Use ArcNLET to Estimate Nitrogen Load from Removed Septic Systems to Surface Water Bodies in City of Port St. Lucie, City of Stuart, and Martin County

> Ming Ye (<u>mye@fsu.edu</u>) Department of Scientific Computing Florida State University

# Background

- Estimate nitrogen loads from removed septic systems to surface water bodies in the City of Port St. Lucie (red), City of Stuart (green), and Martin County (orange)located in the St. Lucie River and Estuary Basin
- The load estimates can be used to calculate credit for septic tank phase-out projects in support of the on-going Basin Management Action Plan (BMAP).



# Outlines

- Introduction of ArcNLET
  - Rational of developing ArcNLET
  - Functions of ArcNLET and associated software
  - Data requirements of using ArcNLET
- ArcNLET modeling for the City of Port St. Lucie, the City of Stuart, and Martin County
- On-going ArcNLET modeling
- suggestions, and comments

# ArcNLET Project Team

- Contract Manager:
  - Rick Hicks (FDEP) (Richard.W.Hicks@dep.state.fl.us)
- Principal Investigator:
  - Ming Ye (FSU) (<u>mye@fsu.edu</u>)
  - Paul Lee (FDEP) (retired in 2012)
- Graduate Students:
  - Fernando Rios (Graduated in 2010)
  - Raoul Fernendes (Graduated in 2011)
- Post-docs:
  - Liying Wang (2010-2012)
  - Huaiwei Sun (2012-2013)

#### Schematic of an Onsite Sewage Treatment and Disposal System (OSTDS) and Subsurface Nitrogen Transformation and Removal Processes



#### From Heatwole and McCray (2007)

# Soil Processes: Simulated using VZMOD

- Unsaturated flow
- Solute transport
- Nitrification and denitrification

#### **Groundwater Process:**

Simulated using ArcNLET

- Groundwater flow
- Solute transport
- Denitrification

ArcNLET-MC: Quantify uncertainty of ArcNLET simulations 5

# Why Developing ArcNLET?

- There is no suitable tool for estimating nitrate load to meet TMDL requirements and perform Nitrogen BMAP. Existing tools are either too simple or too complex.
- Develop a simplified model that consider key hydrogeologic processes of groundwater flow and nitrate fate and transport.
- Implement the model by developing a user-friendly ArcGIS extension to
  - Simulate nitrate fate and transport including the denitrification process
  - Consider either individual or clustered septic tanks
  - Provide a management and planning tool for environmental management and regulation
- Disseminate the software and conduct technical transfer to FDEP staff and other interested parties.

# What is ArcNLET?

#### ArcGIS-based Nitrate Load Estimation Toolkit

- A simplified
   conceptual model
   of groundwater flow
   and solute
   transport
- Implementation as an ArcGIS extension
- Calculation of nitrate plume and nitrate load



Compatible with ArcGIS 9.3, 10, and 10.1

Simplified Conceptual Model to consider key hydrogeologic processes involved in nitrate transport:





- Groundwater flow
   model to estimate
  - flow path
  - flow velocity
  - travel time
- *Nitrate transport model* to consider
  - Advection
  - Dispersion
  - Denitrification
- Load estimation model to estimate nitrate load

# Illustration of simulated nitrate plumes and nitrate load



#### Software Download and References

- ArcNLET: <u>http://people.sc.fsu.edu/~mye/ArcNLET</u>
- VZMOD: <u>http://people.sc.fsu.edu/~mye/VZMOD</u>
- Peer-reviewed publications:
  - Rios, J.F. (*student*), M. Ye, L. Wang, P.Z. Lee, H. Davis, and R.W. Hicks (2013), ArcNLET: A GIS-based software to simulate groundwater nitrate load from septic systems to surface water bodies, *Computers and Geosciences*, 52, 108-116, 10.1016/j.cageo.2012.10.003.
  - Wang, L. (*post-doc*), M. Ye, J.F. Rios, R. Fernandes, P.Z. Lee, and R.W. Hicks (2013), Estimation of nitrate load from septic systems to surface water bodies using an ArcGIS-based software, *Environmental Earth Sciences*, DOI 10.1007/s12665-013-2283-5.
  - Wang, L. (*post-doc*), M. Ye, P.Z. Lee, and R.W. Hicks (2013), Support of sustainable management of nitrogen contamination due to septic systems using numerical modeling methods, *Environment Systems and Decisions*, 33, 237-250, doi:10.1007/s10669-013-9445-6.

