Getting the jump on invasives: Considerations during habitat management and restoration

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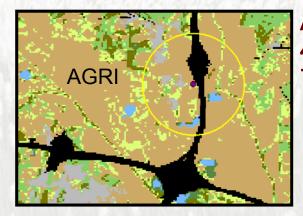
Getting the jump on invasives

Invasion and disturbance
Predicting susceptible areas
An example with cogongrass
Interpreting results

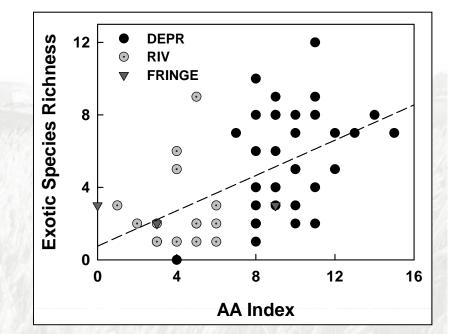
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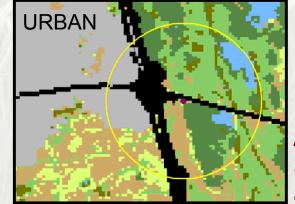
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Human disturbance and exotic species in wetlands



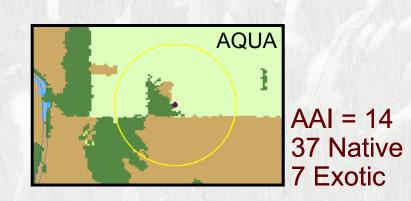
AAI = 13 41 Native 7 Exotic



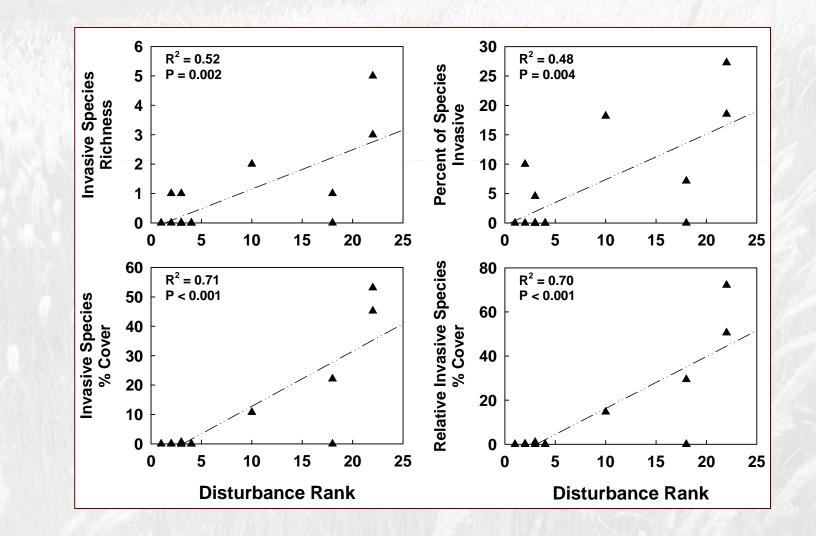


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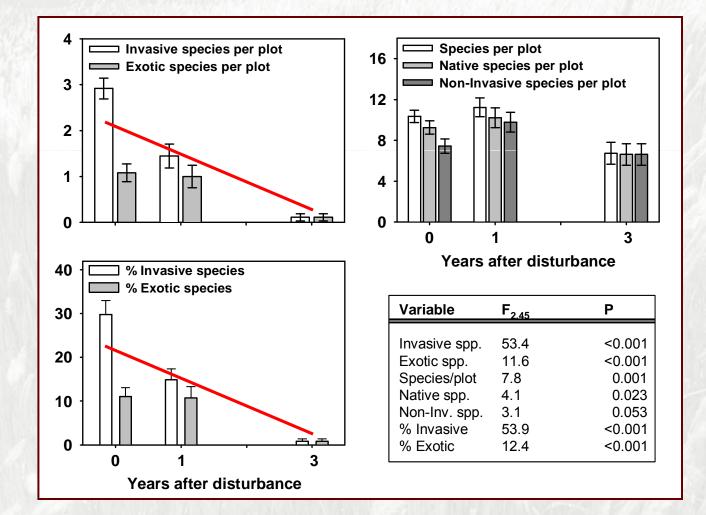
Ervin et al., 2006. Wetlands.



