Modeling ballast water management strategies for slowing the secondary spread of aquatic invasive species on the Laurentian Great Lakes

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Network centrality as a potential method for prioritizing ports for aquatic invasive species surveillance and response in the Laurentian Great LakesJT Kvistad, WL Chadderton, JM Bossenbroek Management of Biological Invasions 10 (3), 403-427

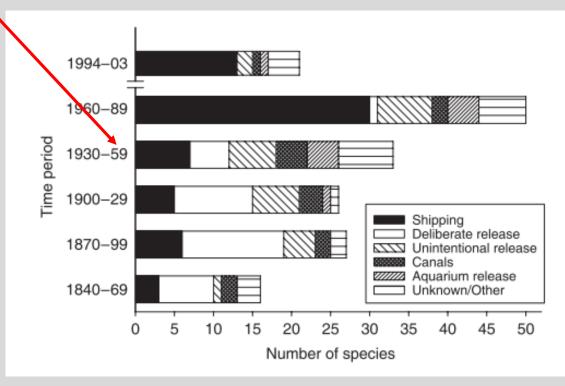


<u>Acknowledgements</u>

- Dr. Patrick Kocovsky USGS
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 Funding- Michigan Dept. of Natural Resources Invasive Species Grant Program

Ballast water is the dominant vector for species introductions and spread in the Great Lakes St. Lawrence Seaway

- 182 recognized non-native species opens the Great Lakes
- ~ 2/3 of invasive species in Great Lakes introduced through ballast



Ricciardi (2006)

Some of our worst invaders were introduced through ballast water



nas.usgs.gov

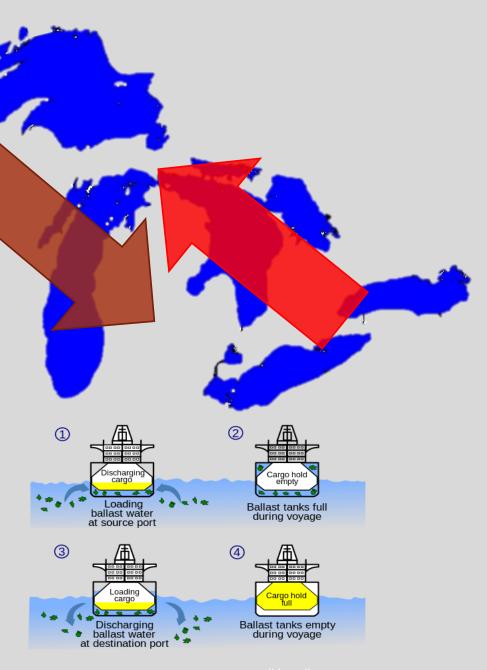


newscientist.com



Secondary spread

- Large domestic shipping network- "Lakers"
- Ships empty of cargo travel upstream- carrying ballast water
- Cargo transported downstream- ballast released upon loading cargo
- Net ballast transport upstream



Lakers exempt from most ballast water regulations set by IMO

Short intraregional trips

- Old ships- up to 60 years in some cases
- Exclusively freshwater- ballast tanks not lined with anticorrosive paint
- 15,000 30,000 m³ ballast capacity

How can risk be minimized?



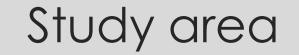
Shoreside ballast water management

- Ships deballast at ballast reception facilities
- 'Clean' ballast can be loaded, 'dirty' ballast can be treated
- Significant investment in construction
- So which ports should construct these facilities?



Goals/Objectives

- Goal: identify ports in the Great Lakes shipping network where hypothetical shoreside de-ballasting could be implemented to slow the secondary spread of potential aquatic invasive species (AIS).
- Objectives:
 - 1. Construct stochastic model of secondary spread by ballast water
 - 2. Predict secondary spread of potential AIS
 - 3. Quantify importance of ports in Great Lakes network
 - 4. Simulate management scenarios





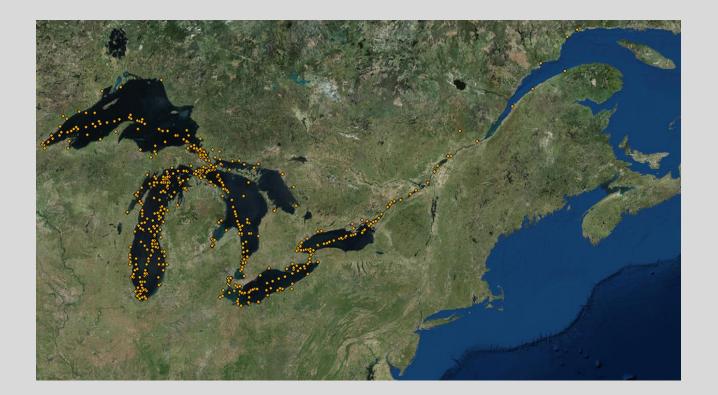
Data description

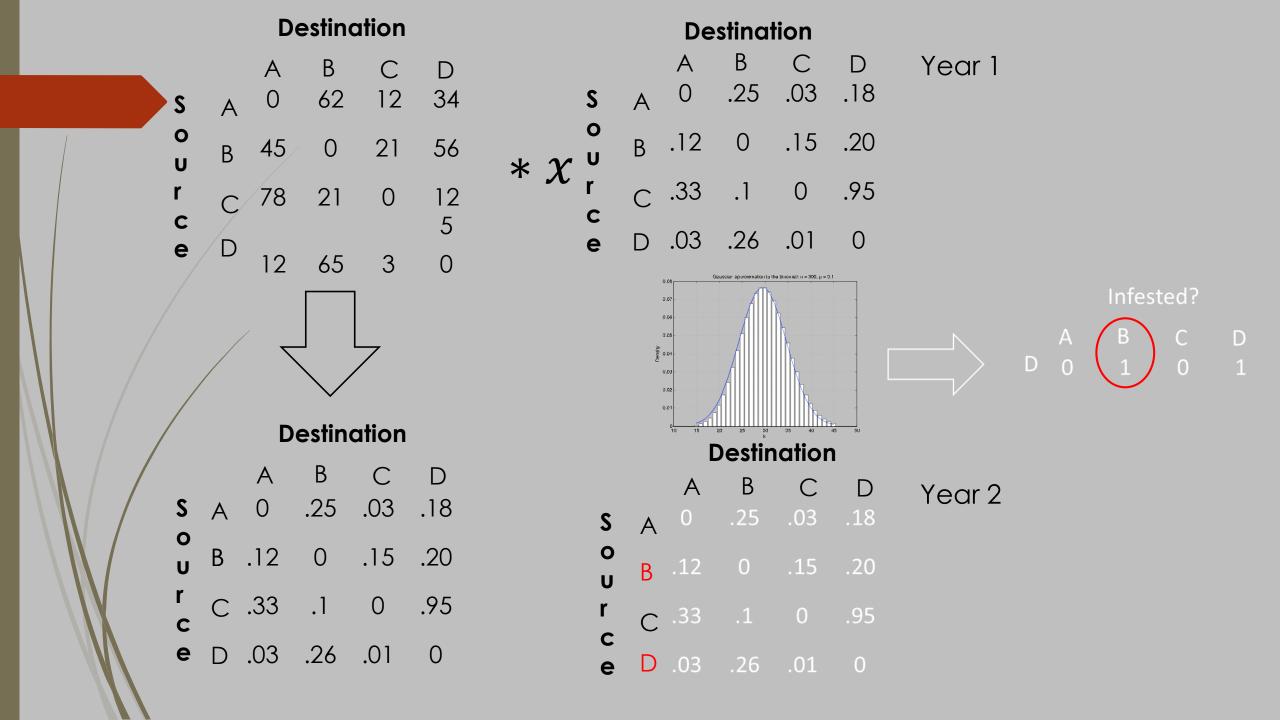
- Public records from National Ballast Information Clearinghouse (NBIC)
 - **2005-2015**
 - Contains detailed records of ballast activity
- Dataset from Fisheries and Oceans Canada
 - **2**005-2007
 - Ship movements
 - Matrix of average ballast water movements between port pairs



