to ***estimate water quality parameters***
  - > at locations without data (or very little data)
  - > for dates without measurements





## Future Use of Water Quality Statistical Prediction Models

- Fill in data gaps with model predictions
  - > EFDC model requires daily inputs of nutrients
    - measurements occur less frequently than daily
    - models should provide unbiased estimates (linear interpolation is likely biased, e.g. Ferguson 1987).
- Statistically test hypotheses about changes in WQ post management action (e.g. WWTP upgrades).
- Estimate trends in WQ through time (e.g. gradual effects from multiple BMP implementations).

# Use of Statistical Models to Support Design of Monitoring Program



## Monitoring Implications

- If a site of interest is well predicted\* by models, UNRBA may not need to sample that location as frequently as others (but still enough to verify model predictions).
- If a site of interest is poorly predicted by models, sampling may need to occur more frequently.

\*Estimates are unbiased with narrow confidence intervals





## Monitoring Implications

- Statistical model provides the ***expected daily mean*** nutrient concentrations
  - > Adjust sampling frequency based on the expected mean concentrations.
    - For example, sites with high nutrient loading are more influential to the Falls Lake Nutrient Response model than sites with low nutrient loading.
  - > Sample sites more frequently where expected (modeled) values are significantly different from those assumed by NC-DWR in the current version of the Falls Lake Nutrient Response model.



## Monitoring Implications

- Model provides estimates of the **prediction interval** around the estimate of the daily mean
  - > sites with narrow prediction intervals may not need to be sampled as frequently as sites with wide CIs
  - > may want to sample sites with narrow prediction intervals quarterly and use model to statistically test whether samples are significantly different from predictions. If so, adjust frequency of sampling accordingly.



## Monitoring Implications

- Decisions to use models in place of sampling can be adaptive; new data which validate (or do not validate) the models for specific sites can support decisions to reduce (or increase) sampling frequency at any point in time.
- Sampling frequency in year-one need not dictate frequency in all subsequent years; as data accumulate and the models' predictive capacities are reassessed, UNRBA may be able to sample less frequently.



## Model Overview

Model 1: applies to locations for which we have historical data. The model predictions are informed by the historical data specific to the location of interest.

Model 2: applies to locations for which we do not have historical data. Predictions are informed by spatial relationships among locations and water quality.

# Statistical Model for Sites with Existing Data (Model 1)





## Model 1 Overview

- Model 1 is a standard regression model that predicts WQ as a function of location, time and predictors such as precipitation and stream flow
- Model 1 applies a natural log transformation to WQ and assumes normally distributed errors

$$\ln(WQ) = \mu + \gamma_l + \tau_y + \delta_m + \mathbf{x}\boldsymbol{\beta} + \varepsilon$$

where  $\mu$  is the **intercept** parameter,  $\gamma_l$  is a **location** effect,  $\tau_y$  is a **year** effect,  $\delta_m$  is a **month** effect,  $\mathbf{x}\boldsymbol{\beta}$  are the effects of **predictors** (next slide) and  $\varepsilon$  is the **error** term

