Characterizing Disaster Impact Using Lidar and AVIRIS: 2011 Mississippi Flood Study



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Characterizing Disaster Impact Using Lidar and AVIRIS: 2011 Mississippi Flood Study

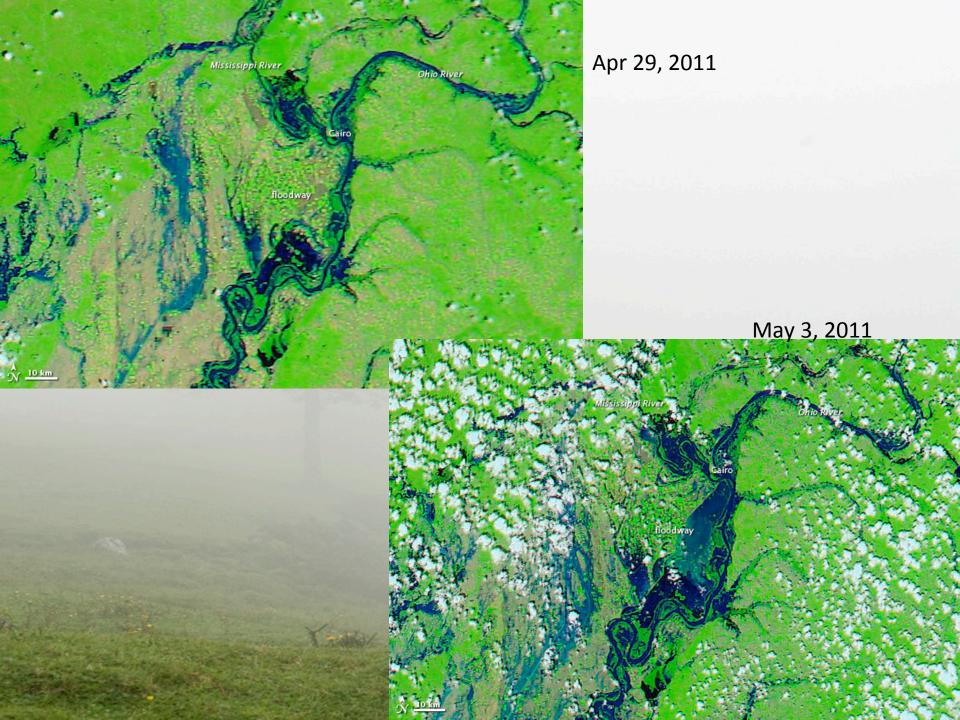
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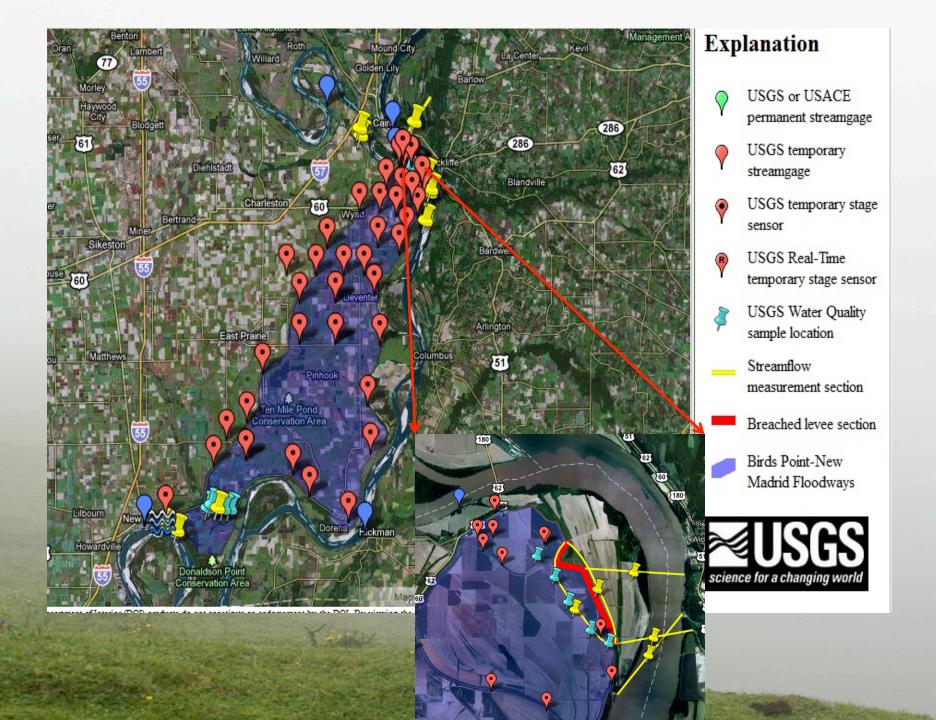
¹Department of Civil & Env. Eng., Univ. of Illinois ²Department of Geography, Univ. of Illinois ³US Army Corps of Engineers, Memphis ⁴USGS

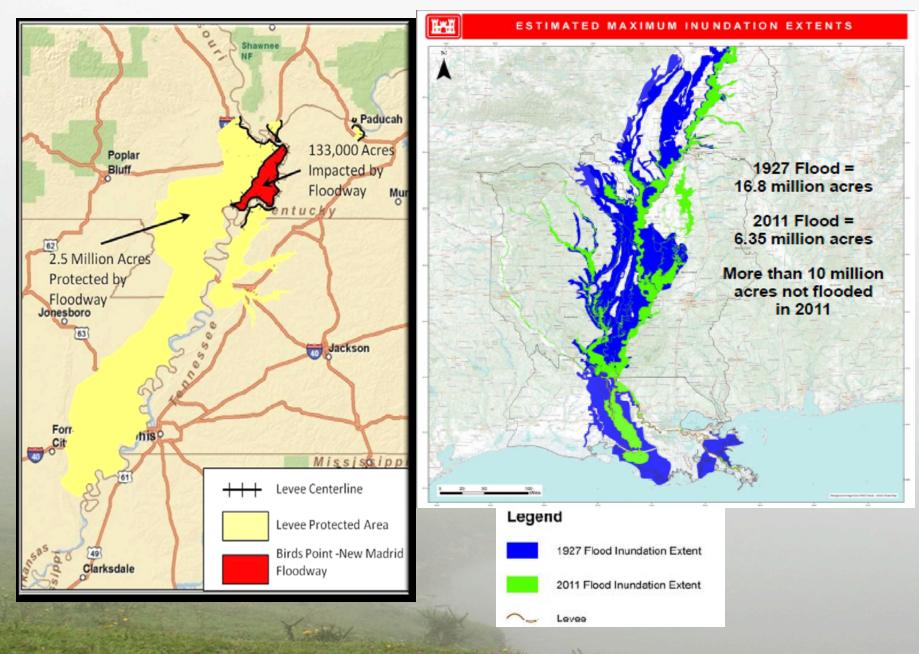
⁵Natural Resource and Env. Science, Univ. of Illinois ⁶Department of Zoology, Southern Illinois Univ.