Canada







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Golden mussel (Limnoperna fortunei)

- Mytilid mussel native to SE Asia
 - ► ~ 25 30mm
- Invasive in South America
- Similar consequences as dreissenids in North America
- Sieracki et al. (2014) predicted Bay City, MI as potential initial infestation



Monkey goby (Neogobius fluviatilis)

- Ponto-Caspian species
- Invasive in parts of Western Europe
- Likely to invade Great Lakes (Stepien and Tumeo, 2006)
- Initial infestation chosen at ports along St. Clair-Detroit Rivers



Needed to parameterize model to accurately reflect likely spread patterns

- Existing examples that could be used?
 - Zebra mussel
 - Round goby
- Once parameterized, realistic scenarios could be forecast

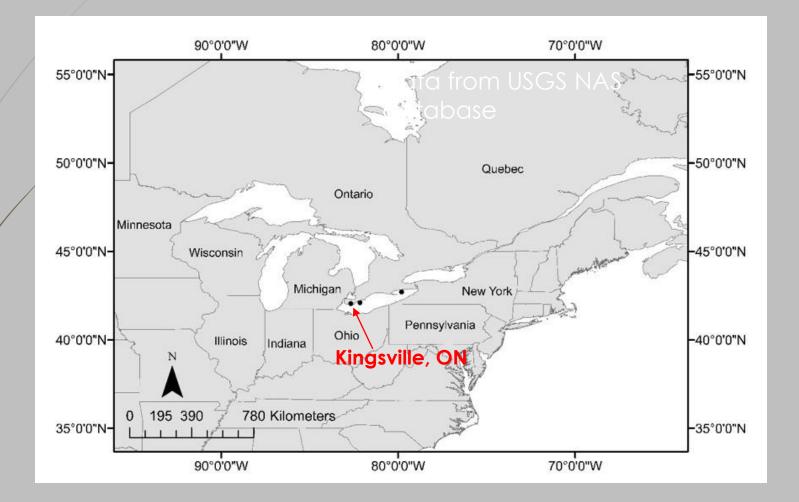


Zebra mussel (Dreissena polymorpha)

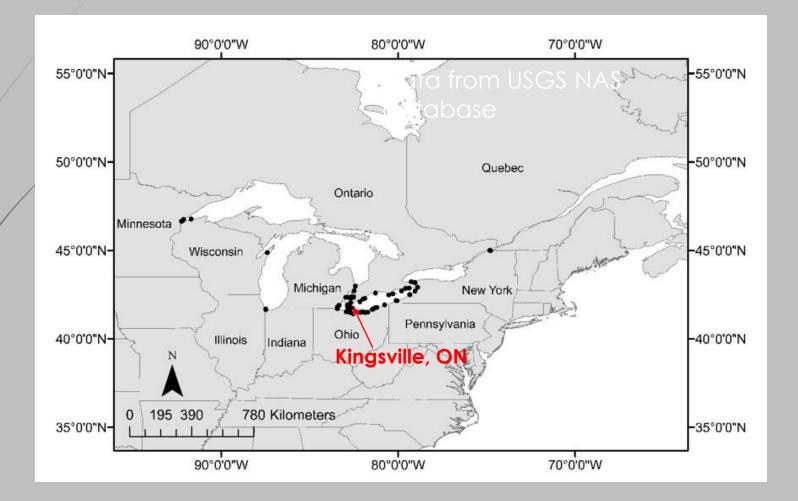
- Introduced in ballast by ocean-going freighters from Black Sea
- Enormous ecological and economic damage
- Rapid secondary spread in the Great Lakes primarily through ballast water