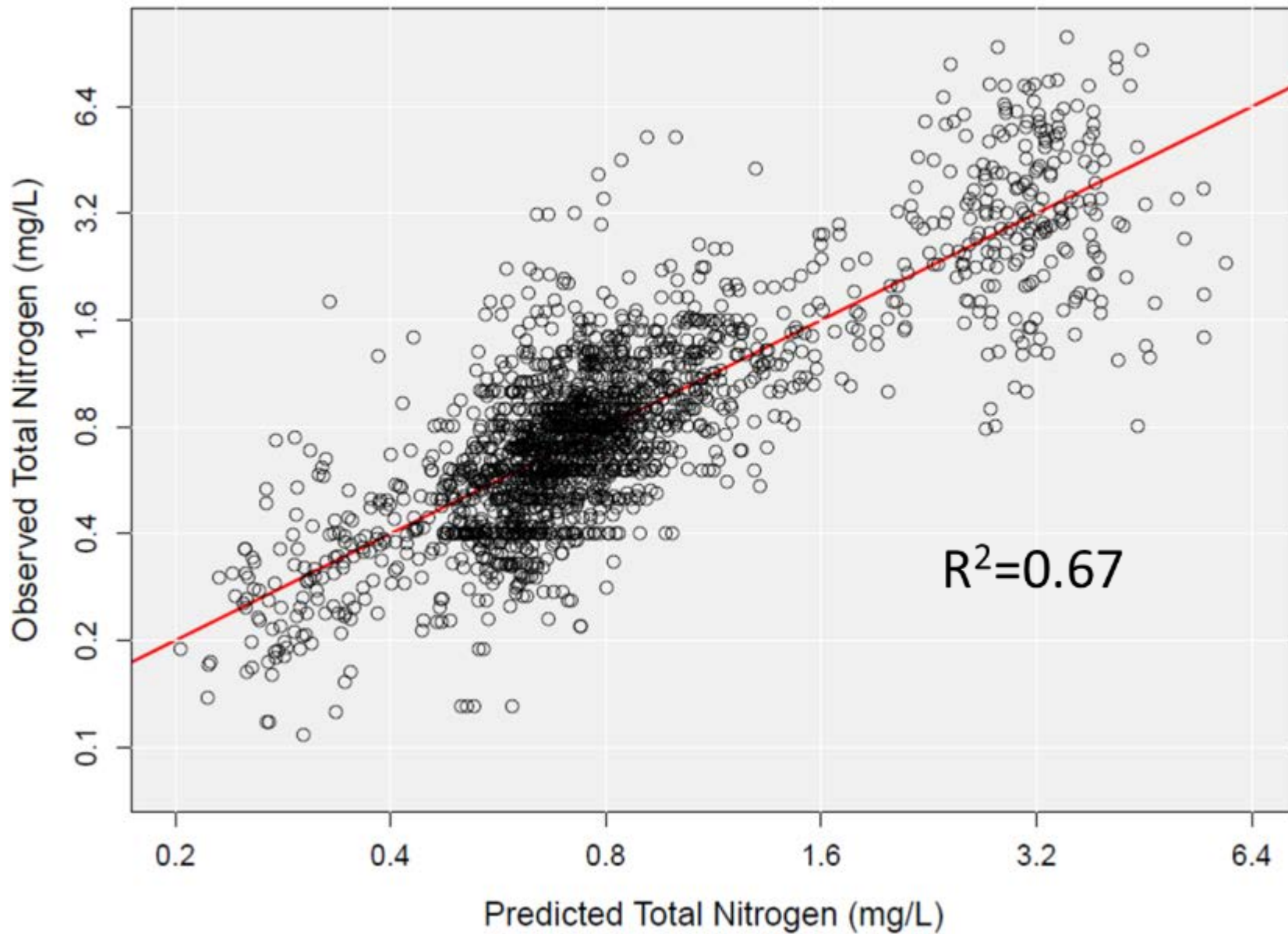




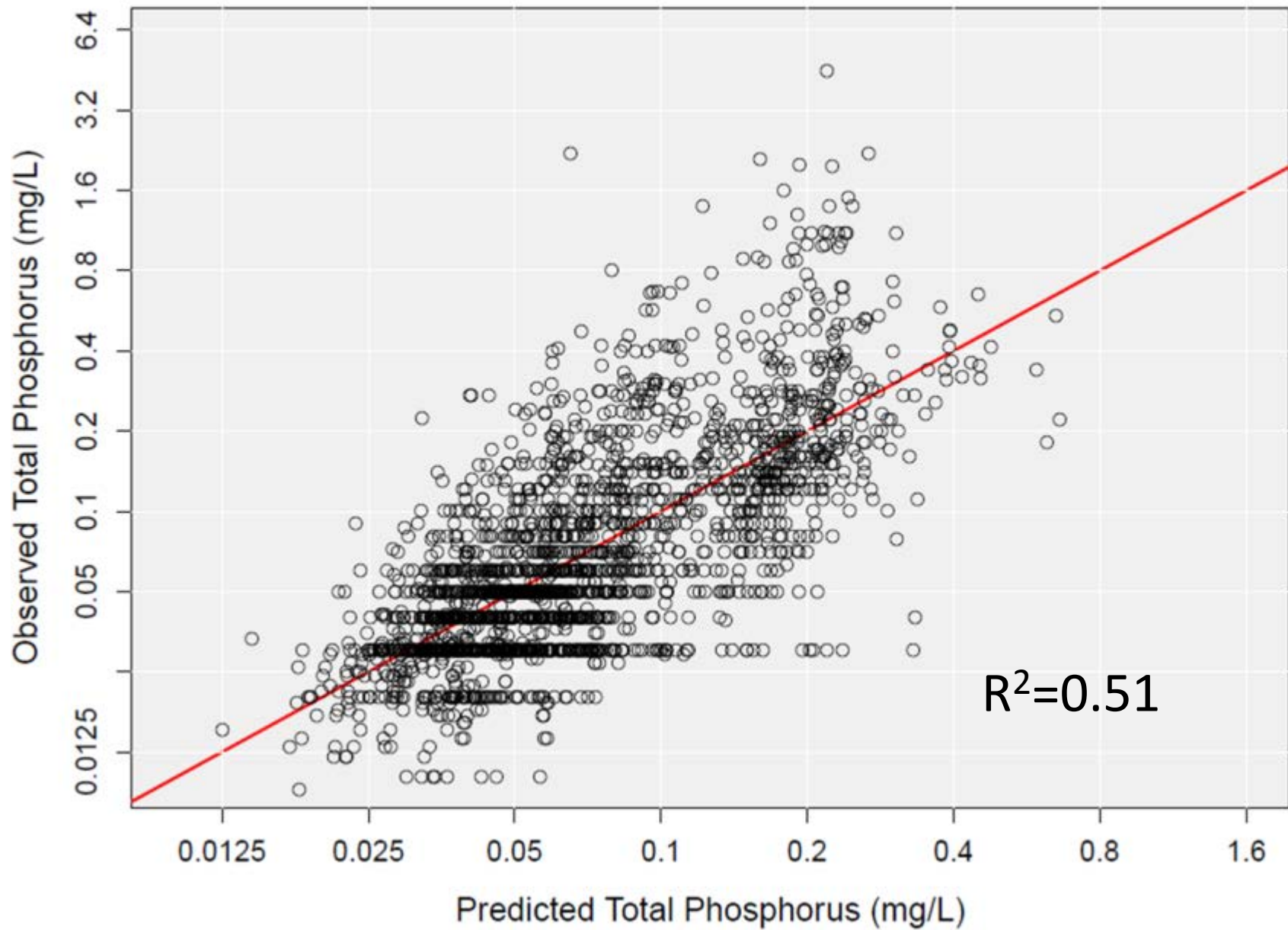
## Model Predictors for TN, TP, TSS and TOC

TN	TP	TSS	TOC
Precipitation	Precipitation	Precipitation	Precipitation
Precipitation* Catchment Area	Precipitation* Catchment Area	Precipitation* Catchment Area	Precipitation* Catchment Area
In(Flow)	In(Flow)	In(Flow)	In(Flow)
[In(Flow)] <sup>2</sup>	[In(Flow)] <sup>2</sup>	[In(Flow)] <sup>2</sup>	[In(Flow)] <sup>2</sup>
Maximum Daily Temperature	Maximum Daily Temperature	Maximum Daily Temperature	
Precipitation on Day Prior	Precipitation on Day Prior	Precipitation on Day Prior	
Precipitation Two Days Prior	Precipitation Two Days Prior	Precipitation Two Days Prior	

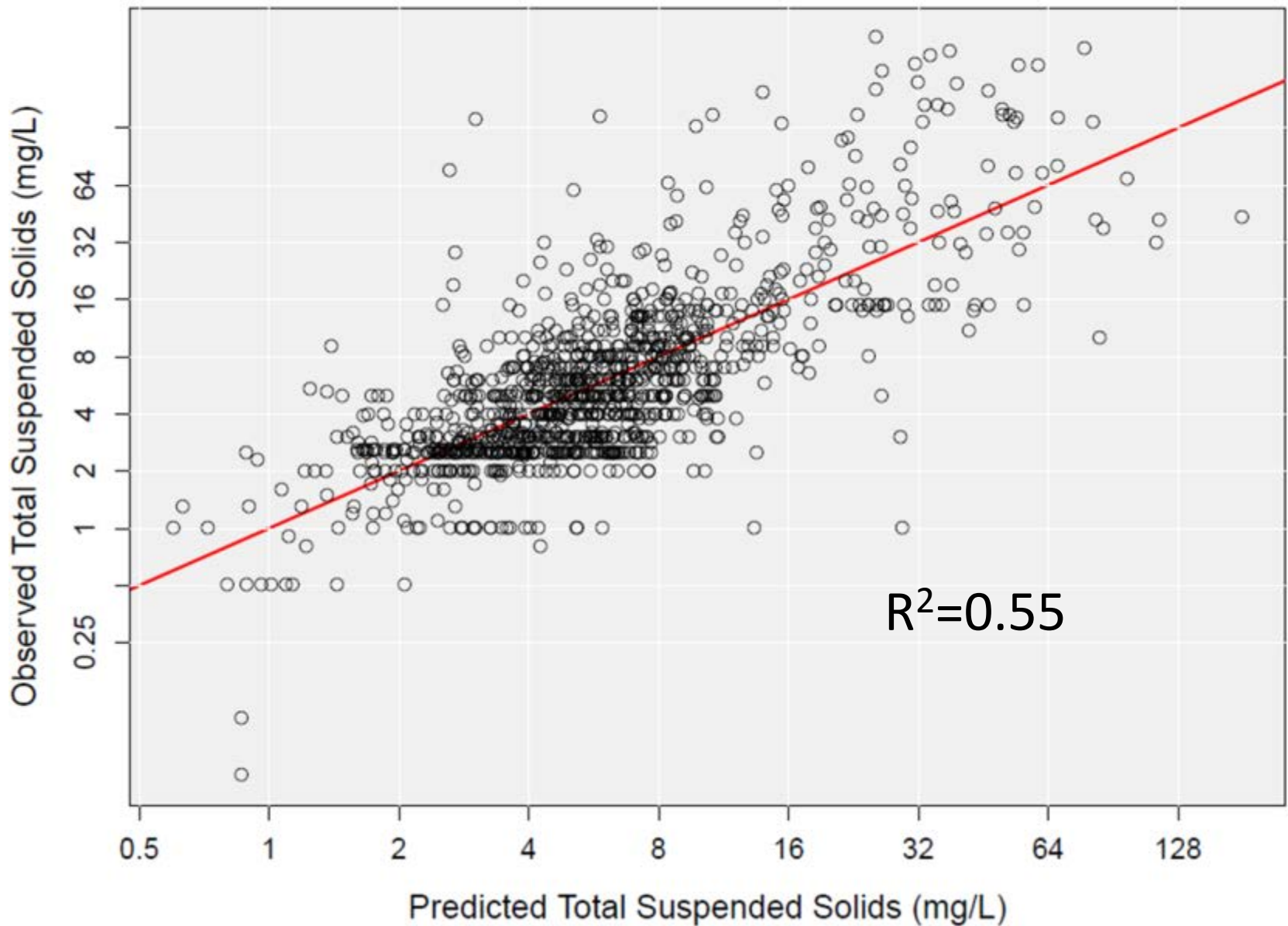
Model 1 Predicted Total Nitrogen Versus Observed Total Nitrogen