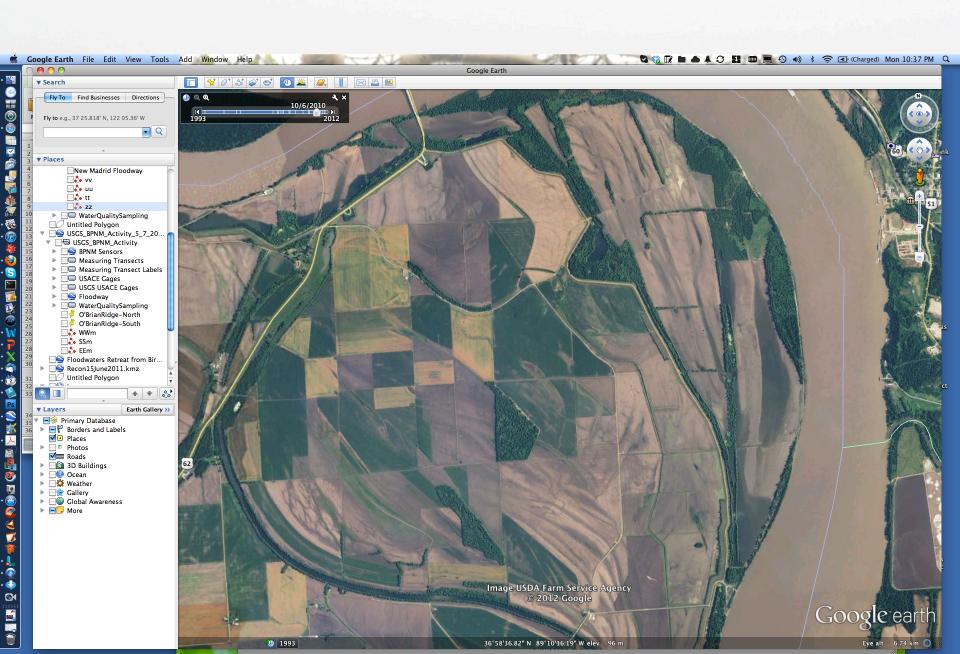


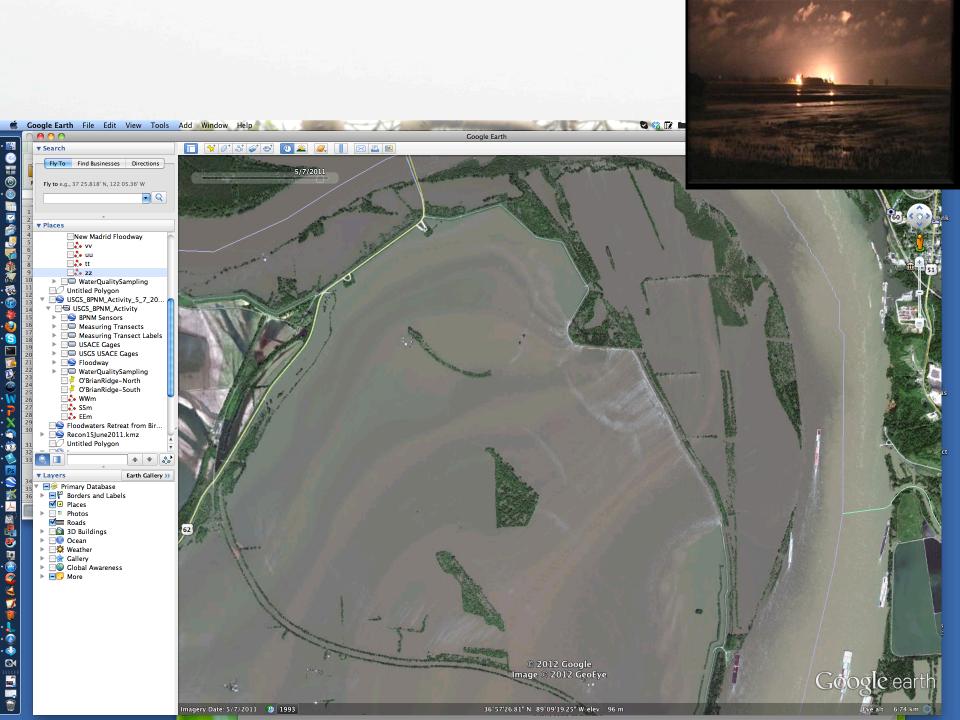


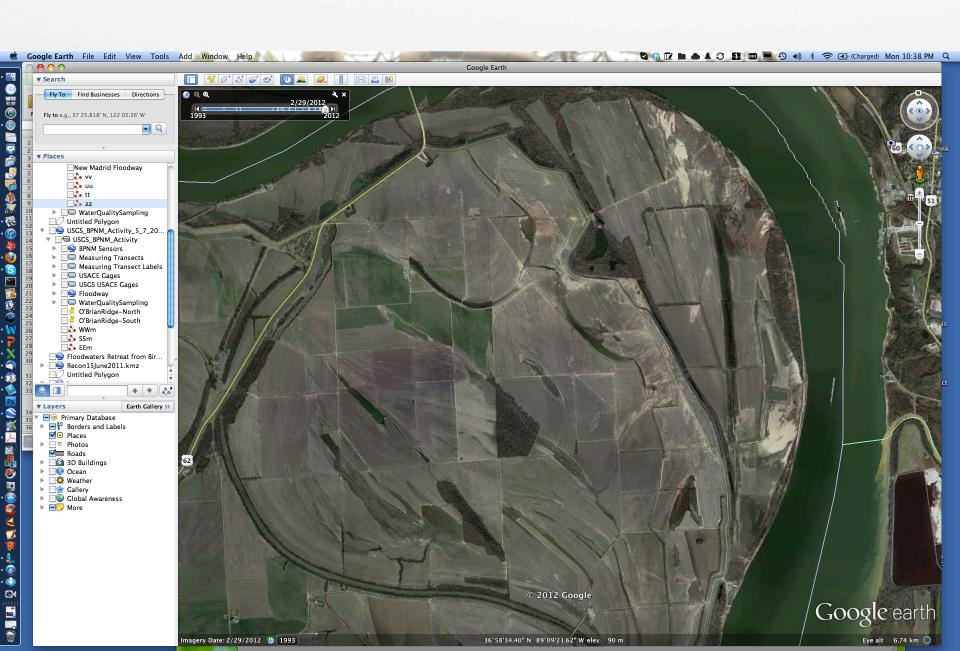




Mississippi River Commission 2011 MR&T Flood Report



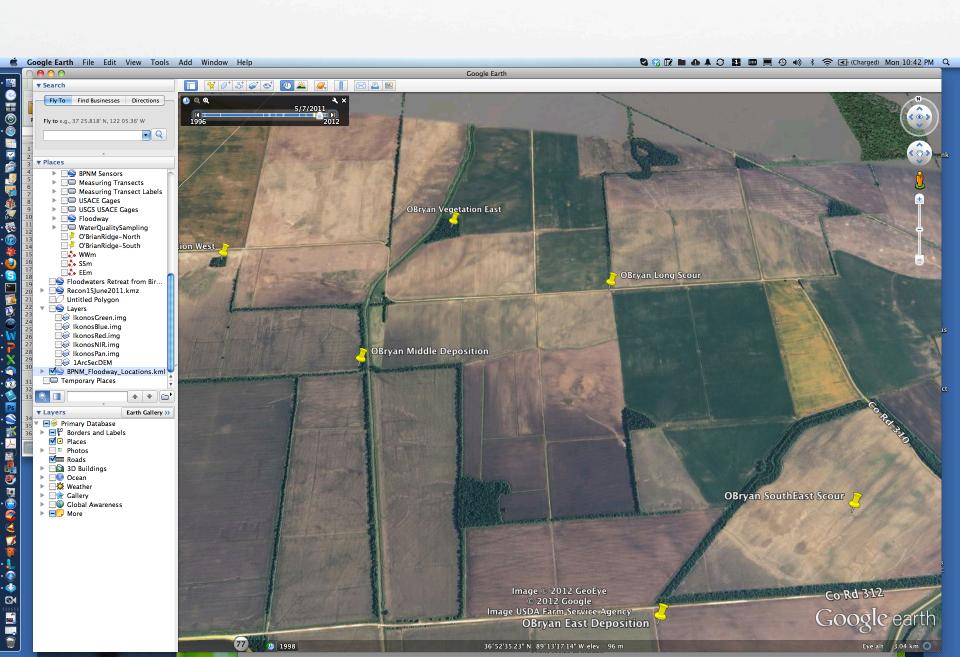


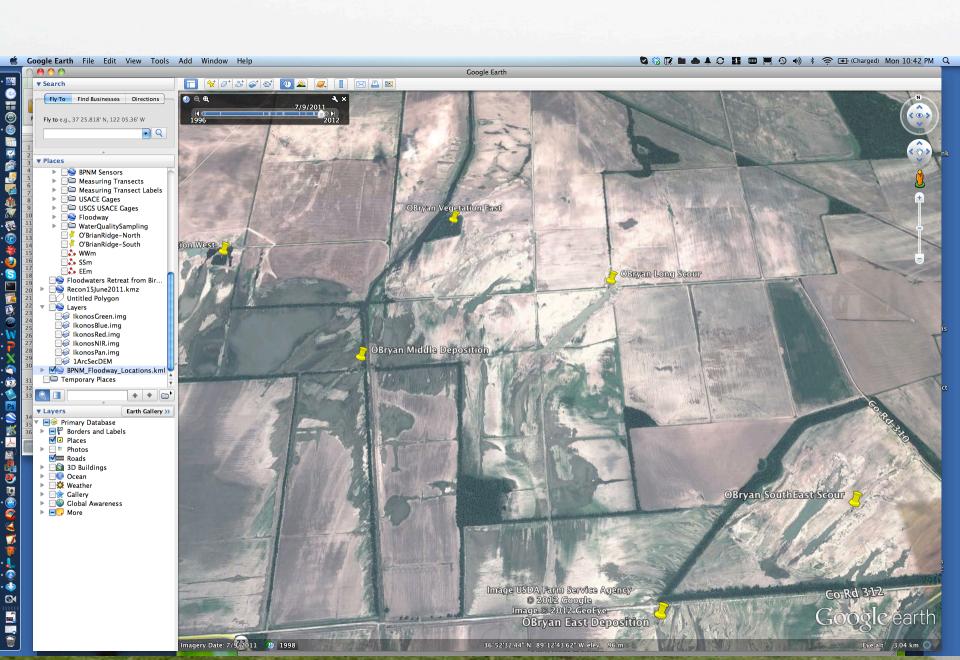


IKONOS (05-12-2011)













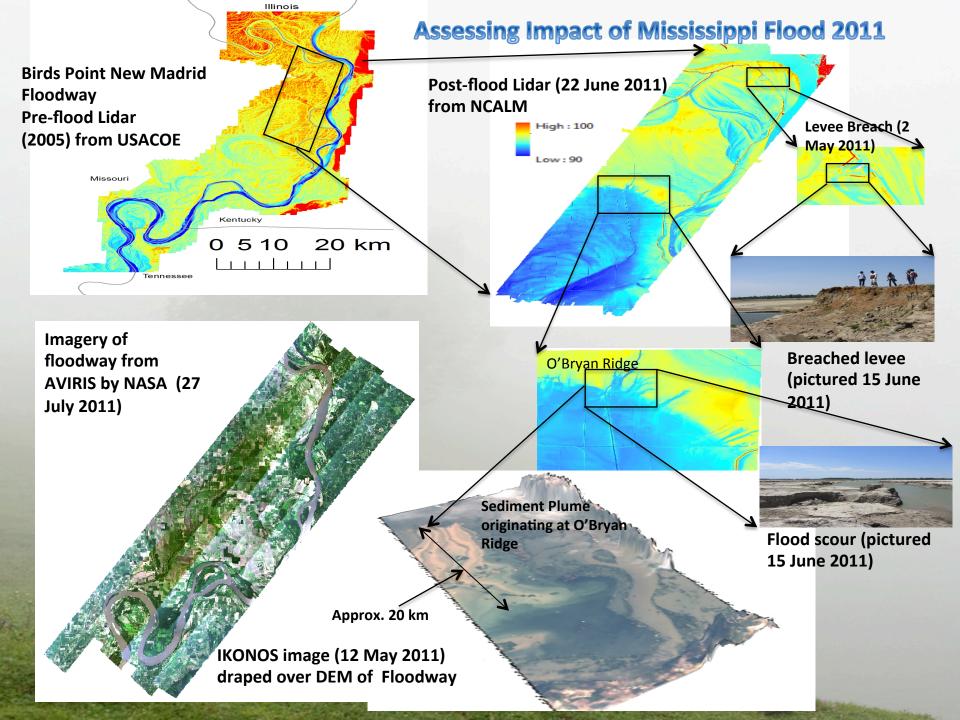




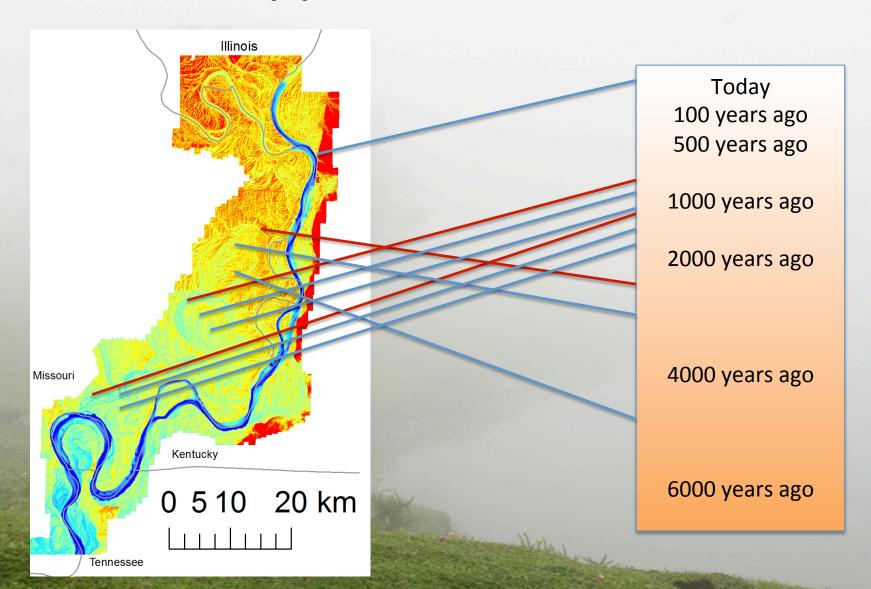
Photos of BPNM Floodway scouring and deposition, 15th June 2011

Problem Statement

- How can we assess the impact of the flood (disaster) at such large spatial scales?
- What are the challenges and what methodological advances are required?
 - Classification
 - Identification
- How can the estimates be scaled for satellite (HyspIRI) based assessment?
- How can this data be used for modeling and prediction?



Mississippi River Time Line



Vegetation and Moisture Indices

- Normalized Difference Vegetation Index (NDVI)
 - -NDVI=NIR-Red/NIR+Red
 - Values between +1 and -1
 - Range for green vegetation is 0.2 to 0.8
- Water Band Index (WBI)