# Input Data of ArcNLET

All input data files are in ArcGIS format.

- Locations of septic tanks
- Locations of water bodies
- Topography (DEM: Digital Elevation Model): Process it to obtain water table
- Hydrogeological and transport parameters
  - Smoothing factor (used to process topography)
  - Hydraulic conductivity (from SSURGO)
  - Porosity (from SSURGO)
  - Dispersivity
  - Decay coefficient of denitrification
  - Source load and concentration



#### **Requirements on Potential Users**

- The GUI make it easier for some with little experience in analyzing groundwater transport problems to apply a solute-transport model to a field problem.
- Users of ArcNLET need to have
  - Basic knowledge of hydrogeology such as concepts of groundwater flow and solute transport
  - Intermediate level of ArcGIS skills for preparing input files and visualizing software output files
- A model (simple or complex) is not an end in itself, but a tool to organize one's thinking and engineering judgment.
- Interpretation and improvement of ArcNLET results require
  - Fundamental understanding of groundwater flow and solute transport
  - Familiarity with site-specific information such as geology and hydrogeology
- It may be useful to test and tune the model for several representative sites to find representative parameter values and use them for prediction.

# **Model Calibration**

- The ArcNLET model requires several model parameters that are largely unknown.
- The parameter values may be obtained from literature review, but the values are not site-specific.
- A better way to determine site-specific parameter values is model calibration to adjust the parameter values to match model simulations to site observations of system state variables such as hydraulic head and nitrate concentration.

#### Manual Model Calibration: Trial and Error



## **Example Model Calibration**

Eggleston Heights with 3,500 OSTDS



- Two neighborhoods in the City of Jacksonville:
  - Eggleston Height
  - Julington Creek
- Relatively large amount of observations of hydraulic head and nitrate concentrations are available.



Average values are used as the calibration targets.

#### Model Calibration Results: Heads



The smoothed DEM agrees well with the mean observed hydraulic head, because the correlation coefficient (0.93) and the slope of linear regression (1.03) are close to one.



### Model Calibration Results: Nitrate Concentrations

- The simulated nitrate concentrations are close to the mean observations.
- Because of the large variability of concentration observations, it happens often that simulated nitrate concentrations deviate from mean observations.
- We consider that the calibration is reasonable if the simulations fall within the inter-quartile of the observed concentrations, which covers 50% of the data.



#### Challenges: Uncertainty in Input Parameters and Load Estimates

Poetry of Donald H. Rumsfeld: Feb. 12, 2002 Department of Defense news briefing

The Unknown As we know, There are *known knowns*. There are things we know we know. We also know There are known unknowns. That is to say We know there are some things We do not know. But there are also *unknown unknowns*, The ones we don't know We don't know.



The calibrated parameters are just one possible combination, and there may be other parameter combinations that give similar model fit but different load estimates.

### An Illustrative Example

Parameter ranges:

Hydraulic conductivity (*K*): 0.0864 ~ 30.4992 m/d Longitudinal dispersivity ( $\alpha_L$ ): 0.21 ~ 21.34 m Horizontal transverse dispersivity ( $\alpha_T$ ): 0.021 ~ 2.134 m First-order decay coefficient (*k*): 0.004 ~ 2.27 /d

Parameter set 1 Load=0.15 lb/day  $\alpha_L$ =2.113m,  $\alpha_T$ =0.234m k=0.008/d Parameter set 2 Load=0.25 lb/day  $\alpha_L$ =2.113m,  $\alpha_T$ =0.234m k=0.004/d Parameter set 3 Load=0.60 lb/day  $\alpha_L$ =21.34m, $\alpha_T$ =0.021m k=0.004/d