Model 1 Predicted Total Phosphorus Versus Observed Total Phosphorus

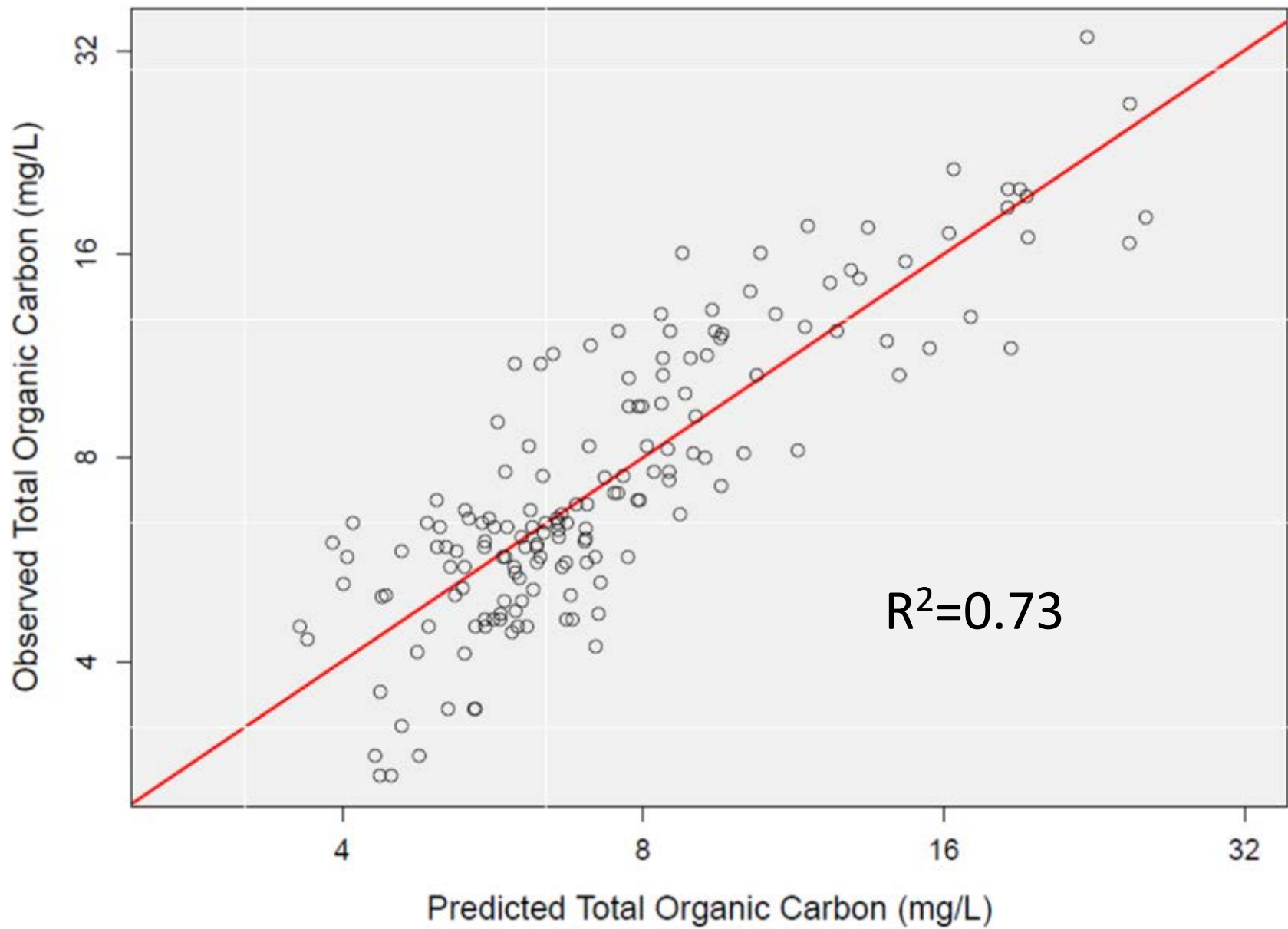


Model 1 Predicted Total Suspended Solids Versus Observed Total Suspended Solids

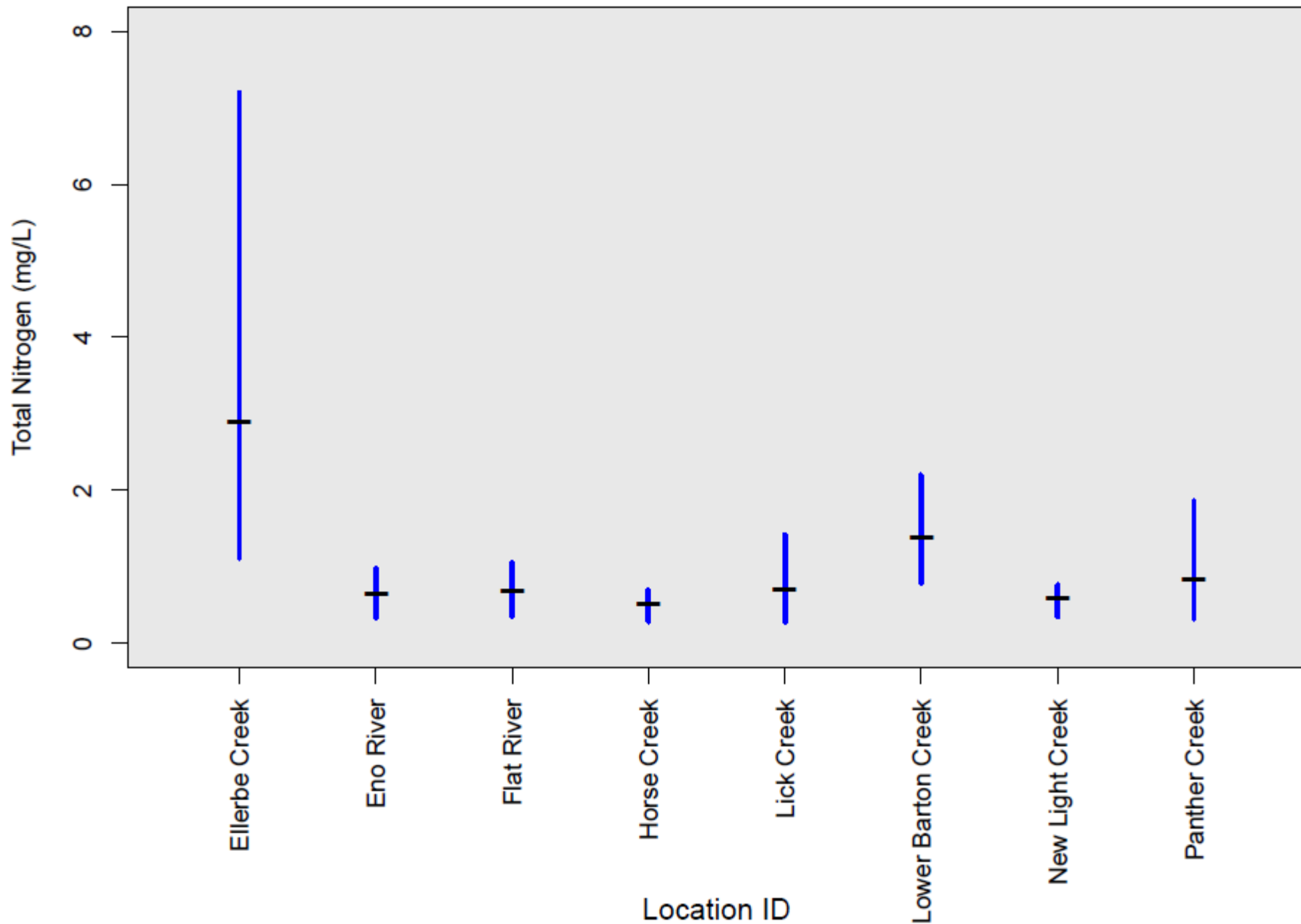




Model 1 Predicted Total Organic Carbon Versus Observed Total Organic Carbon

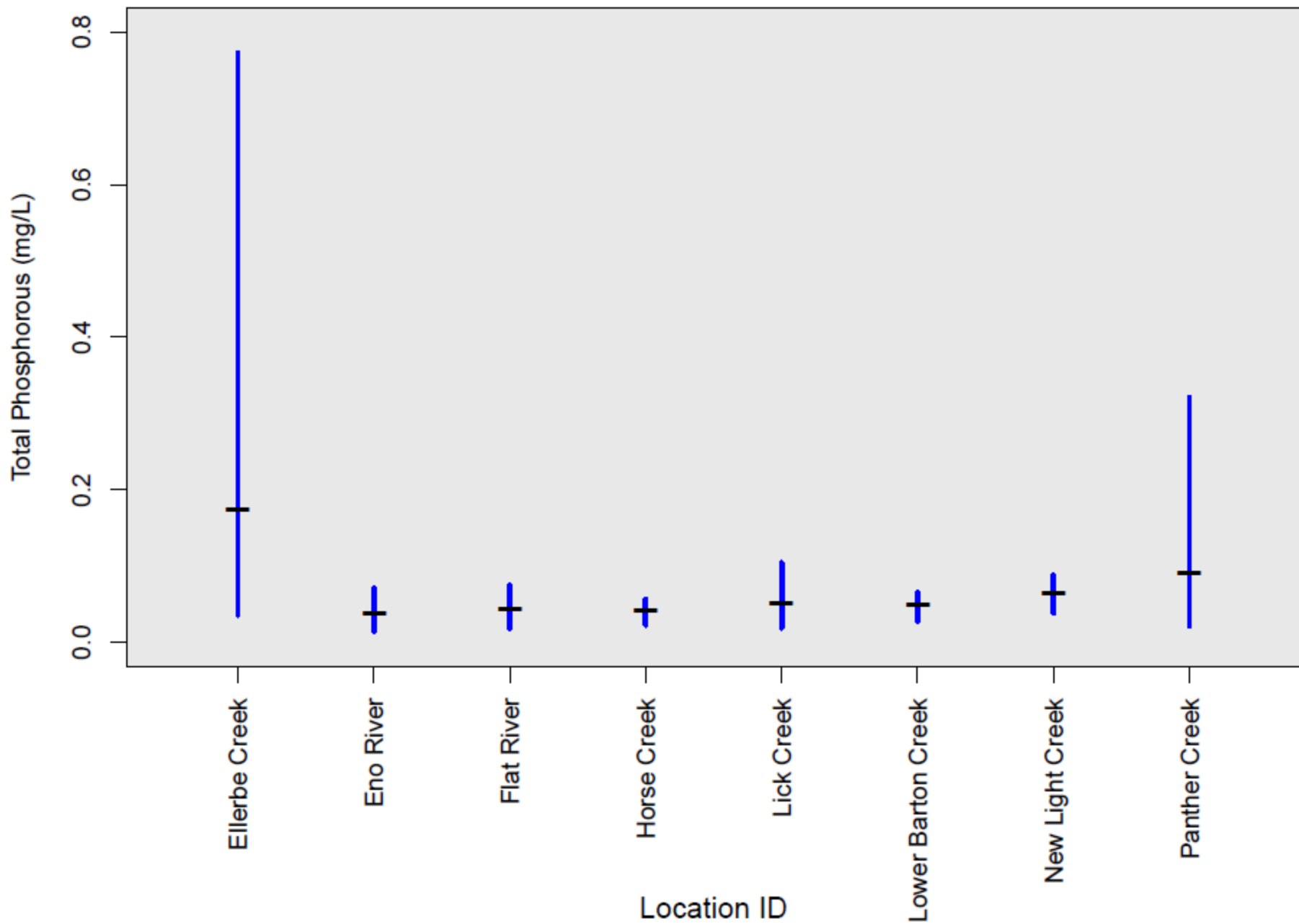


Model-Adjusted 90% Prediction Intervals of Total Nitrogen by Location

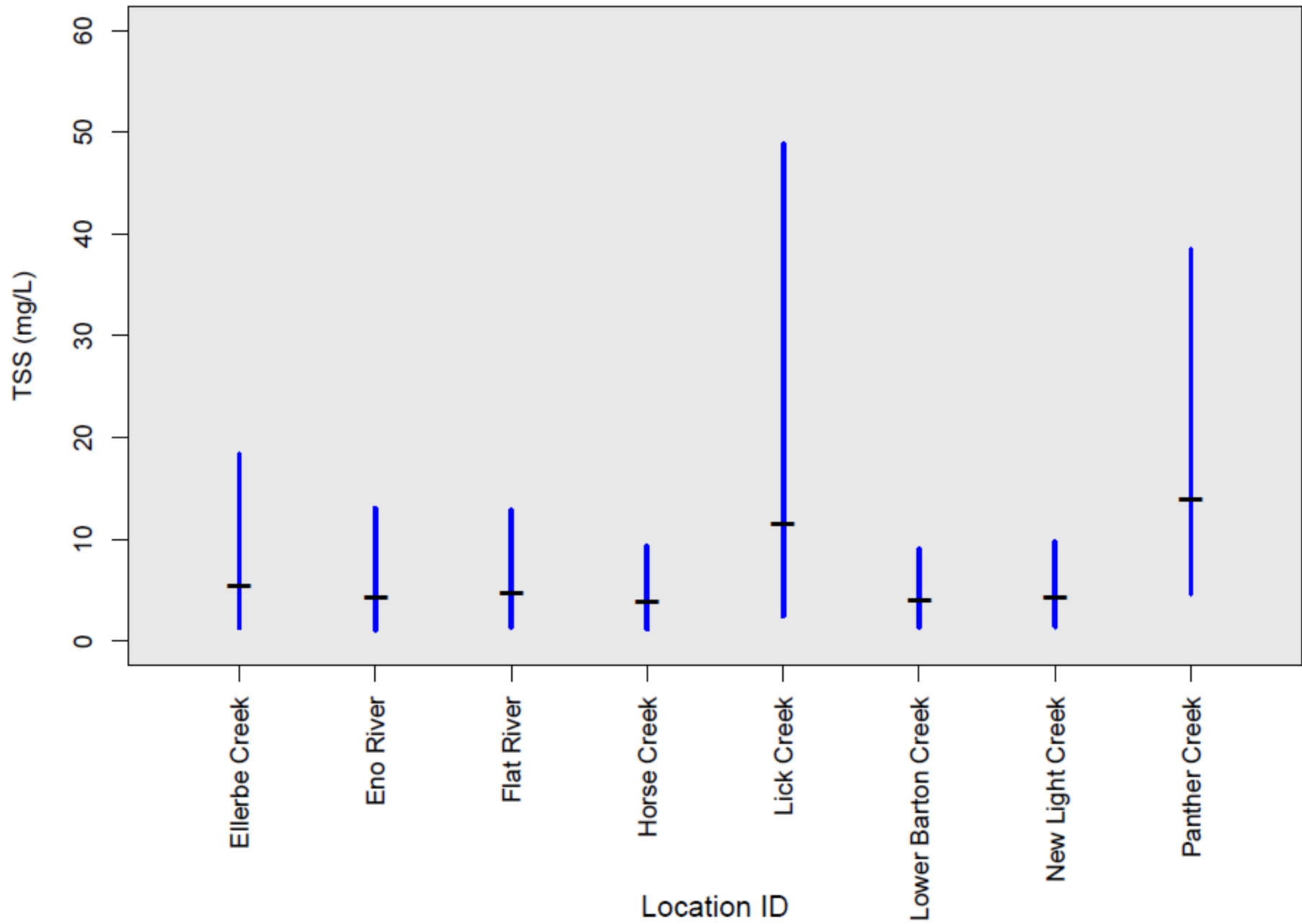




Model-Adjusted 90% Prediction Intervals of Total Phosphorus by Location



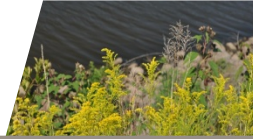
Model-Adjusted 90% Prediction Intervals of Total Suspended Solids by Location





## Using results to inform sampling frequency

- Some sites have large prediction intervals for all three parameters, for example **Panther Creek**, and may be candidates for more frequent sampling than other sites.
- Some sites are well predicted for all three parameters, for example **Horse Creek**, and may not need to be sampled as frequently as other sites.
- Samples can be continually compared to model predictions and sampling frequency adjusted accordingly.



# Using Statistical Model 1 to Determine Monitoring Frequency



## Sampling Frequency Analysis

- The relationship between the number of samples used in model development and uncertainty in model predictions of water quality parameters can be obtained from Model 1