 - -WBI = 900/970
 - As water content increases the strength of absorption around 970 nm increases relative to 900 nm.
 - Green vegetation ranges from 0.8 to 1.2.

References

Penuelas, J., I. Filella, C. Biel, L. Serrano, and R. Save, 1995. *The Reflectance at the 950-970 Region as an Indicator of Plant Water Status*. International Journal of Remote Sensing 14:1887-1905.

Sellers, P.J., 1985. *Canopy Reflectance, Photosynthesis and Transpiration*. International Journal of Remote Sensing 6:1335-1372.

Vegetation and Moisture Indices

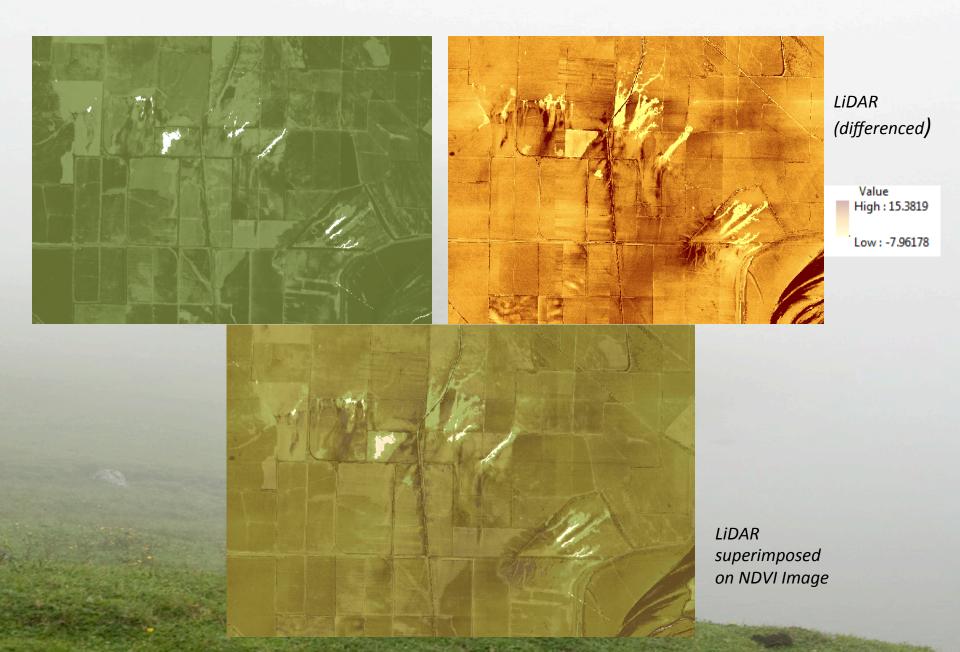
- Moisture Stress Index (MSI)
 - As water content increases the strength of absorption around 1599 nm increases but that of 819 nm is unaffected.
 - Ranges from 0 to more than 3
 - Green vegetation ranges from 0.4 to 2.
- Normalized Difference Nitrogen Index (NDNI)
 - Reflectance around 1510 nm dependent on Nitrogen content as well as foliar biomass
 - Reflectance around 1680 nm dependent on foliar biomass only.
 - Green vegetation ranges from 0.02 to 0.1.