#### ArcNLET-MC for Uncertainty Quantification

#### Recently developed. Have not been released.

N ArcNLET - Monte Carlo	
File Help	
Setup Random Variables Deterministic Parameters Execute	
LHS Setup	
Title julingtonclip.mxd 20120723.1505	
Number of Realizations 10 Realizations Frocess a Subset of Realizations	
Monitoring Points	
Select a layer containing points where raster data will be sampled	
Output Setup	
Output Folder (All generated files will be stored in this folder)	
File Management. Controls how output files will be handled after they are used	
Keep all realizations Delete all realizations	
Keep first n realizations Keep every n'th realization n: 20	
Keep only these realizations (comma separated)	
Message Log	
15:05:50 Filling in values on MainForm 15:05:50 ArcNLET UA load	<u>^</u>
	-
4	۱

#### **Uncertainty Analysis**

#### Histogram of Load 600 Mean: 566 g/d 500 Median: 306 g/d Mode: 200 g/d **Leduency** 400 300 200 Lower quartile:168 g/d Upper quartile: 687 g/d 95% percentile: 1897 g/d Maximum estimate: 4654 g/d 100 0 300 Load (g/d)

- The load estimation has large uncertainty.
- Uncertainty reduction can be achieved if more data and 21 information becomes available.

### ArcNLET modeling for the City of Port St. Lucie, the City of Stuart, and Martin County



A technical report has been submitted to FDEP. It can be requested from me or Katie directly.

# Acknowledgement

We appreciate technical assistance and data collection provided by:

- Katie Hallas (FDEP)
- Marcy Policastro (Wildwood Consulting)
- Dale Majewski (City of Port St. Lucie)
- Dianne K. Hughes (Martin County)
- William Griffin (City of Stuart)
- Kevin Carter and Steve Kruppa (SFWMD) and their colleagues.

# **Modeling Procedure**

For each site, whenever site-specific data are available,

- Compile historical data to understand groundwater flow and nitrogen transport at the modeling sites. (Chapter 3)
- Select calibration data of hydraulic head and nitrogen concentration to estimate ArcNLET flow and transport model parameters. (Chapter 3)
- Calibrate the ArcNLET model. (Chapter 4)
- Simulate nitrogen transport at the modeling site, using the calibrated model. (Chapter 4)
- Estimate the nitrogen load. (Chapter 4)
- Conduct Monte Carlo simulation to address uncertainty in model parameters. (Chapter 5)

### **Compiled Data: Water Level**

The data in the modeling sites are old (measured in the period of 1988-1995), but their average values are still representative of the groundwater conditions of the modeling sites.



### **Compiled Data: Nitrogen Concetration**

- Observations of nitrogen concentrations are extremely scarce.
- Four data are available in the City of Port St. Lucie and one data in Martin County.
- The data at well PG-25 was measured in 1976-1977. The other four data were measured in 2008.



Area	Wells	Data source	NO <sub>x</sub>	NH4	TN/DIN
City of Port	SOFLSUS2-19	USGS	0.040	0.220	0.380
St. Lucie	SOFLSUS2-21	USGS	0.021	0.349	0.520
	SOFLSUS2-23	USGS	0.040	0.900	1.260
	PG-25	USGS	0.005	0.283	0.288
Martin County	SOFLSUS2-17	USGS	0.002	0.210	0.290

More data are necessary to validate the modeling results, improve nitrogen transport modeling, and reduce estimation uncertainty.

# Data for ArcNLET Modeling

- All the GIS data needed for ArcNLET modeling are available in the public domain or from local environmental agencies.
- Local data are important, e.g., the canals in the City of Port St. Lucie.





Local canal data + National NHD data

### Calibration of Flow Model

- Good agreement between smoothed DEM and observed water table elevation was achieved at (a) the City of Port St. Lucie and (b) the City of Stuart.
- There is no water level monitoring data for the four sites in the Martin County.



# Calibration of Transport Model

- Reasonably good agreement is achieved for three data: two in the City of Port St. Lucie and one in Martin County (there is no concentration observation in the City of Stuart).
- Overestimation occurred to two data in the City of Port St.
   Lucie.



### **Simulated Nitrogen Plumes**

A strong correlation is observed between the median values of surface water nitrogen concentration and the nitrogen loads to the corresponding surface water bodies.