- The uncertainty calculation occurs as follows:

Percent relative error =

$$\frac{\text{Upper bound of } X\% \text{ CI for } WQ_{\text{predicted}} - \text{average } WQ}{\text{average } WQ} \times 100\%$$

where the upper bound is from a Confidence Interval on the model estimated WQ



For example

- If average WQ is 10 and for a given sample size we expect a 90% CI from 5 to 15, then the expected relative error rate (for 90% confidence) is:

$$\left( \frac{15 - 10}{10} \right) \times 100\% = 50\% \text{ error}$$





# Total number of samples needed

Approximate number of samples needed to achieve specified confidence/relative error rates

		Relative Error Rate				
		10%	20%	30%	40%	50%
<b>90% Confidence</b>	TN	99	27	13	8	6
	TP	218	60	29	18	13
	TSS	>260	105	51	31	22
		Relative Error Rate				
		10%	20%	30%	40%	50%
<b>95% Confidence</b>	TN	140	39	19	12	8
	TP	>260	85	41	25	18
	TSS	>260	149	72	44	30

Example: monthly sampling for 5 years (n=60) would allow the model to estimate daily values of TP (and TN) for which we could be 90% certain that the predictions are within 20% of the true value. For TSS the same monitoring frequency would result in estimates which have a higher (30%) relative error.