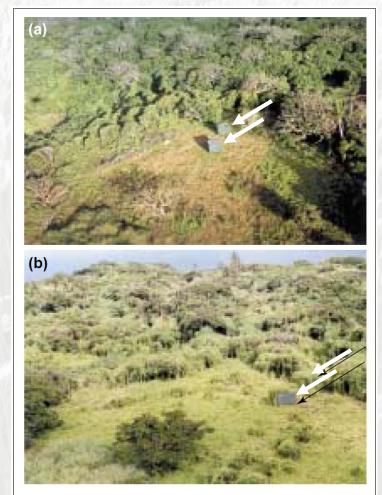
Human disturbance and exotic species in wetlands



Human disturbance and exotic species in wetlands



An example of invasion after restoration -Release of an exotic vine after feral goat and pig removal, Mariana Islands



TRENDS in Ecology & Evolution

Zavaleta et al., 2001

"Spread by Cogongrass...has doubtless been due to wind-blown seed and stolons transported along highways by road machinery" (Tabor 1952).





John D. Byrd, Mississippi State University, Bugwood.org

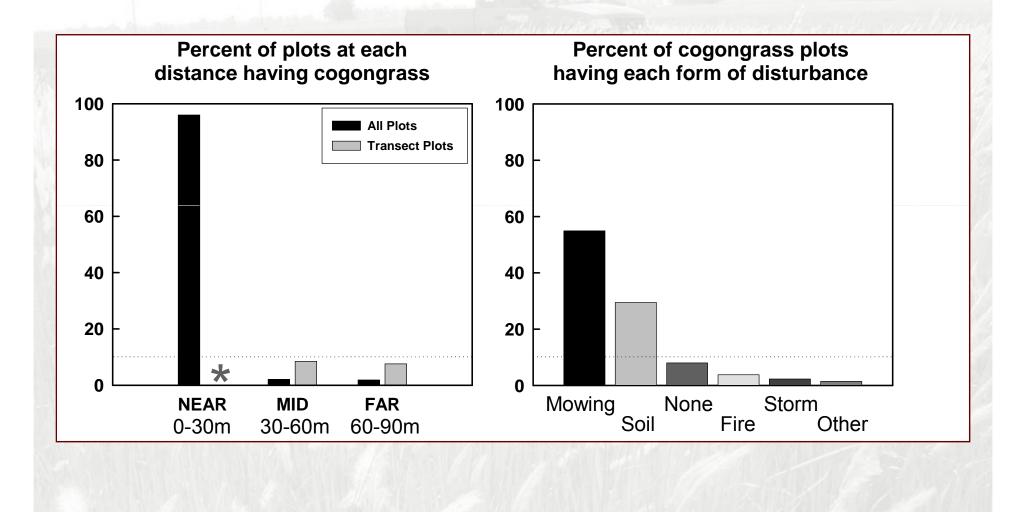


Mark Atwater, Weed Control Unlimited, Inc. Bugwood.org

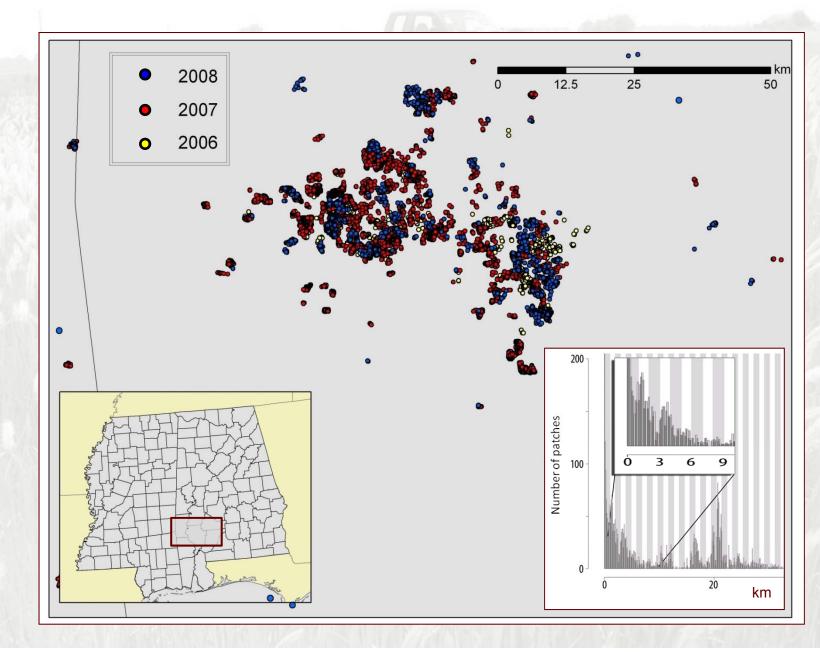


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Cogongrass and disturbance