#### Spatial Variability of Nitrogen Plumes

Spatial variability is obvious at different modeling sites, e.g., Seagate Harbor (left) (reduction ratio of 10.8%) and Hobe sound (right) (reduction ration of 70.5%) in the Martin County.



### Factors Controlling Load Estimate

- Mean length of flow path (left): long mean length of flow path corresponds to more denitrification and thus less load estimate.
- Mean velocity (right): larger mean velocity results in shorter travel time, less denitrification, and thus more load estimate.





32

### Factors Controlling Load Estimate

In the City of Port St. Lucie, the load estimate increases when the drainage condition changes from very poorly drained to excessively drained, because nitrogen transport is faster in well-drained soil is faster than in poorly drained soil.



VPD: very poorly drained PD: poorly drained SPD: somewhat poorly drained MWD: moderately well drained WD: well drained SED: somewhat excessively drained ED: excessively drained

The number of septic systems corresponding to each drainage condition is given in the parentheses

### **Comparison with Literature Data**

The nitrogen reduction ratios in this study have a large range but are comparable with the literature data, especially with that of Roeder (2008) obtained in the Wekiva Study.

Reference	Site Location	Daily nitrogen loads per septic system (g/d)	Daily nitrogen loadings to surface water per septic system (g/d)	Nitrogen reduction ratio
<b>Roeder</b> (2008)	Wekiva Study Area, FL	21.7		70.0% <sup>a</sup>
Valiela et al. (1997)	Waquoit Bay, MA	23	9.87 <sup>b</sup>	57.1%
<b>Meile et al. (2010)</b>	McIntosh County, GA			65-85 % <sup>c</sup>
This study	Port St. Lucie, FL	23	7.60	67.0%
	Stuart, FL	23	11.4	50.4%
	North River Shores, FL	23	20.3	11.7%
	Seagate Harbor, FL	23	20.5	10.8%
	Banner Lake, FL	23	8.15	64.6%
	Rio, FL	23	4.80	79.1%
	Hobe Sound, FL	23	6.78	70.5% 34

The septic system removal (actual and hypothetical) is

- absolutely worthy for the North Fork and Basin 4-5-6 sub-basins,
- (somewhat) worthy for the South Fork sub-basins,
- unworthy for C-24, C-23 and C-44/S-135 sub-basins.



	Basin 4-5-6	C-23	C-24	C-44/ S-153	North Fork	South Fork
Percentage of nitrogen load from septic systems to BMAP estimated load	22.87%	0.03%	1.66%	0.00%	31.20%	10.33%
Percentage of load reduction of removed septic systems to BMAP required reduction	33.67%	0.05%	1.71%	0.00%	17.02%	1.35%
Percentage of load reduction to BMAP required reduction	81.02%	0.06%	3.25%	0.00%	85.75%	<b>25.76%</b>

#### Uncertainty Analysis: Compare with Field Observations

- A monitoring well is available at the site.
- Random parameters based on literature data

Parameter	Distribution	Minimum	Mode	Maximu m
Smoothing Factor	Uniform	20	N/A	80
Longitudinal Dispersivity	Normal	1	N/A	100
Source Plane Concentration	Normal	25	N/A	80
Decay Coefficient	Lognormal	5.4E-5	N/A	0.015

 Random parameters (hydraulic conductivity) based on site-specific data.

Soil Zone FID	Minimum	Mode	Maximum
5	3.629	7.949	12.18
8	12.18	18.14	24.36
9	12.18	18.14	24.36





The simulated concentration at the monitoring location follows a lognormal distribution, which is attributed to the lognormal distribution of the first-order decay coefficient of denitrification, the most influential parameter to nitrogen concentration.

- The histogram indicates that, with the parameter distributions considered in this study, it is significantly more likely for the model to simulate low concentration values than to high values.
- This is consistent with the low nitrogen concentration of 0.29 mg/L observed at the monitoring well, suggesting that the calibrated model is likely to reflect nitrogen transport at the calibration site.

#### Relation between Concentration and (a) Load Estimate



300

- The estimated loads corresponding to the calibration data is relatively large.
- The overall positive correlation indicates that larger nitrogen concentration corresponds to larger load.
- However, larger load estimate may be still possible for low concentration, because uncertainty in the load estimate increases when the simulated concentration decreases.
- The uncertainty can be reduced by collecting more field observations (e.g., continuous monitoring at the well), as more monitoring data can remove the realizations that cannot simulate the monitoring data.