# Statistical Model for Prediction of Water Quality at Locations with No Data (Model 2)



## Model 2 Characteristics

- Model 2 extends Model 1
- Instead of location-specific parameters, Model 2 is a spatial model able to consider historically unmonitored location
- Model incorporates thin-plate splines which are used in spatial statistical modeling
- Spline modeling is a modern alternative to kriging methods



## Model 2 Characteristics

The model is similar to Model 1:

$$\ln(WQ) = f(lat, long) + \tau_y + \delta_m + \mathbf{x}\boldsymbol{\beta} + \varepsilon$$

where now  $f(lat, long)$  is the spatial component, modeling WQ as a function of latitude and longitude; the other components are analogous to Model 1



Model 2 includes additional physical predictor variables that were not useful for Model 1

**Precipitation**

**Precipitation\*Catchment Area**

**In(Flow)**

**[In(Flow)]<sup>2</sup>**

**Terrain slope index 1**

**Terrain slope index 2**

**Elevation index 1**

**Elevation index 2**

**Percent impervious surface in 2006**

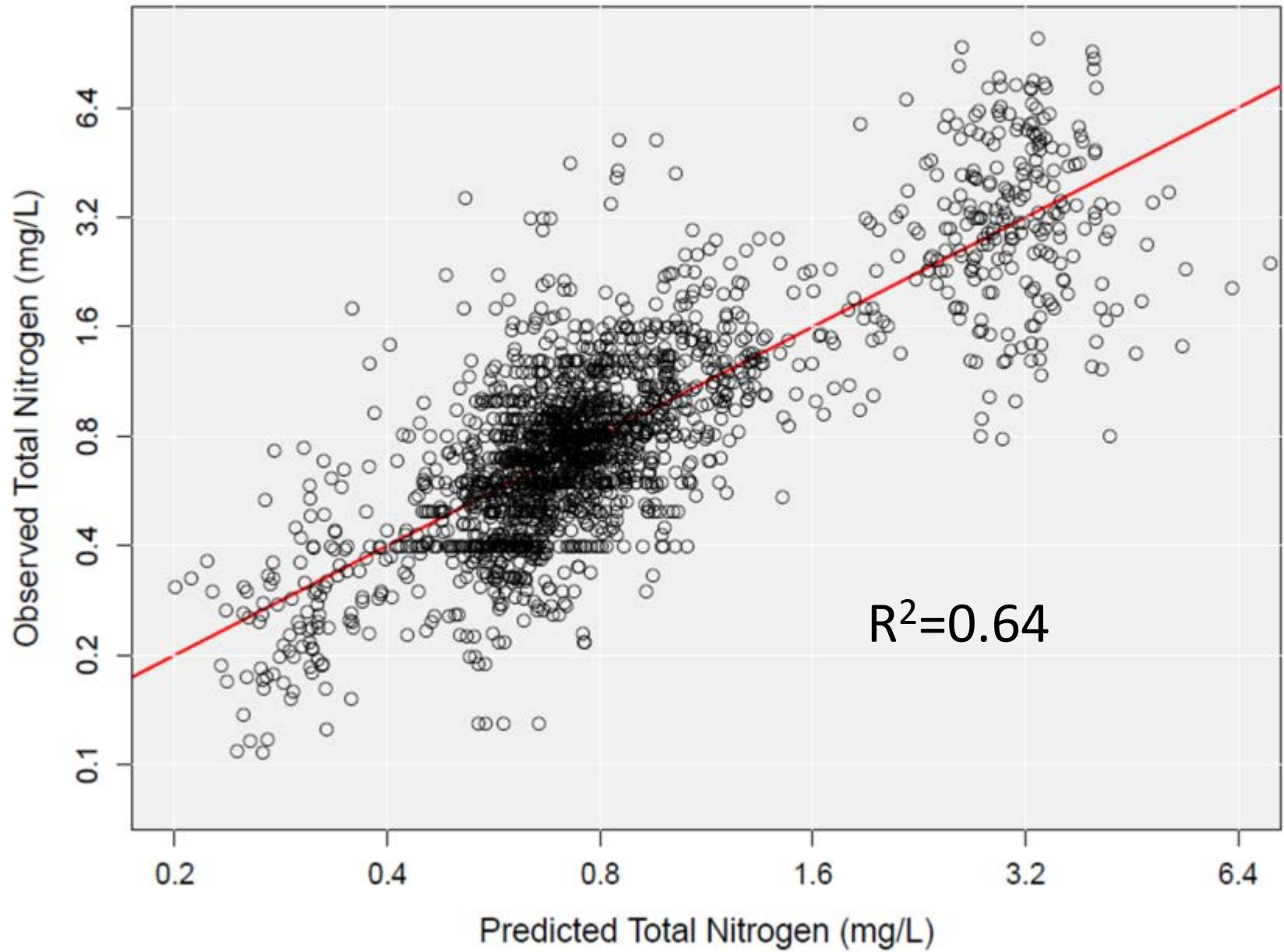
**Percent wetlands in 2006**

**5-year peak flow level**

**Percent forest in 2006**

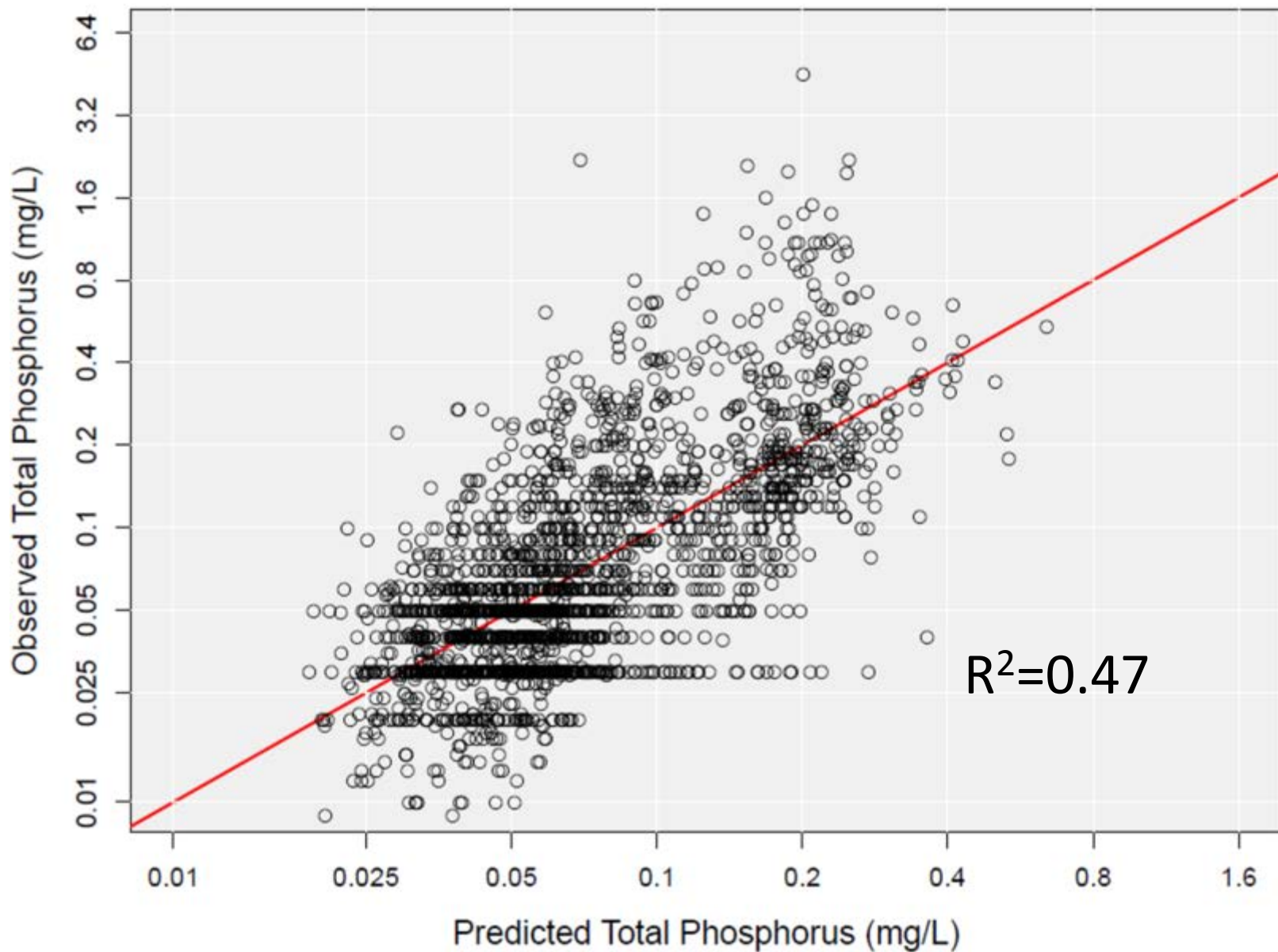


Model 2 Predicted Total Nitrogen Versus Observed Total Nitrogen

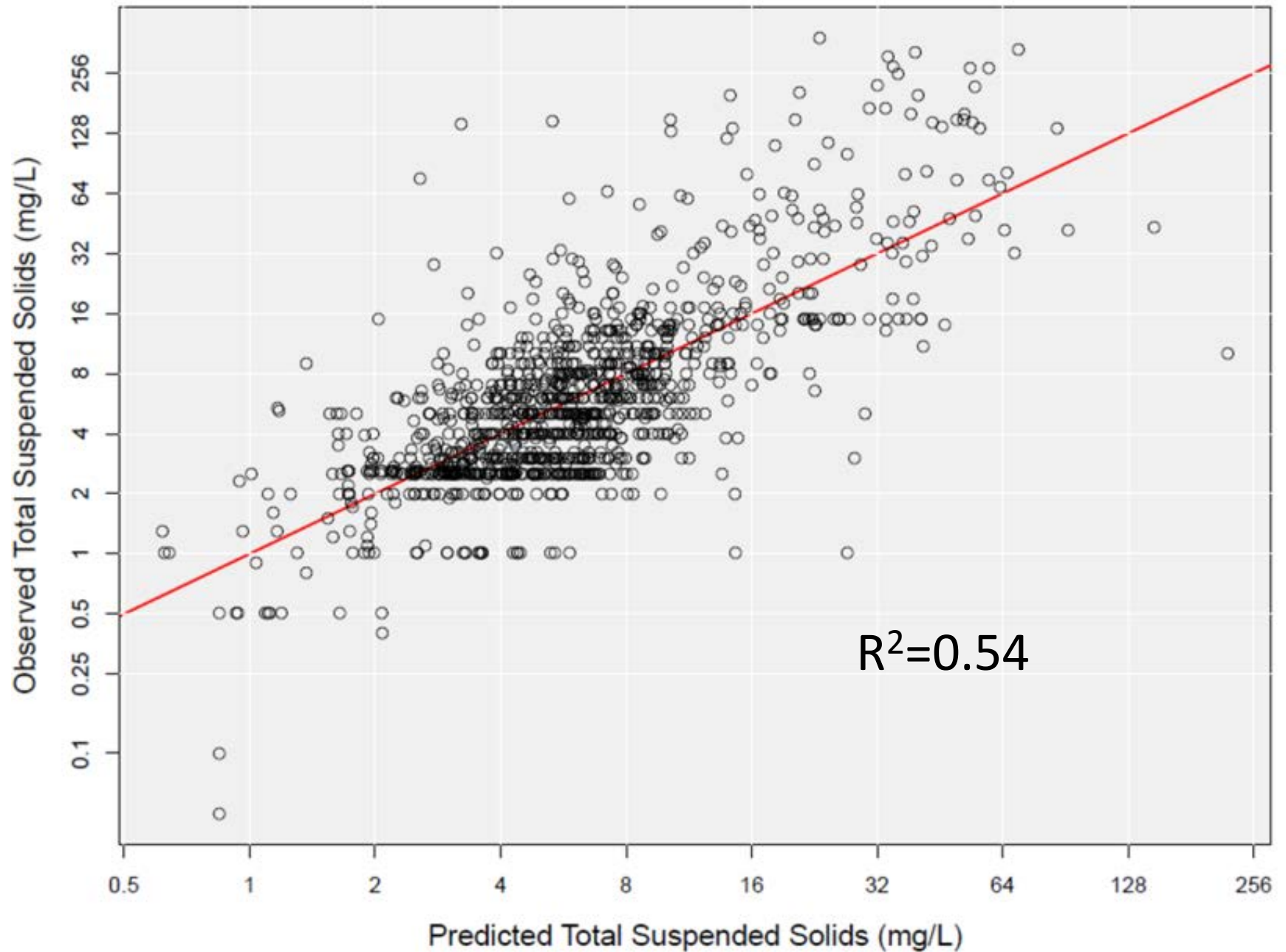




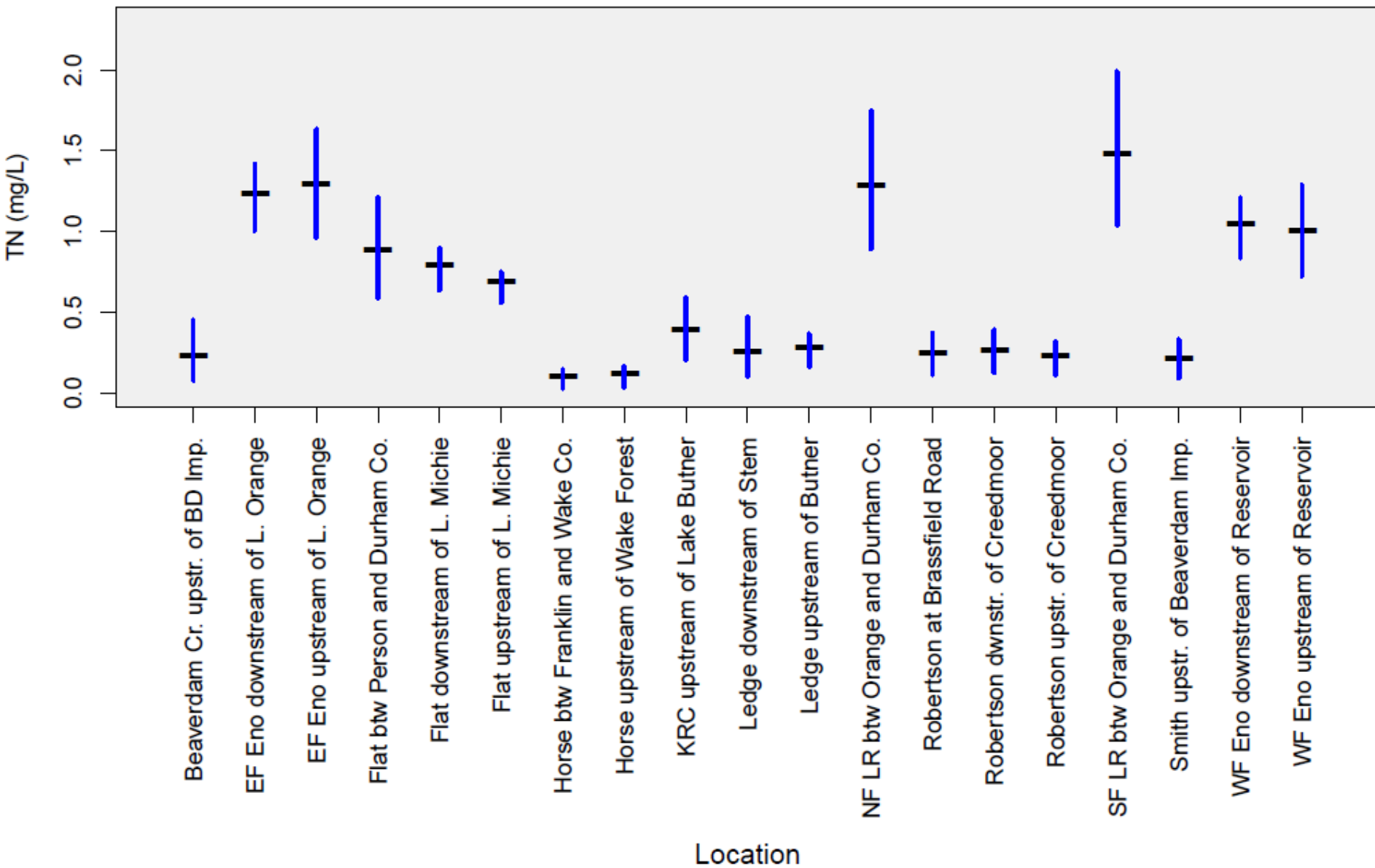
Model 2 Predicted Total Phosphorus Versus Observed Total Phosphorus