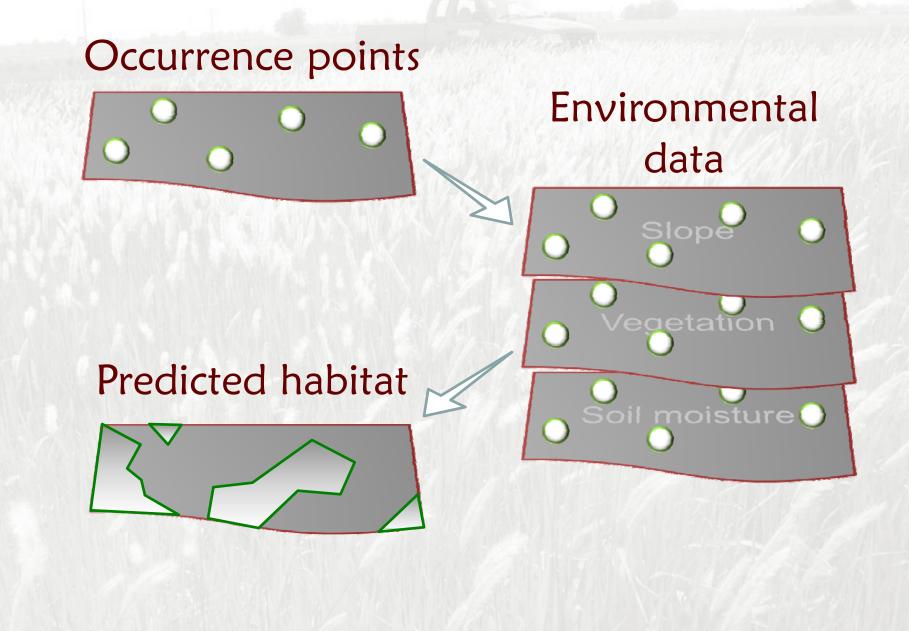
Cogongrass patch density



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Habitat modeling - General approach

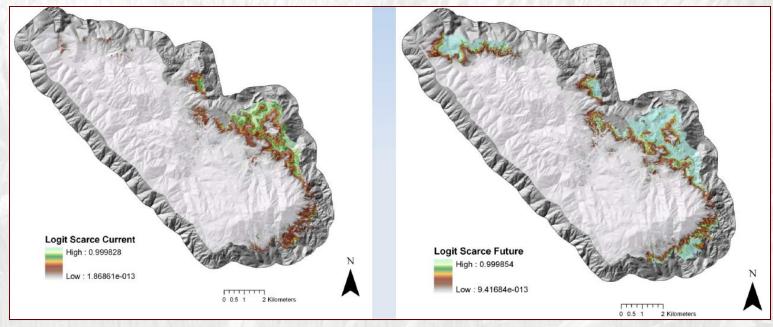


Application of habitat modeling for rare/threatened species

Rare broadleaf trees in Utah Zimmermann et al. 2007 Endangered *Eryngium* in Switzerland Engler et al. 2004 Monarch butterflies Oberhauser & Peterson 2003 Application of habitat modeling for invasive species

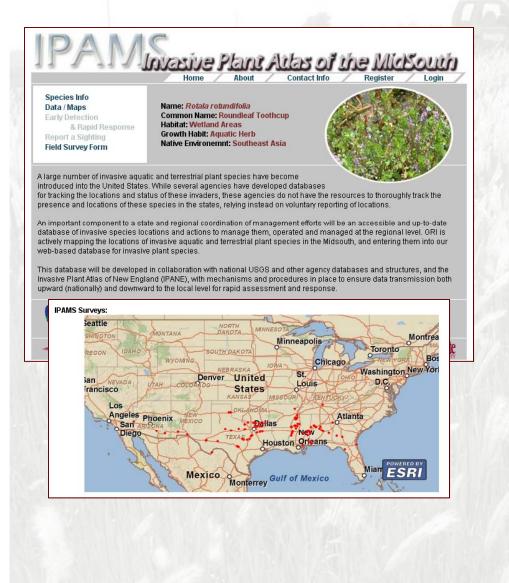
Eurasian watermilfoil in Wisconsin Buchan & Padilla 2000 Purple loosestrife in North America Welk 2004 Invasive plants across North America Peterson et al. 2003

