

Identifying Critical Source Areas to Enhance Water Quality: Integrated Analysis to Reveal Key Variables and Thresholds in Choctawhatchee

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Introduction

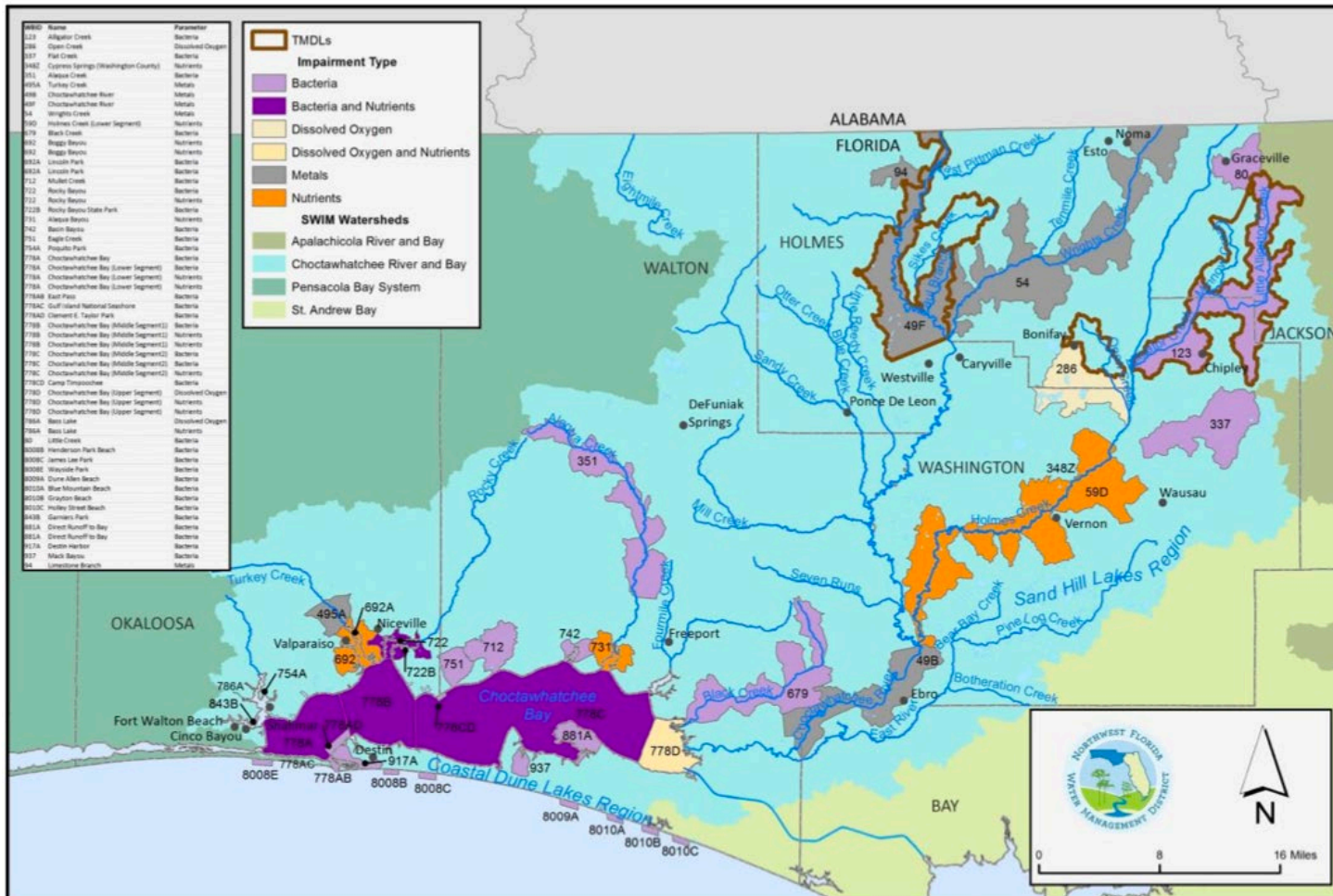
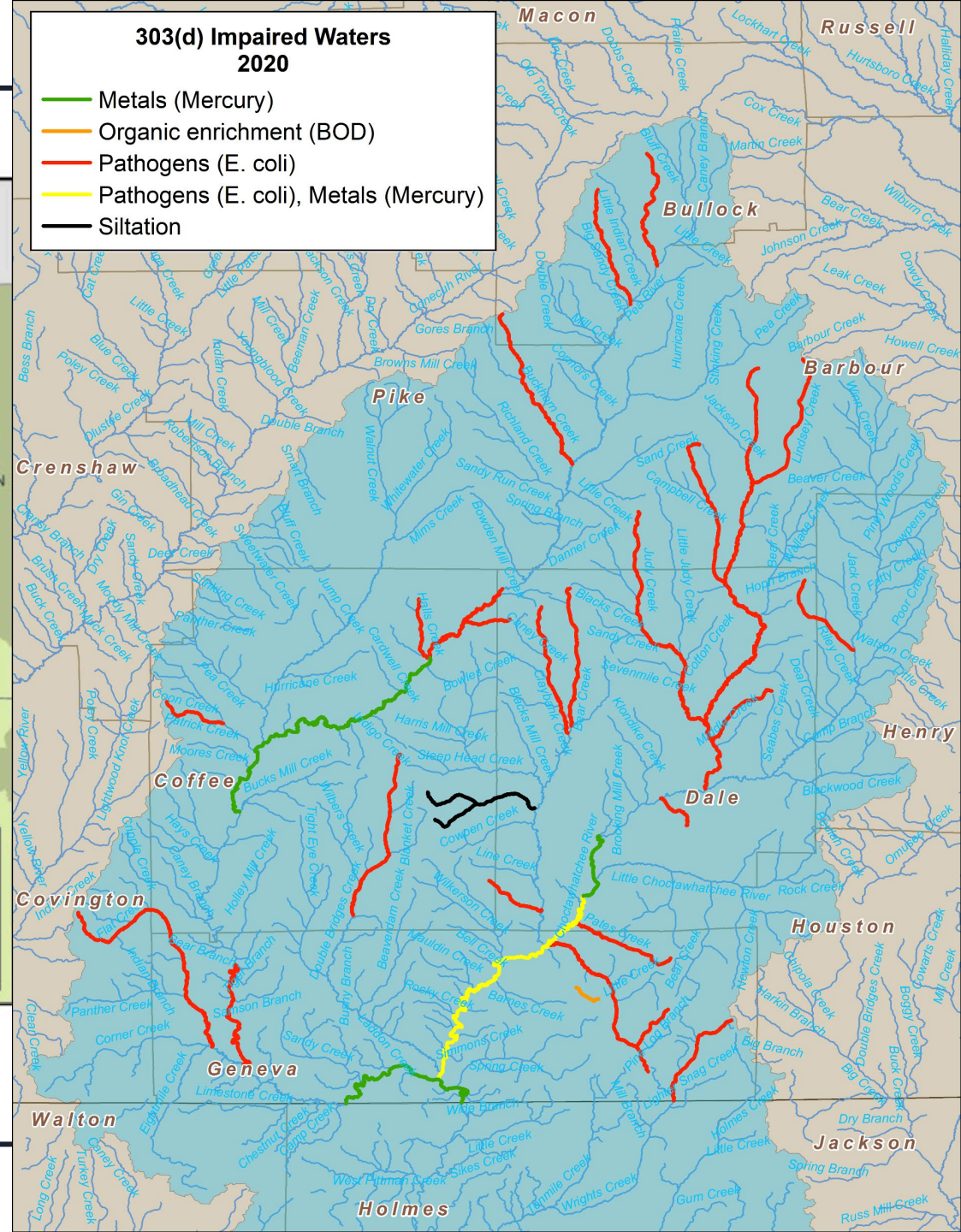


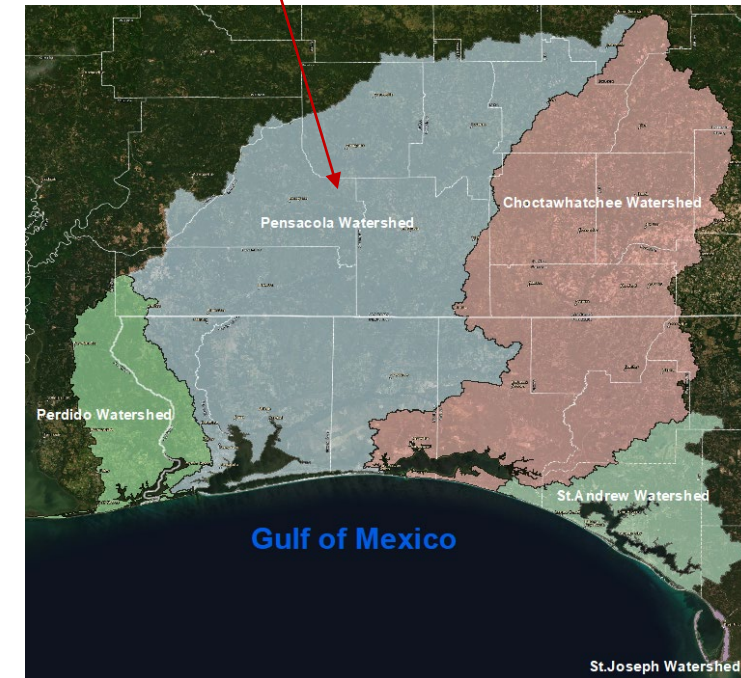
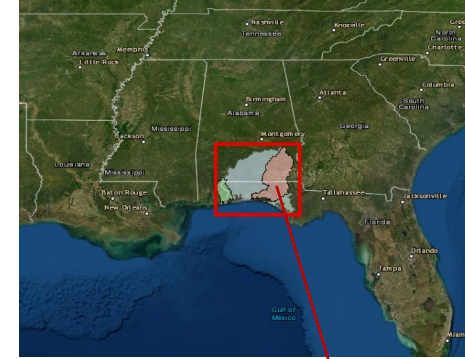
Figure 3-1 Impaired Waters in Florida's Portion of the Watershed

By: *Choctawhatchee Bay Estuary Program*
Technical Advisory Committee, 2021



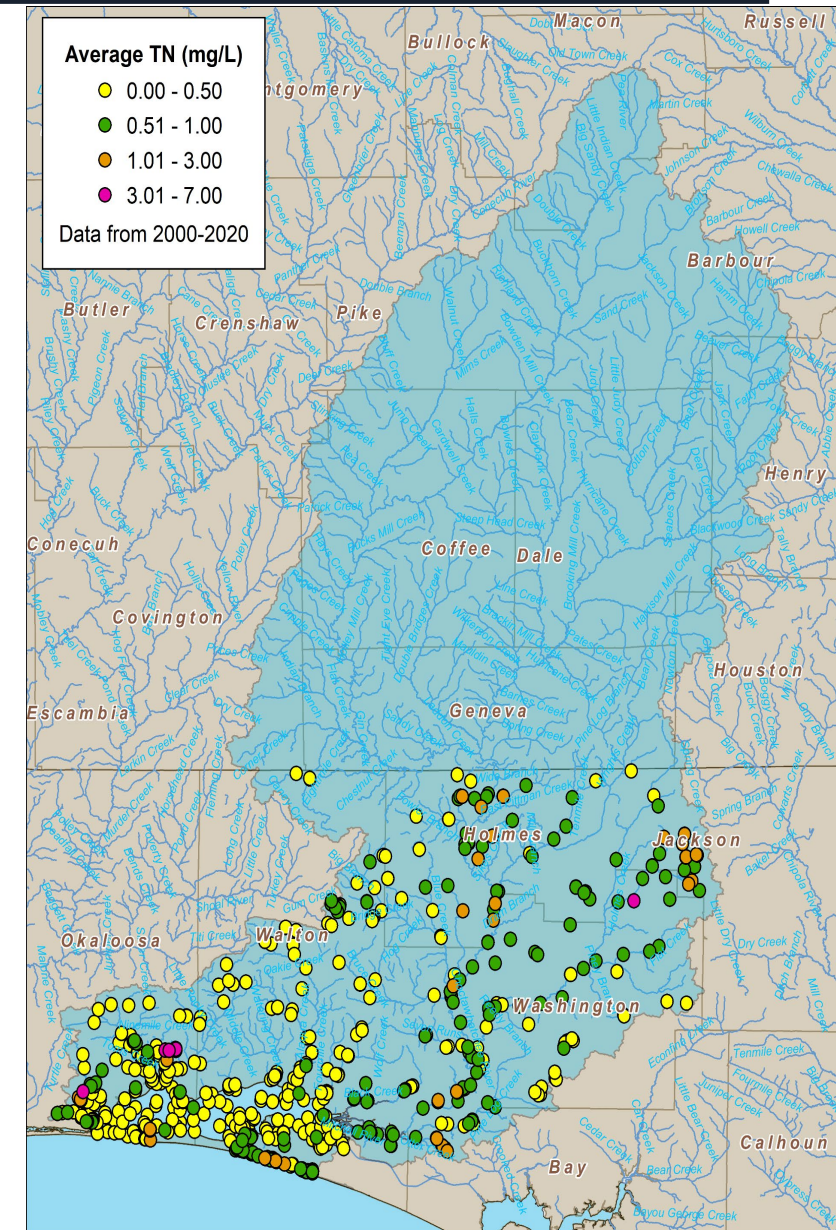
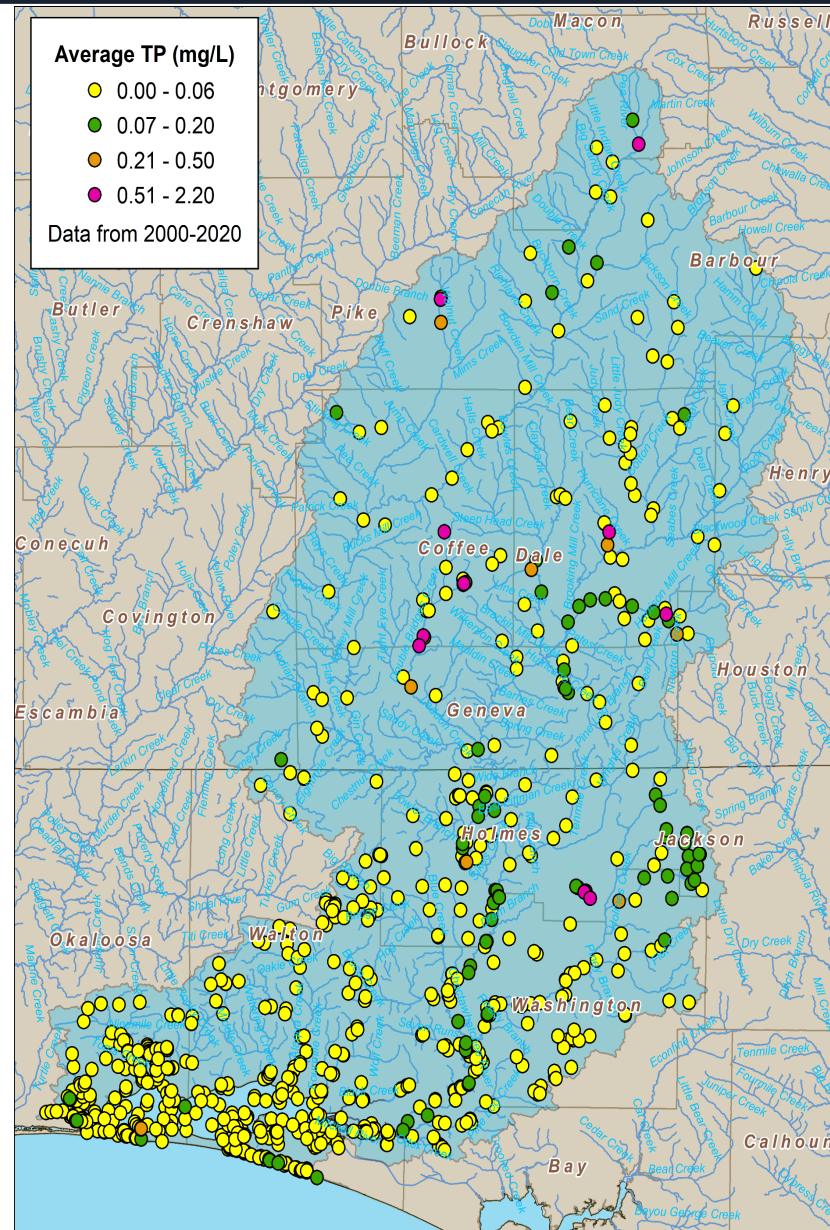
Introduction