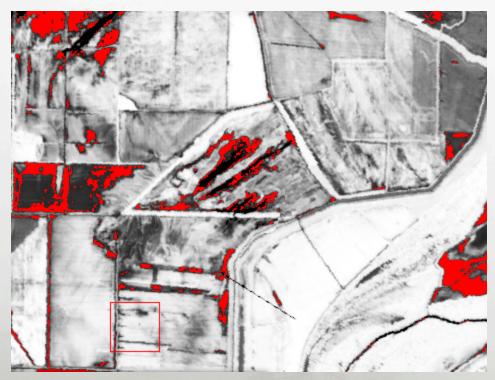
References

Hunt Jr., E.R. and B.N. Rock, 1989. *Detection of Changes in Leaf Water Content Using Near- And Middle-Infrared Reflectances*. Remote Sensing of Environment 30:43-54.

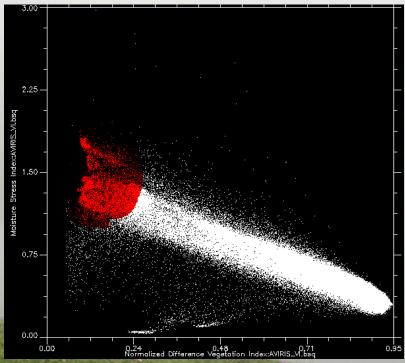
Serrano, L., J. Penuelas, and S.L. Ustin, 2002. *Remote Sensing of Nitrogen and Lignin in Mediterranean Vegetation from AVIRIS Data: Decomposing Biochemical from Structural Signals*. Remote Sensing of Environment 81:355-364.

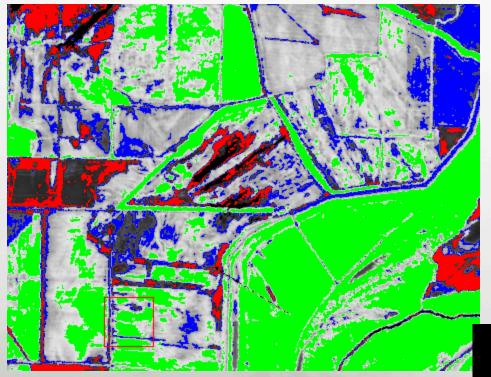
NDVI



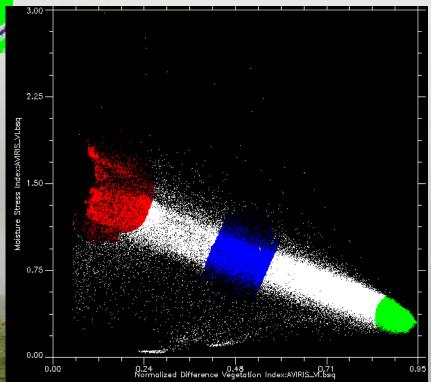


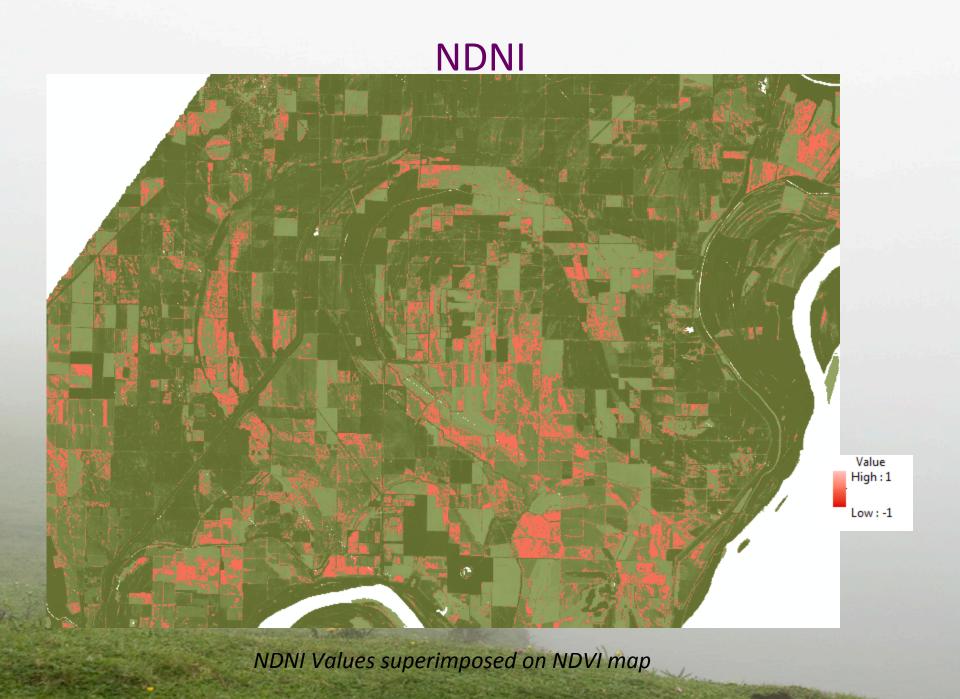
Zones of high scouring show strong signals of High moisture stress and low NDVI values





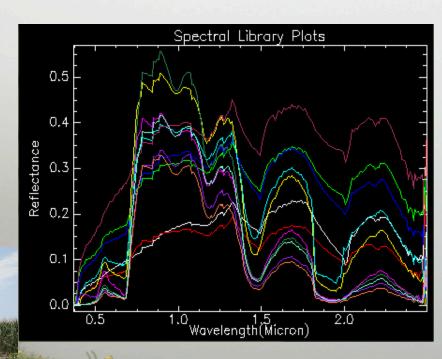
Zones of very High NDVI and low moisture stresses are the Forest Regions





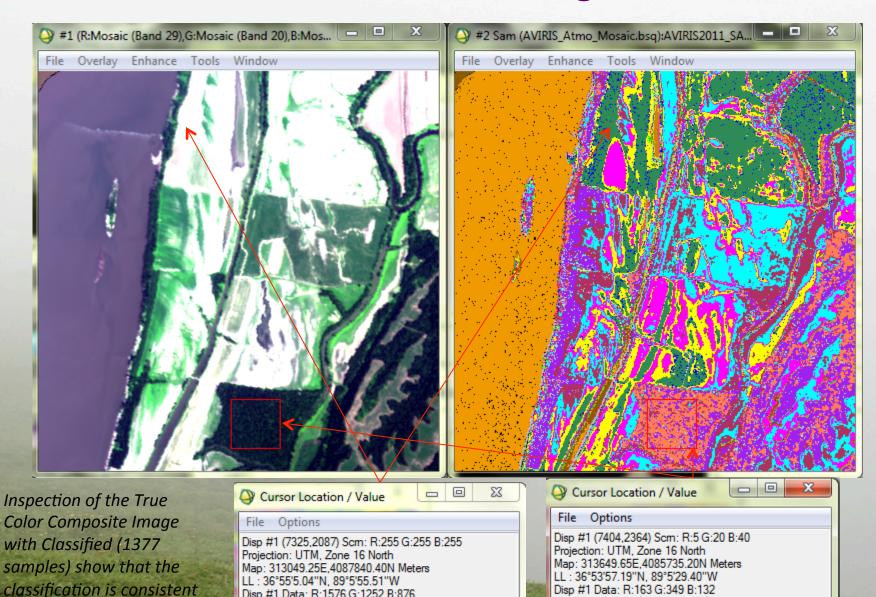
Spectral Angle Mapper (SAM) Classification using Endmember Collected from Data Combined with USGS Library