## Use of Monitoring Data



#### For Calibration:

- Are the one-time measurements of nitrogen concentration representative of nitrogen concentration in time?
- Are the measurements at the several locations representative in space?
- The model calibration can be updated by assimilating the new data.

#### For Uncertainty Reduction:

- If observed nitrogen concentrations are continuously higher than the simulated value, the bottom figure indicates that the load estimate will be higher with smaller uncertainty.
- If the opposite, we can update the modeling results by removing the realizations that give higher concentration, which will also reduce the uncertainty and give more certain load estimate.

# Conclusions

- Data and information needed to establish ArcNLET models for nitrogen load estimation are readily available in the modeling areas.
- Although there is no groundwater monitoring network, historical data are available from public-domain databases (e.g., DBHYDRO and USGS websites). However, calibration data is limited.
- After calibrating the ArcNLET flow and transport models, model simulations can reasonably match corresponding field observations.
- ArcNLET estimated nitrogen loads in the modeling sites vary substantially in space, and the spatial variability is useful to management of nitrogen pollution.
- The load estimates can be used directly to facilitate BMAP planning.
- Uncertainty in the load estimate is different at different sites. The overall positive correlation between the load estimate and the simulated concentration at the three sites of MC simulation indicates that larger nitrogen concentration corresponds to larger load.
- In the context of site monitoring, if higher concentrations are continuously observed at the monitoring well, the load estimate should be larger than the deterministic estimate given by the calibrated model.

#### Questions, Suggestions, and Comments?



#### ArcNLET Functions: Graphic User Interface

	Aquifer Denitrification			
	File Execute Settings Help			
	Groundwater Flow Particle Tracking Tra	ansport Denitrificatio	on	Abort
	DEM surface elevation map [L] (raster)	Laver Info	Hydraulic conductivity [L/T] (raster)	Layer Info
	Water bodies (polygon)	Layer Info	Soil porosity (raster)	Layer Info
dashed line: nitrate depleted no impact to lake	Drain field of Onsite Wastewater Treatment System (OWTS	) othing Amount	7 🗧 Z-Factor	1
lake shore	E Contraction of the second se			
				<u> </u>
target lake	Overlapped plume	S		~
	11			

#### Simplifications and Limitations in Groundwater Flow Modeling

#### Simplifications:

- Treat water table as subdued replica of topography (Process topographic to approximate shape of water table)
- Use Dupuit assumption to simulate 2-D, horizontal groundwater flow

#### Limitations:

- Steady-state flow
- 2-D flow instead of fully 3-D flow



#### Simplifications and Limitations in Nitrate Transport Modeling



# Simplifications and Limitations in Nitrate Transport Modeling

#### • Simplifications:

- Analytical solution of transport model with uniform flow
- Linear kinetic reaction for denitrification process
- Limitations:
  - Only consider nitrate (a new module is being developed to simulate ammonium)
  - Pseudo-3D model
  - Steady state model
  - Use of empirical or calibrated value of decay coefficient

# Application at Eggleston Heights and Julington Creek Neighborhoods, Jacksonville



Reasons of selecting the two sites:

- Nitrate due to septic systems is believed to be one of the reasons of nutrient enrichment in surface water bodies (Leggette et al., 2004)
- Relatively large amount of observations of hydraulic head and nitrate concentrations are available.

# Food for Thought

Victor Baker, the former President of the Geological Society of America, Member of Academy of Sciences, once said:

"Allowing the public to believe that a problem can be resolved ... through elegantly formulated ... models is the moral equivalent of a lie."

Pilkey, O.H. and L.P. Javis, 2007. Useless Arithmetic – Why Environmental Scientists Can't Predict the Future, New York, Columbia University Press.

Leonard Konikow (2011, Ground Water): "the secret to successful solute-transport modeling may simply be to lower your expectations."

# **Challenges of Applications**

• Keith Beven (2001): The Dalton Lecture How far can we go in distributed hydrological modeling?