- A comprehensive project was launched in 2020 to identify and implement methods to assess impacts and stressors in the estuary systems of Florida Panhandle.
- The project has been working with stakeholders to
 - Identify main stressors
 - Develop conceptual models,
 - Develop adaptive management frameworks,
 - Understand how land management impacts the water quality of estuaries
 - Develop predictive tools to examine how management actions could reduce impacts



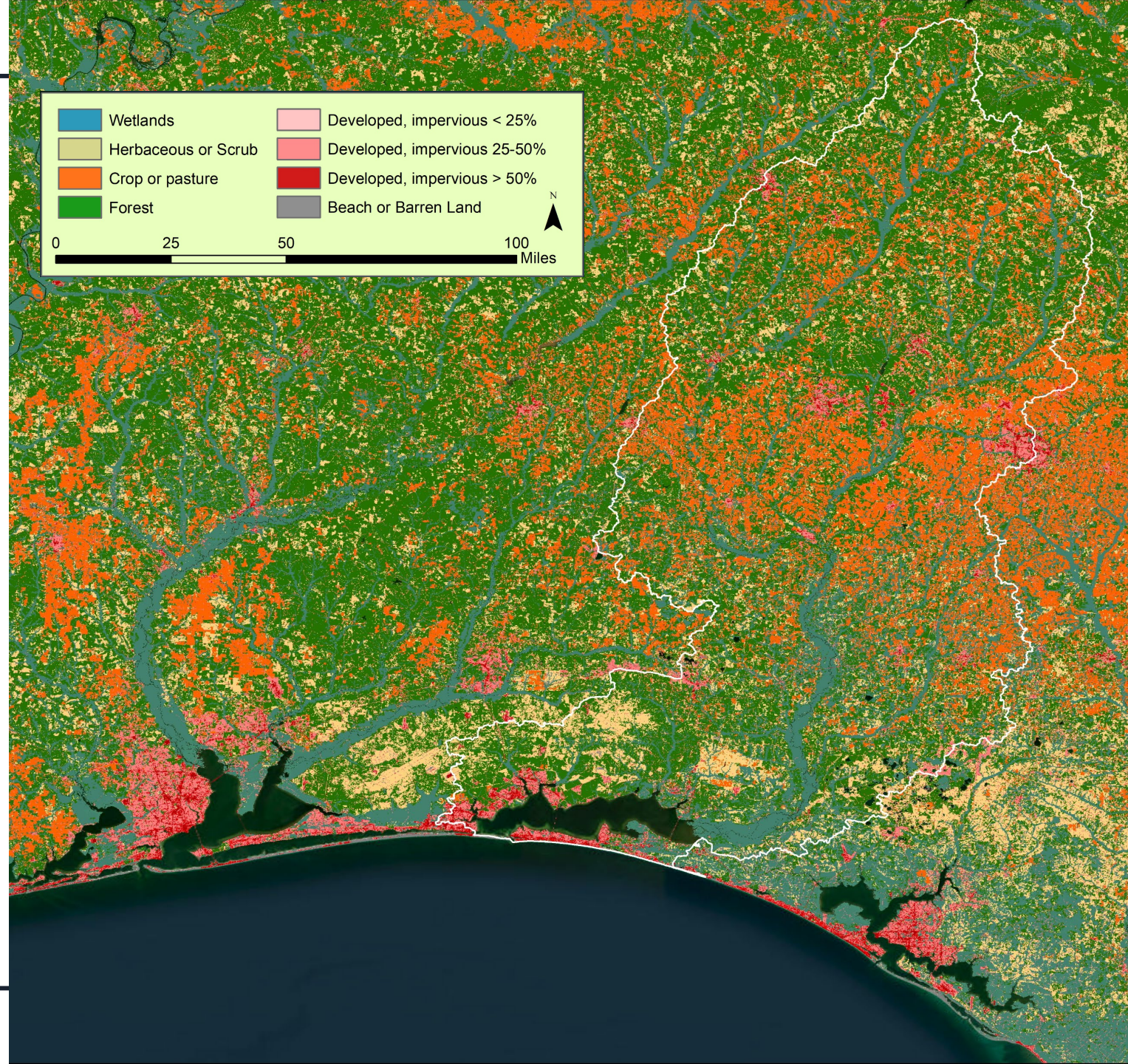
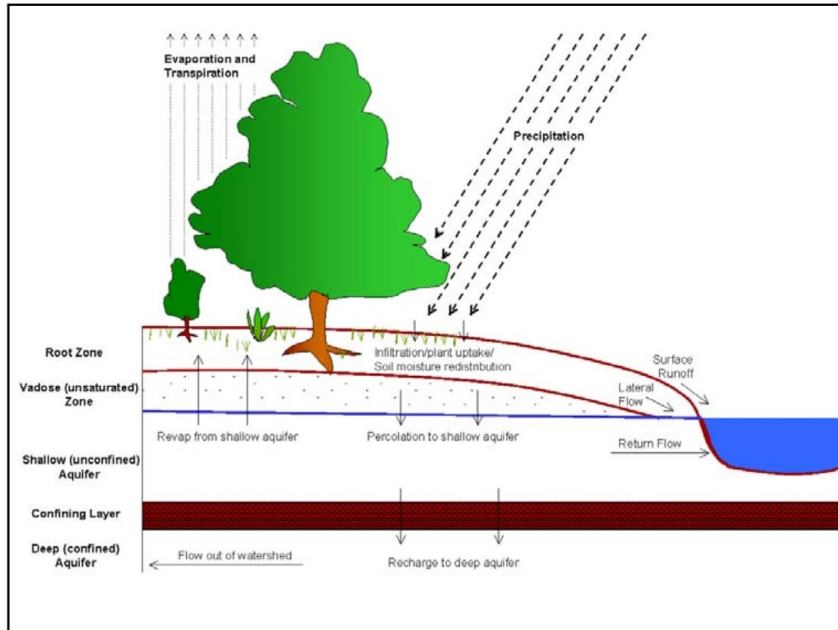
Distributed hydrological model

- Identifying the critical areas of non-point sources of pollutants from watersheds is important to reduce inflows
- Develop and prioritize best management practices and restoration projects to reduce nutrient flows in water bodies



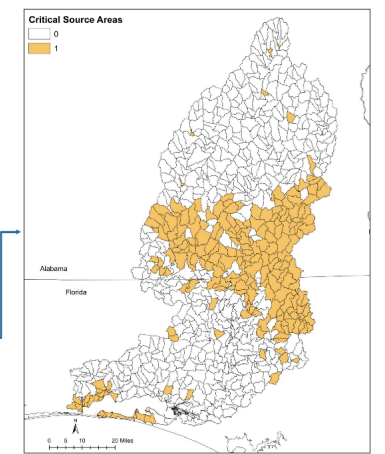
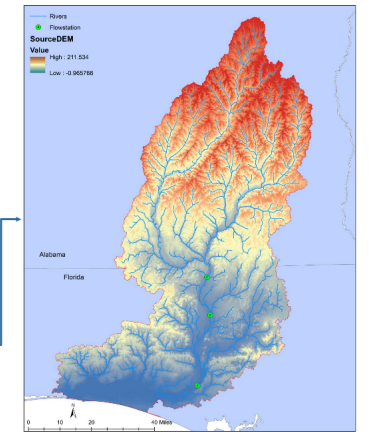
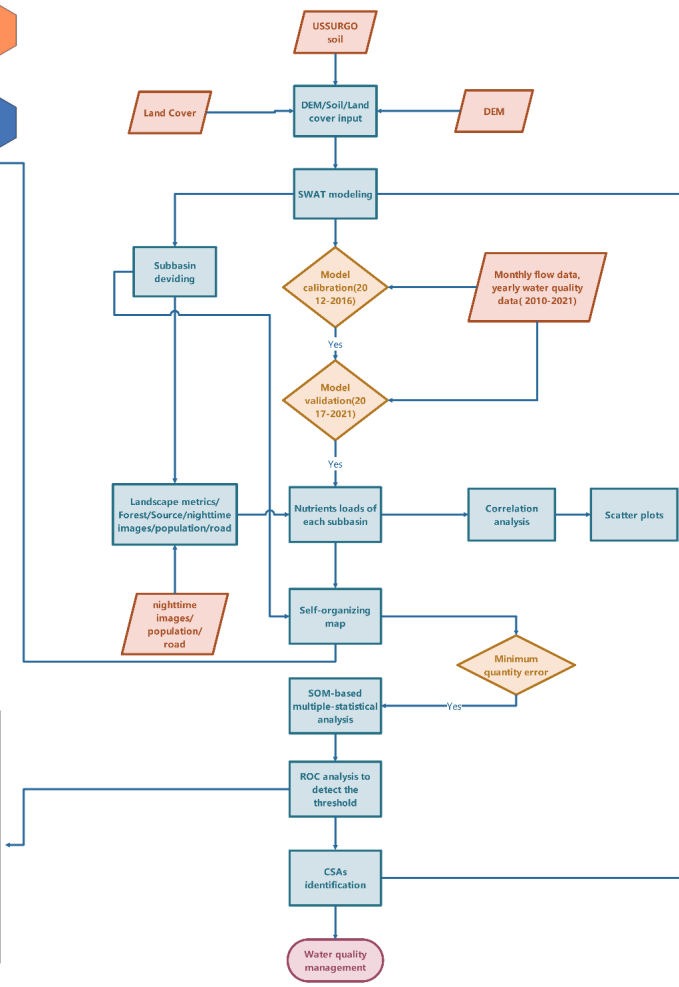
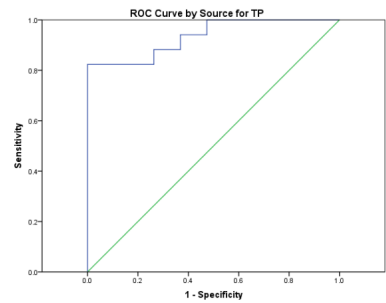
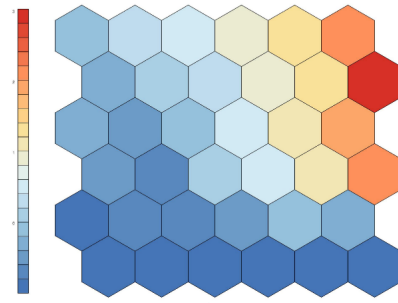
Soil and Water Assessment Tool (SWAT) is widely used to estimate nutrient origin and delivery from the landscape.

Model includes adjustable parameters for internal algorithms



Working flow

- 1, SWAT modeling
- 2, SWAT-based subbasin analysis
- 3, Self-organizing mapping
- 4, SOM-based subbasin analysis
- 5, ROC analysis
- 6, CSAs identification

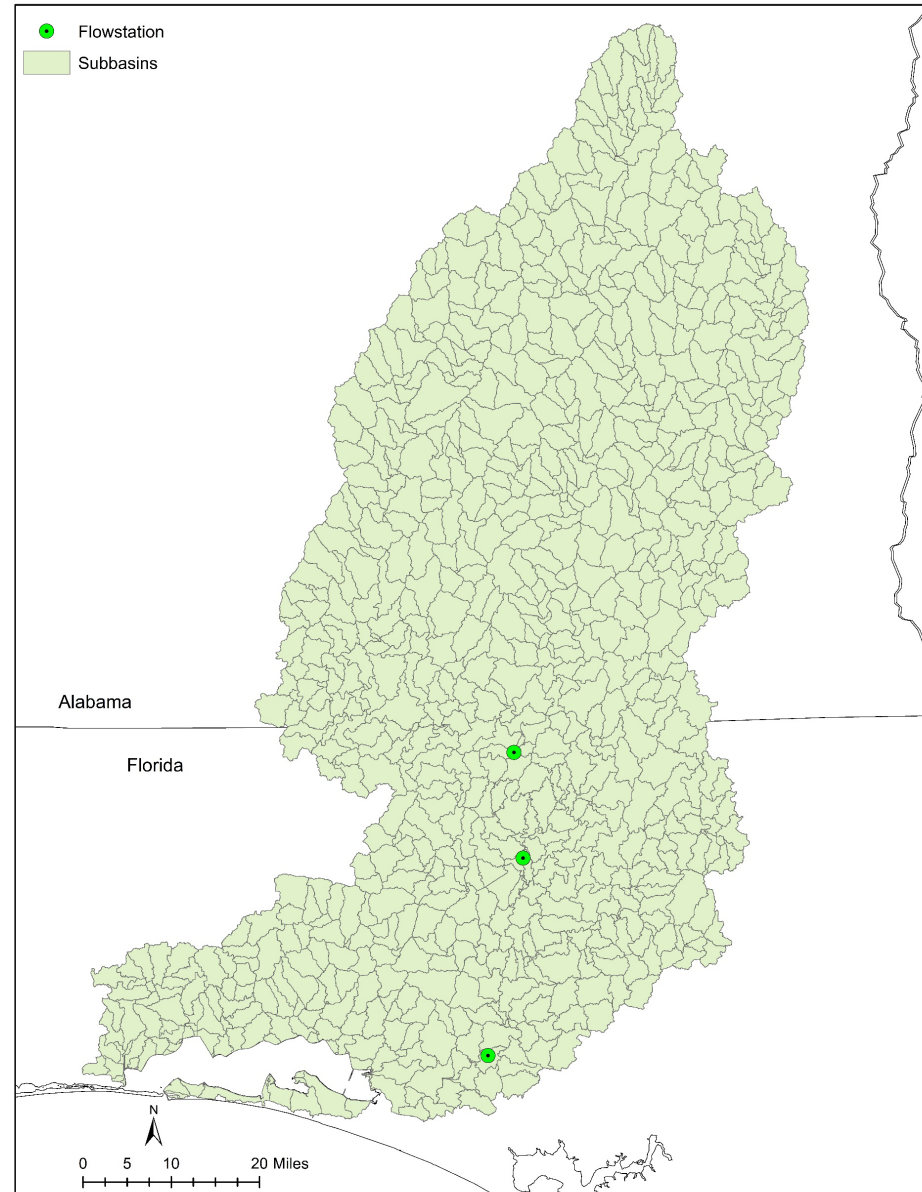


Model development (SWAT)

Model inputs

- DEM of STRM 30 meter
- Land use data NLCD (2016)
- Soil data from SURGO
- Climate data (2010 – 2021)
- Discharge data (2010-2021)
- Water quality data (2010-2021)