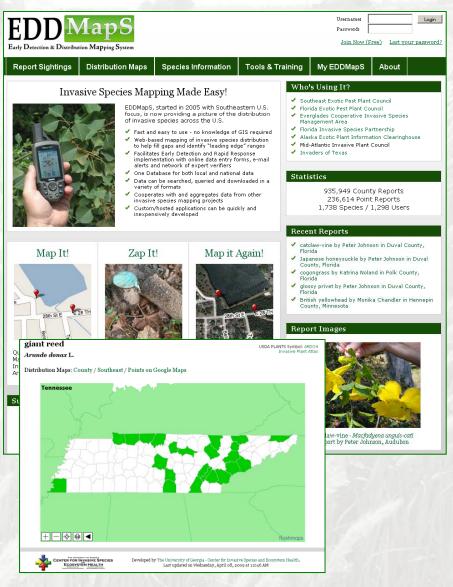
Example Modeling of *Genista monspessulana* spread in association with prescribed burning, Marin County, CA



Hollander and DiPietro, 2010

Databases for invasive species





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Imperata cylindrica - cogongrass

First introduced through the port of Mobile Bay, AL during early 20th century

One of worlds "Ten Worst Weeds"

Infests between 500,000 to 1 million acres of land in MS, AL, and FL

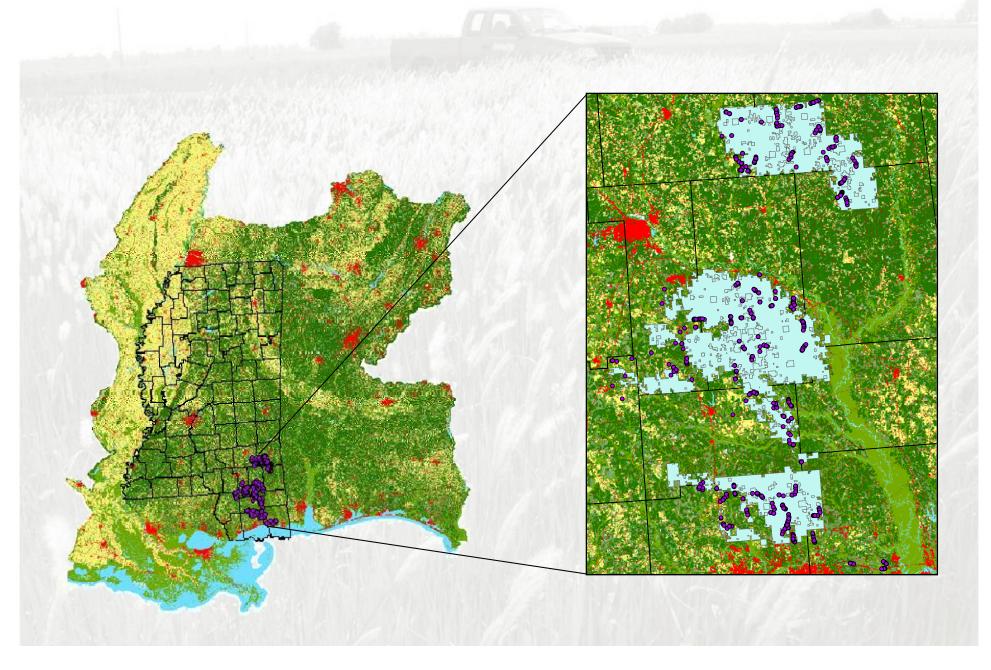
Causes significant economic costs for land managers

