

Remote Sensing for Lake Management

Assessment of Water Quality of a Eutrophic Lake in North Central Florida

Prepared for:

**Shruginar 2012
14 May 2012**

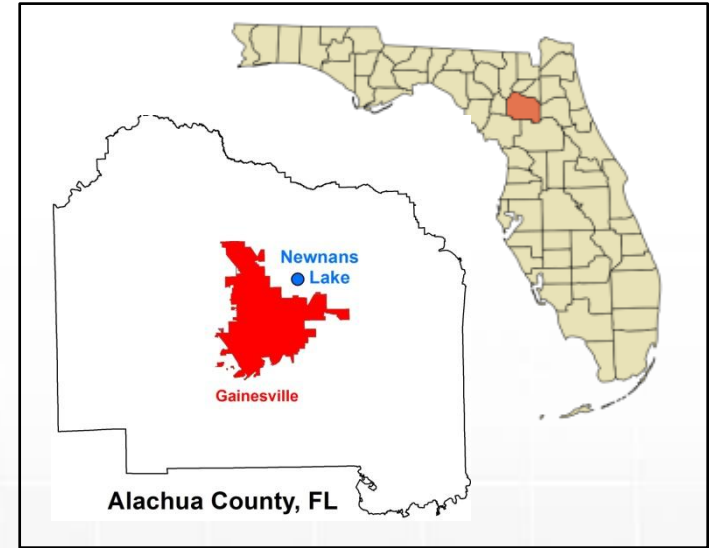
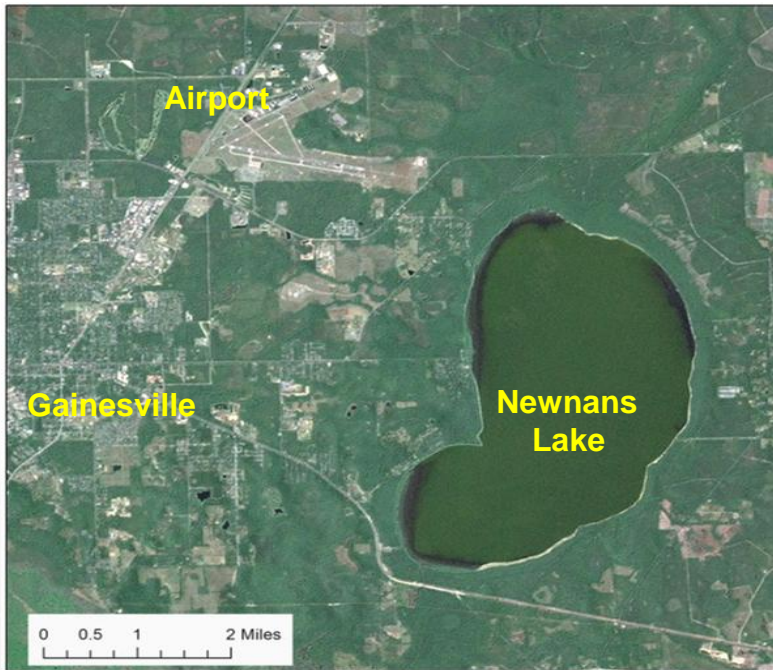
**Kevin Shortelle
System Dynamics International, Inc
kshortelle@sdi-inc.com**

Overview & Objective

Determine if remote sensing data can be used to assess water quality for shallow eutrophic lakes in North Central Florida

- ***In Situ* data collection**
- **Satellite imagery**
- **Remote sensing used effectively to assess water quality in deep (> 15 ft) northern lakes ***
- **Determine applicability of deep water model to shallow lakes (< 10 ft)**

