Occupancy Modeling as a Tool to Delineate Invasive Species Distribution in Reservoir Riparian Areas



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03/30/2016

Layout

- Invasive Species
- Occupancy Modeling

- Asian Swamp Eels
- Methods
- Results
- Conclusions



Invasive Species

Reservoirs are more susceptible to the introduction and proliferation of non-native species (Johnson et al. 2008)

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- Threaten Native Species
 - Predation
 - Competition
 - Hybridization
 - Loss of Biodiversity
 - Suite of Indirect Ecological Effects
 Davis (2009)

Occupancy Modeling

- Occupancy estimates can be severely biased when species are rare or elusive (McKenzie et al. 2003)
- Account for imperfect species detection through repeat sampling

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- Assumptions
 - Site Closure
 - No False Positive Detections
 - No unexplained heterogeneity
 - Survey Independence



Asian Swamp Eels

- Native to tropical and temperate climates of southeast Asia, Indonesia, and Australia
- Obligate Air breathers
- Capable of overland migration
- Cryptic
- Sedentary behavior (Likely high site fidelity)
- 4 known populations introduced into the continental U.S.
- Genetic variation within M. albus equivalent to that observed in some families of fish. (Collins et al. 2002)





ASEs in Georgia

- GA population has persisted since approximately 1990 (Freeman and Burgess 2000)
- Known to reside in 2 ponds and adjacent marsh area of the Chattahoochee River (Freeman et al. 2005)
- Possibly an undescribed species due to genetic analysis and observed variation in behavior (Collins et al. 2002)



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Long and LaFleur (2011)

Research Objective

Determine current ASE distribution within the Chattahoochee River







Methods

- 111 Sampling Locations within a 2km radius of known ASE occurrence
 - Sampling locations randomly selected from NHD layer broken into 5 m segments
 - 5 m segments were sampled with a standardized array of leaf litter traps
- Each site visited on 10 occasions from June August 2015
- Sampling-level covariates collected on every occasion
 - Temperature
 - Depth at each leaf litter trap



Methods Cont.

- Site-level covariates collected once
 - Aquatic Vegetation (as proportion of area)
 - Silt substrate (as proportion of area)
- Other site-level metrics
 - Distance from proposed invasion point
 - Mean transect depth over sampling season
 - Mean transect temperature over sampling season
 - Temperature variance over sampling season
- Single season models of detection and occupancy
- Interpolate covariate values for locations not sampled
- Apply model averaged estimates of occupancy to entire study area



Study Area









Leaf Litter Traps

- Unpublished data on LLT's indicated detection probability is low (11%)
- Alternative capture techniques have shown even lower detection probabilities







Results

> 31 ASEs captured at 14 unique transects

Site ID	Detection History	Total # of Detections	
184	0000001000	1	
181	0110010000	3	
180	1110000101	5	
172	000000001	1	
171	0000001000	1	
169	001000000	1	
168	000000001	1	
154	010000000	1	
153	0010100001	3	
152	0100101011	5	
147	010000000	1	
146	010000000	1	
143	1010011010	5	
108	100000000	1	





Detection Models

Model	K	AIC	ΔAIC	AIC Weight
psi(.)p(Temp+Mtd)	4	225.06	0	0.89
psi(.)p(Mtd)	3	229.67	4.61	0.08
psi(.)p(.)	2	233.37	8.31	0.01
psi(.)p(Temp)	3	234.77	9.71	< 0.01

Detection Estimate (at mean Temp+Mtd) = 0.174 SE = 0.039

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Detection Model Fit

- C-Hat = 1.00
- Likelihood Ratio Test
 - psi(.)p(Temp+Mtd) , psi(.)p(.)