Threatens native biodiversity and ecosystem function

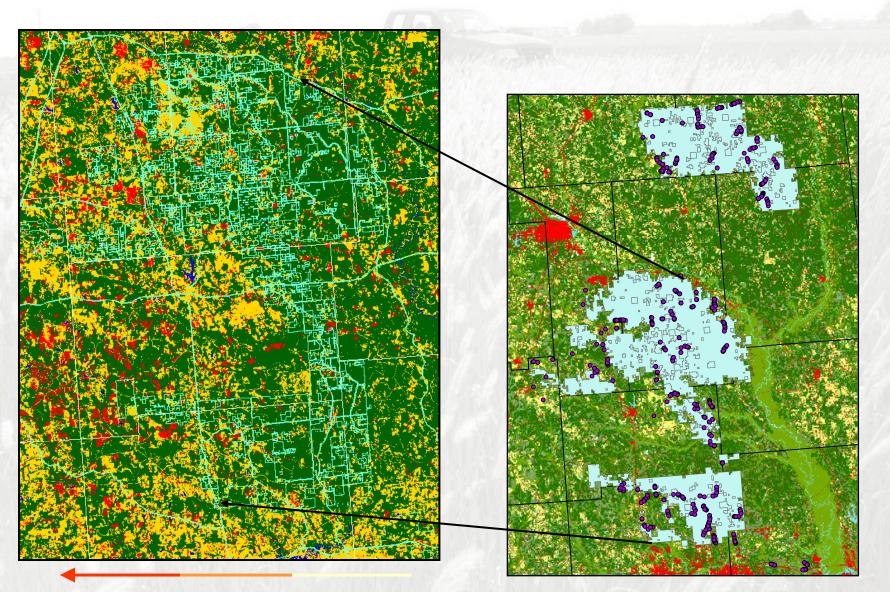
Completely alters native south MS *Pinus palustris* fire regimes



Study area

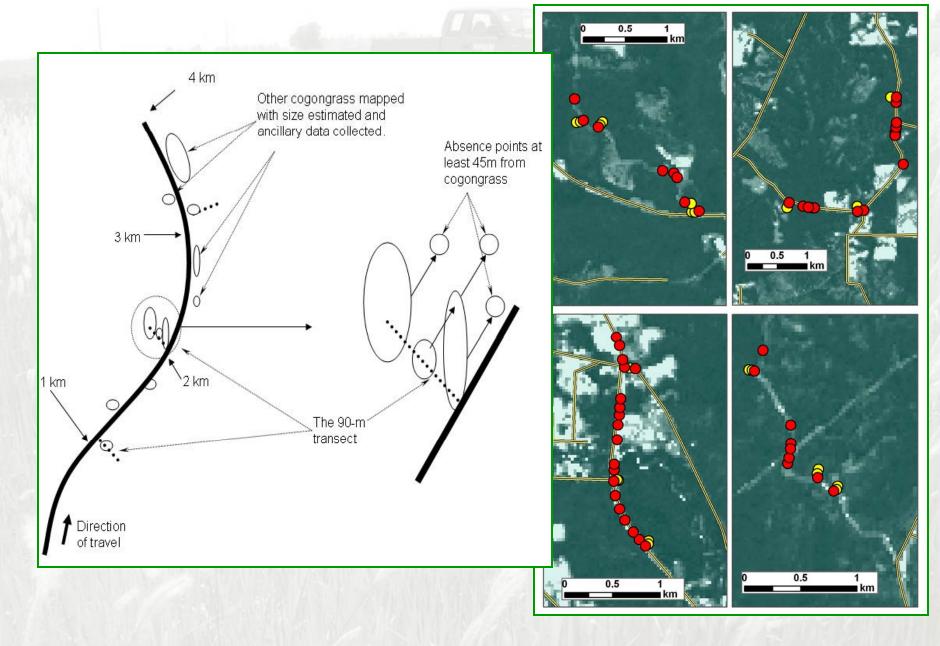


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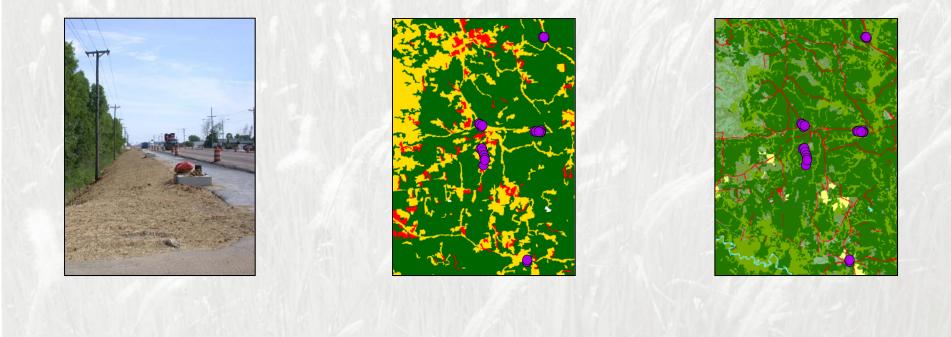
Increasing deforestation 2004-06

