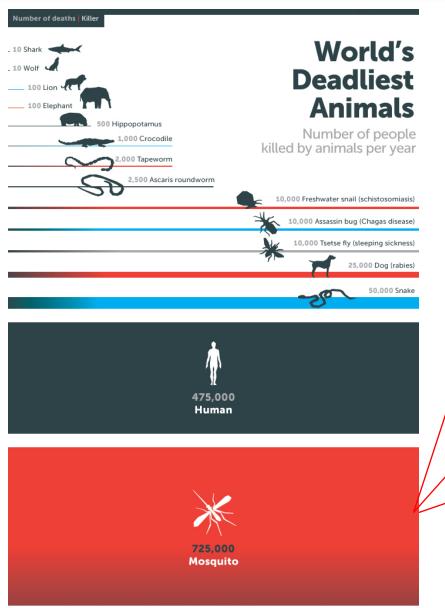




BIG Meeting
June 5th 2018
Sarah M Gunter, PhD, MPH

Mosquito-Borne Diseases



Aedes spp.

- Chikungunya
- Dengue fever
- Lymphatic filariasis
- Rift Valley fever
- Yellow fever
- S Zika

Anopheles

- Malaria
- Lymphatic filariasis

Culex

- Japanese encephalitis
- Lymphatic filariasis
- West Nile fever

SOURCES: WHO; crocodile-attack info; Kasturiratne et al. (doi.org/10.1371/journal.pmed.0050218); FAO (webcitation.org/60gpS8SVO); Lignell et al. (webcitation.org/60RL7DBUO); Packer et al. (doi.org/10.1038%2F436927a); Alessandro De Maddalena. All calculations have wide error major et al. (doi.org/10.1038%2F436927a); Alessandro De Maddalena. All calculations have wide error major et al. (doi.org/10.1038%2F436927a); Alessandro De Maddalena. All calculations have wide error major et al. (doi.org/10.1038%2F436927a); Alessandro De Maddalena. All calculations have wide error major et al. (doi.org/10.1038%2F436927a); Alessandro De Maddalena. All calculations have wide error major et al. (doi.org/10.1038%2F436927a); Alessandro De Maddalena. All calculations have wide error major et al. (doi.org/10.1038%2F436927a); Alessandro De Maddalena. All calculations have wide error major et al. (doi.org/10.1038%2F436927a); Alessandro De Maddalena. All calculations have wide error major et al. (doi.org/10.1038%2F436927a); Alessandro De Maddalena. All calculations have wide error major et al. (doi.org/10.1038%2F436927a); Alessandro De Maddalena. All calculations have wide error major et al. (doi.org/10.1038%2F436927a); Alessandro De Maddalena. All calculations have wide error major et al. (doi.org/10.1038%2F436927a); Alessandro De Maddalena. All calculations have wide error major et al. (doi.org/10.1038%2F436927a); Alessandro De Maddalena. All calculations have wide error major et al. (doi.org/10.1038%2F436927a); Alessandro De Maddalena. All calculations have wide error major et al. (doi.org/10.1038%2F436927a); Alessandro De Maddalena. All calculations have wide error major et al. (doi.org/10.1038%2F436927a); Alessandro De Maddalena. All calculations have wide error major et al. (doi.org/10.1038%2F436927a); Alessandro De Maddalena. All calculations have wide error major et al. (doi.org/10.1038%2F436927a); All calculations have wide error major et al. (doi.org/10.1038%2F436927a); All calculations have wide error major et al. (doi.org

Baylor College of Medicine

Major Limitations of Mosquito Borne Disease Prevention

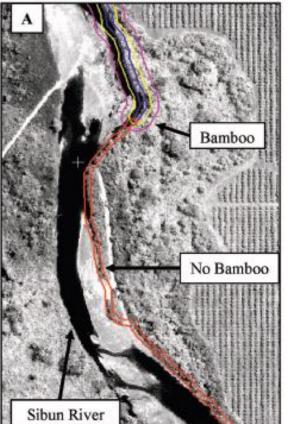
- 1. Majority of these diseases originate in infrastructure-poor, resource-limited countries
 - I. Hard to predict spread of new Mosquito-Borne Diseases
 - a. Arboviral mutations
 - Unpredictable jump to new mosquito species-animal hosts
 - b. Lack of surveillance
 - i. Can't identify new epidemics
 - ii. Can't track spread
 - iii. Unaware of highest-risk populations
- 2. Globalization contributes to spread of disease
- 3. Paucity of available diagnostics, vaccines, and therapeutics