Study Area – Newnans Lake



Hydrological Features

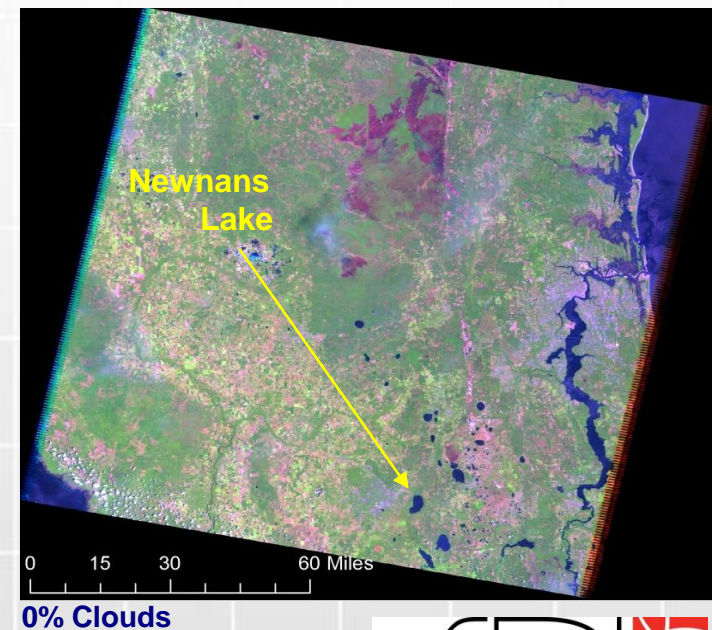
FEATURE	VALUE
Area (acres)	6,600
Average Depth (ft)	4.4
Maximum Depth (ft)	11.5
Volume (ac-ft)	29,000
Average Stage (ft NGVD)	66.5



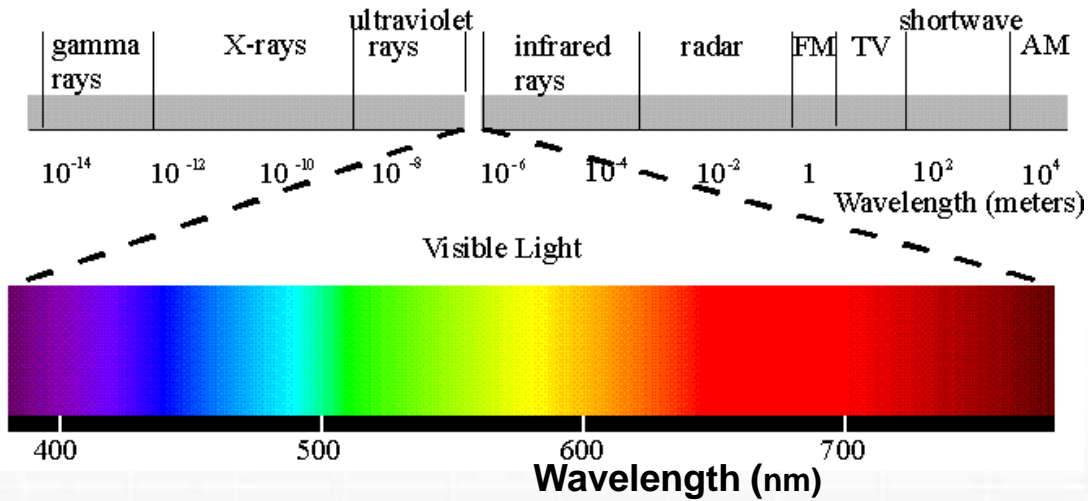
Landsat 5 Thematic Mapper Imagery

Spectral Band (μm)	Band Descriptor	Spatial Resolution (m)	Radiometric Resolution (bits)	Swath Width (km)	Repeat Orbit (days)	Cost per Image (\$)
0.45 - 0.52	Blue (TM1)	30	8	180	16	Free
0.52 - 0.60	Green					
0.63 - 0.69	Red (TM3)					
0.76 - 0.90	Near IR					
1.55 - 1.75	SWIR					
2.08 - 2.35	LWIR	120				
10.4 - 12.5	Thermal IR					