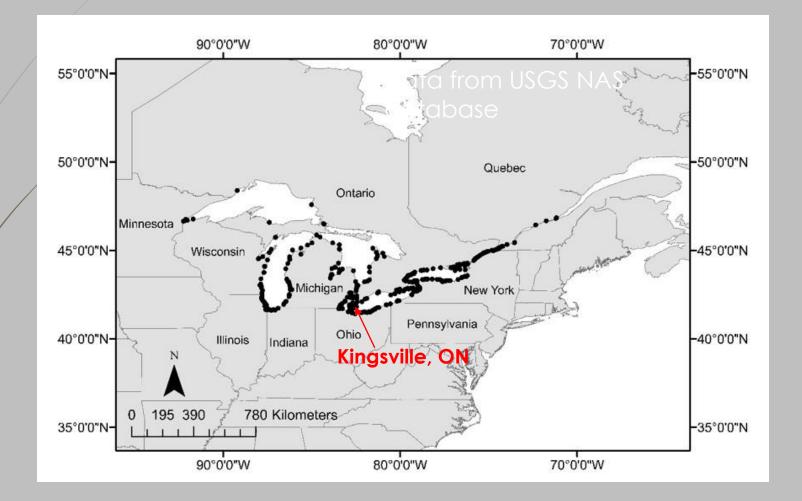
Zebra mussel 1987



Zebra mussel 1989



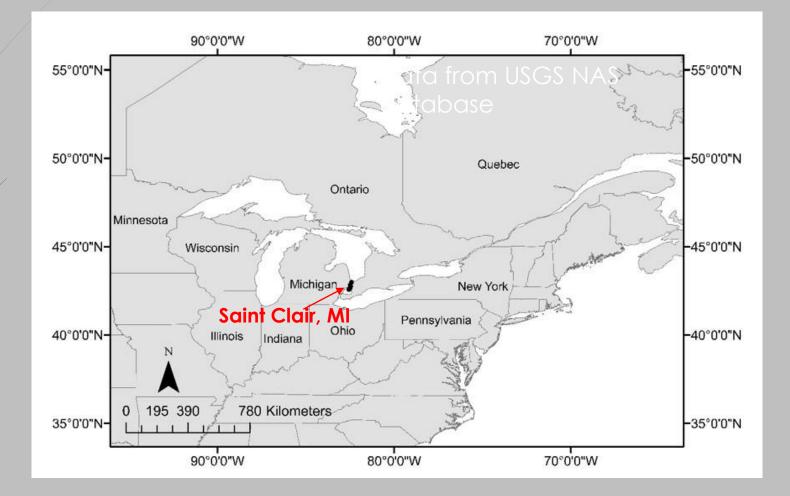
Zebra mussel 1992

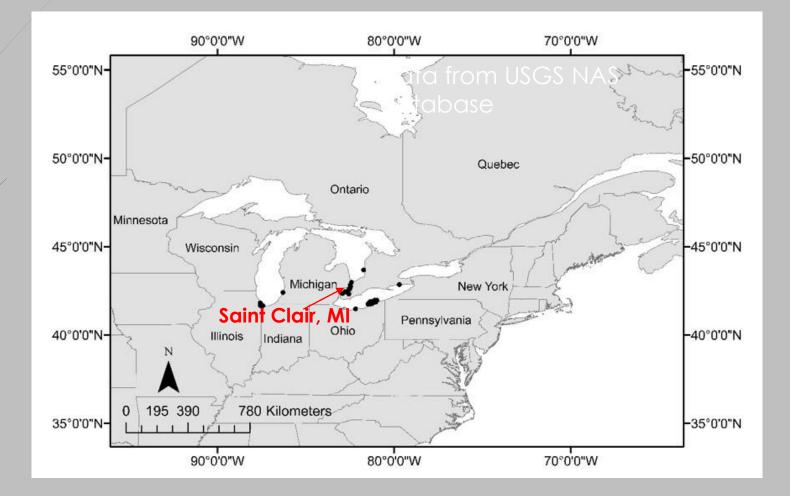


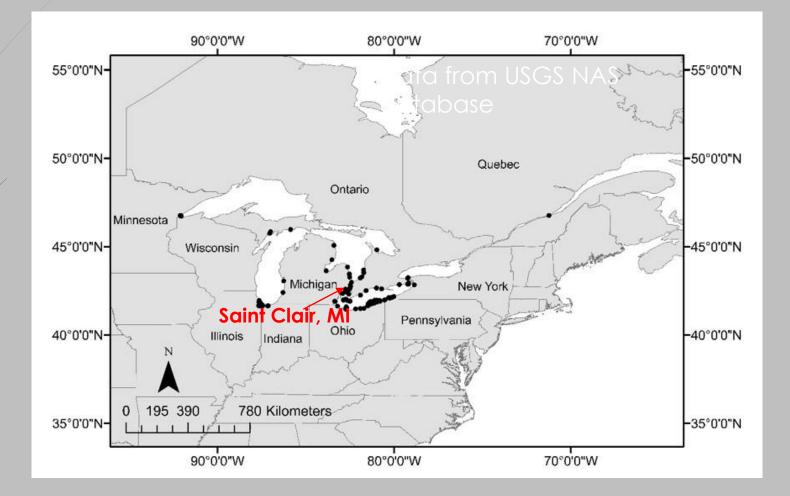
Round goby (Neogobius melanostomus)

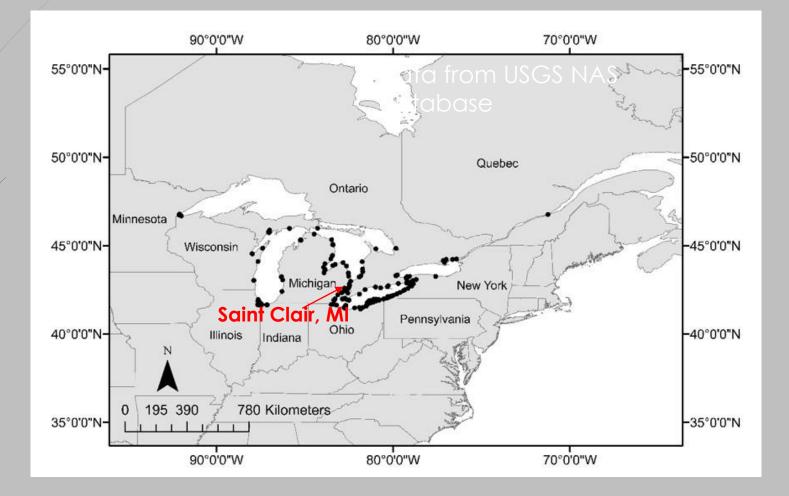
- Introduced by ballast water from the Baltics
- Ballast water largely responsible for secondary spread
 - Genetic evidence for multiple introductions
- Found throughout Great Lakes

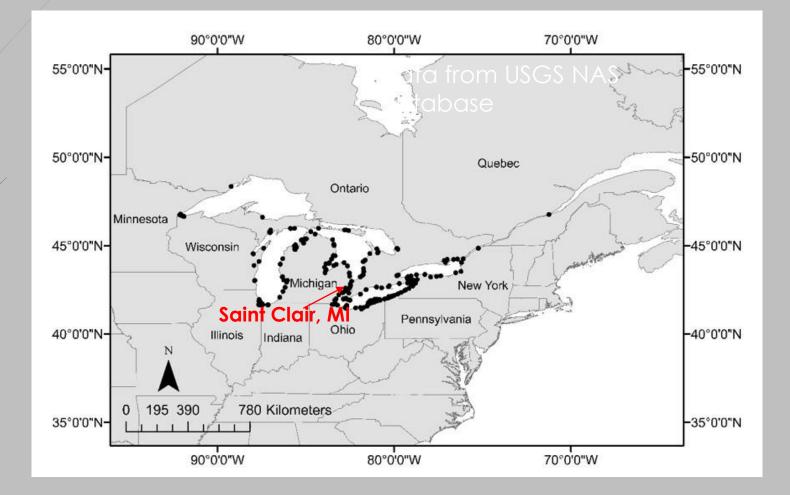




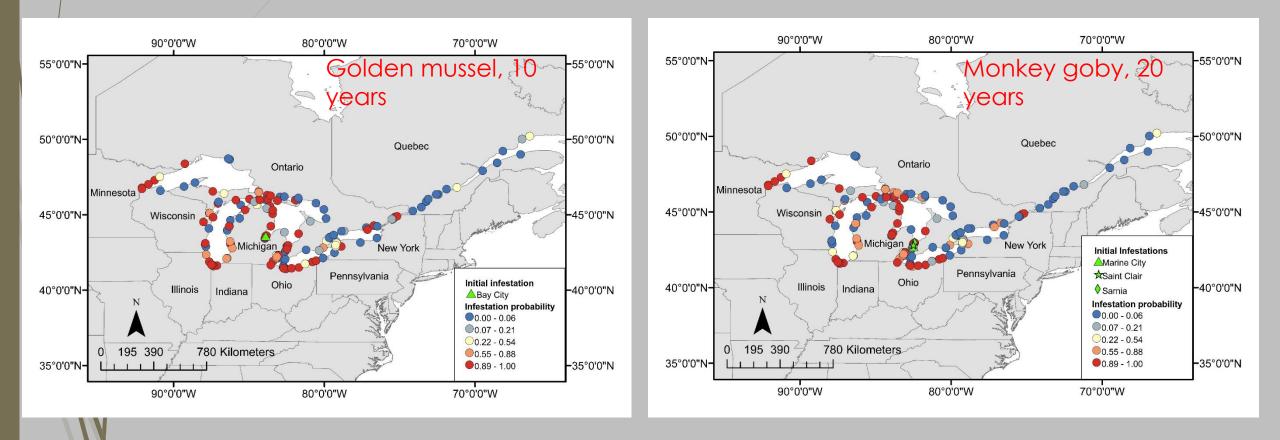








Secondary spread forecasts



Goals/Objectives

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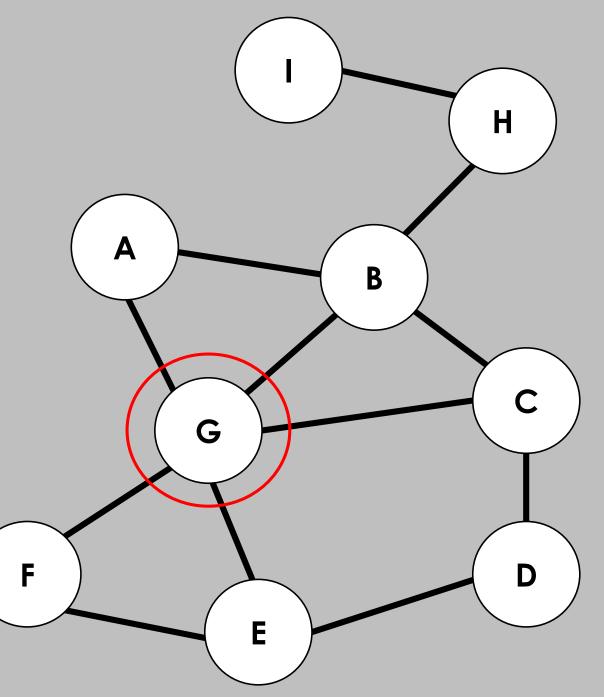
How do you quantify importance?