863 Sub-basin
3195 HRUs



Calibration and Validation

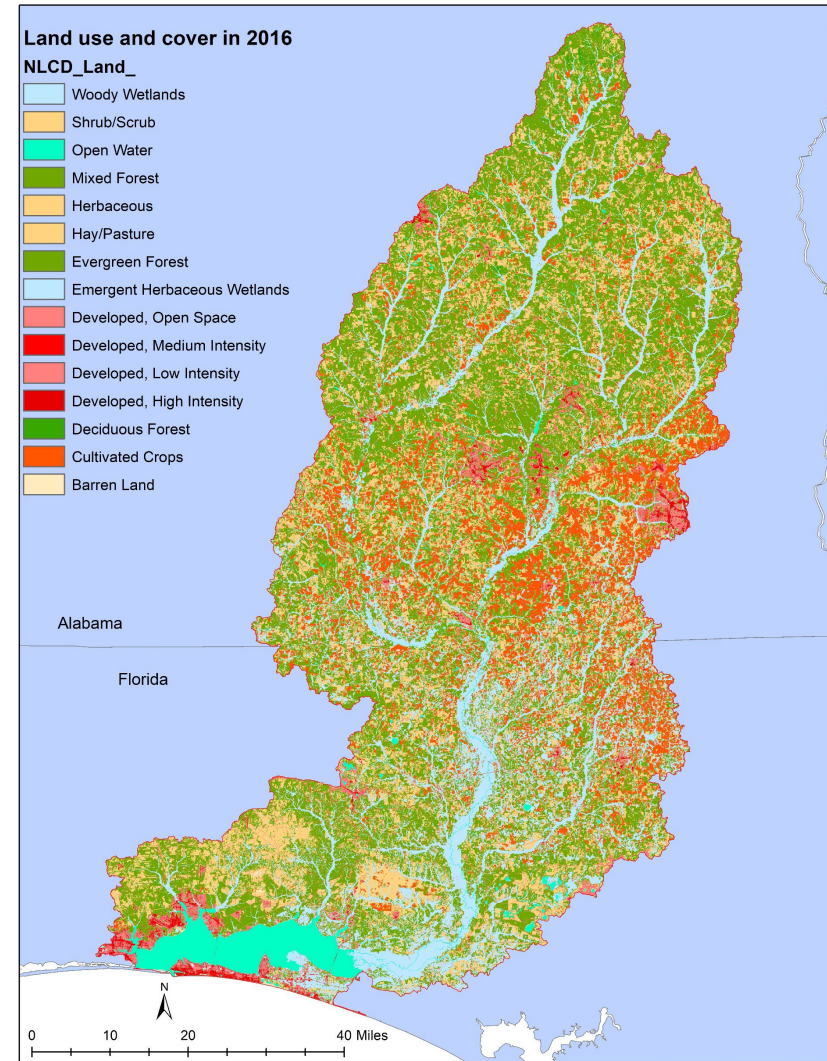
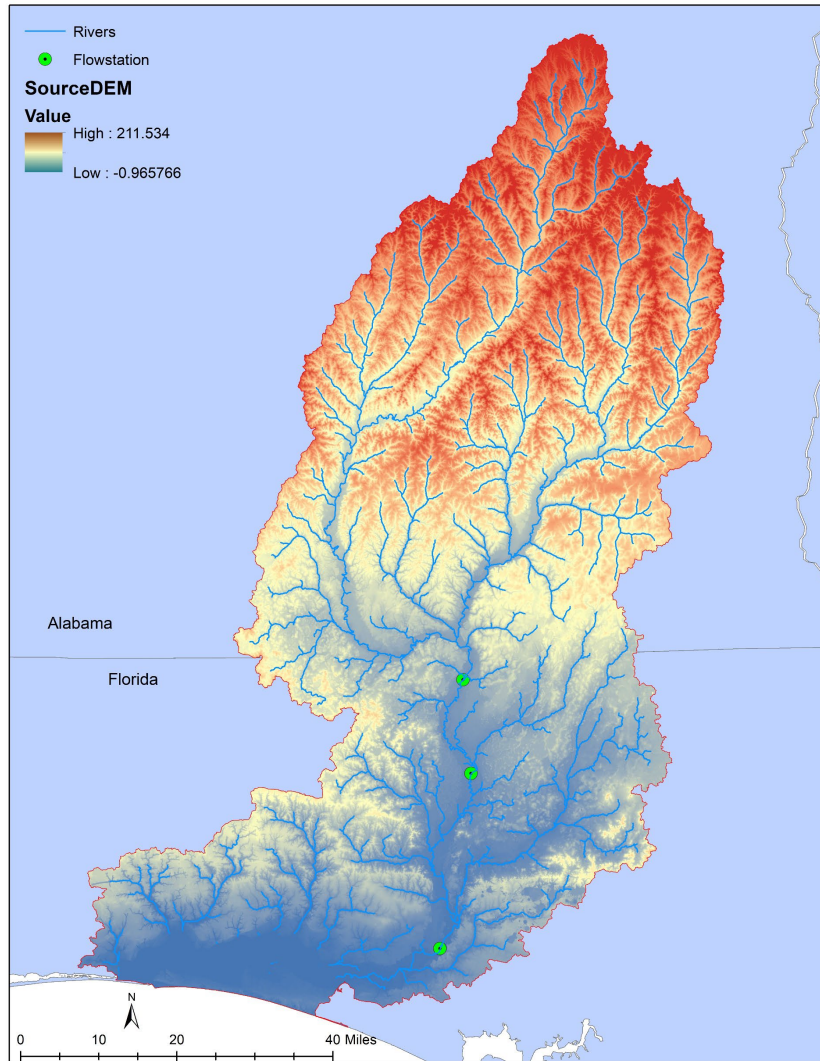
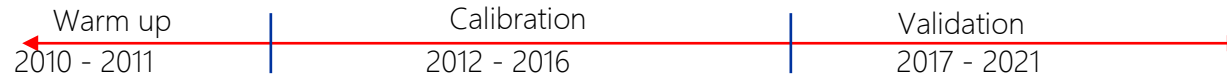
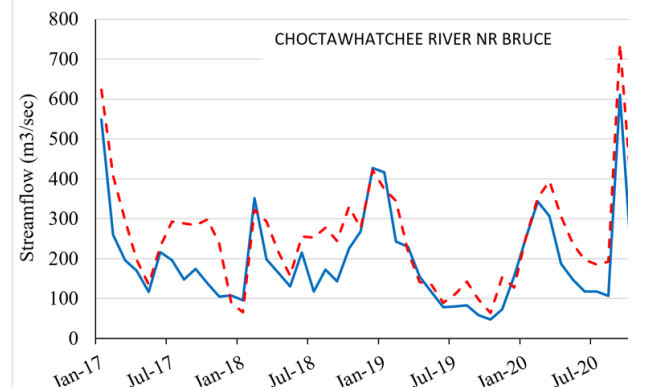
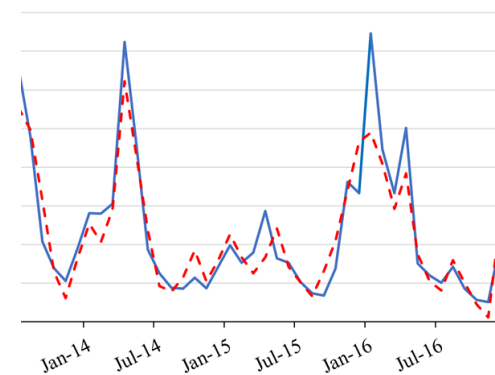
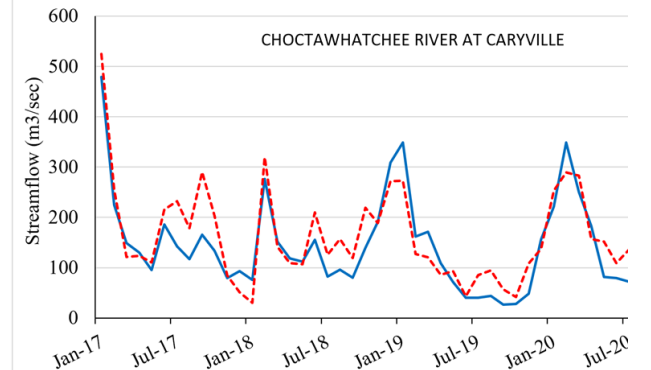
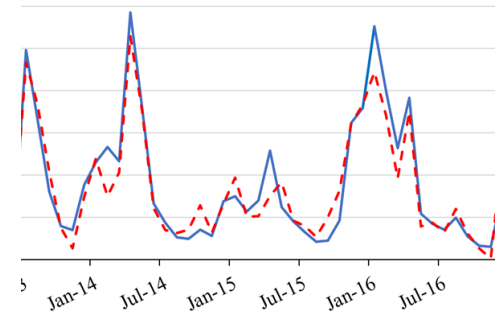
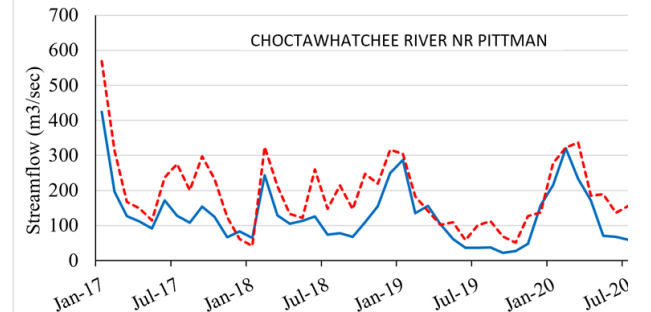
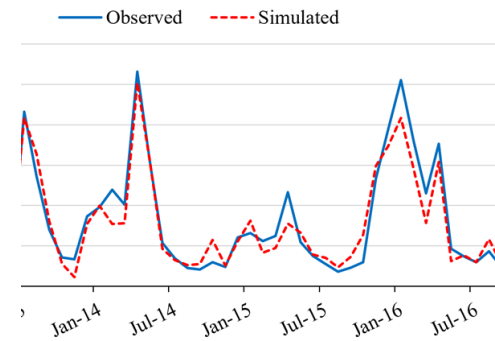


Table 1 Model performance on monthly streamflow and annual water quality variables at multi-locations in the Choctawhatchee watersheds

Variables	Sub-basin	Calibration		Validation	
		NSE	R^2	NSE	R^2
Streamflow	CHOCTAWHATCHEE RIVER NR PITTMAN (02365200)	0.81	0.84	0.47	0.82
	CHOCTAWHATCHEE RIVER AT CARYVILLE (02365500)	0.81	0.86	0.84	0.86
	CHOCTAWHATCHEE RIVER NR BRUCE (02366500)	0.84	0.84	0.66	0.81
Total Nitrogen	CHOCTAWHATCHEE RIVER NR PITTMAN (02366500)	0.57	0.62	0.69	0.73
Total	CHOCTAWHATCHEE RIVER NR BRUCE (02366500)	0.4	0.47	0.57	0.61
Phosphorus					

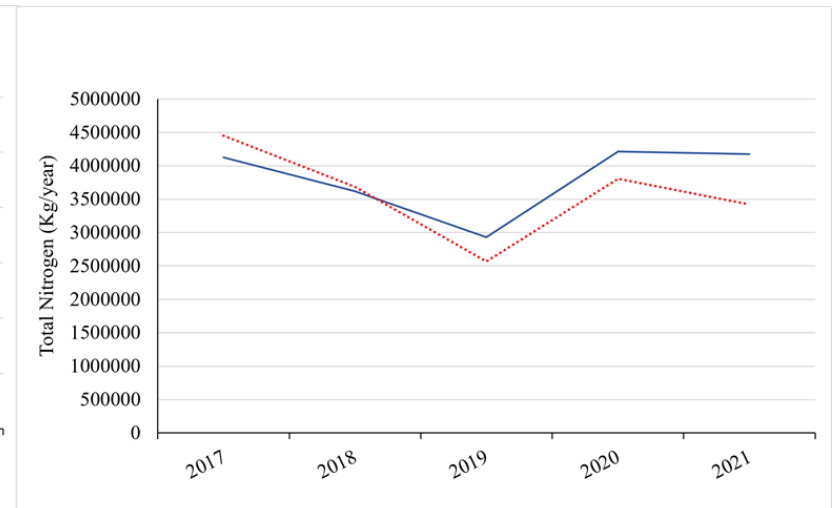
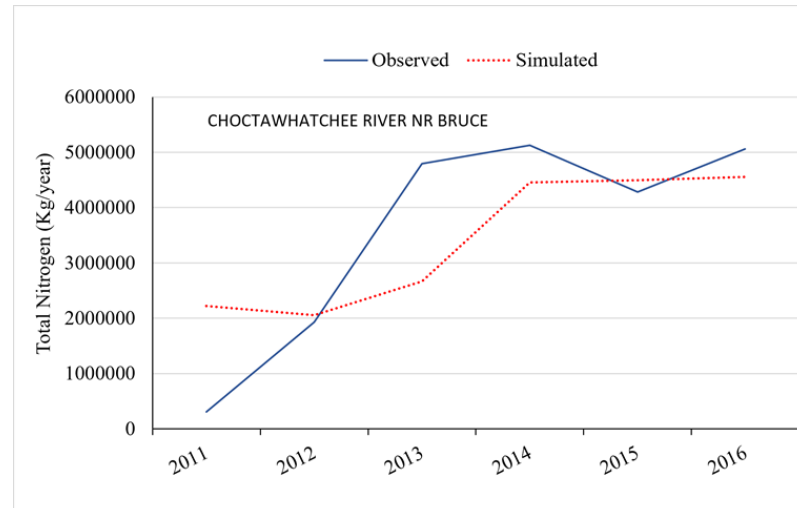
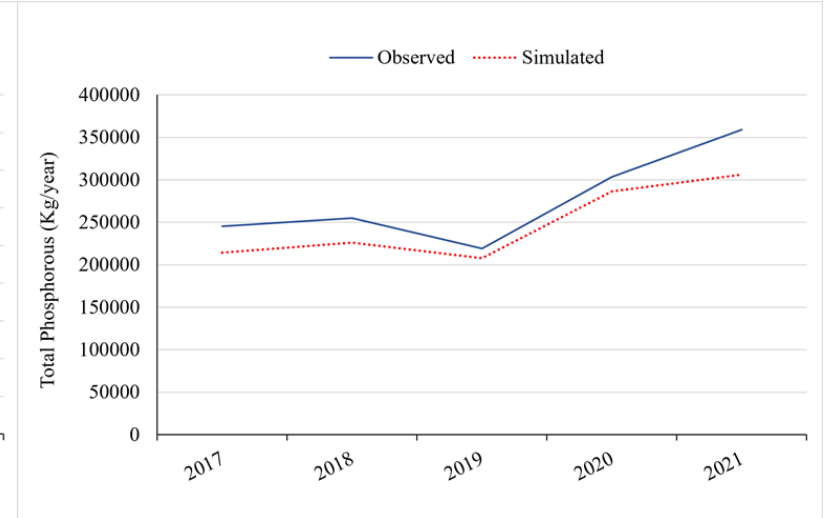
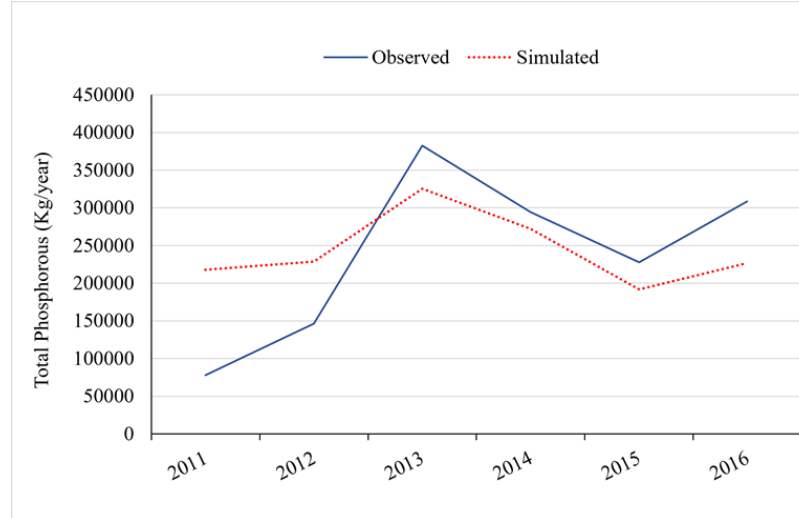
SWAT modeling