"The principles are general and we have at least a qualitative understanding of their implications, but the difficulty comes in the fact that we are required to apply hydrological models in particular catchments, all with their own unique characteristics."

 For a fixed model structure, model outputs are determined by model parameters. However, in most applications, there is no site-specific measurements of model parameters.

### Monte Carlo Method to Address Parametric Uncertainty

- Identify random parameters X and their distributions p(x) (uncertainty characterization)
- Draw samples (x) from the distributions
- Run the model, y=f(x), for each sample
- Obtain probability density function, p(y) of desired predictions



#### What do we have?

- SSURGO database for Duval County that contains hydraulic conductivity and porosity
- Observations of hydraulic head and nitrate concentration
- Evidence of denitrification

#### What do we need?

- Smoothing factor
- Dispersivity
- Decay coefficient of denitrification
- Source plane size and concentration





#### Schematic of an Onsite Sewage Treatment and Disposal System (OSTDS) and Subsurface Nitrogen Transformation and Removal Processes



#### From Heatwole and McCray (2007)

# Soil Processes: Simulated using VZMOD

- Unsaturated flow
- Solute transport
- Nitrification and denitrification

#### **Groundwater Process:**

Simulated using ArcNLET

- Groundwater flow
- Solute transport
- Denitrification

ArcNLET-MC: Quantify uncertainty of ArcNLET simulations 51 Illustration for loamy soil

- Share data files of ArcNLET such as raster files of DEM, hydraulic conductivity and porosity.
- Model parameters for various soil types.
- Estimate nitrate load to groundwater for multiple septic tanks.

% Nitrate Fate And Transp	oort In Soi	I						
Select Soil Types	Hydrau	lic Parameters	Nitrificati	on Parameters	Denitrifica	tion Parameters	29.4534195829 mg/l	
C Clay	HLR	2.0	Kr-max	56.0	Vmax	2.56	Nitrate concentration of Septictank 579 is 28.0753677696 mg/l	
C Clay Loam	aG	0.021	Km-nit	5.0	Km-dnt	5.0	Nitrate concentration of Septictank 580 is 26.0708917094 mg/l	[
• Loam	aVG	0.011	e2	2.267	e-dot	3 774	Nitrate concentration of Septictank 581 is 25.639511792 mg/l	
C Loamy Sand	Ks	12.04	еЗ	1.104	e-unit	0.247	Nitrate concentration of Septictank 582 is 28.7852402434 mg/l	
C Sand	θr	0.061	βnit	0.347	ßdnt	0.347	Nitrate concentration of Septictank 583 is 26 2275453238 mg/l	
O Sandy Clay	θs	0.399	fs	0.0	sdn	0.0	Nitrate concentration of Septictank 584 is	
C Sandy Clay	n	1.474	fwp	0.0	Water Tab	le Depth	Nitrate concentration of Septictank 585 is	
C Sandy Clay Loam	m	0.321	swp	0.154	Distance	288	Nitrate concentration of Septictank 586 is	
C Sandy Loam	I	0.5	sl	0.665			Copying the source point file to the workspace	
⊂ Silt	Temper	rature Parameters	sh	0.809	Output Co	oncentrations	and adding the calculation results to it A new shape file has been created with	
O Silty Clay	Т	18.5	Effluent (	Concentrations	C-NH4	1e-05	calculated nitrate concentrations added to the field "N0_Conc"	
O Silty Clay Loam	Topt-ni	t 25.0	C0-NH4	60.0	C-NO3	30.764884896	Calculation is done, you can check the concentration profile of individual septic tank by	
C Silty Loam	Topt-dr	nt 26.0	C0-NO3	1.0			the FID	
Multiple source		Heterogeneous	hydraulic cond	uctivity and soil p	orosity 🔽 Us	ing smoothed DE	M to calculate WTD	
Source locations file(poin	t)	E:/julingtonUA/sul	- b_septictank.shp	)	, _		Browse	
Hydraulic conductivity file	e(raster)	) E:/julingtonUA/hydr_cond_t.img					Browse	
Soil porosity file(raster)		E:/julingtonUA/porosity_heter.img					Browse	
Smoothed DEM file (raste	r)	E:/temp/smoothedDEM.img					Browse	
DEM file (raster)		E:/julingtonUA/lidardem.img					Browse Quit	