Tonnage handled at the top 10 U.S. Great Lakes ports in 2010. Data from www.midamericafreight.org

PORT NAME	Total Tonnage (U.S. Tons)	Domestic (U.S. Tons)
Duluth-Superior, MN and WI	36,598,247	26,936,111
Chicago, IL	18,534,237	15,381,973
Two Harbors, MN	13,877,097	13,391,512
Detroit, MI	13,406,493	10,792,960
Cleveland, OH	10,791,326	9,218,274
Toledo, OH	10,720,187	3,927,139
Indiana Harbor, IN	10,168,960	9,901,838
Presque Isle, MI	8,720,506	6,447,080
St. Clair, MI	7,988,201	7,988,201
Gary, IN	7,831,215	7,607,812

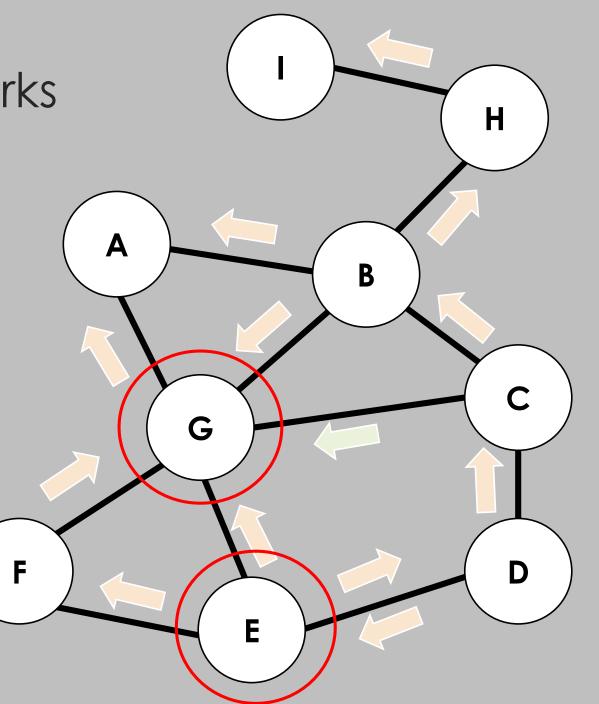
Network centrality

- Most influential?
- Identify key 'actors' in the shipping network.
- More central nodes have greater influence over network
 - Many ways to define this
 - Always context dependent
- Most basic: Freeman's degree centrality



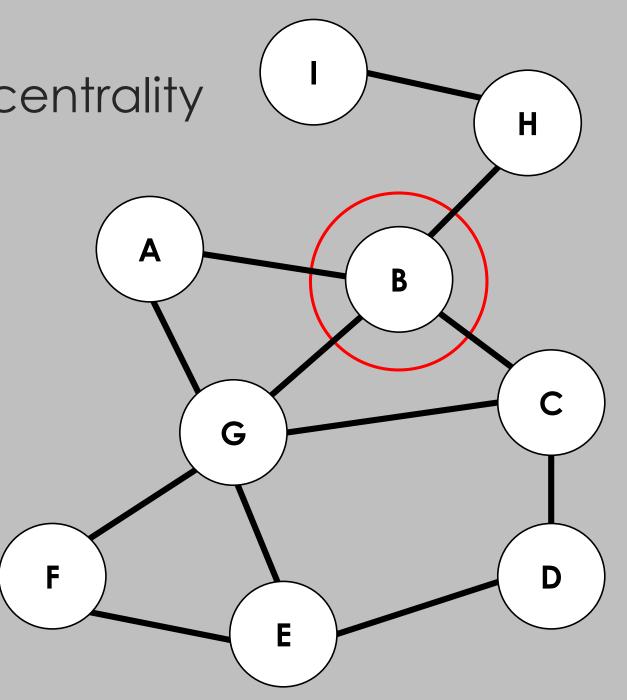
Directed networks

- Direction of connections is important to consider
- Indegree (incoming connections)
 - Receivers
- Outdegree (outgoing connections)
 - Distributors



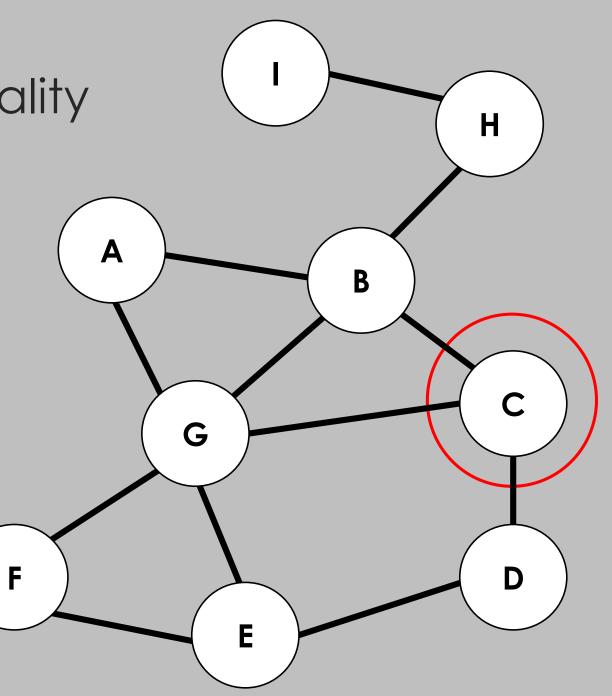
Betweenness centrality

- Which nodes act as bridges along the shortest paths to all other nodes?
- Critical links in the network
- Can determine fate of entire sections of the network
 - Gateways



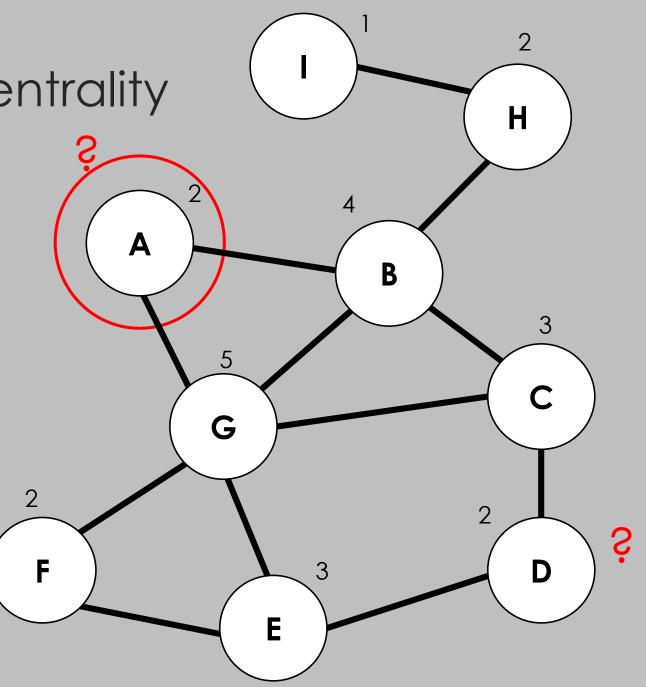
Closeness centrality

- Lowest number of steps to reach other nodes
- Facilitates rapid transport within the network
 - Sort of like geographic centers



Eigenvector centrality

- Nodes that are connected to nodes with high connections, which in turn share many connections and so on...
- Consider two nodes with 2 ties each