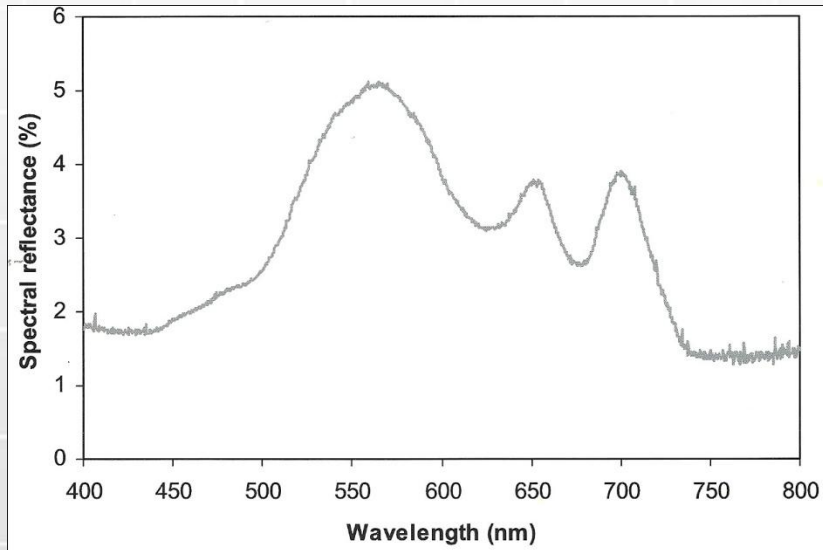
- Available at <http://glovis.usgs.gov> (as GeoTiff)
- Spatial resolution (\approx 0.1 hectare) suitable for mapping in-lake variability
- Wide spatial extent
- Relatively high temporal resolution
- False Natural Color image shown – bands 5,4,3 (SW IR Color Composite)



Electromagnetic Spectrum



Spectral Reflectance of Eutrophic Lake

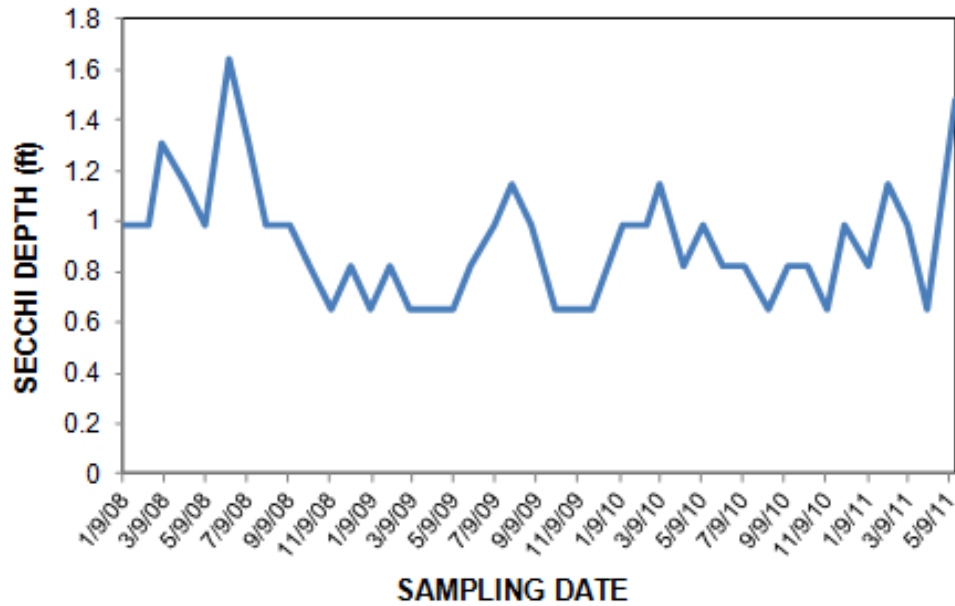


Landsat Processing Workflow

1. Access USGS web site and download zipped Landsat data file
2. Unzip and import seven *.TIF files in ERDAS Imagine and convert to *.img files
3. Form composite image from seven *.img files
4. Create “chip” of composite image of just Newnans Lake and surrounding area
5. Perform **unsupervised classification** on image chip using eight classes
6. Add image chip and classified image chip into ArcMap. Perform **Select-by-Attribute** query on classified chip to select only classified water feature
7. Use water feature as mask to perform **Extract-by-Mask** on image chip, thereby rendering a water-only raster comprising the seven spectral TM bands
8. For water-only raster, determine mean values of TM1 (blue) and TM3 (red) bands (Properties → Symbology → Statistics)



In Situ Data (Secchi Depth, SD)



- Secchi depth, turbidity, chlorophyll *a*, total suspended solids
- High temporal resolution