Development of BCM-ExxonMobil Collaboration

MODELING/GIS, RISK ASSESSMENT, ECONOMIC IMPACT

Use of Remote Sensing and Geographic Information Systems to Predict Locations of *Anopheles darlingi*-Positive Breeding Sites Within the Sibun River in Belize, Central America

NICOLE L. ACHEE, 1 JOHN P. GRIECO, 1 PENNY MASUOKA, 1 RICHARD G. ANDRE, 1 DONALD R. ROBERTS, 1 JAMES THOMAS, 1 IRENEO BRICENO, 2 RUSSELL KING, 2 AND ELISKA REJMANKOVA3



J. Med. Entomol. 43(2): 382–392 (2006)

ation of the confusion matrix indicated a 75.9% accuracy rate by which all land cover categories were classified. Bare ground, forest, and pasture/low grass land cover categories had the highest accuracy rates with 98.8, 97.0, and 94.9% of the pixels being correctly classified, respectively. The orchard and sandbar land cover classes suffered from the worse classification confusion, with 58.8 and 58.6% of the pixels, respec-

Collaborative Project Goals

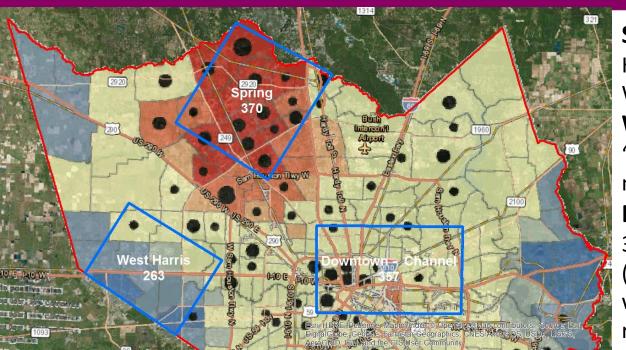


1) Develop a image analysis workflow that can identify mosquito breeding habitats

Evaluate efficacy of our model with real-world validation

3) Determine public health impact with arboviral surveillance

Project Overview



Spring– 370 sq.km.:

High WNV + mosquito & High WNV+ human incidence 2014

West Harris – 263 sq.km.: "control" area, Low WNV+

mosquitos & human cases

Downtown/Ship Channel—

357 sq.km.: Mixed use areas (industrial & residential) which should provide a widest range of habitats



Satellite Imaging Provider Selection

Satellite	Pixel Size (m)	# pixels that fit into a single Landsat-8 pixel	Number of Bands		Heri	mann Pa	K (e)
WorldView-3	0.31	2341.3	16			1 / YO	
WorldView-2	0.46	1063.3	8				(1)
QuickBird	0.65	532.5	4				
SPOT-6	1.50	100.0	4	*		47	
Sentinel-2	10.00	2.3	13	W. W.			18 1 1 1 1
Landsat-8	15.00	1.0	11				0 25 50 100 150
WorldView Panchromat Multispectra	ic						
	400	500 60	0 70	0 80	00	900	1000

Image Analysis Workflows



Visual Inspection:

- Abandoned tires: Look for 'dark pixels" using automated classification refined by visual inspection of images and spectral readings
- OSSF: look for clustering of permitted systems

Color Band Ratios:

- Normalized difference vegetation index (NDVI) to find areas with a high density of healthy vegetation
- Normalized difference water index (NDWI) to find areas with standing water