Simulating management scenarios

6 network centrality metrics plus busiest ports, total of 7 scenarios

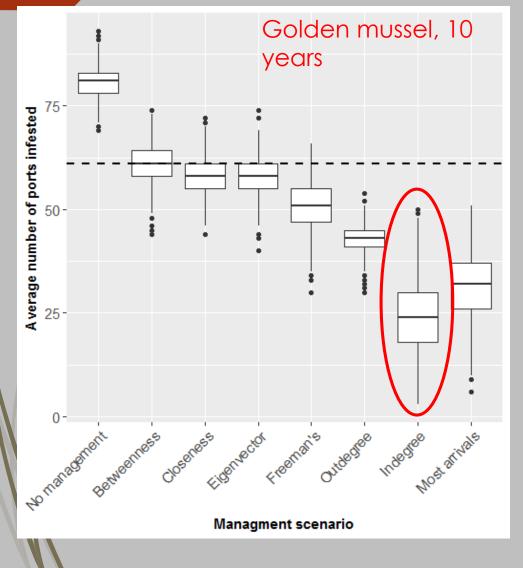
- 1. Freeman's degree total connections
- 2. Indegree receivers
- 3. Outdegree distributors
- 4. Closeness central hubs
- 5. Betweenness gateways
- 6. / Eigenvector prestige
 - Busiest ports by ship arrivals

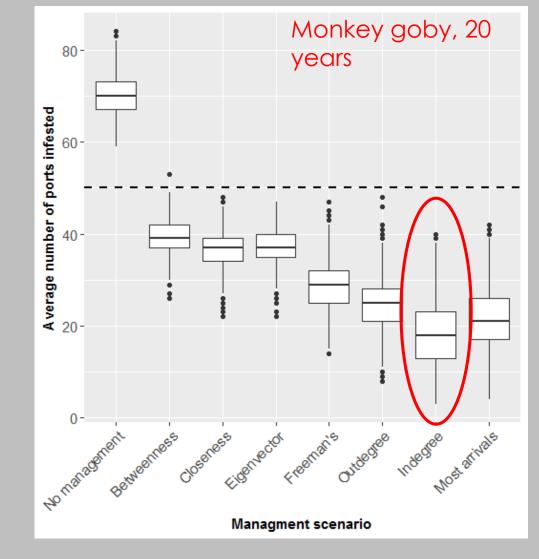
Simulating management scenarios

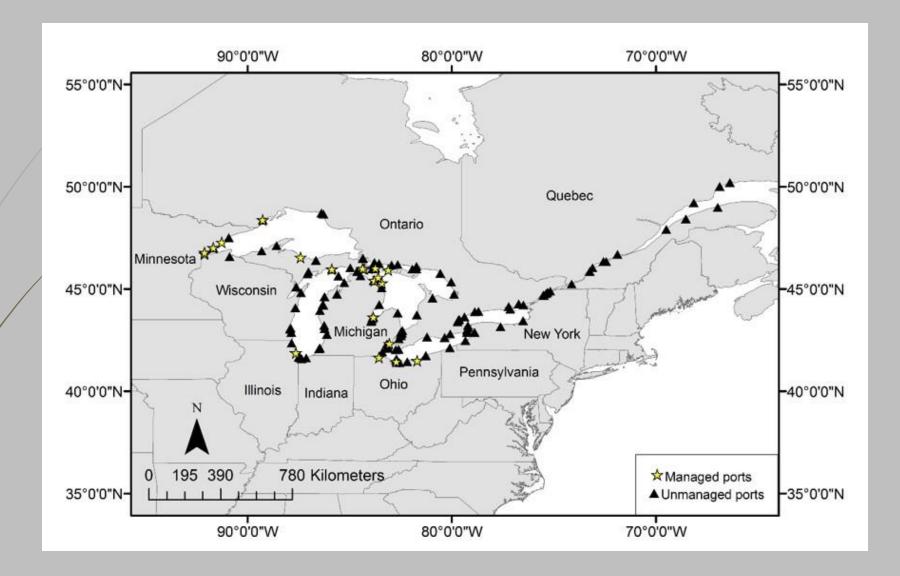
6 network centrality metrics plus busiest ports

- 1. Freeman's degree total connections
- 2. Indegree receivers
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- 6. / Eigenvector prestige
 - Most arrivals (general measure of business)
- Modeled management at top 20 scoring ports in each metric; assumed 0% infestation probability
- 1000 times at 10 years of spread for golden mussel, 20 years for monkey goby

Management at top 20 scoring ports for each species produced similar results



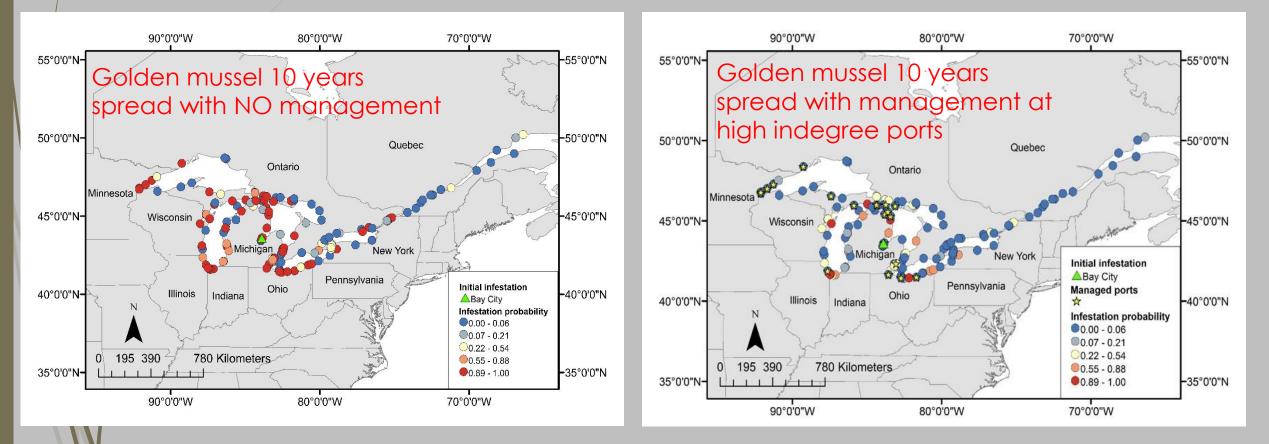




Secondary spread results

Average number of ports infested: 62%!