#### Illustration for sandy soil





#### **Random Parameter and Their Distributions**



#### Maximum, minimum and representative values of hydraulic conductivity is derived from soil data

Parameter	Distribution	Max	Representative	Min
hy_con23	TRIANGULAR	3.6593	7.9488	12.1976
hy_con65	TRIANGULAR	3.6593	7.9488	12.1976
hy_con71	TRIANGULAR	0.122	0.6705	1.2198
hy_con73	TRIANGULAR	0.122	0.6705	1.2198
hy_con116	TRIANGULAR	3.6593	7.9488	12.1976
hy_con117	TRIANGULAR	0.122	0.6705	1.2198
hy_con120	TRIANGULAR	1.2198	6.696	12.1976
hy_con164	TRIANGULAR	0.122	0.6912	1.2198
hy_con165	TRIANGULAR	12.1824	21.3408	30.4992
C <sub>0</sub>	NORMAL	25		80
α,	NORMAL	0.21		21.34
k .	LOGNORMAI	0.004		1.08
smthF	UNIFORM	20		80

#### Distributions of LHS Samples

#### Identify Influential Model Parameters





#### Uncertainty Reduction by Field Observations

- The parametric uncertainty can be reduced dramatically by incorporating the field observations into model calibration.
- Take the first-order decay coefficient as an example.



#### **Uncertainty Reduction of Load Estimation**

Load estimates before incorporating field observations.



<b>Frequency</b> 5 4 2 5 3 5 4 5 4														
1	_													
0	1000	1100	1200	1300	1400	1500	1600	1700	1800	1900	2000	2100	More	I

Load (g)

Mean	1334.48
Median	1225.43
Standard	
Deviation	652.61
Minimum	177.62
Maximum	5655.87
Realizations	2000
95 <sup>th</sup> percentile	2581.89
5 <sup>th</sup> percentile	513.28

Mean	1504.24
Median	1466.39
Standard Deviation	257.08
Minimum	1048.57
Maximum	2078.18
Realizations	19

### **On-going and Future Research**

- Analyzing monitoring data collected with SJRWMD support to better understand hydrogeology and nitrogen dynamics in neighborhoods of Jacksonville (e.g., controlling factors of nitrogen concentrations)
- Validating ArcNLET using existing and new data such as groundwater baseflow collected by the City
- Supporting FDEP on an effort of selecting representative sites to better characterizing model parameters
- Developing new functions of ArcNLET, e.g., simulating ammonium concentrations
- Applying ArcNLET to other sites in Florida

### **On-going and Future Research**

- Analyzing monitoring data collected with SJRWMD support to better understand hydrogeology and nitrogen dynamics in neighborhoods of Jacksonville (e.g., controlling factors of nitrogen concentrations)
- Validating ArcNLET using existing and new data such as groundwater baseflow collected by the City
- Supporting FDEP on an effort of selecting representative sites to better characterizing model parameters
- Developing new functions of ArcNLET, e.g., simulating ammonium concentrations
- Applying ArcNLET to other sites in Florida

### Summary of Load Estimation

	City of	City of	Martin County				
	Port St. Lucie	Stuart	North River Shores	Seagate Harbor	Banner Lake	Rio	Hobe Sound
Total Load (kg/d)	42.48	1.665	8.346	9.255	0.856	0.317	0.346
Total load (lbs/yr)	34206.5	1340.7	6720.5	7452.5	689.3	255.3	278.6
Number of Septic Systems	5592	146	411	451	105	66	51
Load per Septic System (g/d)	7.60	11.40	20.31	20.52	8.15	4.80	6.78
Nitrogen Reduction Ratio (%)	67.0	50.4	11.7	10.8	64.6	79.1	70.5

### Compiled Data: Surface Water Quality

- Dale Majewski from the City of Port St. Lucie provided the surface water quality data measured at 21 stations.
- The data from fourteen stations, located in the septic tank removal area, were analyzed.



- At all the stations, the TN concentrations are higher than the TMDL target for most of the monitoring period.
- This is mainly caused by the high TKN concentrations, which are significantly higher than NO<sub>x</sub> concentrations.