- The USGS digital spectral library 06a has 1365 spectral signatures (0.2 150 micron):
 - Minerals
 - Soils, Rocks, and Mixtures (except those with vegetation)
 - Coatings
 - Liquids, Liquid Mixtures, Water, Other Volatiles, and Frozen Volatiles
 - Artificial (Man Made) Including Manufactured Chemicals
 - Plants, Vegetation Communities, Mixtures with Vegetation, and Microorganisms
- These have been convolved to the AVIRIS wavelength range (0.4 – 2.5 micron).
- A number of endmember spectra of trees, bare soils, small plants and corn have also been collected from the calibrated image and used for classification.



Spectral Plots from Endmember Collection

Ground Truthing



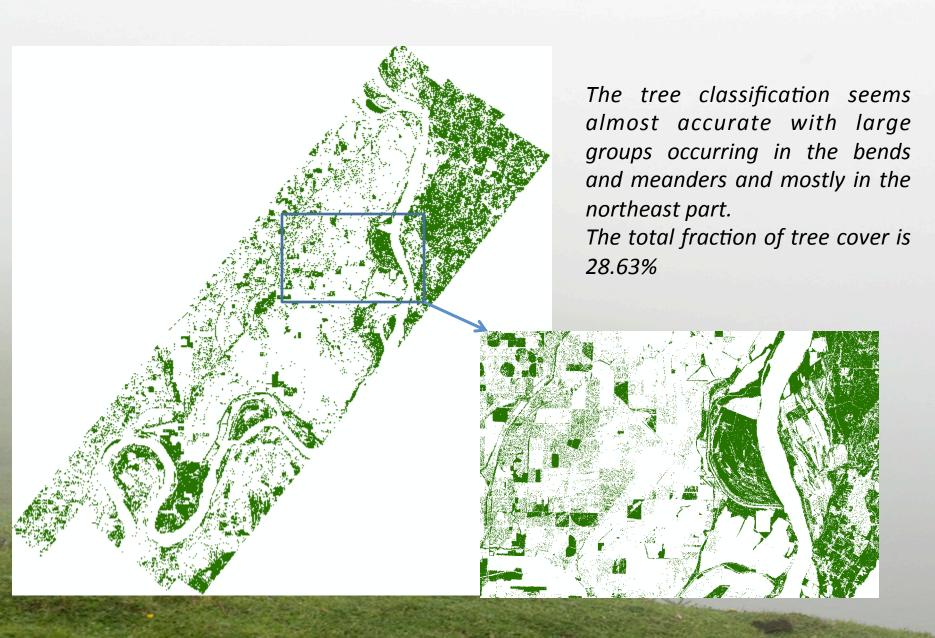
Disp #2 Data: 10 {Treespecies2}

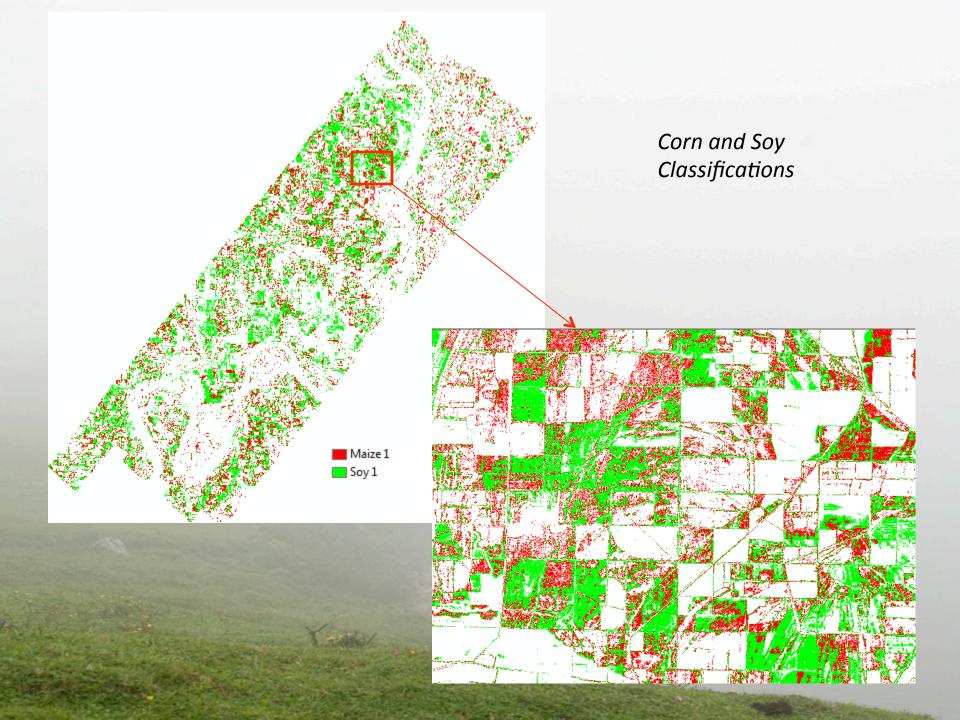
Disp #1 Data: R:1576 G:1252 B:876

Disp #2 Data: 8 (Obryansoil)