Flow Calibration and validation



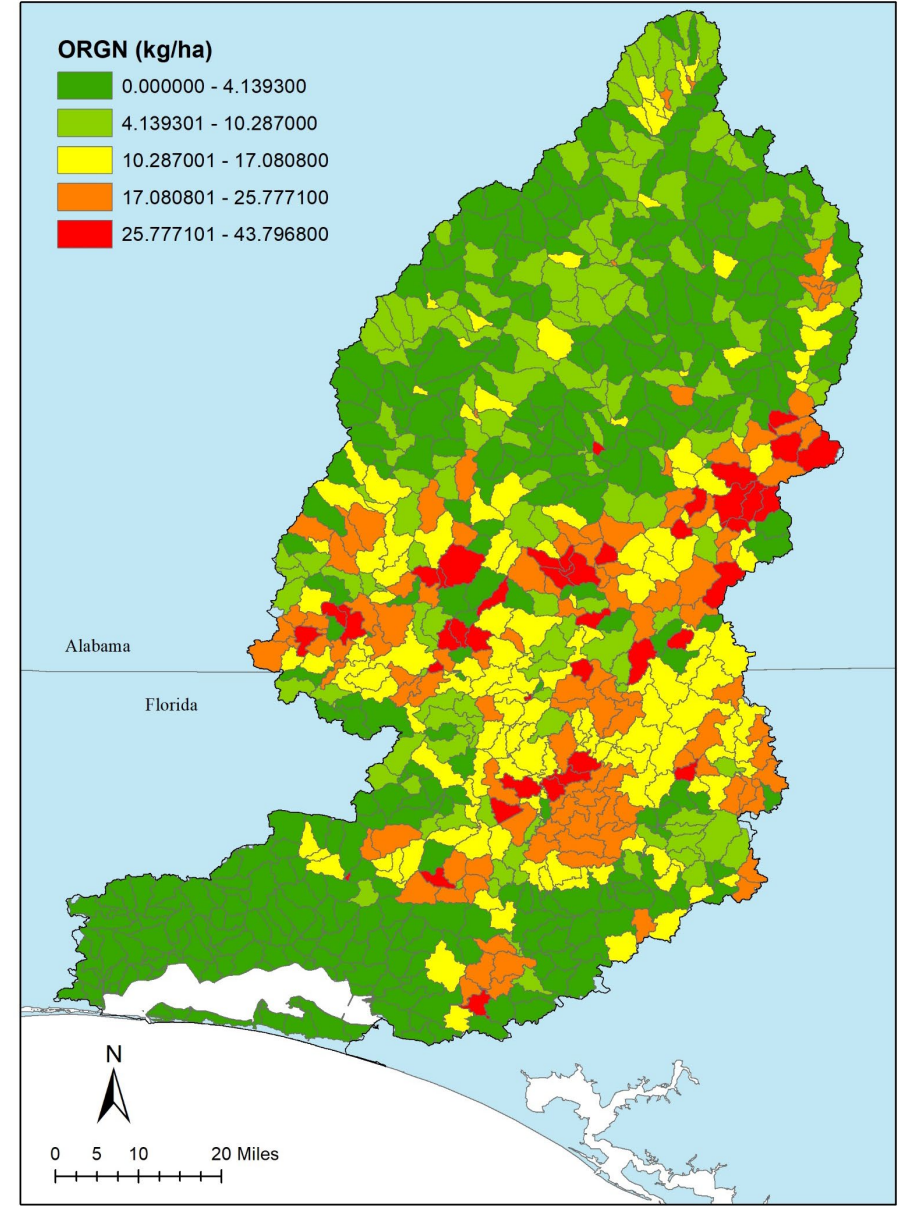
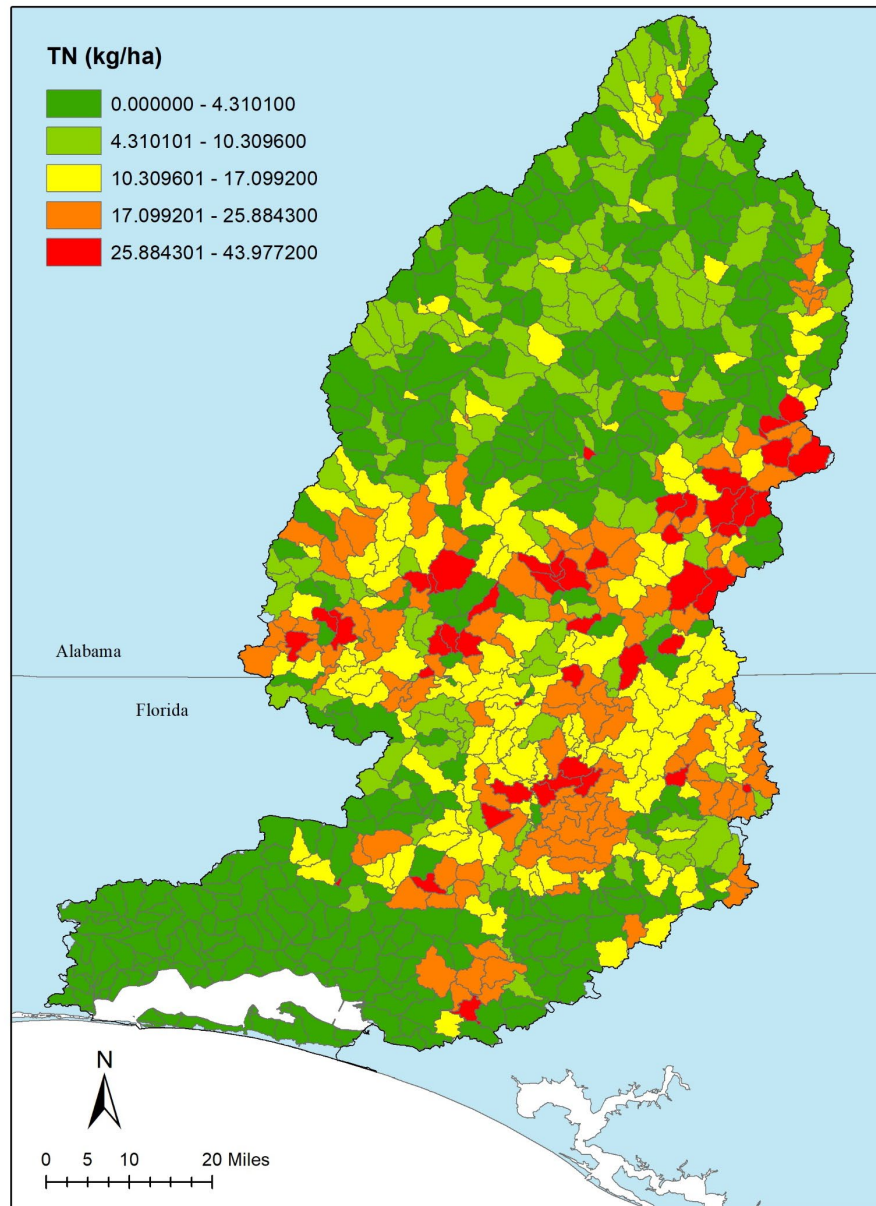
SWAT modeling

Water quality
Calibration and
validation



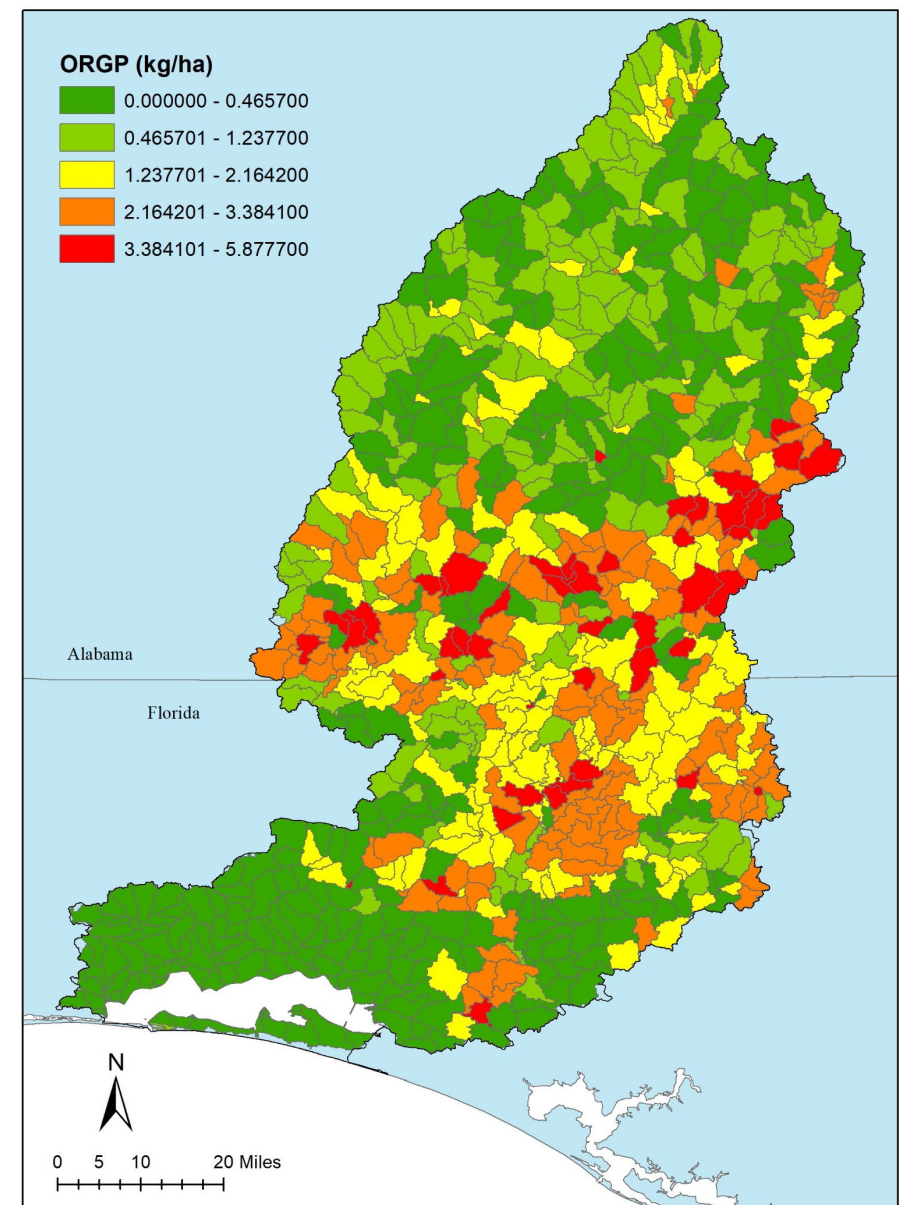
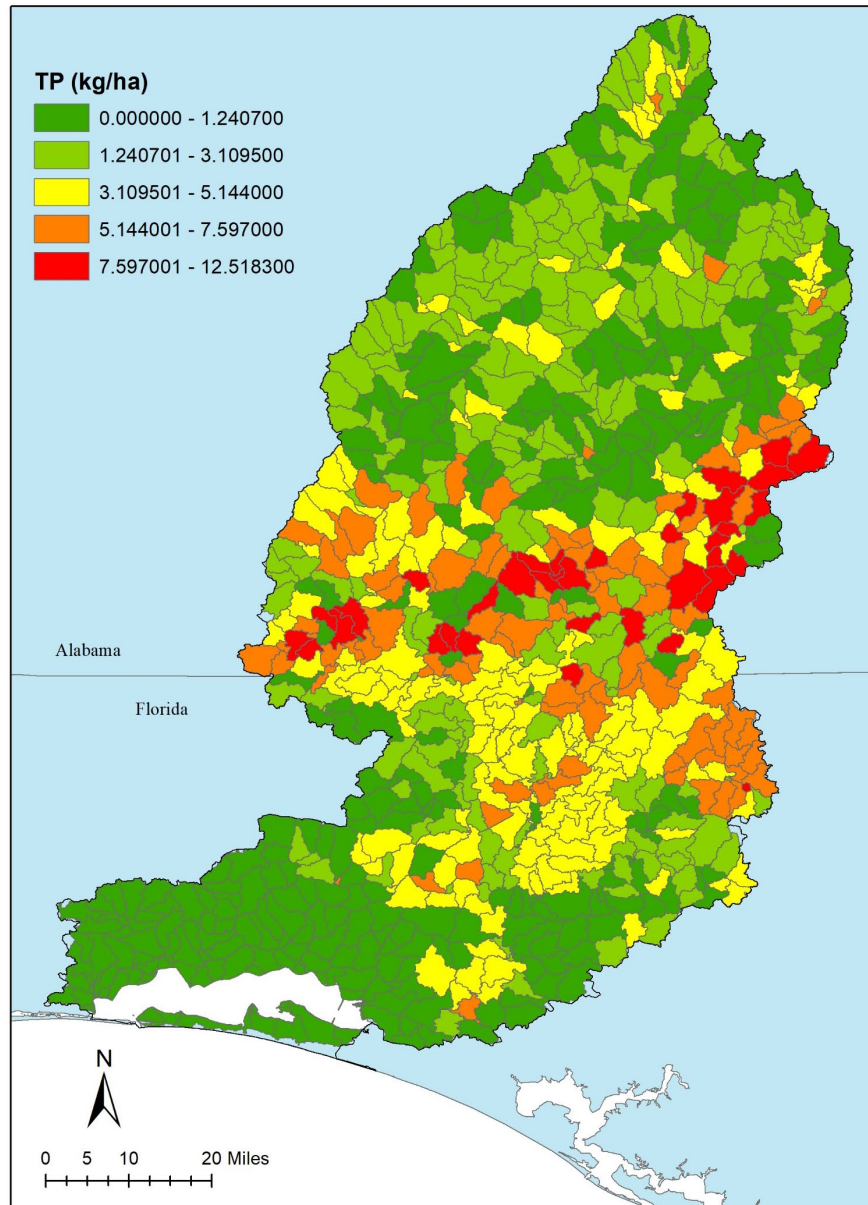
Nitrogen

TN and Organic nitrogen of subbasin(average of 2012-2021)