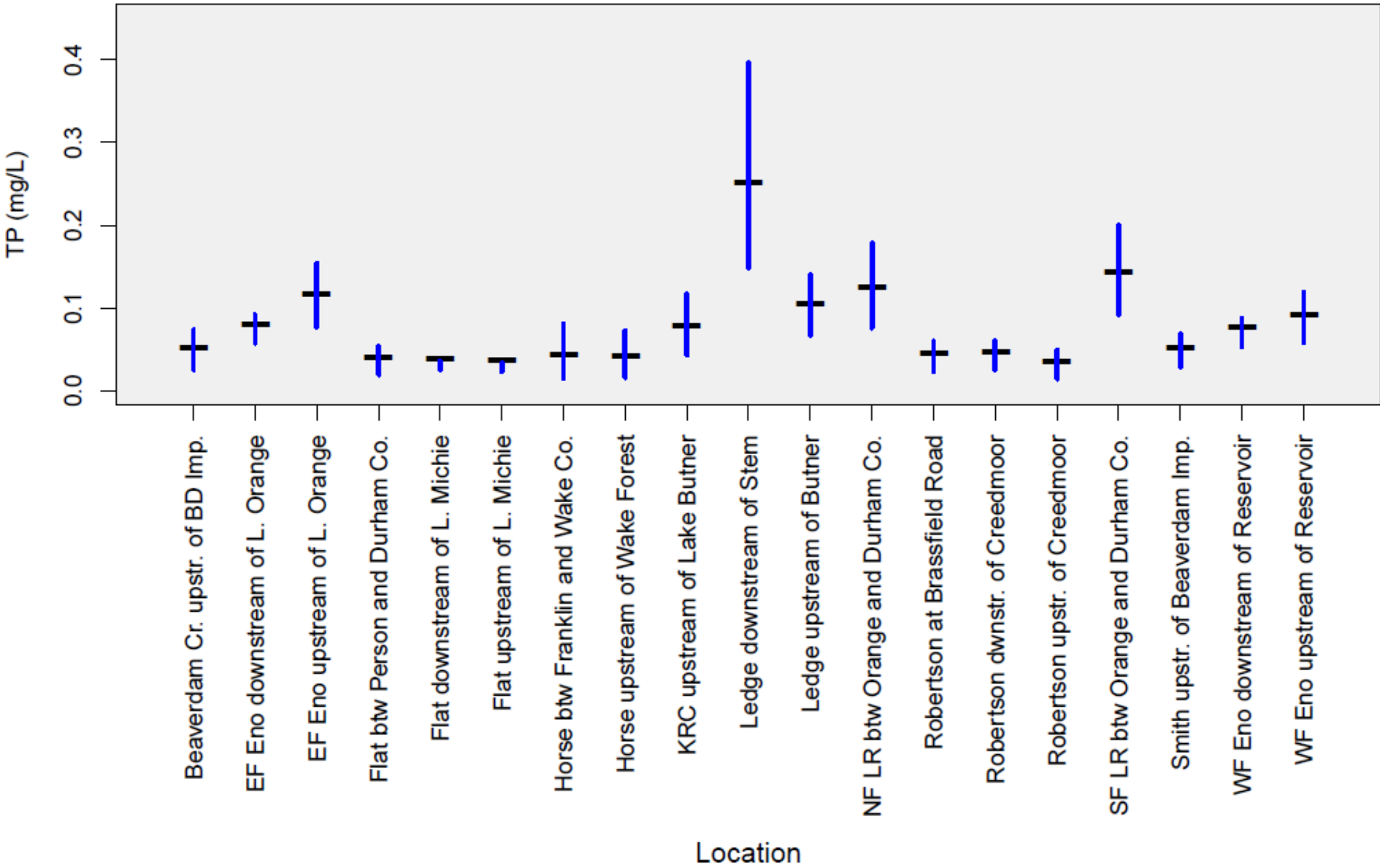
Model 2 Predicted Total Suspended Solids Versus Observed Total Suspended Solids



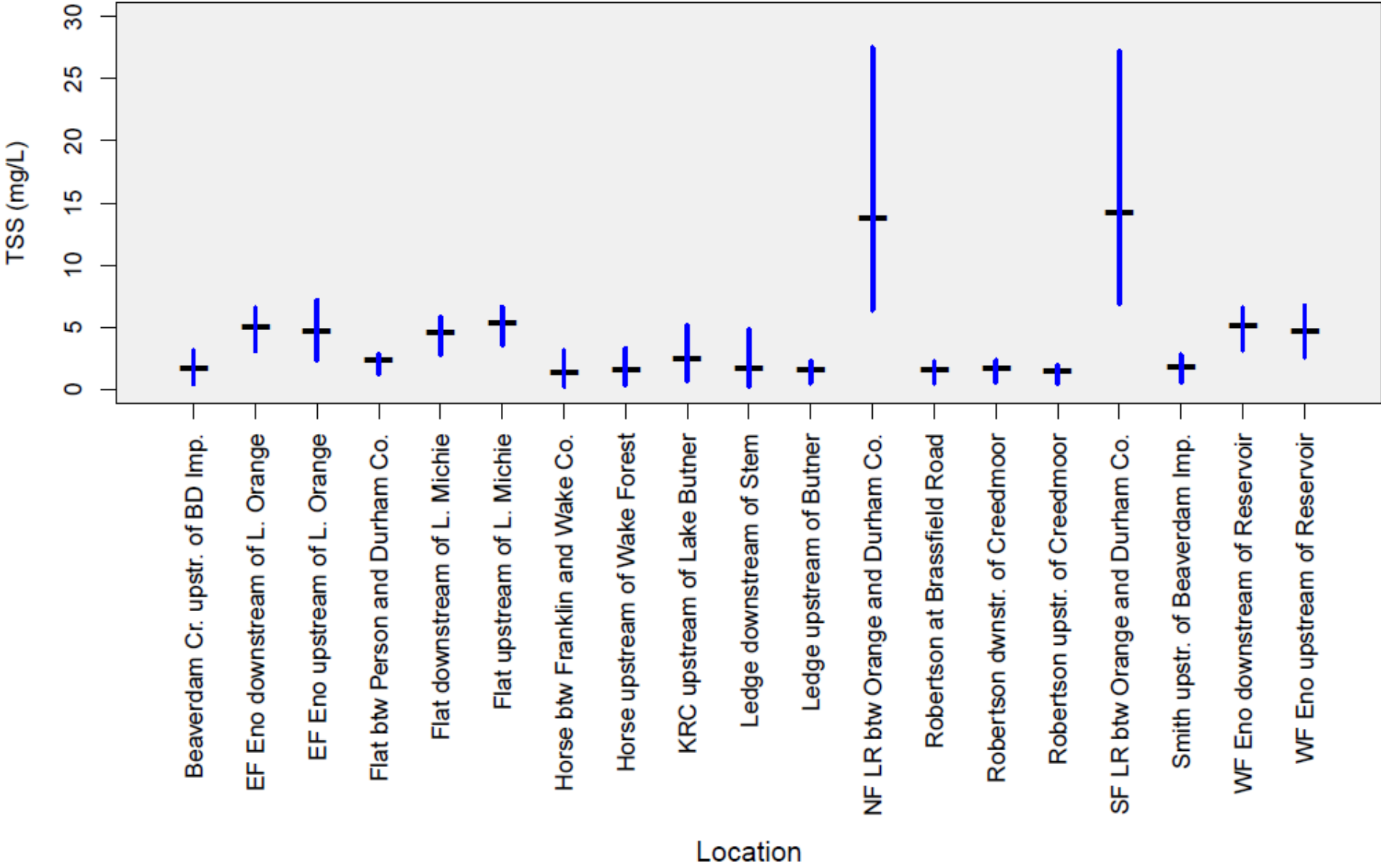
## Spatial Model Predicted Total Nitrogen and 90% Prediction Intervals at Jurisdictional Boundary Locations