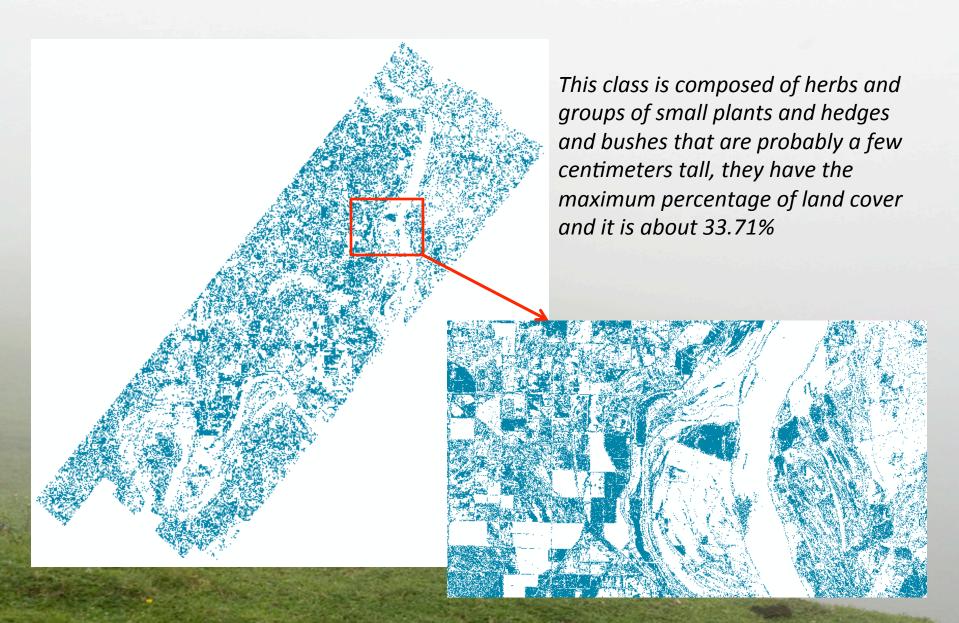


Trees

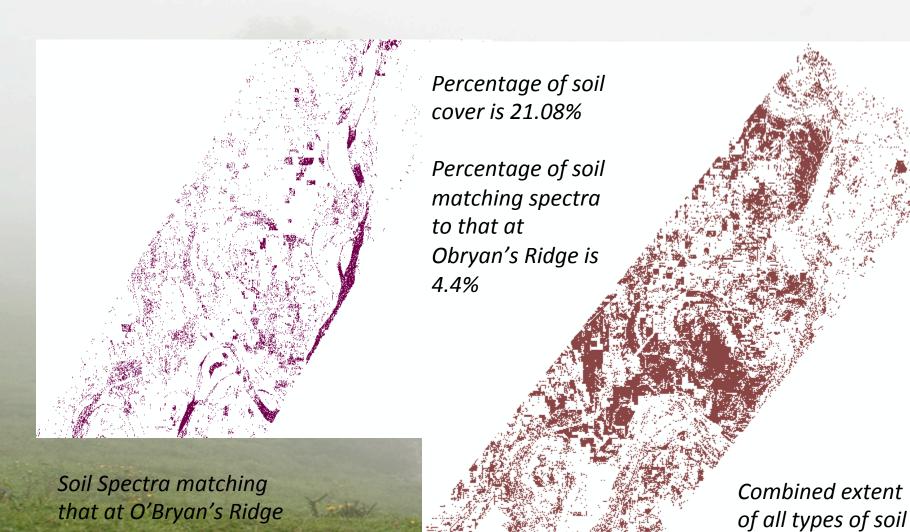




Herbs and Small Plants



Soils



cover

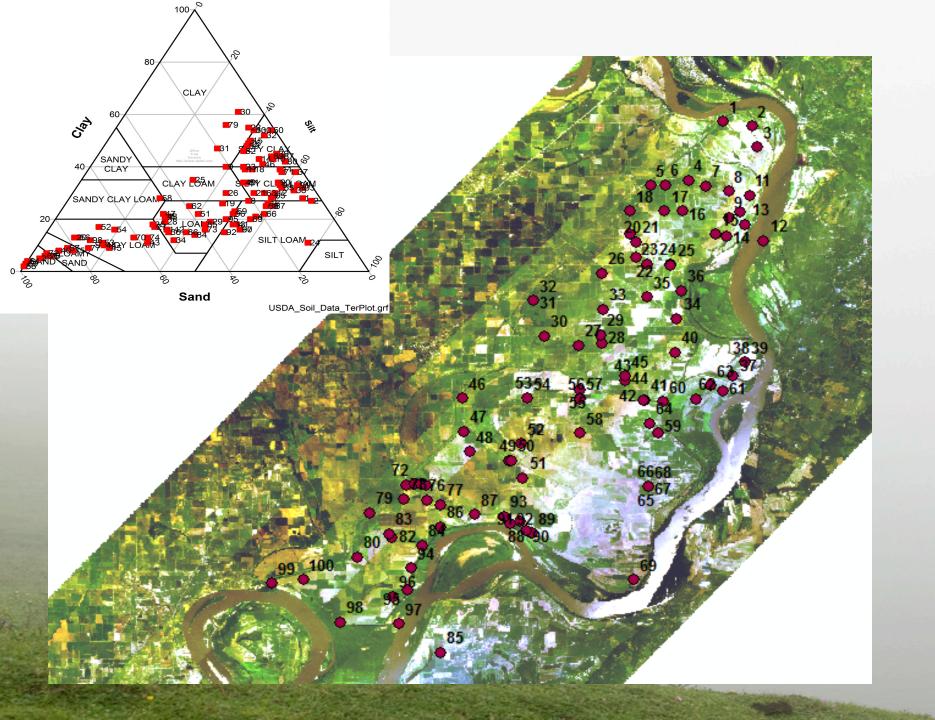
Soil Sample Data

- Soil grab sample from the floodway from about 130 locations were collected around the same date of the AVIRIS flights (27th July 2011).
- Results from the textural and chemical analysis of the samples are available.
- Physical textural analysis results include the percent sand, silt, clay, moisture content etc.
- Chemical analysis results include pH, SOM, S, P, Na, Ca, Mg, K, B, Mn, Cu,
 Zn, Al, other bases, base saturation and soluble salts.

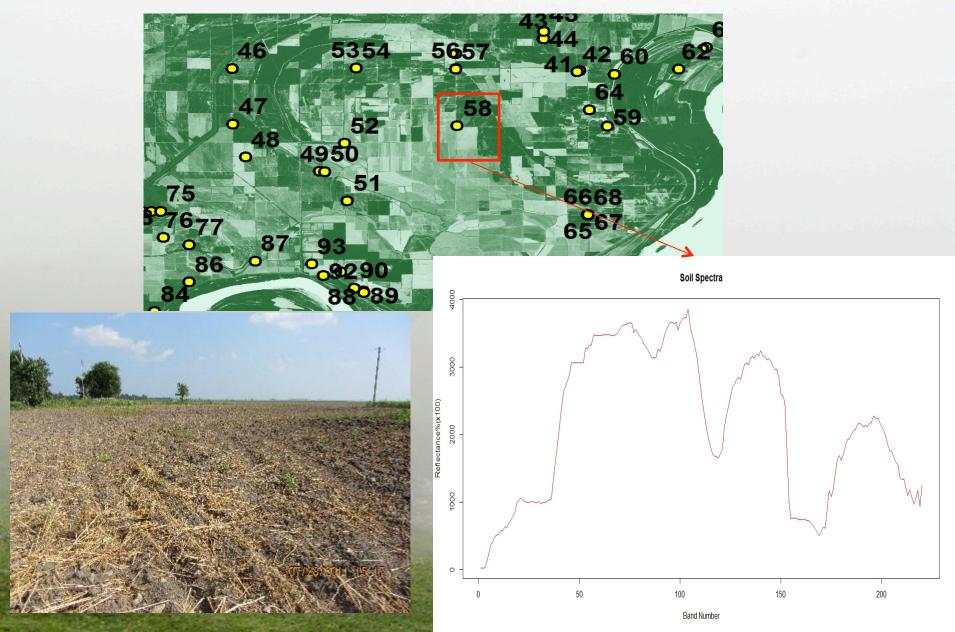








Extracting the Soil Spectra from AVIRIS Data



$$Percent_{sand/silt/clay} = A_0 + A(\alpha_1 f_1 + \alpha_2 f_2 + \dots + \alpha_n f_n)$$

$$A_0 = Intercept$$

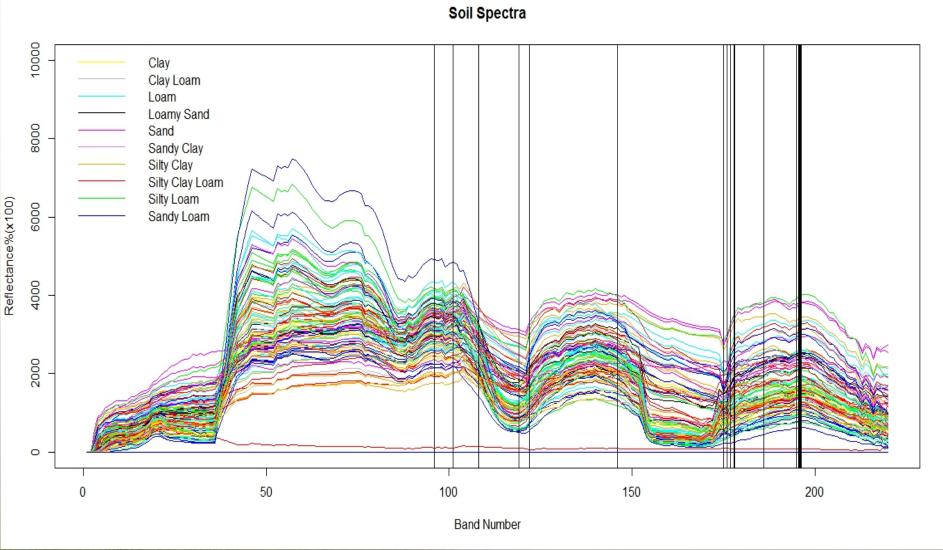
$$\alpha_k = \frac{a_k}{A}$$

$$A = \sum_{i=1}^{k} |a_k|$$

Value of Alpha k	Sand	Silt	Clay
(Intercept)	0.104	0.507	0.414
band.96	0.001	0.024	-0.028
band.101	0.006	-0.031	0.022
band.108	-0.003	0.011	-0.003
band.119	0.037	-0.042	-0.031
band.122	-0.096	0.103	0.072
band.146	0.014	-0.031	0.008
band.175	-0.013	-0.005	0.030
band.176	0.033	-0.010	-0.051
band.177	-0.138	0.112	0.139
band.178	0.134	-0.117	-0.124
band.186	0.005	-0.024	0.022
band.195	0.270	-0.240	-0.262
band.196	-0.252	0.251	0.207