Phosphorous

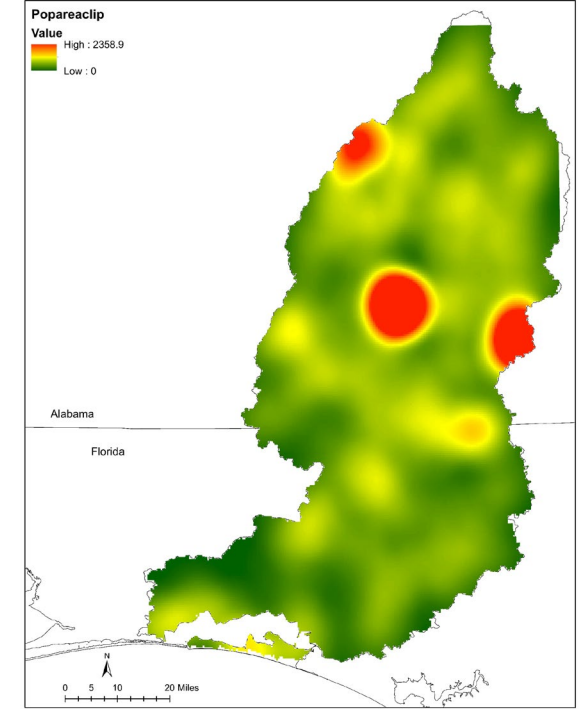
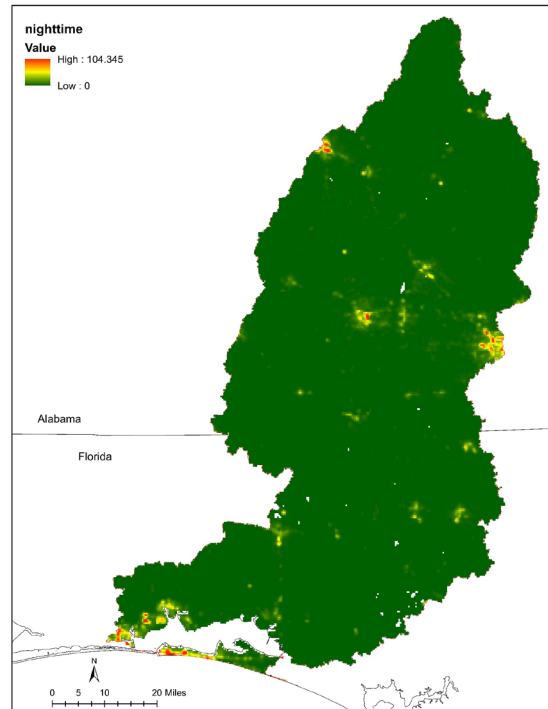
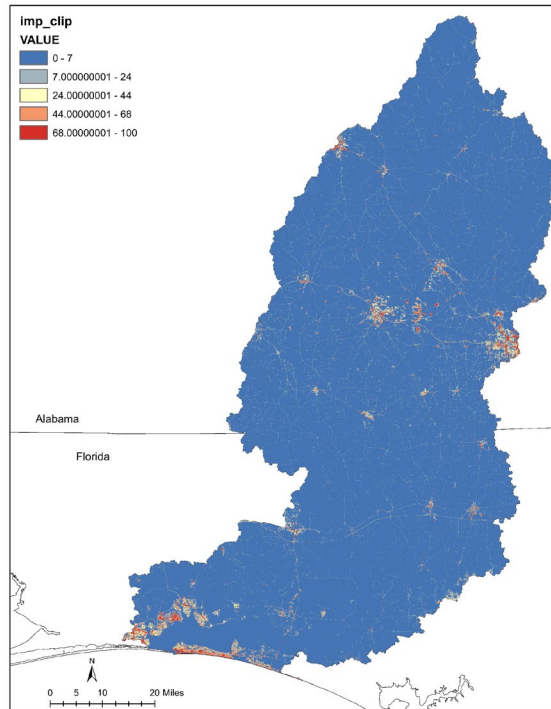
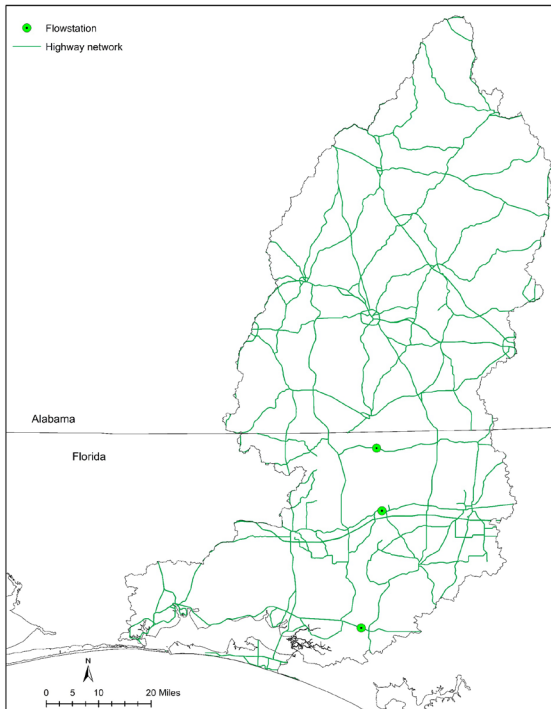
TP and Organic phosphorous of subbasin(AV of 2012-2021)



What factors influencing water quality

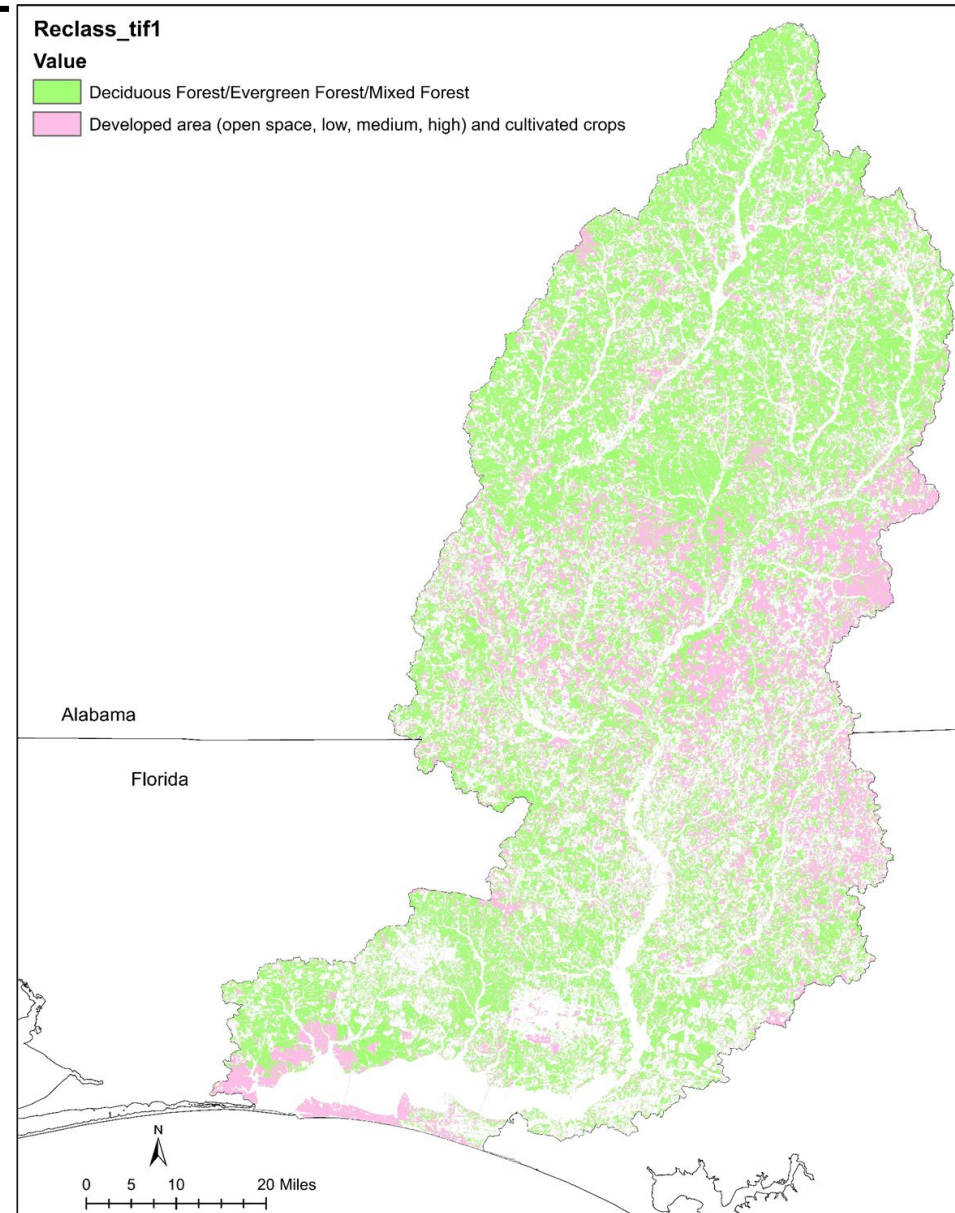


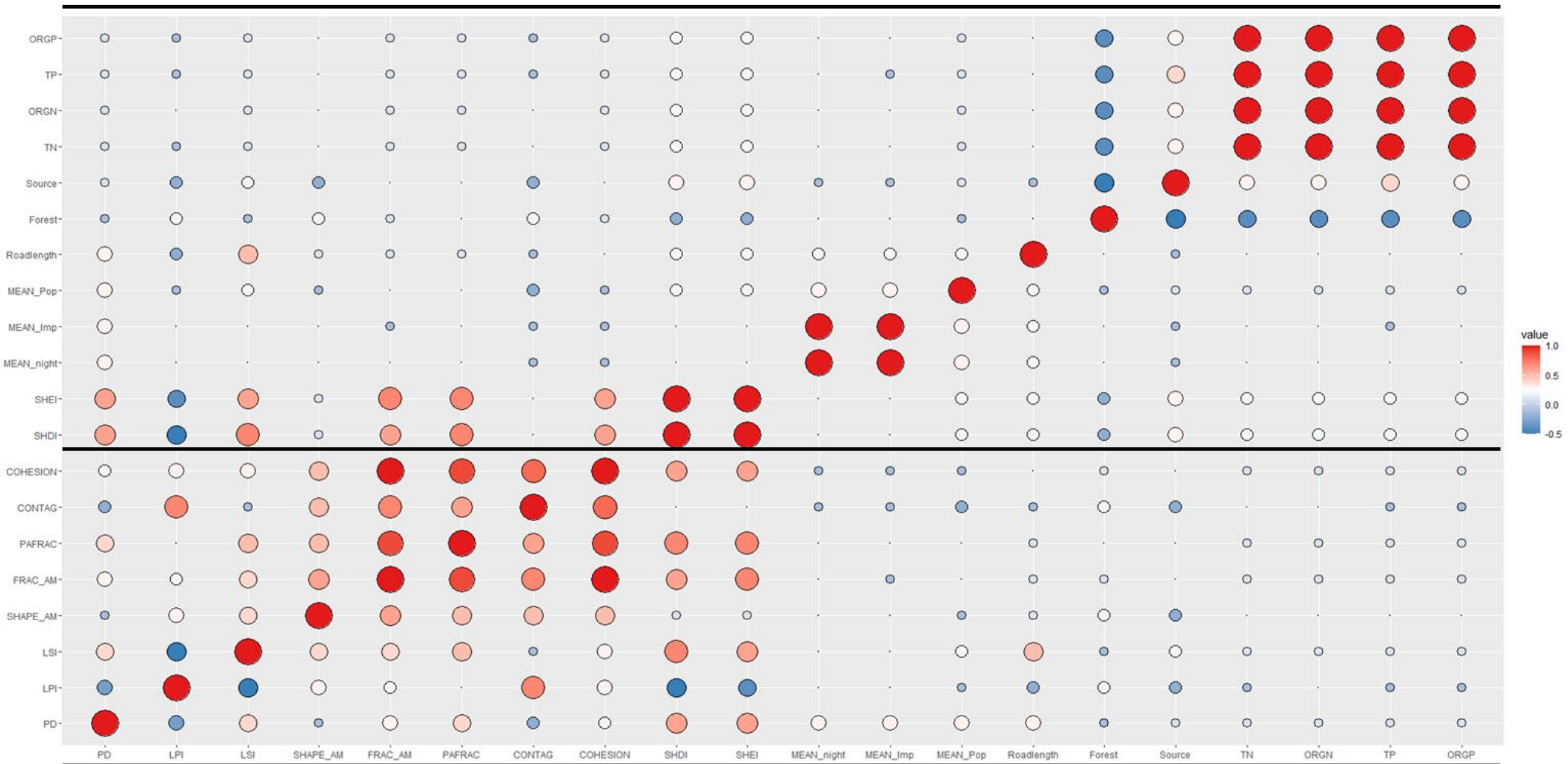
Road, impervious surface, nighttime RS, and populated area

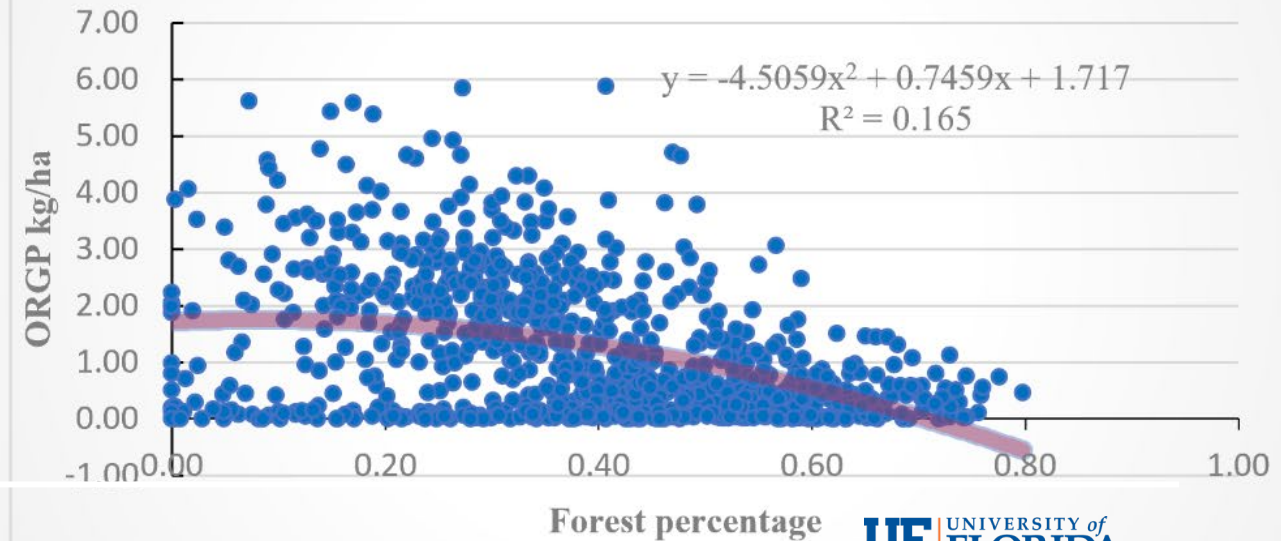
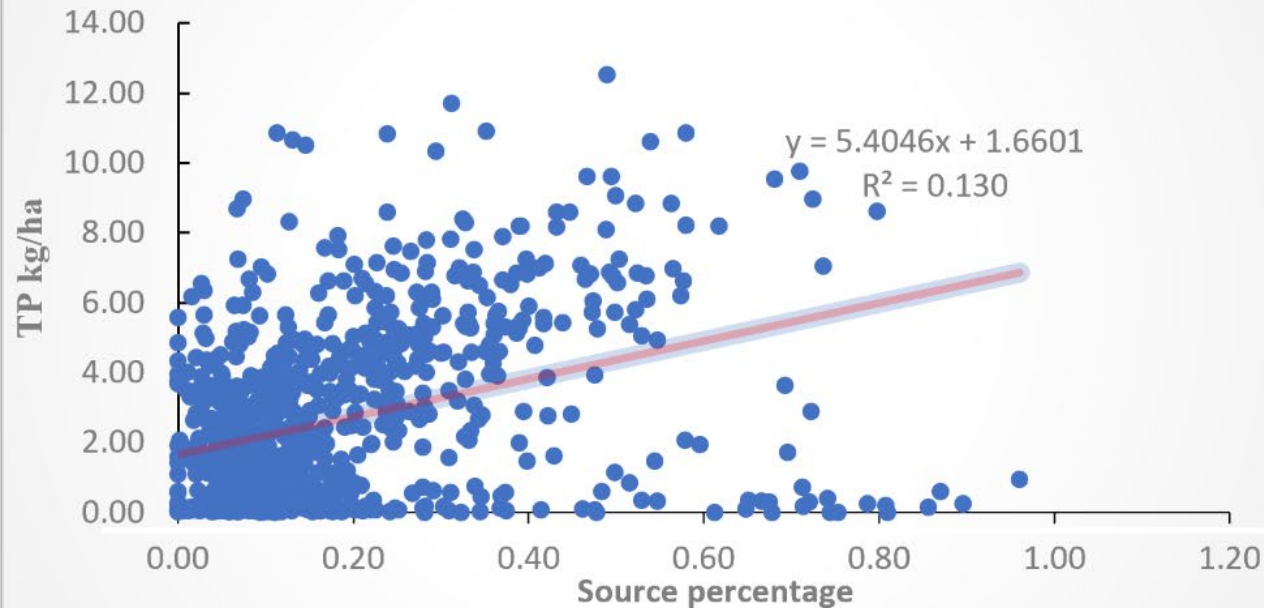
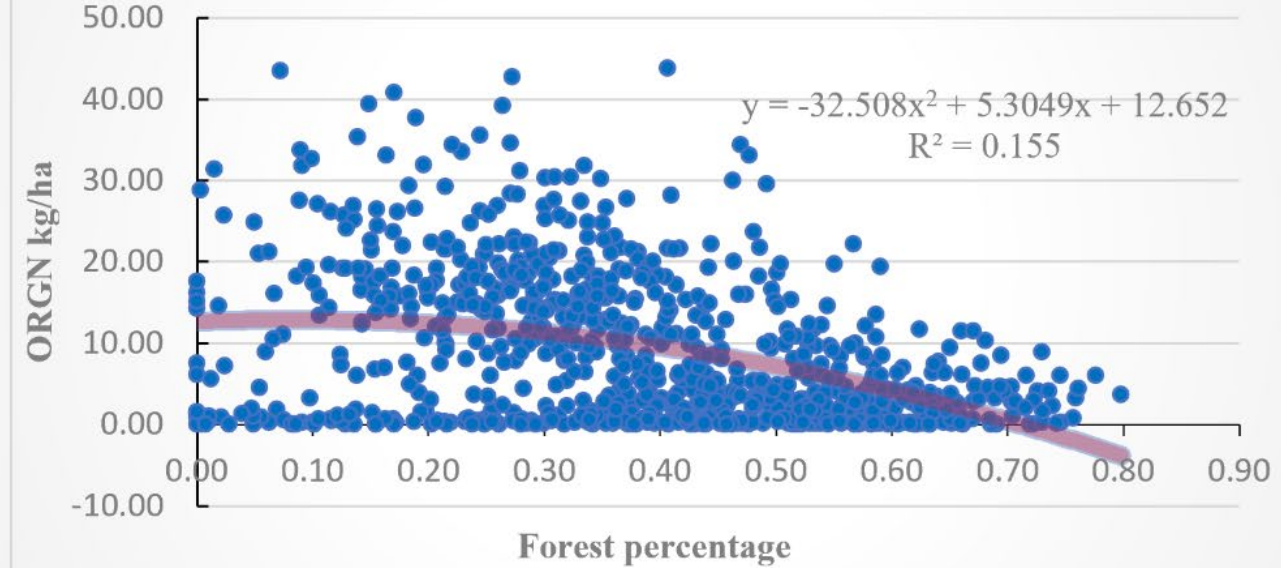
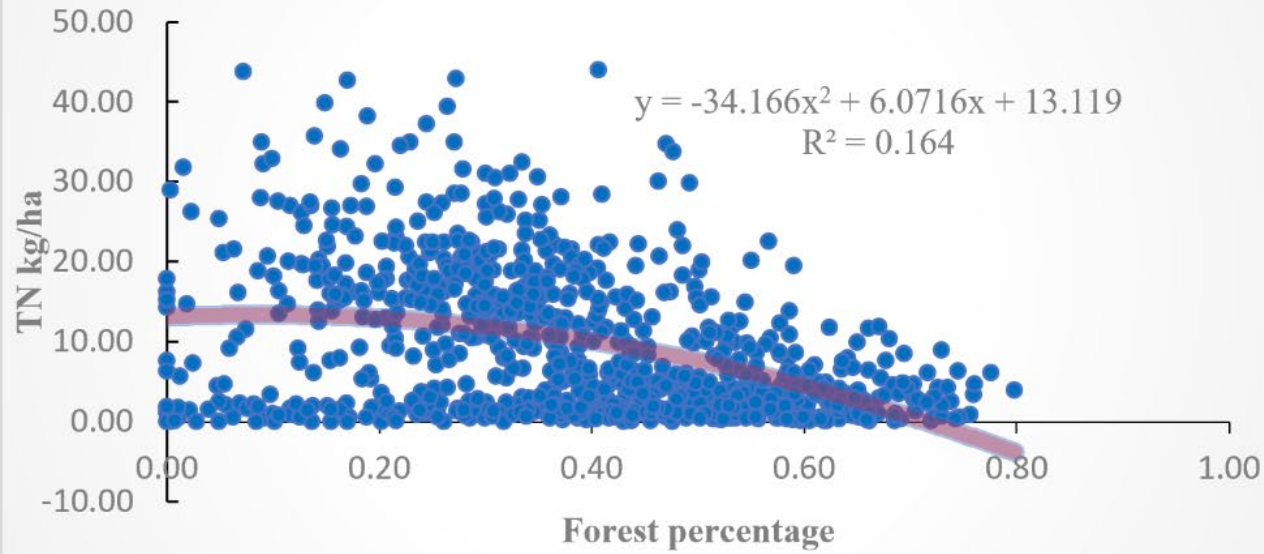