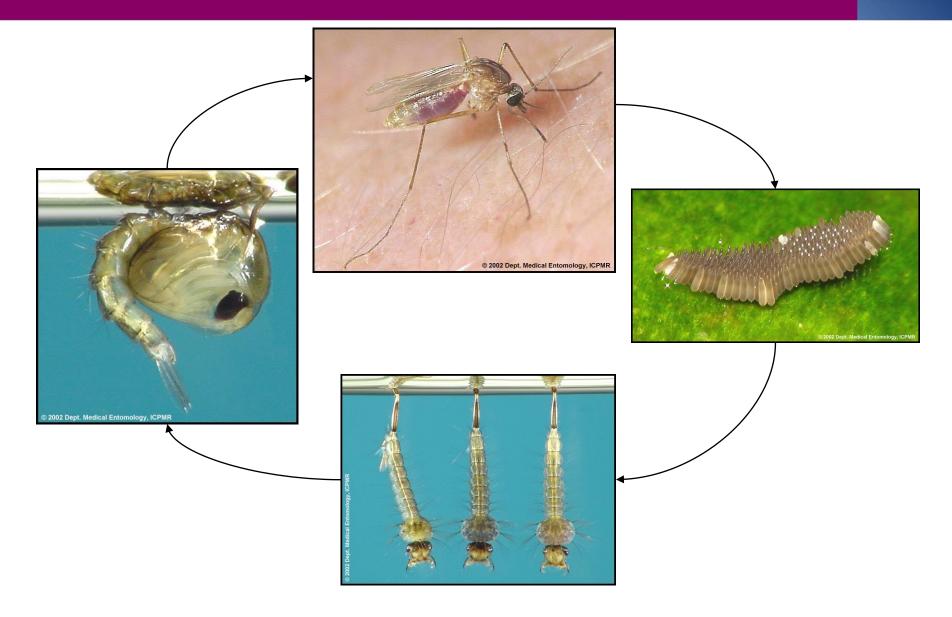
Image Classification "object oriented":

 If we know where good habitats for mosquito growth exist, we can use pixels from specific components of those habitats to predict where similar pixels exist

LIDAR

Find roadside ditches and classify by depth

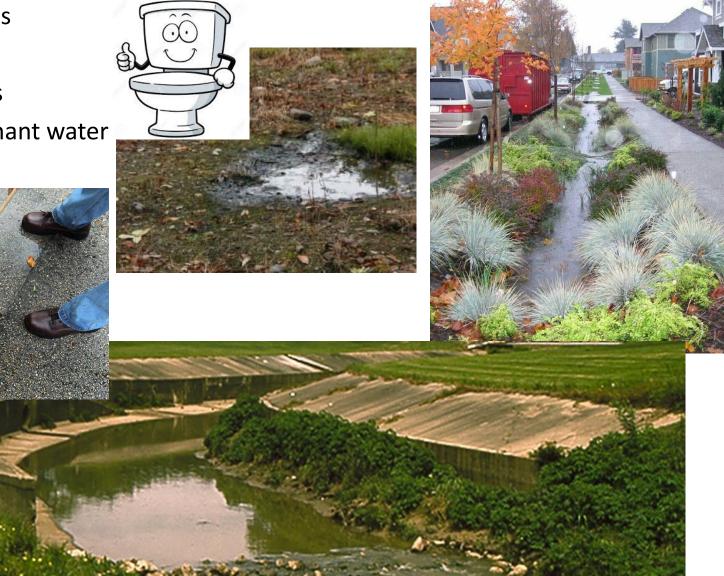
Mosquito Life Cycle



Mosquito Breeding Habitats & Model Identification Plan

Culex quinquefasciatus

- 1. Drainage ditches
- 2. Septic leaks
- 3. Manhole covers
- 4. Vegetated stagnant water



Mosquito Breeding Habitats & Model Identification Plan

Aedes aegypti & A. albopictus

- 1. Tire grouping-ASDI HandHeld2 spectroradiometer
- 2. Trash/container index (junk)
- 3. Construction sites- *master plan communities*
- 4. Industrial yards
- 5. Cemeteries