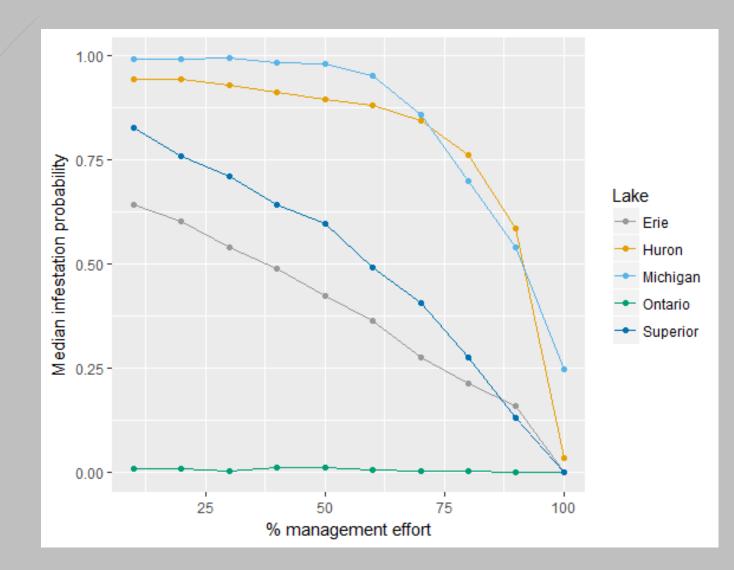
Average number of ports infested: 23.62



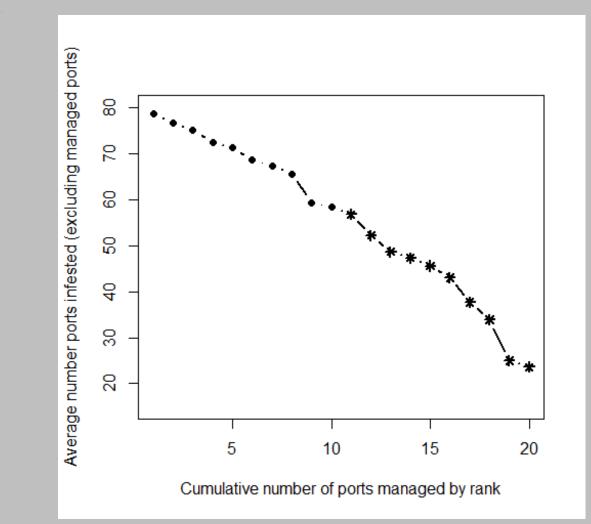
How much effort is needed to make a difference?

- Tested effort in two ways:
 - How intensely do ports need to be managed?
 - How many ports need to be managed?
- Simulated range of management intensity scenarios
- Simulated spread by adding management incrementally
 - Mann-Whitney U test to determine when number of ports infested became significantly different from 'No management'

Management should be intensive and the effects of management are not even between lakes

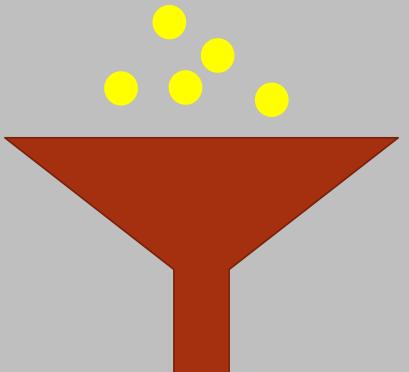


How many ports need to be managed?



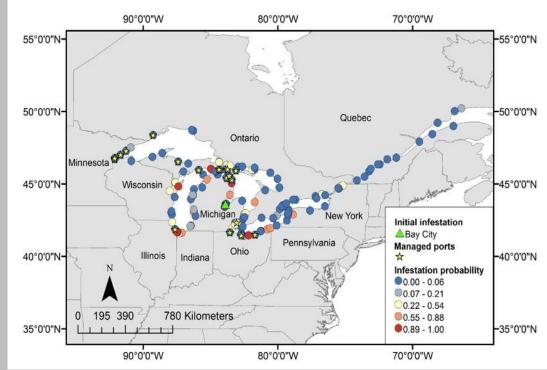
High indegree ports are like a funnels for AIS

- The higher the indegree score, the bigger the funnel
- Therefore provide the greatest chance to encounter and stop new AIS early in the invasion process
- Once they are through the funnel, they become hard to contain

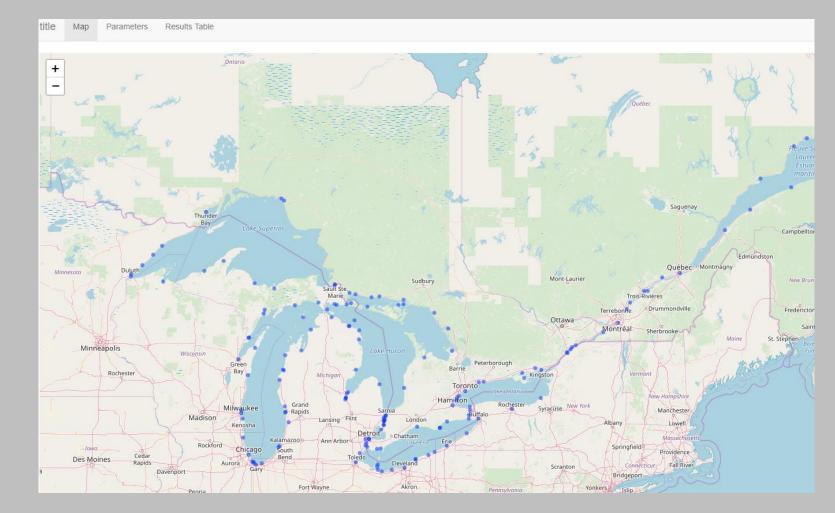


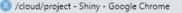
Conclusions

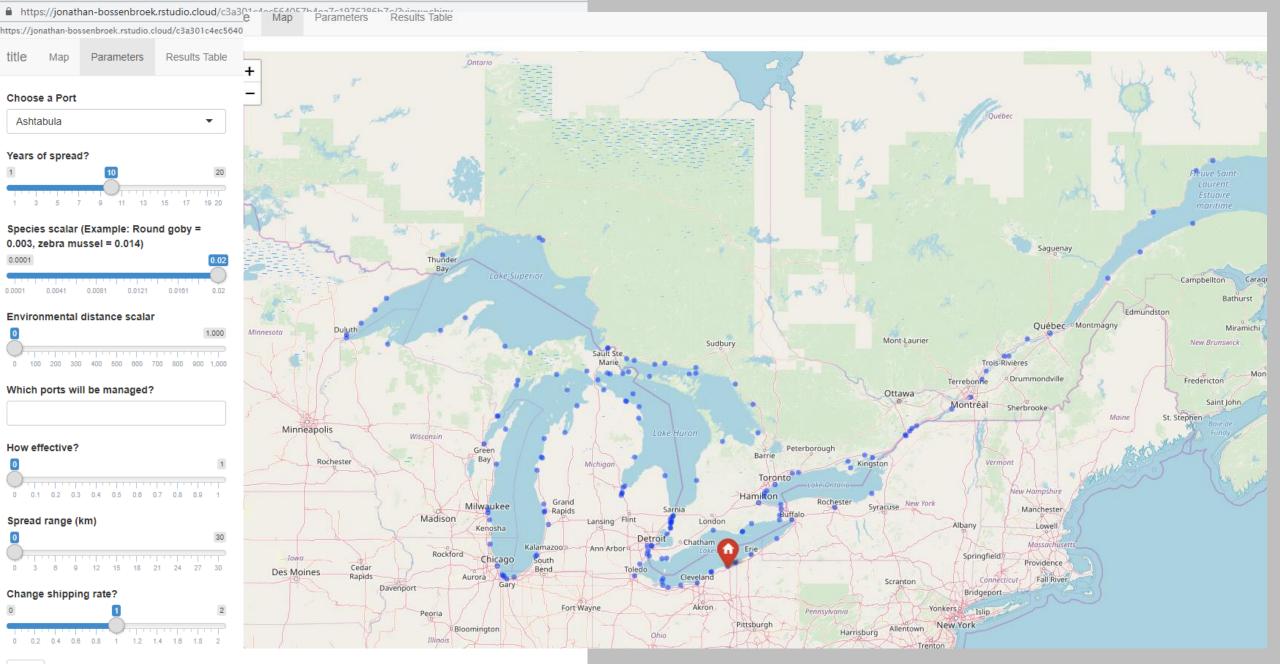
- Model capable of predicting secondary spread through ballast water
 - Best available data from U.S. and Canada
 - Parameterized for a range of taxa
- Demonstrated the utility of network centrality for AIS management
- Mønagement should be both:
 - Intensive
 - Collaborative