LULC reclassification

- . Forest
- . Developed areas and cultivated crops (Source)





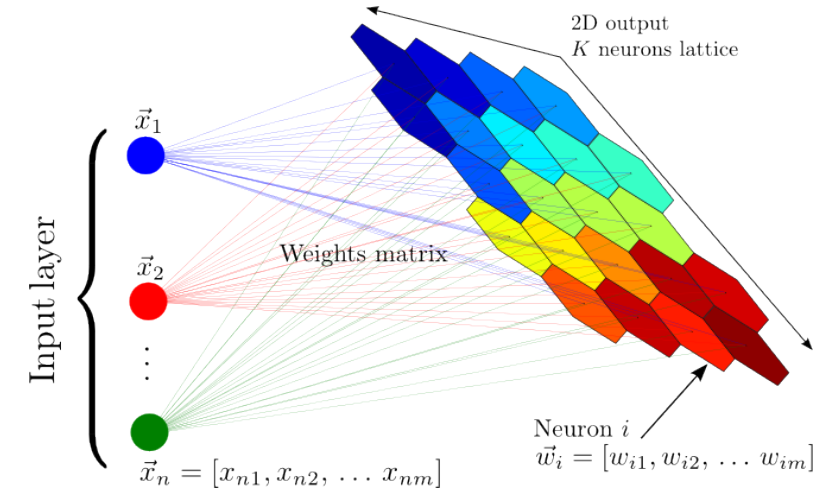
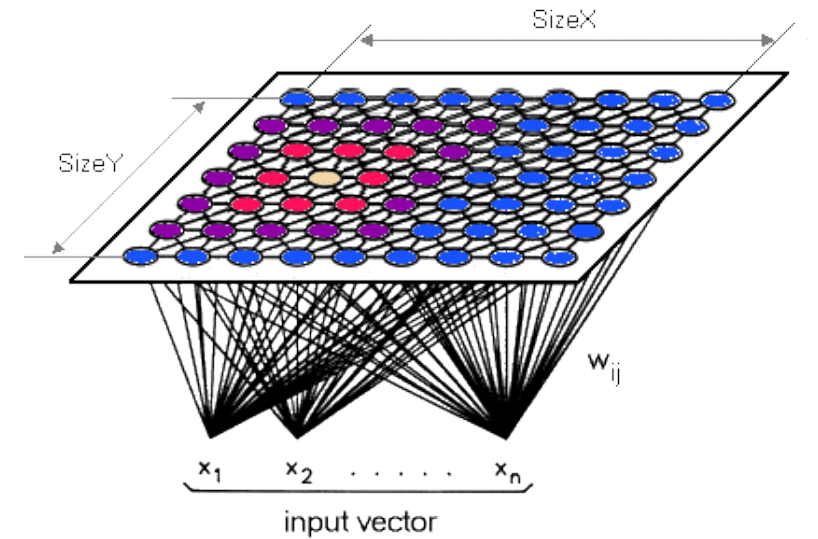


Self-organizing maps



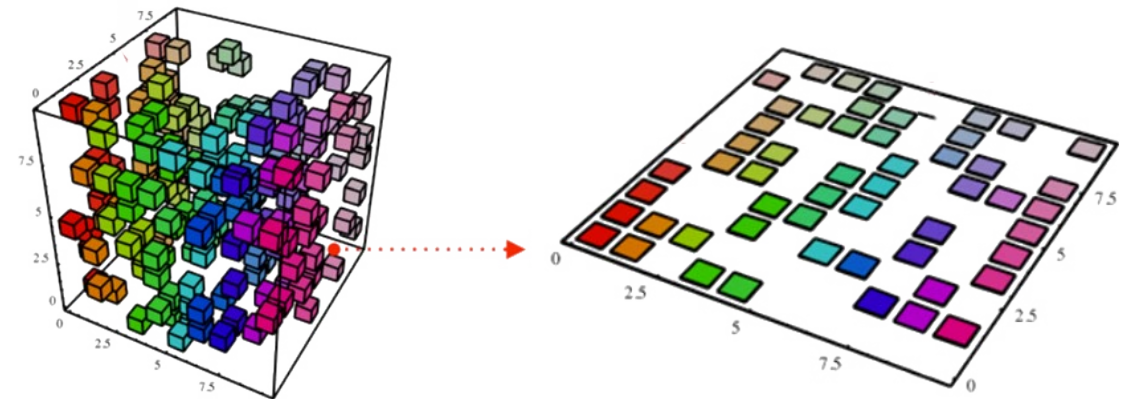
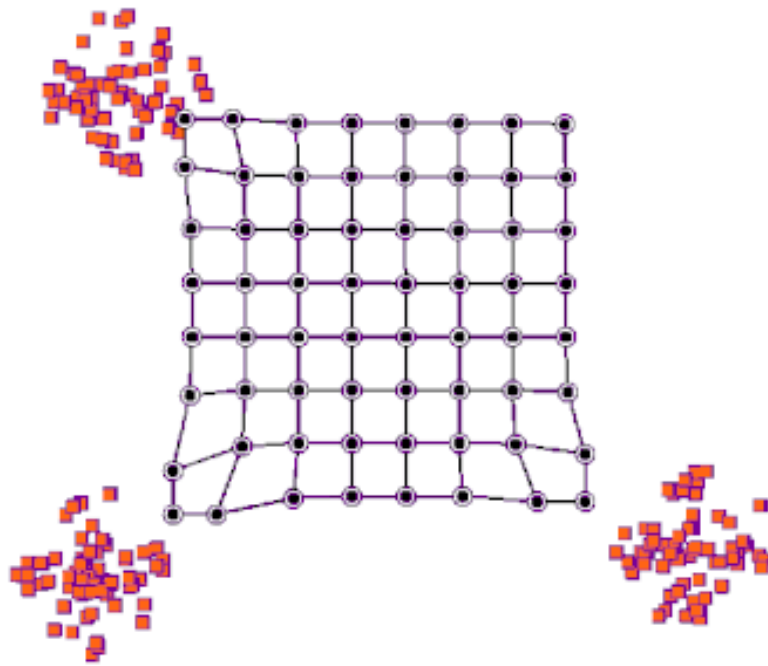
Self Organizing Maps

- SOMs are neural networks that employ unsupervised machine learning methods, mapping their **weights** to conform to the given input data with a goal of representing multidimensional data in an **easier** and **understandable** form for the human eye.
- Neurons that lie close to each other represent clusters with similar properties



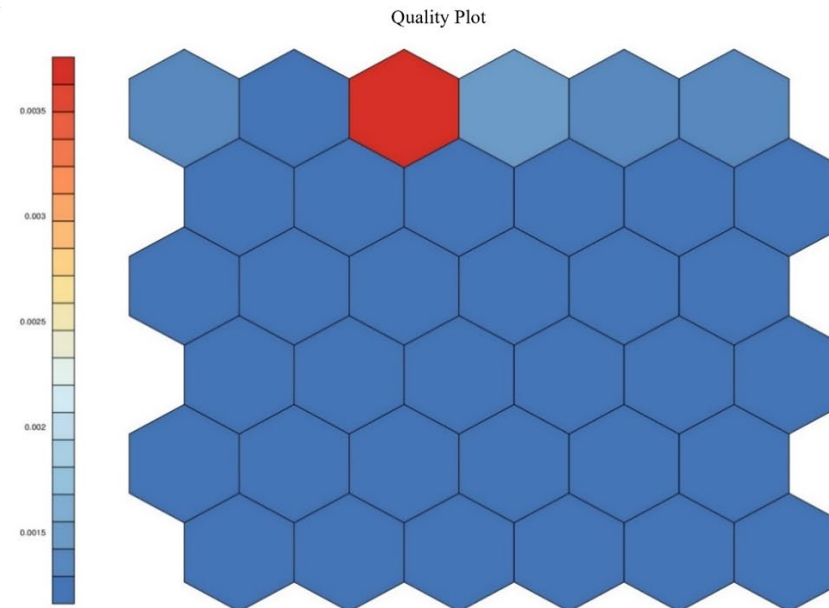
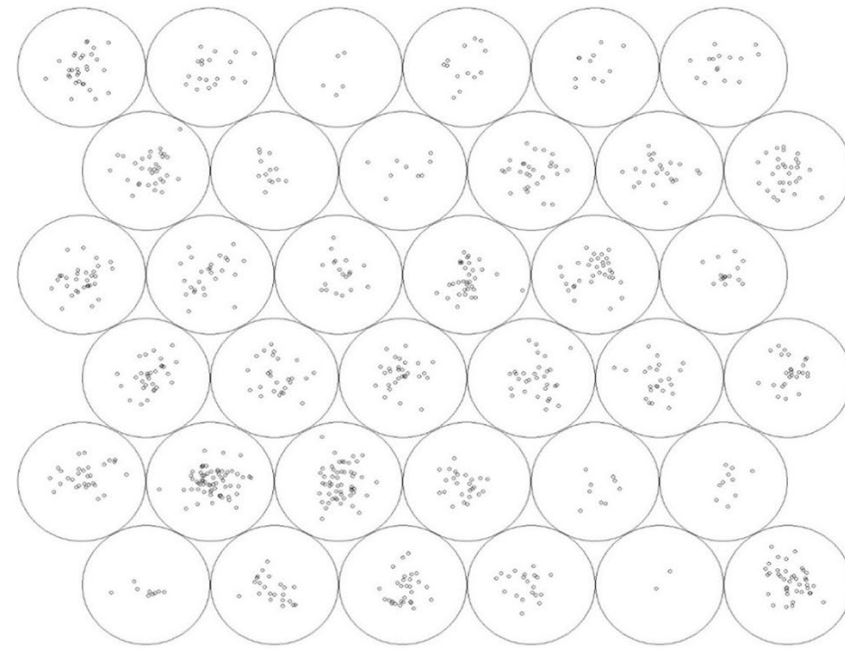
SOM training