SAMPLEID	COLLECTDATE	SAMPLETYPE	ANALYTENAME	RESULTVALUE	UNITS
L20080785-005	3/6/2008	VERT-INT	Secchi	0.4	m
L20080975-005	4/9/2008	VERT-INT	Secchi	0.35	m
L20081130-001	5/8/2008	VERT-INT	Secchi	0.3	m
L20081298-005	6/12/2008	VERT-INT	Secchi	0.5	m
L20081441-005	7/10/2008	VERT-INT	Secchi	0.4	m
L20081612-001	8/7/2008	VERT-INT	Secchi	0.3	m
L20081786-005	9/10/2008	VERT-INT	Secchi	0.3	m
L20090052-002	10/9/2008	VERT-INT	Secchi	0.25	m
L20090272-001	11/11/2008	VERT-INT	Secchi	0.2	m
L20090423-004	12/9/2008	VERT-INT	Secchi	0.25	m
L20090566-001	1/8/2009	VERT-INT	Secchi	0.2	m
L20090696-001	2/5/2009	VERT-INT	Secchi	0.25	m
L20090845-002	3/4/2009	VERT-INT	Secchi	0.2	m



Courtesy of SJRWMD

Data Summary

Landsat Image Filename (Path 17/Row 39)	Image Acquisition Date	Mean TM1	Mean TM3	<i>In Situ</i> Data Collection Date	<i>In Situ</i> Secchi Depth Value (ft)
LT50170392009034GNC01	2/3/2009	43.95	14.33	2/5/2009	0.82
LT50170392010309GNC01	11/5/2010	46.54	15.62	11/9/2010	0.66
LT50170392011008EDC00	1/8/2011	40.95	13.29	1/11/2011	0.82
LT501703920110040GNC01	2/9/2011	47.48	14.59	2/8/2011	1.15
LT50170392011072GNC01	3/13/2001	53.43	19.12	3/8/2011	0.98

LT50170392011072GNC01

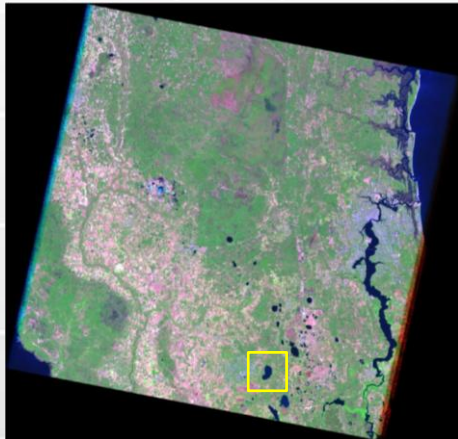
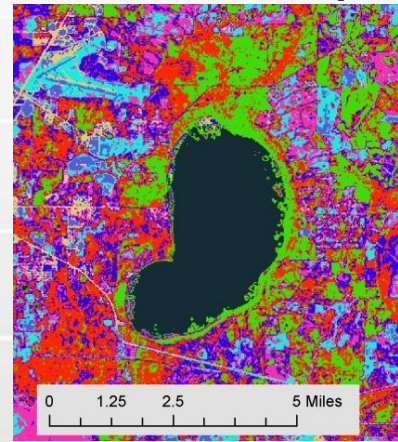


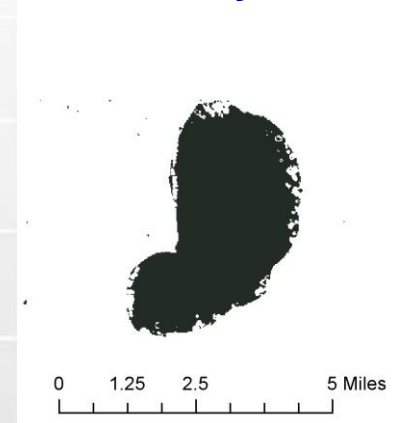
Image Chip



Classified Chip



Water-Only Raster



Calibrate TM data with *in situ* SD measurements and use that relationship to predict SD from other TM data

Model Formulation

- Determine applicability of model used to predict water quality for deep Northern lakes

$$\ln(\text{SD}) = a(\text{TM1}) + b(\text{TM1}/\text{TM3}) + c$$

- Perform multiple regression analysis in Excel to estimate model coefficients based on TM and *in situ* calibration data



Regression Statistics	
Multiple R	0.997262864
R Square	0.99453322
Adjusted R Square	0.983599661
Standard Error	0.029374361
Observations	4