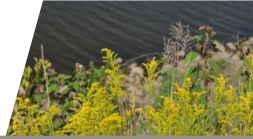


## Spatial Model Predicted Total Phosphorus and 90% Prediction Intervals at Jurisdictional Boundary Locations



## Spatial Model Predicted Total Suspended Solids and 90% Prediction Intervals at Jurisdictional Boundary Locations



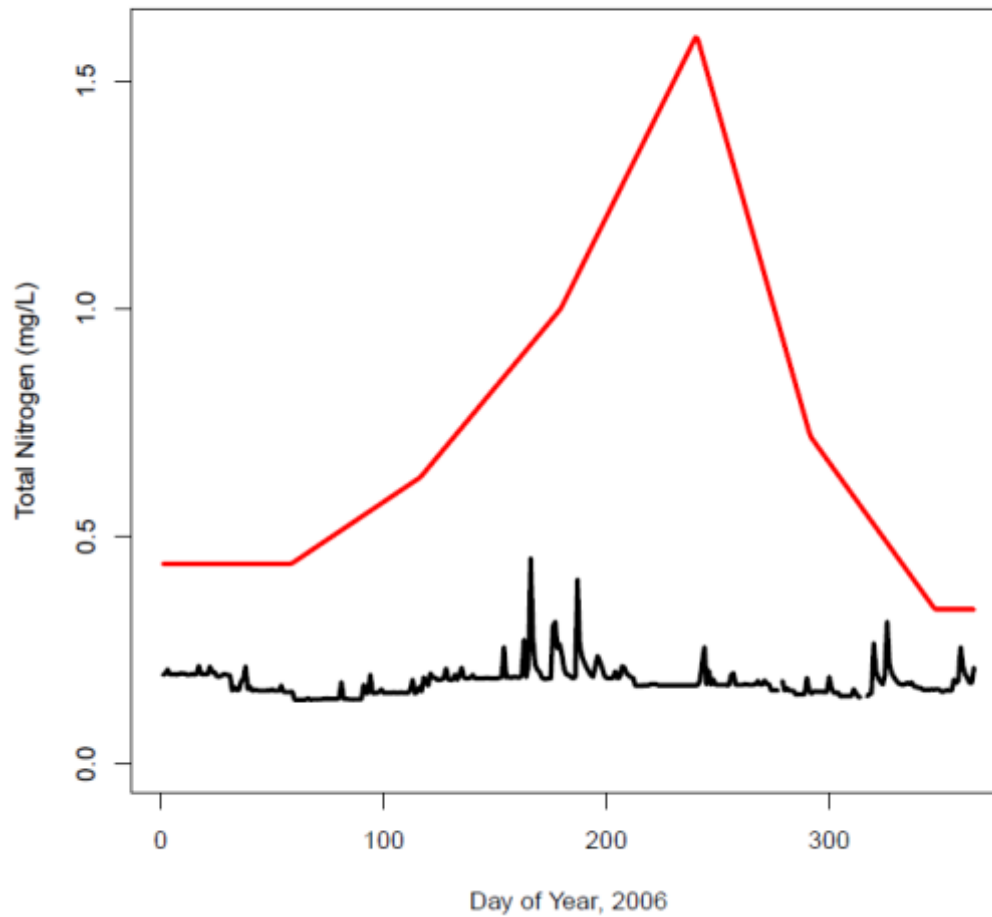


# Comparisons of Model Predictions with Input Nutrient Concentrations for the DWR Falls Lake Nutrient Response Model (EFDC)



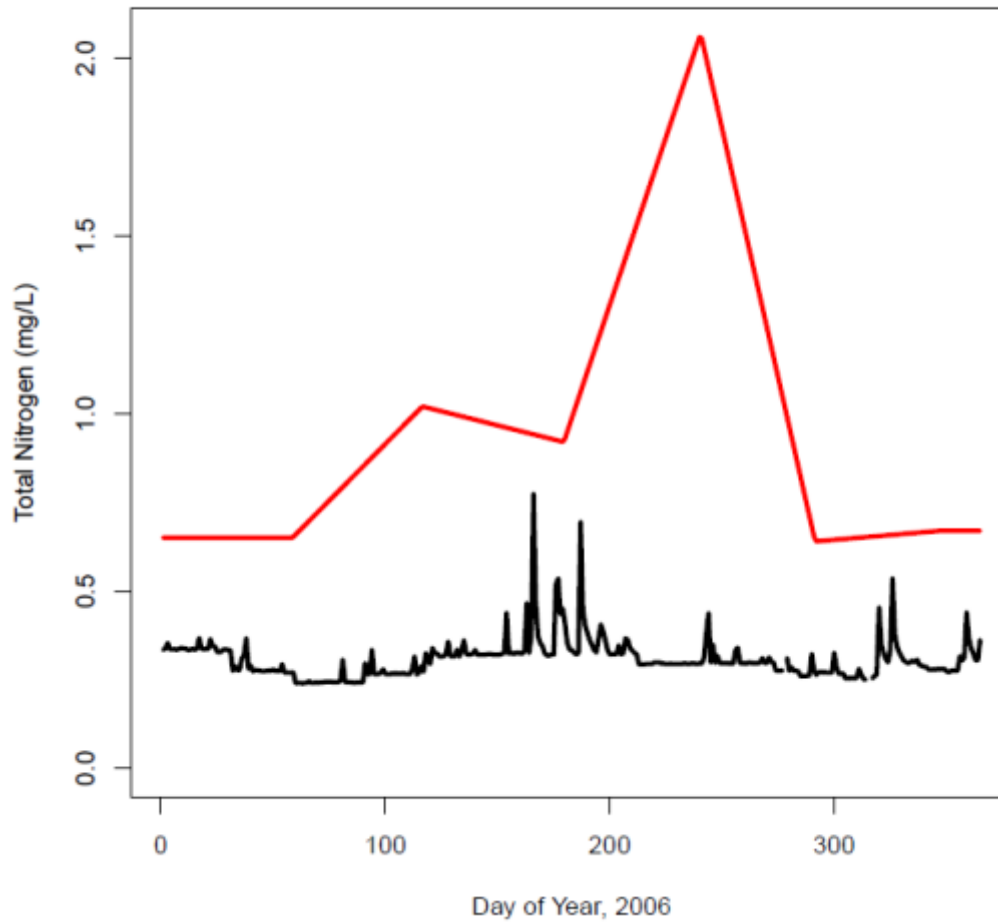


EFDC Input Total Nitrogen versus  
Statistical Model Predictions at F019\_RobertsonAtBrass



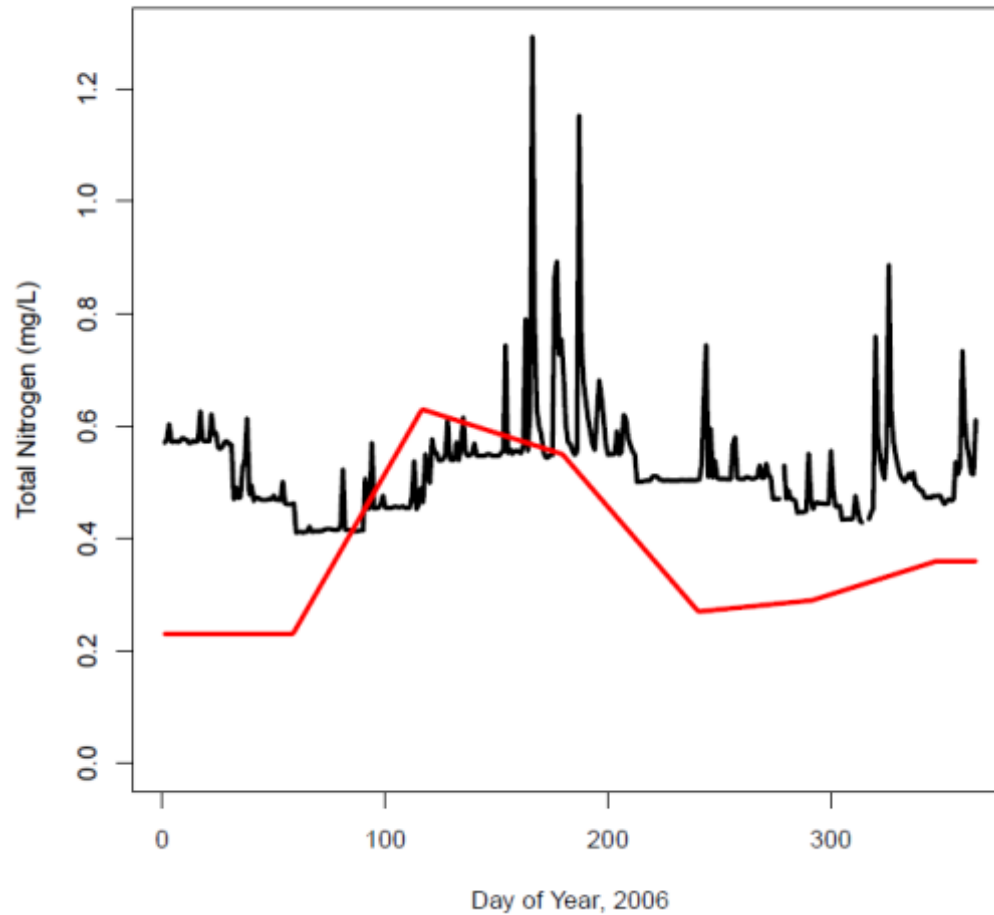


EFDC Input Total Nitrogen versus  
Statistical Model Predictions at F021\_LedgeCr





EFDC Input Total Nitrogen versus  
Statistical Model Predictions at F025\_HoneycuttCr





**EFDC Input Total Nitrogen versus  
Statistical Model Predictions at F020\_UnnamedTrib**