Training process of SOM on a two-dimensional data set



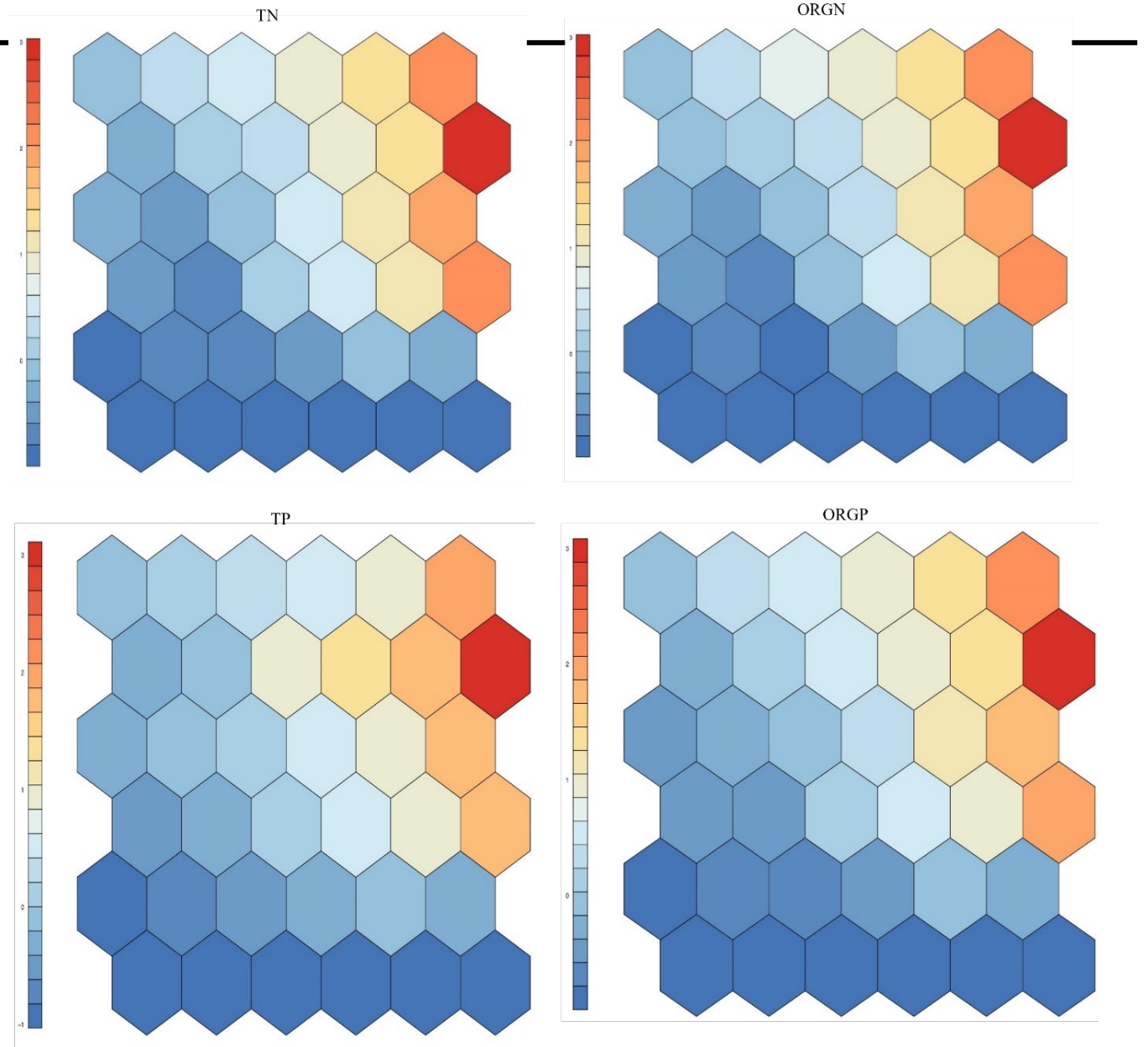
SOM optimizing

Self-organizing map (SOM) of the total 863 subbasins. The quality plot of the trained SOM showed that the SOM with 6*6 neurons was reliable for representing the water quality data (**the lowest value of quantitative error, QE**)

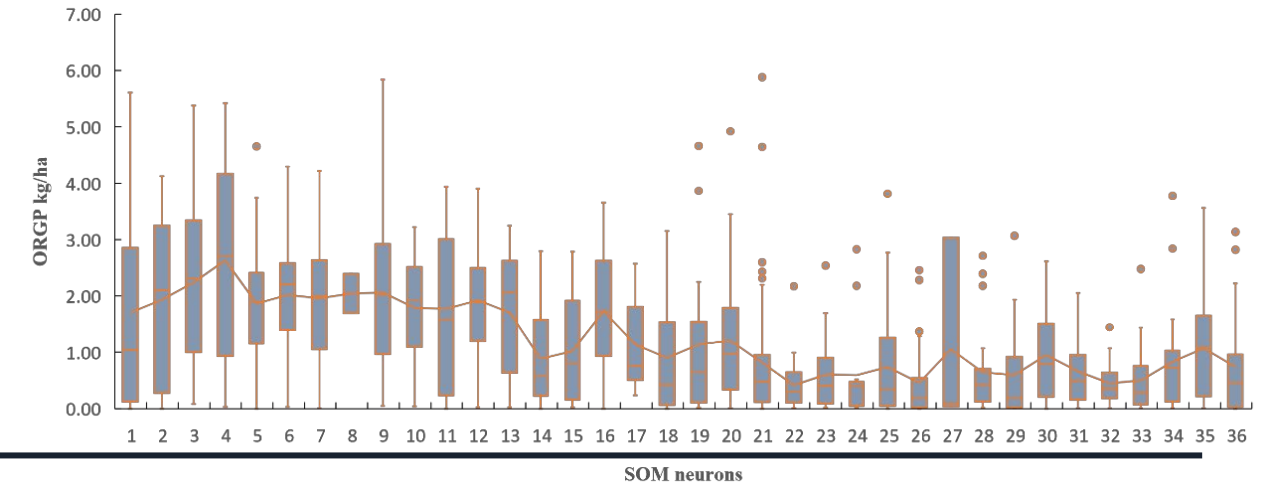
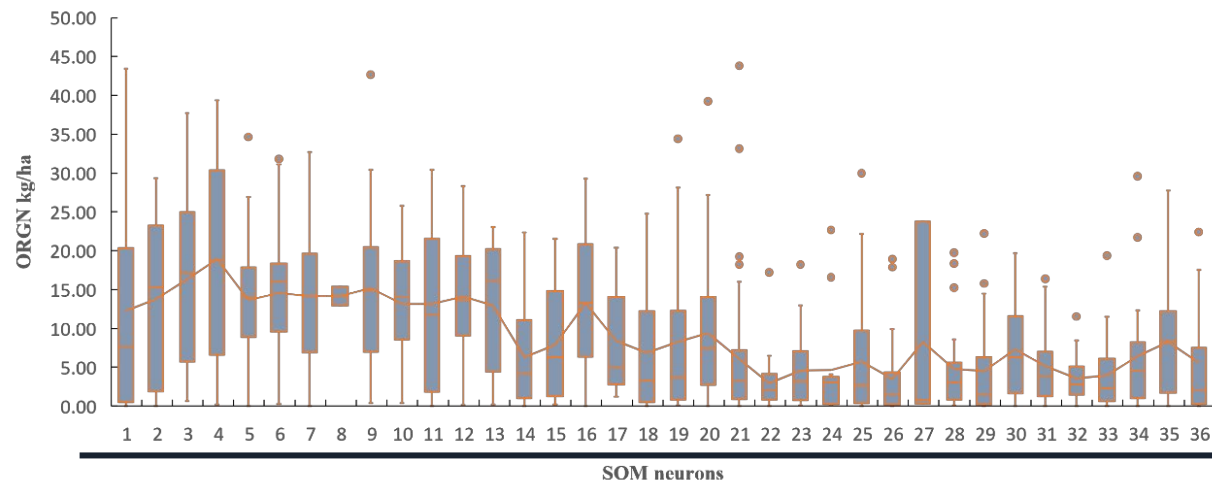
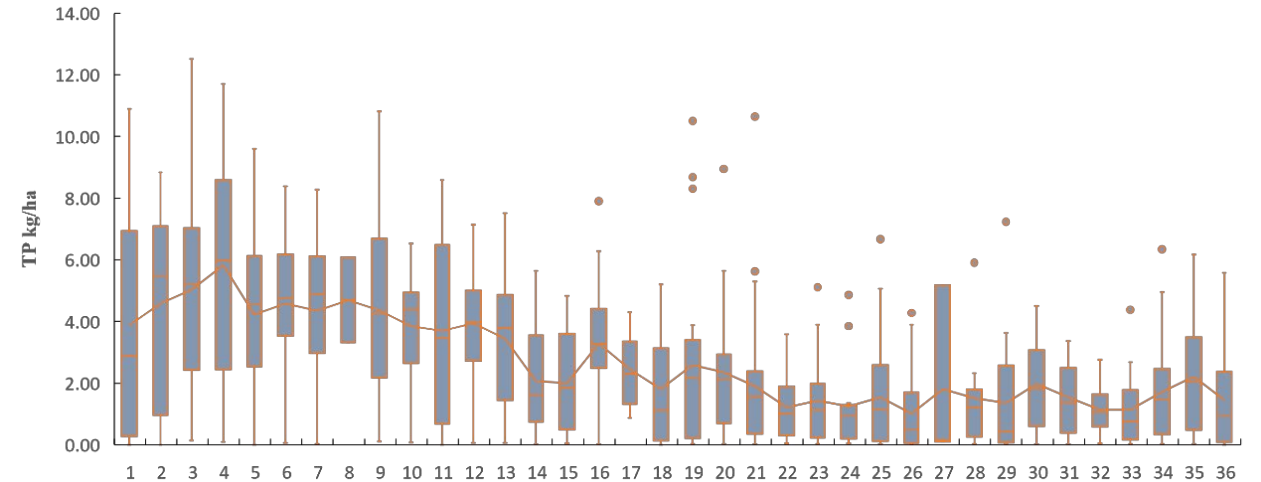
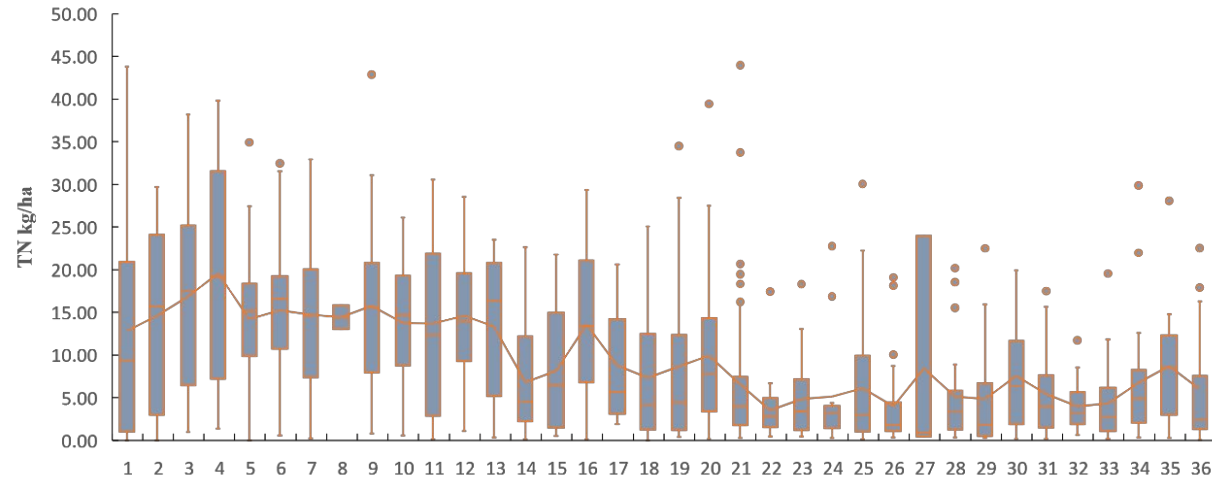


Water quality

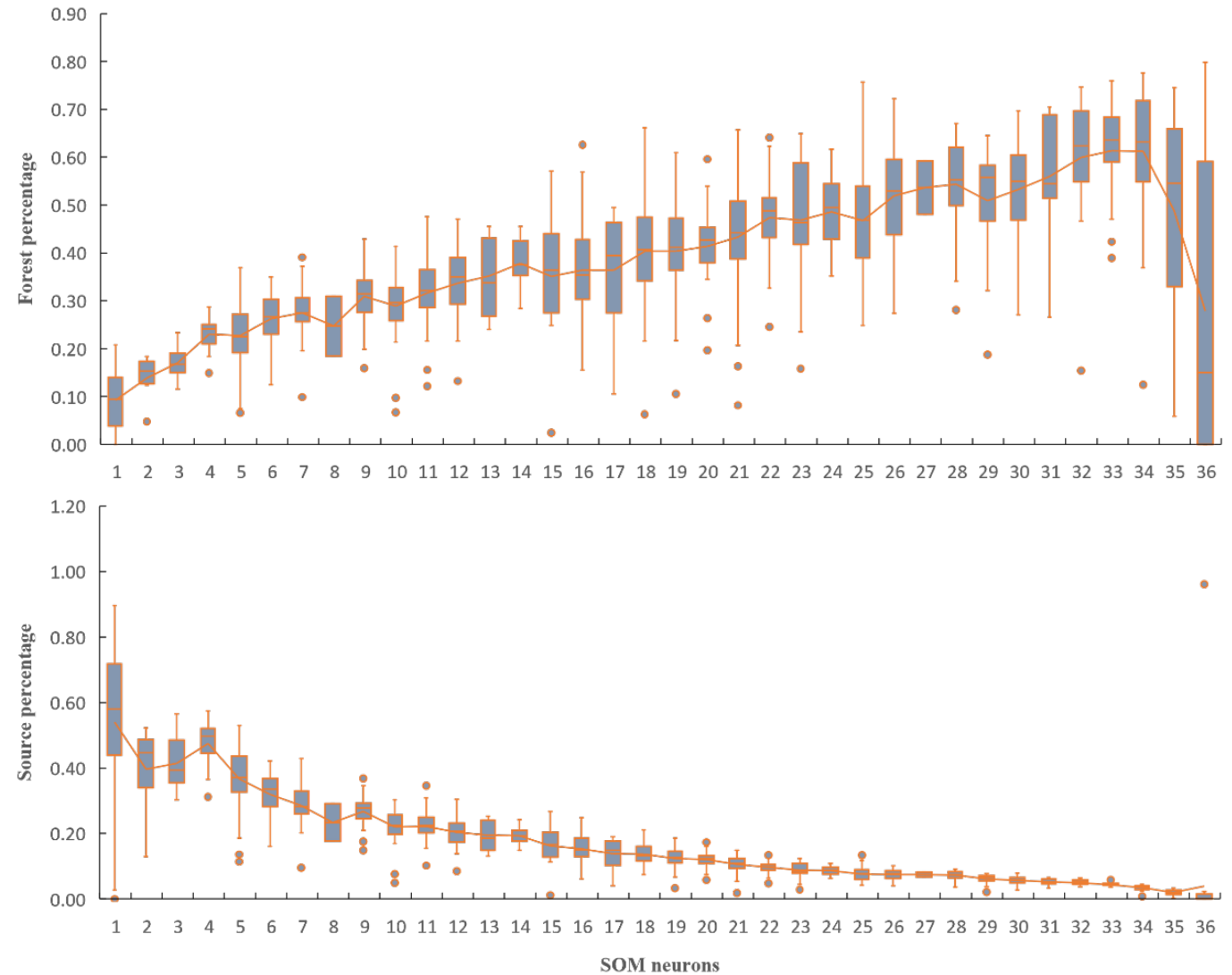
The SOM of the water quality datasets revealed that TN, ORGN, TP, and ORGP exhibited similar patterns.



Box plots for nitrogen and phosphorus of each SOM neurons.



Box plots for forest land cover and Source land use percentage of each SOM neurons.