Only One Set of Equations shown here say for Sand, we can similarly write it for Silt and Clay

There are two distinct set of bands which emerge clearly out of the analysis showing strong gradient type influences and which are found to be physically meaningful also

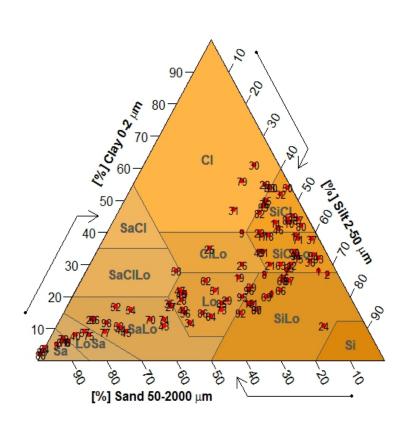


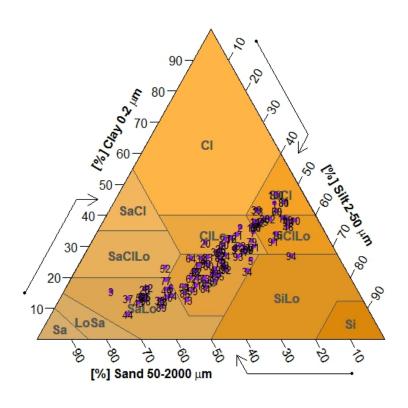
The Band Positions of the selected variables shown with a relative weighting of the coefficients, the bands are representative of some specific signature found in minerals

Observed and Predicted Classes

Observed Soil Texture Data

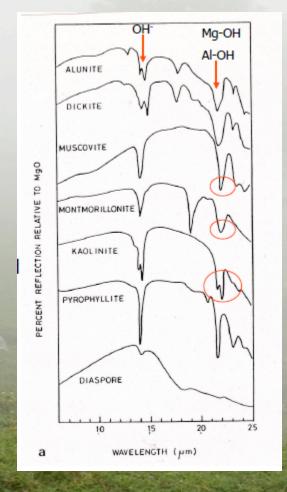
Predicted Soil Texture Data

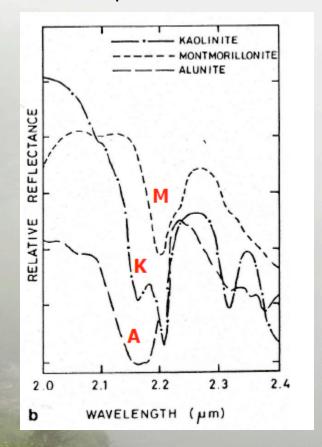




Physical Significance of Results

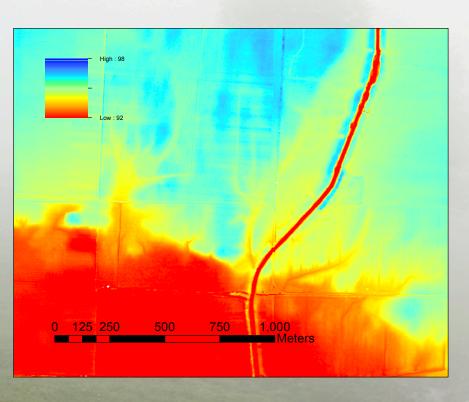
- Silt-granular between sand and clay and mineral origin is quartz and feldspar
- Clay-minerals are kaolin group(kaolinite, dickite etc), montmorillonite, smectite etc
- Sand-most common constituent is silica and quartz.



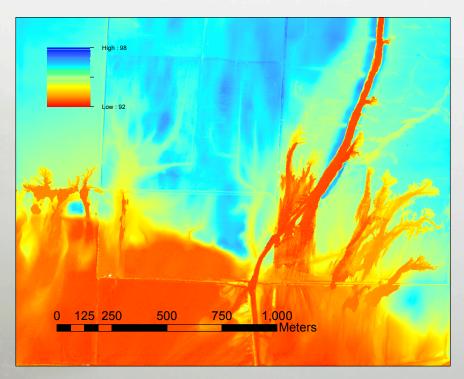


Spectral
Features of clay
in SWIR region

Modeling Study

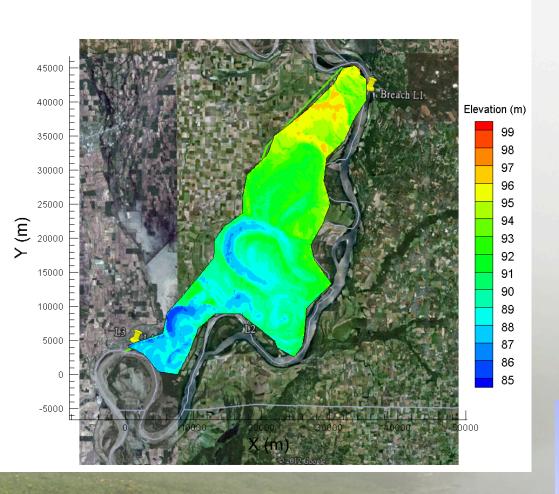


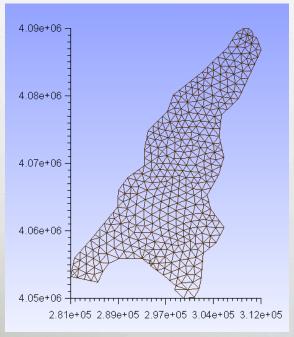
O'Bryan Ridge 2005 Lidar DEM (blue indicates higher elevation)



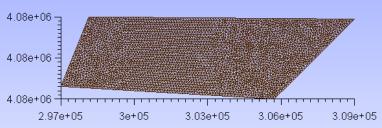
O'Bryan Ridge 2011 Lidar DEM: scours up to 600 m long, 200 m wide, 3 m deep