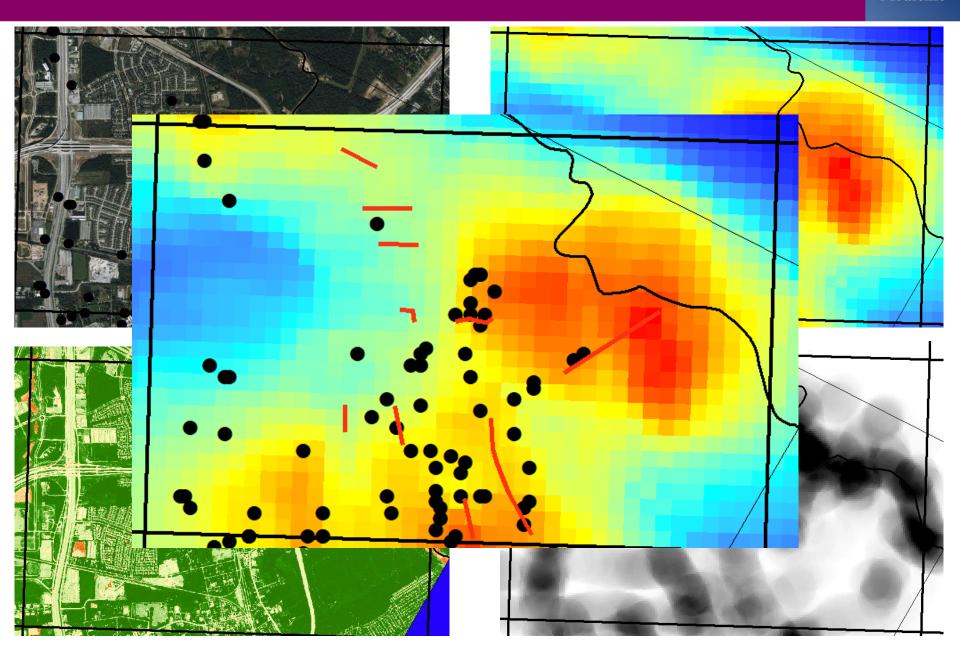


Public Health Relevance

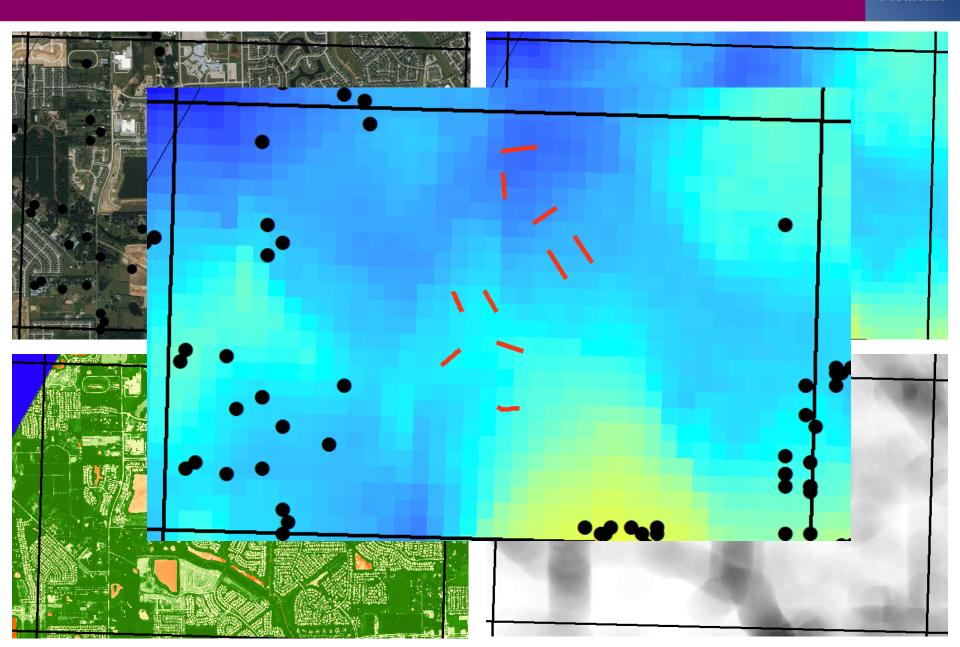
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Trap Locations



Trap Locations



Future Directions

- Refinement of Mosquito Breeding Habitat Model
 - Integration of Dog Detection as a Validation Measure
 - Artificial Intelligence (Neural network analysis), LiDAR data, Texture filters
- Habitat Prediction Models and Potential Applications
 - Afghanistan/Iraq Sandfly (Leishmaniasis)-DoD
 - Africa Anopheles sp. Mosquito (Malaria)-Gates

• Integrated Vector Management for *Aedes, Culex,* and *Ixodes* (Zika, Dengue, Chikungunya, West Nile, and St. Louis Encephali

viruses, and Lyme disease)-NIH









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Study Team

- ExxonMobil Upstream Division
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