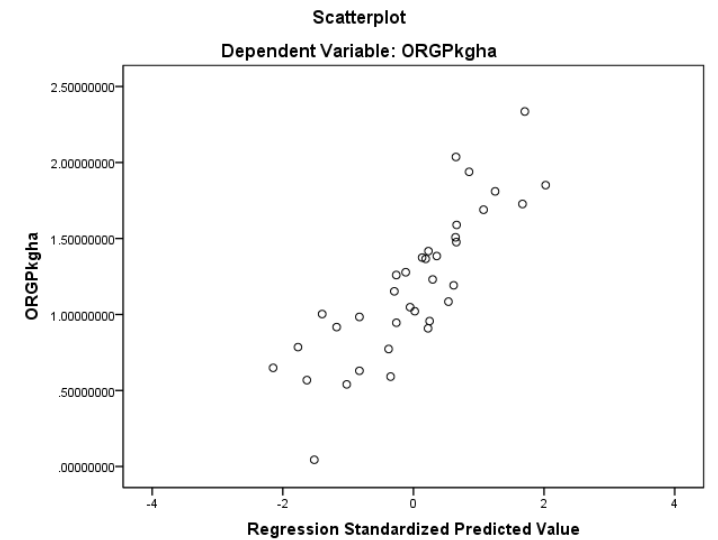
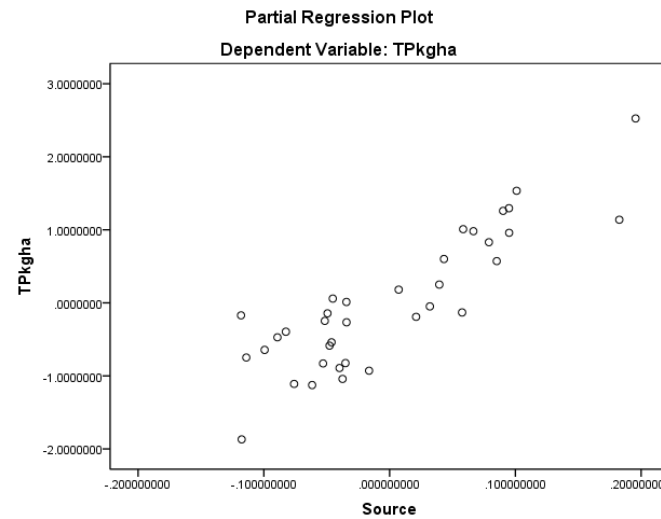
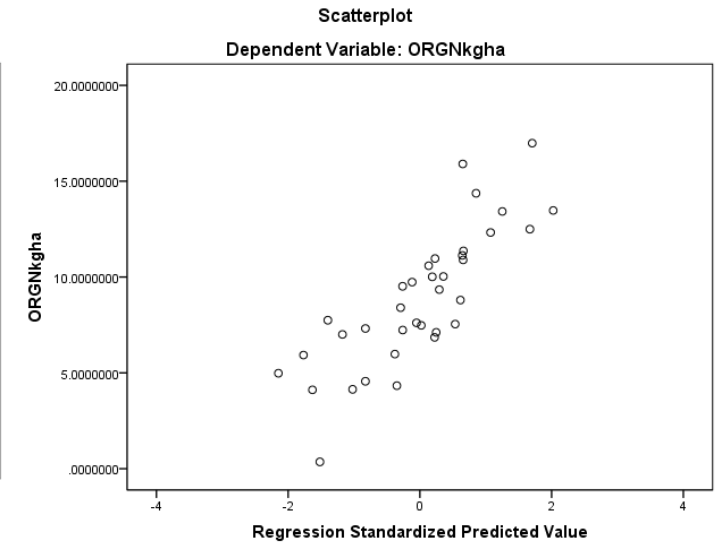
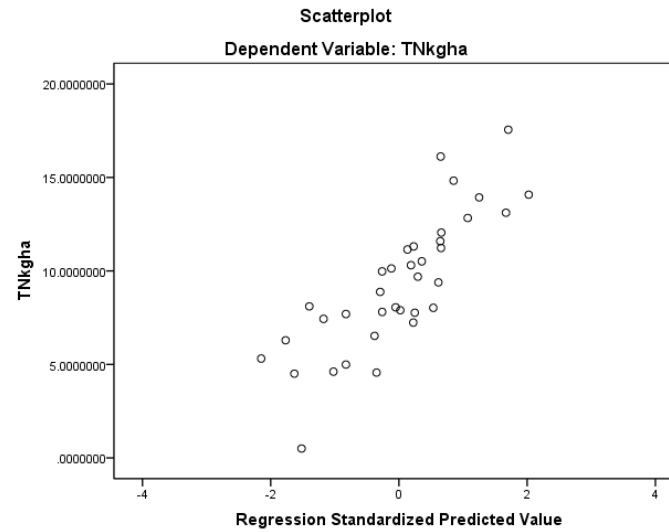
Dependent	Independent	Ad_R2	F	Equation
TN	Forest	0.672	72.786***	$TN = -37.557 * Forest + 23.635$
ORGN	Forest	0.661	69.222***	$ORGN = -36.796 * Forest + 22.902$
TP	Source	0.752	107.132***	$TP = 10.465 * Source + 0.741$
	Source, MEAN_pop	0.786	65.189***	$TP = 9.864 * Source + 0.001 * MEAN_pop + 0.400$
ORGP	Forest	0.695	80.081***	$ORGP = -5.113 * Forest + 3.144$

SOM-based regression analysis

Table 1 The self-organizing maps (SOM) based linear regression analysis between water quality and the independent variables

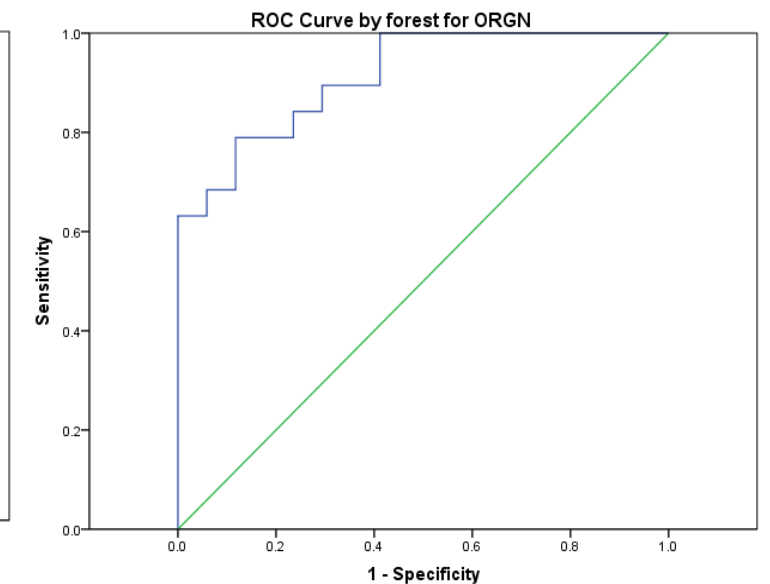
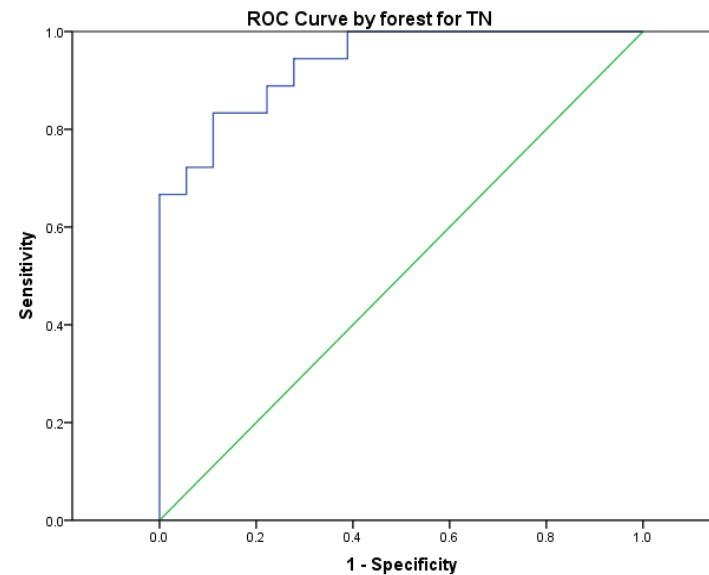
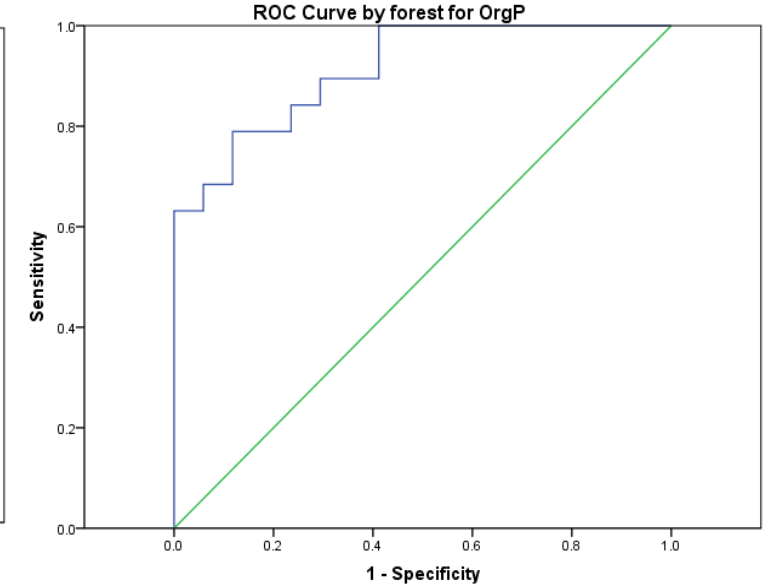
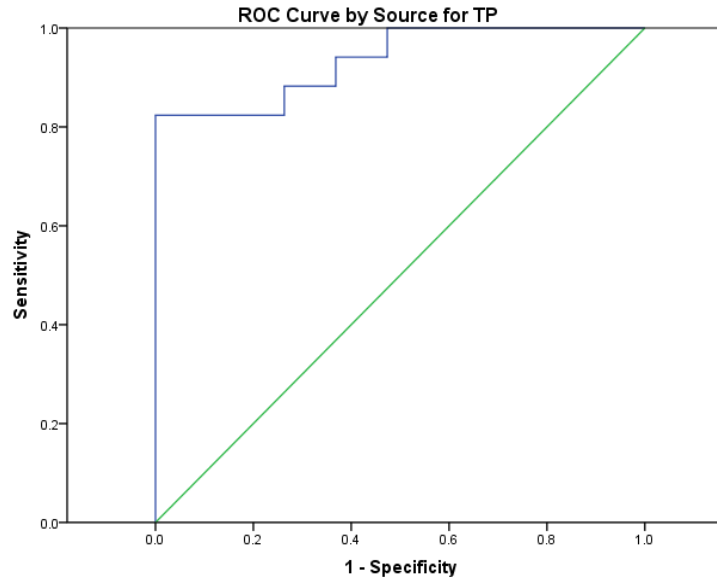
Note: * $p < 0.05$, *** $p < 0.001$; TN, total nitrogen; ORGN, organic nitrogen; TP, total phosphorus; ORGP, organic phosphorus; Forest, percentage of forest in a subbasin; Source, percentage of developed areas (including high, medium and low density) and cultivated crops in a subbasin; MEAN_pop, popularized areas in a subbasin; PAFRAC, perimeter-area fractal dimension

SOM-based scatterplots for water quality indicators

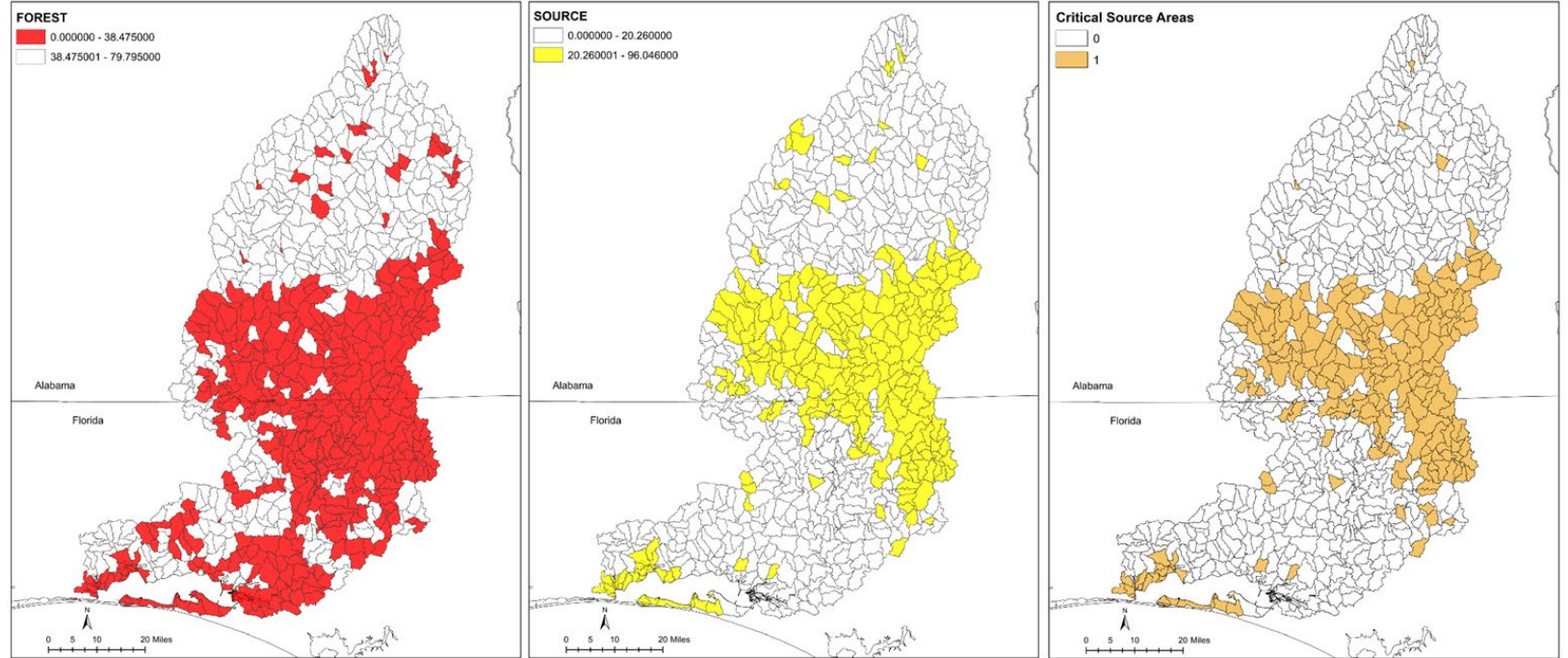


Thresholds detection by ROC analysis

- . Forest 37.47%
- . Source 20.26%



CSAs and performance



Potential critical sources areas (CSAs) identification as those with forest percentage $< 37.47\%$ (red), those with Source percentage $> 20.26\%$ (yellow), and those meeting both criteria (orange)

Water Research (2024), doi:

<https://doi.org/10.1016/j.watres.2024.121286>

- ✓ **Red** 46% of the total area, but 67% of total TN loads for the whole basin
- ✓ **Yellow** 33% of the total area, but 54% of the total loads
- ✓ **Orange** 28% of the total area, but 47% of the TN, and 50% of the TP loads of the whole

LULC of Choctawhatchee

It seems the story is not so bad according to our research!!

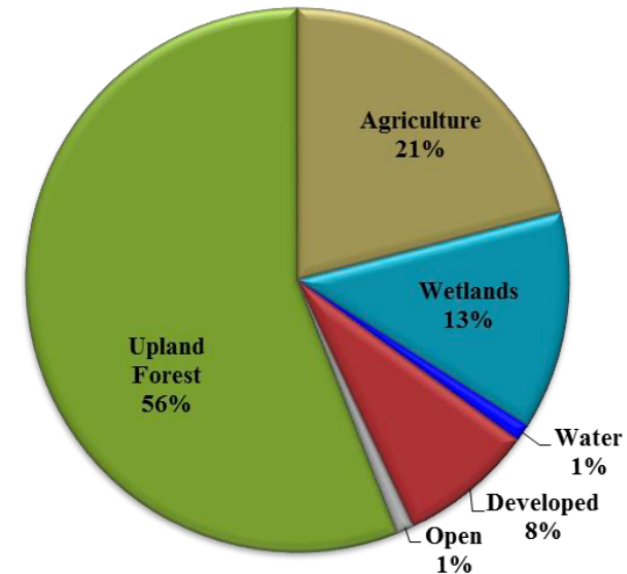


Figure 2-6 Land Cover in the Greater Choctawhatchee Watershed (Alabama and Florida)

Sources: FDEP 2015a; MRCL NLCD 2011

Table 2-1 2012-2013 Land Use and Land Cover in the Choctawhatchee River and Bay Watershed (Florida Only)

Land Use Category	Square Miles	Percent of Basin
Agriculture	334	1.3
Developed	191	9.1
Open Land	42	2.0
Upland Forests	1,032	49.2
Water	28	15.9
Wetlands	469	22.4

Source: FDEP 2015a

By: *Choctawhatchee Bay Estuary Program*
Technical Advisory Committee, 2021

Take away

- ✓ Forest and Source percentages play a critical role in regulating water quality with a threshold effects
- ✓ SOMs provided a powerful tool for data clustering and dimension reduction, aiding in the identification of driving factors and spatial patterns of water quality
- ✓ By the methods initiated in this study it is found 37.46% and 20.26% are the detected thresholds of forest and Source respectively

Thank you!

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