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▶ $x^2 = 12.31 \text{ DF} = 2 \text{ P} < 0.01$



Occupancy Model Selection

Pearson Correlation Matrix						
	Dst	Veg	Slt	Msd	Mst	Tvar
Dst	1.00	-0.50	-0.55	-0.01	0.01	-0.39
Veg		1.00	0.53	-0.31	0.51	0.46
Slt			1.00	-0.23	0.05	0.32
Msd				1.00	-0.32	-0.11
Mst					1.00	0.39
Tvar						1.00

Pearson's Correlation Coefficient (<|0.50|) used to eliminate multicollinearity among site-level covariates



Exploratory model selection approach due to lack of prior research

27 Candidate Models

All possible non-collinear combinations

psi(Veg)p(Temp+Mtd) psi(Veg+Tvar)p(Temp+Mtd) psi(Veg+Msd)p(Temp+Mtd) psi(Veg+Msd+Tvar)p(Temp+Mtd) psi(Slt+Msd)p(Temp+Mtd) psi(Slt+Mst)p(Temp+Mtd) psi(Slt)p(Temp+Mtd) psi(Dst+Msd)p(Temp+Mtd) psi(Slt+Msd+Mst)p(Temp+Mtd) psi(Slt+Msd+Tvar)p(Temp+Mtd) psi(Tvar)p(Temp+Mtd) psi(Dst+Msd+Mst)p(Temp+Mtd) psi(Dst+Msd+Mst)p(Temp+Mtd) psi(Slt+Msd+Tvar)p(Temp+Mtd)

psi(Slt+Mst+Tvar)p(Temp+Mtd) psi(Dst+Mst+Tvar)p(Temp+Mtd) psi(Slt+Msd+Mst+Tvar)p(Temp+Mtd) psi(Dst+Msd+Tvar)p(Temp+Mtd) psi(Dst)p(Temp+Mtd) psi(Dst)p(Temp+Mtd) psi(Msd+Tvar)p(Temp+Mtd) psi(Msd)p(Temp+Mtd) psi(Msd)p(Temp+Mtd) psi(Mst+Tvar)p(Temp+Mtd) psi(Msd+Mst+Tvar)p(Temp+Mtd) psi(Msd+Mst)p(Temp+Mtd) psi(Msd+Mst)p(Temp+Mtd) psi(Msd+Mst)p(Temp+Mtd)

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Models <10∆AIC

Model	Κ	AIC	ΔAIC	AIC Weight
psi(Veg)p(Temp+Mtd)	5	201.29	0.00	0.45
psi(Veg+Tvar)p(Temp+Mtd)	6	202.96	1.67	0.19
psi(Veg+Msd)p(Temp+Mtd)	6	203.01	1.72	0.19
psi(Veg+Msd+Tvar)p(Temp+Mtd)	7	204.84	3.55	0.08
psi(Slt+Msd)p(Temp+Mtd)	6	207.97	6.67	0.02
psi(Slt+Mst)p(Temp+Mtd)	6	208.41	7.11	0.01
psi(Slt)p(Temp+Mtd)	5	208.50	7.21	0.01
psi(Dst+Mst)p(Temp+Mtd)	6	209.43	8.14	0.01
psi(Dst+Msd)p(Temp+Mtd)	6	209.75	8.46	0.01
psi(Slt+Msd+Mst)p(Temp+Mtd)	7	209.77	8.48	0.01
psi(Slt+Msd+Tvar)p(Temp+Mtd)	7	209.89	8.60	0.01
psi(Tvar)p(Temp+Mtd)	6	209.99	8.70	0.01

Veg Tvar Msd Slt Mst



Occupancy Model Fit

- MacKenzie-Bailey Goodness-of-Fit test on two sub-global models of occupancy
- Parametric bootstrap with 10,000 simulations
 - Psi(Slt+Msd+Mst+Tvar)p(Temp+Mtd): c-hat = 0.86 p=0.44
 - Psi(Dst+Msd+Mst+Tvar)p(Temp+Mtd): c-hat =0.63 p=0.42

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ASE Distribution

Occupancy Probabilities based on model averages

Warmer Colors = Higher Probability



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Conclusions

- ASE's appear to have a limited distribution within the Chattahoochee river
- Marsh areas have a higher probability of occupancy
- Distribution models can inform future sampling for monitoring, eradication, or suppression efforts



Broader Implications

- Occupancy models provide an informed analysis of species distribution that is robust to imperfect detection
- Detection probability should be incorporated into distribution models when feasible
 - Important for invasive species due to the potential implications of assuming absence
- Able to investigate relationship of covariates to detection and occupancy probabilities
- Occupancy modeling method can be applied to other invasive species and habitats



Acknowledgements

- Advisor Dr. James M. Long
- Technician Colt Holley
- National Park Service
- OSU Graduate Students





Questions?