Surveys



Candidate predictor variables of invasion

Disturbance	Proximity to Road	Hurricane Associated Deforestation	Forest Community	Abiotic	
Fire	Far 60 - 90m	Non-Forest	Evergreen	Sand (%)	
Mow	Mid 30 - 60m	Forest	Mixed	Organic Matter (%)	
Soil	Near 0 - 30m	Changed	Developed (absent)	pH	
Storm		(2004-2006)		Canopy Cover (%)	
None					



Results

Mo del Step	Model Effect	Wald Chi-Square	P-value	Nagelkerke R ²	SC	AUC
<i>Eff</i>	ects Removed	_				
)	Global Model	45.0254	<.0001	0.6392	222.623	0.894
l	Forest Type	0.1187	0.9424	0.6388	212.148	0.895
2	OM	0.2003	0.6545	0.6381	207.048	0.894
3	Canopy	0.4112	0.5213	0.6366	202.161	0.893
4	рН	1.4789	0.2239	0.6313	198.362	0.879
5	Deforestation	4.2242	0.1210	0.6165	191.865	0.878
<i>Eff</i>	ects Retained	-				
*	Disturbance	10.1282	0.0383			
*	Prox. RD	17.2936	0.0002			
*	% Sand	4.5930	0.0321			
***	Final Model	43.0960	<.0001	0.6165	191.865	0.878

Modeling Approach

<u>Data</u>

Imperata presence-absence (360 points):

205 presence & 155 absence from six counties in southern Mississippi

Soil (SSURGO geospatial data layers):

available water capacity, bulk density, clay & sand content, effective CEC, organic matter content, hydraulic conductivity, pH (all mapped as "representative value" per mapping unit)

Canopy cover (from MRLC database)

Distance to roads (measured in GIS vs. primary and secondary roads)

Analyses

Correlation analyses among soil parameters to exclude correlated variables

→ resulted in 34 candidate logistic regression models

Accuracy criteria vs. training data

Parameters in model	succ	sens	spec	kappa	TSS		
vs. Training data							
– Canopy	0.69	0.72	0.64	0.36	0.36		
– Canopy – BDens	0.69	0.72	0.64	0.35	0.36		
vs. Validation data							
– Canopy	0.76	0.79	0.72	0.51	0.51		
– Canopy – BDens	0.75	0.77	0.73	0.49	0.50		

Equations used to generate raster layer in GIS

Canopy only model:

Probability of occurrence = $\frac{e^{(-0.021*Canopy + 1.43)}}{1 + e^{(-0.021*Canopy + 1.43)}}$

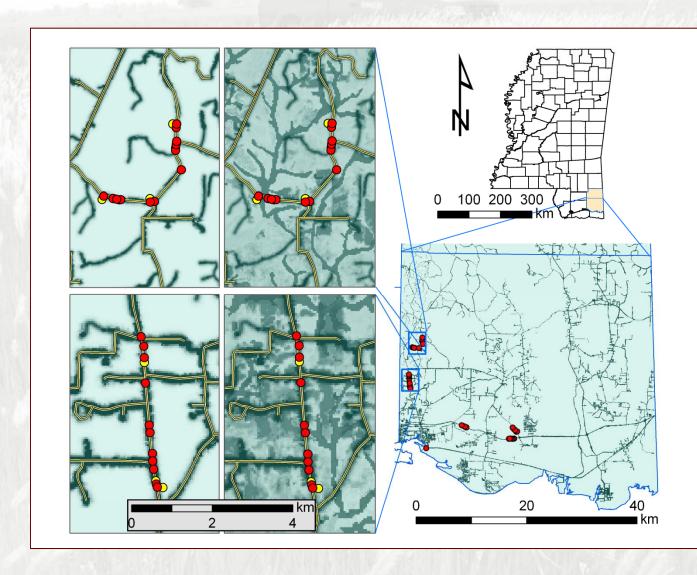
Canopy and bulk density model:

Probability of occurrence =

e (-0.021*Canopy – 2.172*BD + 4.602)

1 + e (-0.021*Canopy - 2.172*BD + 4.602)

Model surfaces



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Interpreting outside a GIS environment

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Predicts \geq 50% probability of suitable habitat at Canopy cover of less than 70%

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- Could be used to set target for canopy density
- Could be used to select areas for monitoring

Questions? **MISSISSIPPI STATE** UNI VERSITY WATER RESOURCES RESEARCH INSTITUTE Science for a changing world NGI INSTITUT Bie INFARTMENT OF INCLUDENCE IN THE INCLUSION