User Friendly Model – RCloud, Shiny Package





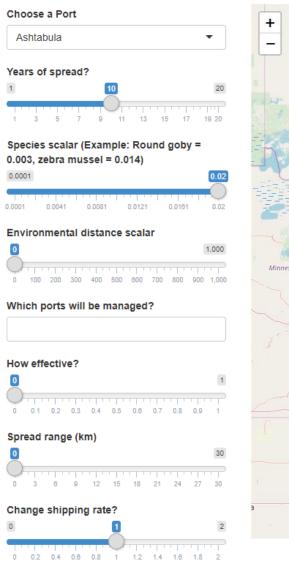


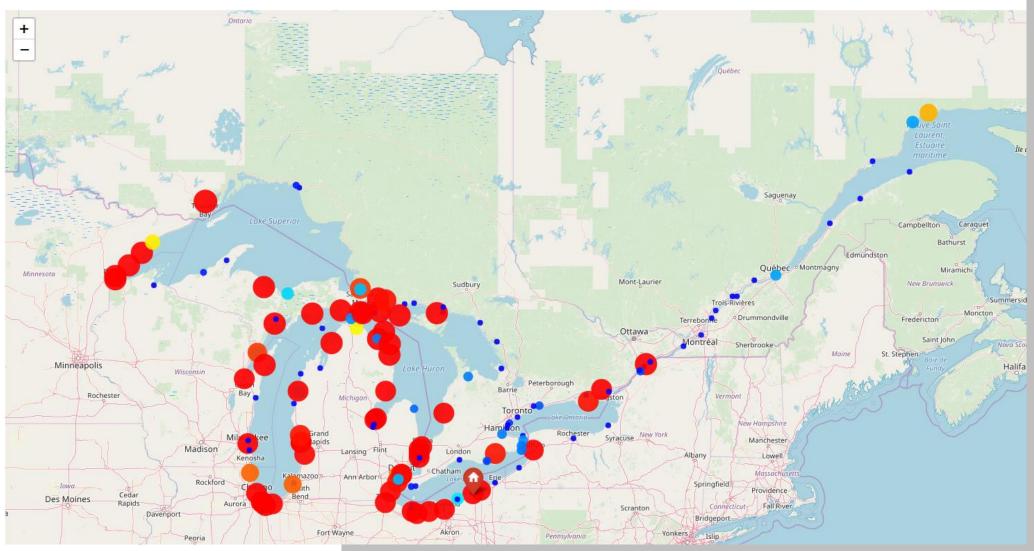
R /cloud/project - Shiny - Google Chrome

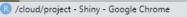
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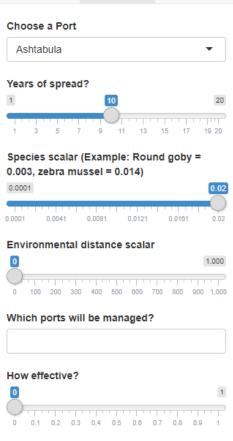




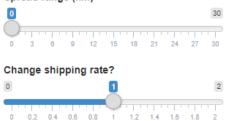
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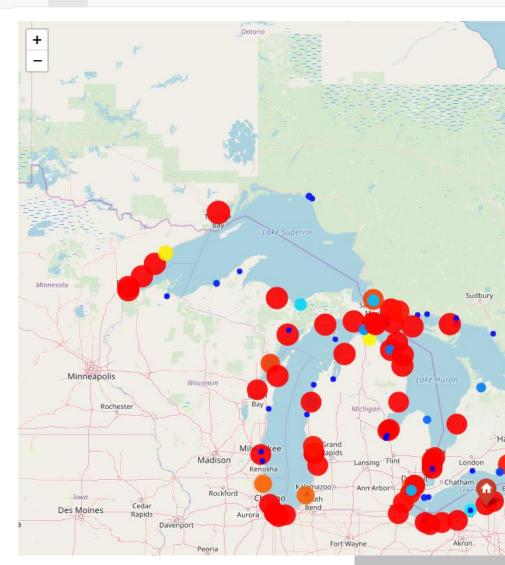
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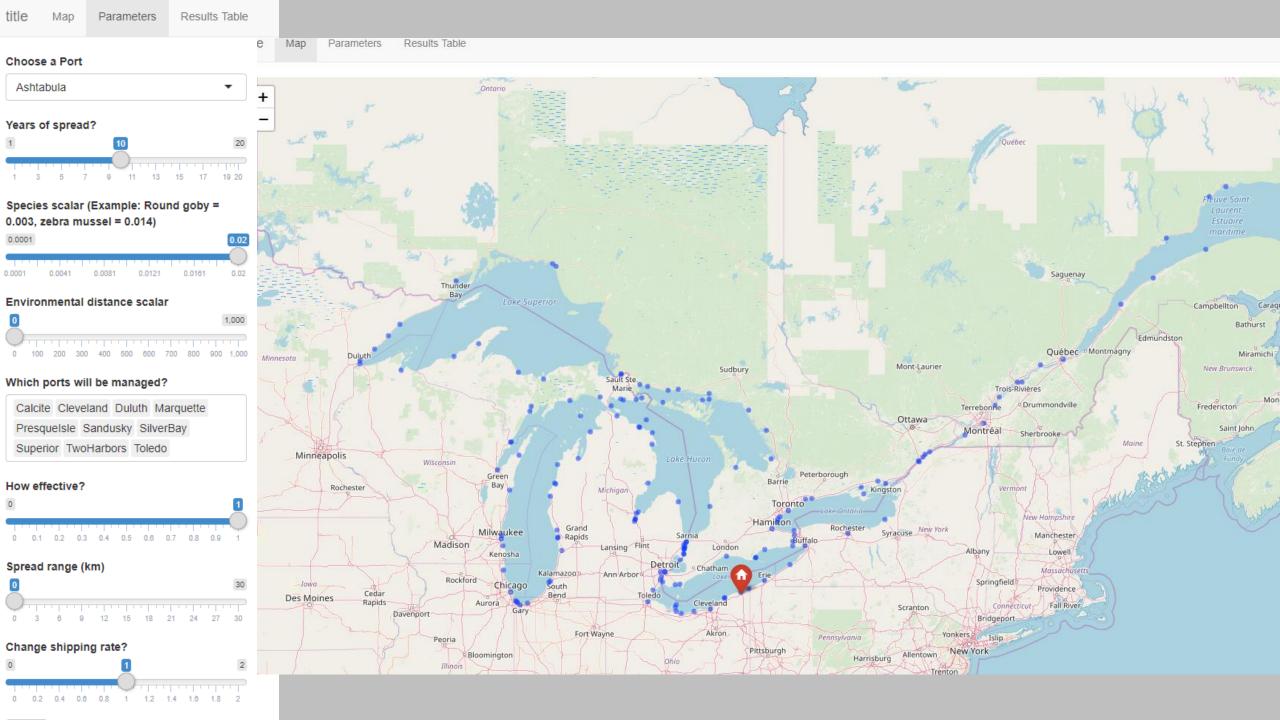
Spread range (km)