Modeling Landscape Changes



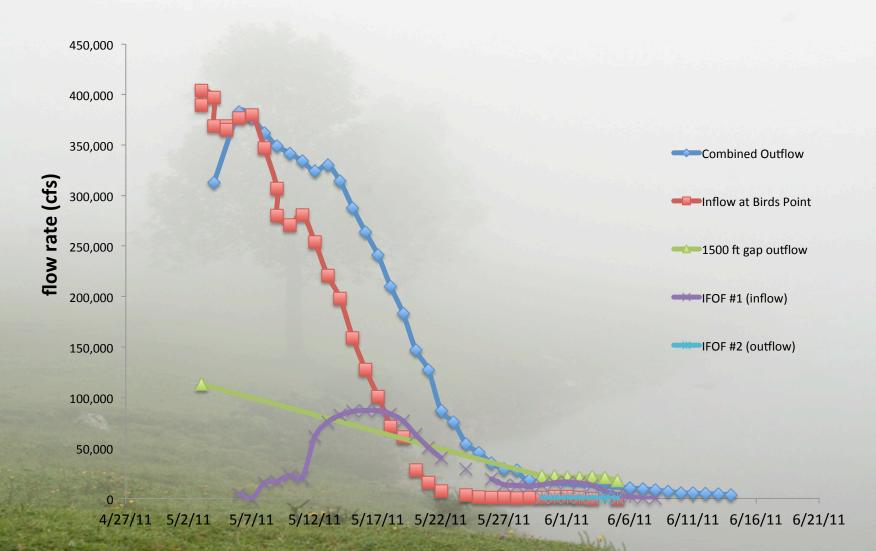


BPNM 1000 m mesh

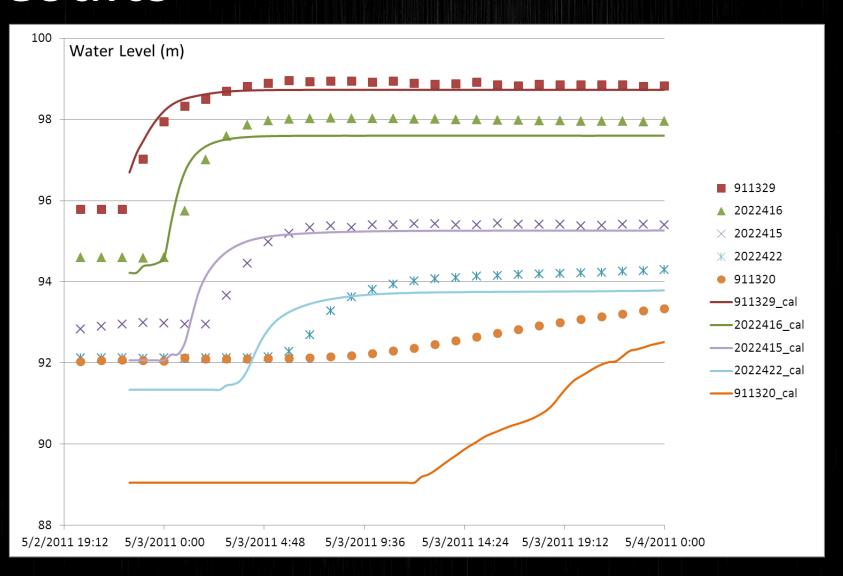


O'Bryan Ridge 100 m mesh

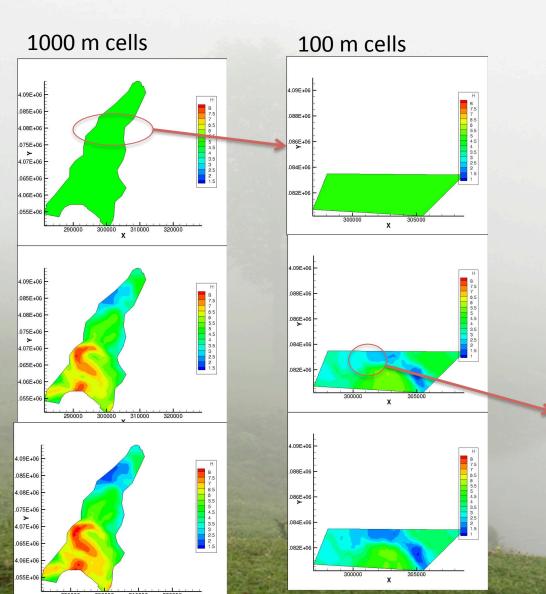
Flow Data from USGS

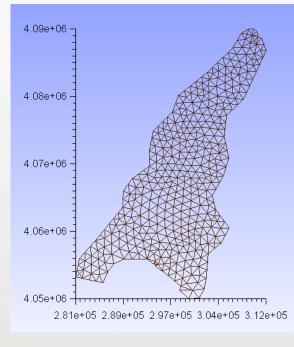


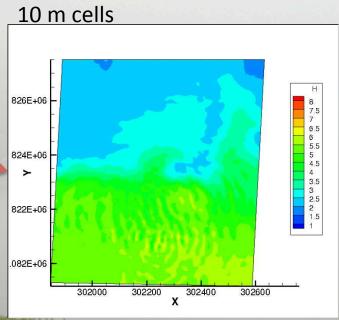
Results

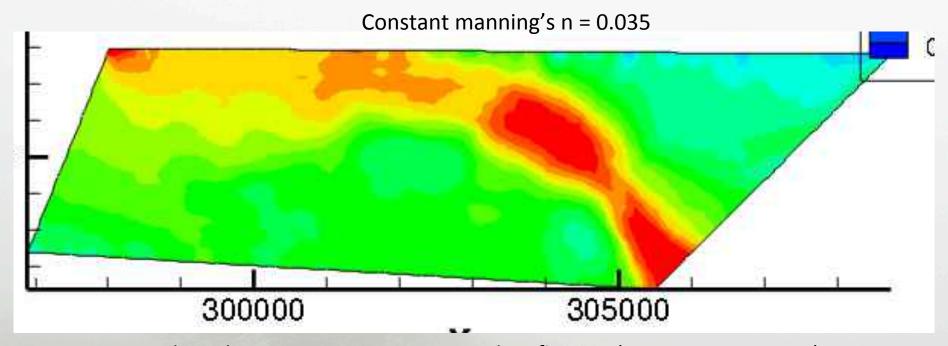


BPNM – OBryan Ridge Nesting

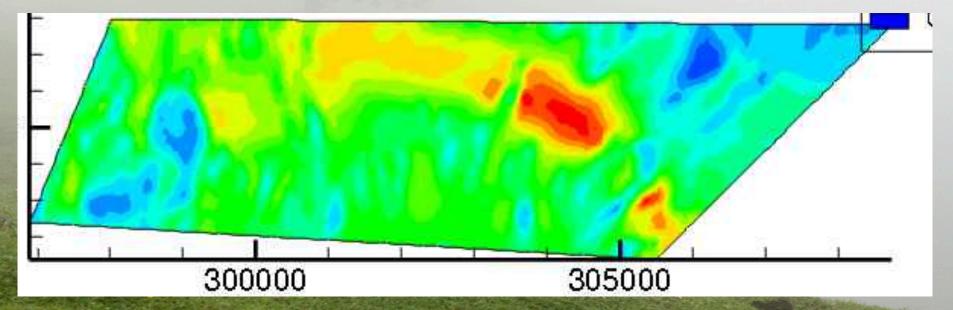


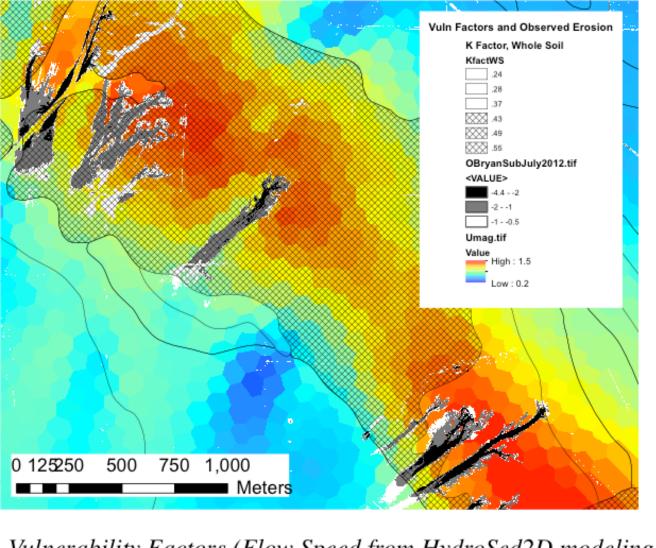


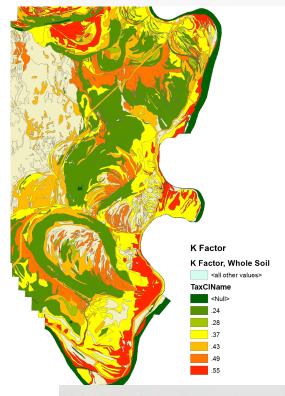




Mannings N based on AVIRIS tree vs no-tree classification (max = .2 min = .035)







Soil Erosion Factor Map from Web Soil Survey

Vulnerability Factors (Flow Speed from HydroSed2D modeling and K Factor from USDA Web Soil Survey) overlaid by observed erosion at O'Bryan Ridge. Regions of high vulnerability are expected to have high exposure (high flow), high sensitivity, and low adaptive capacity (not shown). Observed erosion generally matches well with highly vulnerable areas.

Conclusions

- Accurate identification of land cover classes and more precise vegetation species and soil types is possible. This helps in modeling flow over the landscape accurately.
- Some of the historic meanders of Mississippi were highlighted in indices and classifications from the AVIRIS data showing evolutionary history between topography and vegetation dynamics.
- Simple statistical models can be used to model and interpret soil properties
- Sophisticated statistical models may be able be used for the exploration and study of various physical and chemical properties using imaging spectroscopy data.

Thanks!



Questions?