	Coefficients	Standard Error	t Stat	P-value
Intercept	-6.263222	0.474338156	-13.2041	0.048122
X Variable 1	-0.0027597	0.006069129	-0.45472	0.728309
X Variable 2	2.00887574	0.154562162	12.9972	0.048885

Estimated Model Coefficients:

$$a = -0.0027597 \quad b = 2.00887574 \quad c = -6.26322$$

Model Prediction

- Use estimated coefficients to predict Secchi Depth based on TM1 and TM3 data
- Model yielded poor prediction of Secchi Depth

$$\ln(\text{SD}) = -0.002759(53.43) + 2.00887(53.43/19.12) - 6.26322$$

$$\ln(\text{SD}) = -1.022, \text{ so } \text{SD} = 0.36 \text{ ft (vs. 0.98 ft, actual value)}$$

- Why did model perform poorly?
 - Insufficient number of calibration samples?
 - Northern lake model not appropriate for Florida lakes?
 - Reflectance from vegetation, sediment, and lake bottom affects spectral signatures
 - Time interval between Landsat image acquisition and *in situ* data collection more critical for shallow lakes since water quality can change abruptly based on weather conditions (thunderstorms, runoff, etc.)

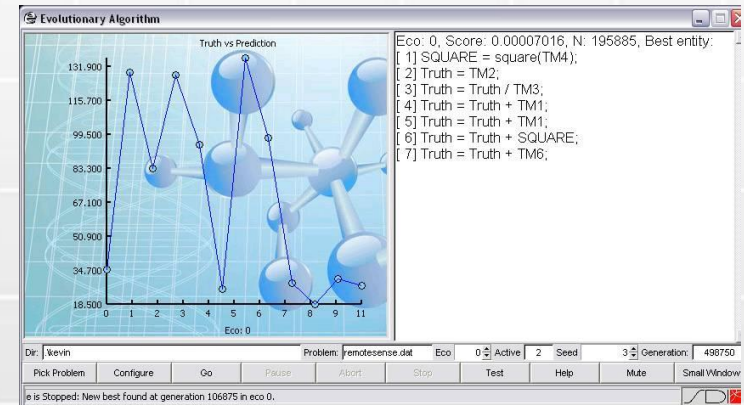
What's a Possible Solution?

- Develop model that considers:
 - All seven TM spectral bands
 - Alternative water quality metrics (e.g., chlorophyll *a*, turbidity, TSS)

$$\text{Water Quality Metric} = f(\text{TM1}, \text{TM2}, \dots, \text{TM7})$$

- Use **genetic algorithm** to estimate functional relationship (i.e., model) between TM data and *in situ* water quality metric

- Hypothetical example shown on next chart



Genetic Algorithm Illustration

Hypothetical Data

Water Quality Metric (Truth)	TM1	TM2	TM3	TM4	TM5	TM6	TM7
29.9	2	6	7	4	3	9	1
94.8	4	9	12	9	7	5	5
63.0	8	4	4	6	7	10	1
93.7	5	5	7	9	3	2	3
70.8	8	11	4	7	9	3	6
23.3	4	7	3	2	8	9	3
99.5	8	6	4	9	2	1	5
73.0	9	9	3	7	9	3	3
25.3	7	5	4	2	4	6	6
18.5	4	3	2	1	9	8	7
26.7	1	2	3	4	7	8	9
24.4	7	4	9	1	5	9	6

Genetic Algorithm Solution

```
Eco: 0, Score: 0.00007016, N: 195885, Best entity:  
[ 1] SQUARE = square(TM4);  
[ 2] Truth = TM2;  
[ 3] Truth = Truth / TM3;  
[ 4] Truth = Truth + TM1;  
[ 5] Truth = Truth + TM1;  
[ 6] Truth = Truth + SQUARE;  
[ 7] Truth = Truth + TM6;
```

se.dat Eco 0 Active 2 Seed 3 Generation: 498750
Stop Test Help Mute Small Window

Pseudo-code reduces to:

$$\text{Truth} = 2(\text{TM1}) + (\text{TM2}/\text{TM3}) + (\text{TM4})^2 + \text{TM6}$$

Questions?