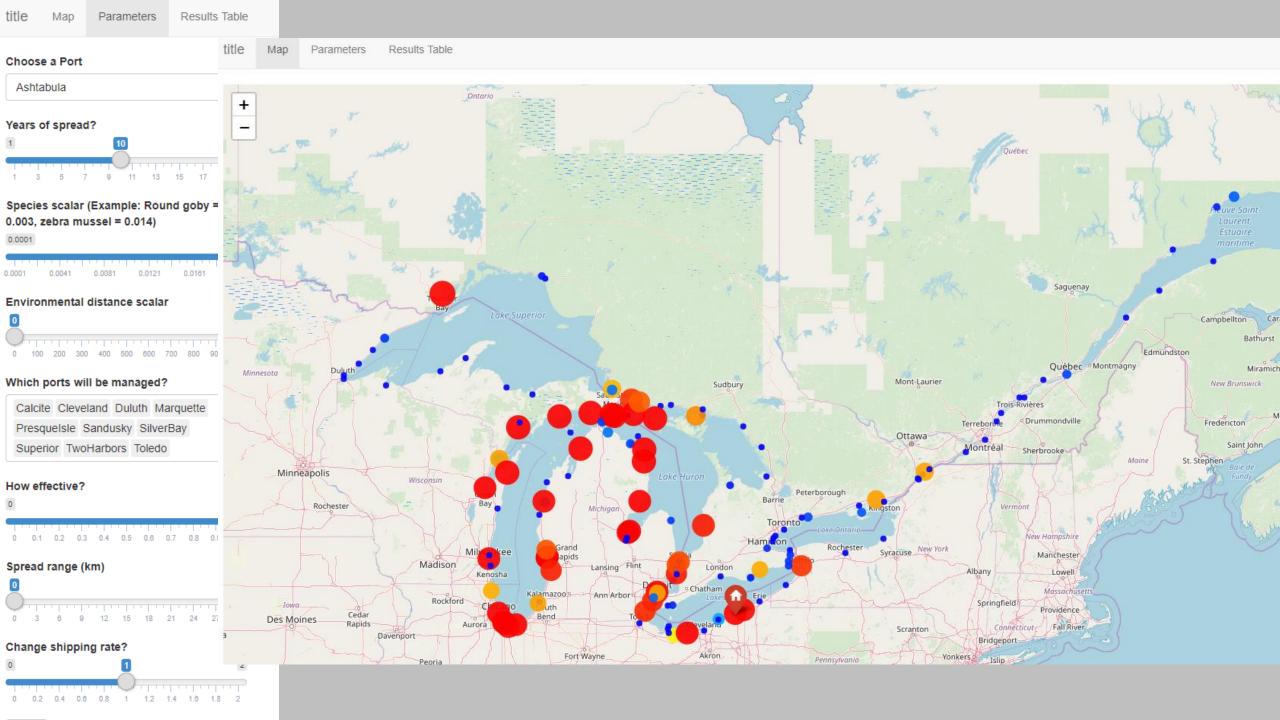


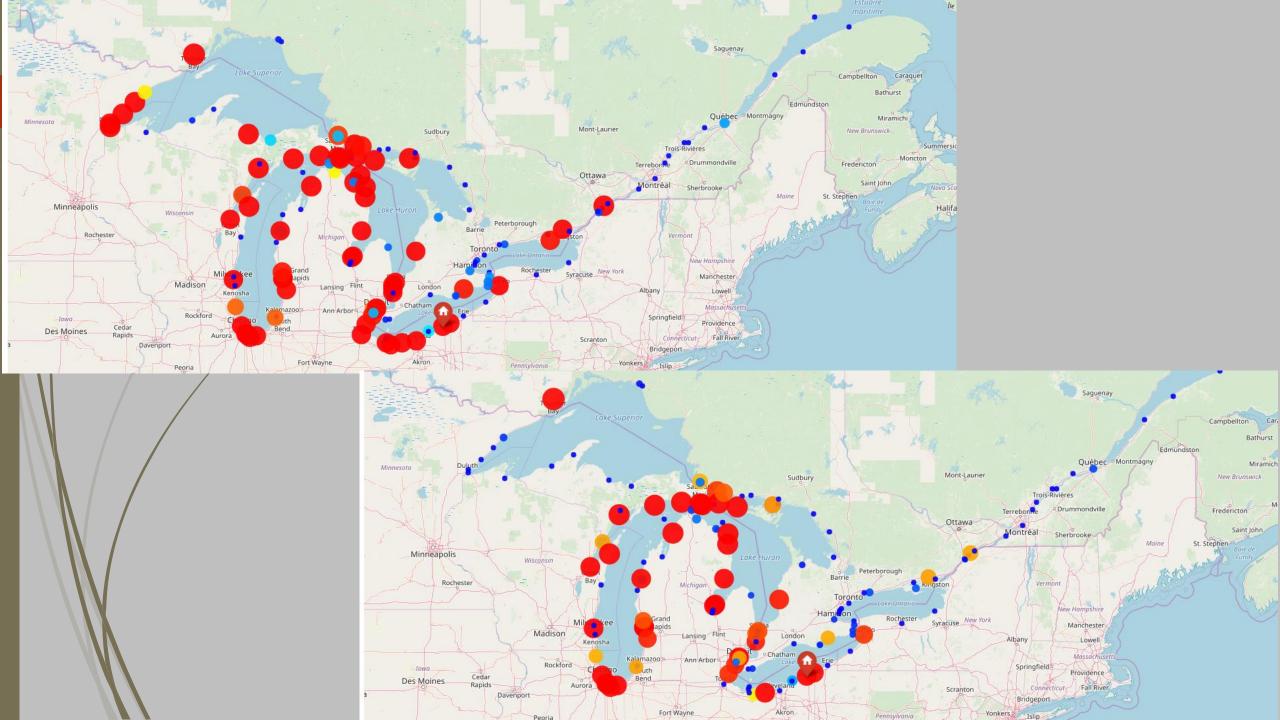


Results Table

title	Мар	Parameters	Results Table				
Port		Pro	bability	lat	long	Lake	State_code
Alpen	а		1.00	45.06	-83.43	Huron	MI
Ashta	bula		1.00	41.87	-80.79	Erie	OH
Bath			1.00	44.28	-76.72	Ontario	CAN
BayCi	ity		1.00	43.59	-83.89	Huron	MI
Brevo	rt		1.00	46.04	-84.97	Michigan	MI
Bruce	Mines		1.00	46.30	-83.79	Huron	CAN
Buffal	0		1.00	42.89	-78.88	Erie	NY
Buffin	gton		1.00	41.64	-87.43	Michigan	IN
Burns	Harbor		1.00	41.63	-87.13	Michigan	IN
Calcit	е		1.00	45.41	-83.79	Huron	MI
Ceda	rville		1.00	46.00	-84.36	Huron	MI
Charle	evoix		1.00	45.32	-85.26	Michigan	MI
Chica	go		1.00	41.88	-87.63	Michigan	IL
Cleve	land		1.00	41.50	-81.69	Erie	OH
Conne	eaut		1.00	41.95	-80.55	Erie	OH
Detroi	it		1.00	42.33	-83.05	LakeStClair	MI
Drum	mond		1.00	46.02	-83.73	Huron	MI
Duluti	h		1.00	46.79	-92.10	Superior	MN
Escar	naba		1.00	45.75	-87.06	Michigan	MI
Essex	wille		1.00	43.62	-83.84	Huron	MI
Ferry	sburg		1.00	43.08	-86.22	Michigan	MI
Gary			1.00	41.59	-87.35	Michigan	IN
Gode	rich		1.00	43.74	-81.71	Huron	CAN
Greer	nBay		1.00	44.52	-88.02	Michigan	WI
Indian	aHarbor		1.00	41.67	-87.44	Michigan	IN







User Friendly Model – RCloud, Shiny Package

- Useful for hypothetical scenarios
- Can incorporate environmental similarity.
- Working with TNC to determine where to host.

Thank you!

Questions

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