

Pushing Past AIS Paradoxes Via R & R Shiny

Dr. Alex Bajcz, Quantitative Ecologist Minnesota Aquatic Invasive Species Research Center (MAISRC) @ The University of Minnesota, USA



Goals

- 1. Reveal woes of being an "AIS* warrior."
- 2. Celebrate the *majesty* of R for us!
- 3. Provide a **whirlwind** tour of how R has empowered us *over just the past year or so*.



*AIS = Aquatic Invasive Species

Acknowledgements

- Thanks to Epi-Interactive for the invite, as well as for support, guidance, and direction for AIS Explorer and for building our own R Shiny capacity!
- Thanks to MAISRC researchers John Fieberg, Dan Larkin, Mike Verhoeven, Amy Kinsley, Bob Haight, Nick Phelps, and Gretchen Hanson for letting me share their work.
- The University of Minnesota sits on ancient lands of the Dakota. In all we do, we must be stewards of the land to protect and enrich all its peoples and lifeforms.

Invasive Species Science's Many Paradoxes

Invasive species: Non-native organism excelling so much in a new spot it causes ecological /economic harms.

- Many are *endangered* at home!
- We must *punish* their *success*, even as *we* likely facilitate(d) it!

Starry stonewort, invasive in MN, is endangered in its native Eurasia.



Zebra mussels spread via boats and thrive in waters we pollute.

Eurasian watermilfoil has

invaded > 500 MN lakes,

lowering property values!

Invasive Species Science's Many More Paradoxes

Invasives harm, resist, and spread better when *established*. Early detection is *key*! But this is saying "Find things at their *rarest*." Where? How?!

- Where you *expect* to find it?
 - You'll only "confirm" what you already think you know.
- Where a model tells you to look?
 - What *data* will you use to fit it??



AIS are often detected using rakes or nets.



Recent study: Only 30% chance of finding starry stonewort when it's <u>present</u>!

Only look in X, only find in X, only continue to look in X...

Still More Invasive Species Science's Paradoxes

Preventing *spread* is just as key. Here, the reverse: Often far too *many* data!

- Consider just Cass County, MN:
 - 514 lakes (264K possible lake pairs)!
 - 30 already infested (that we know of!)
 - Imagine: Enough inspectors for *just* 25 lakes.
 Which do **YOU** choose??
 - 2.11E+42 possible plans!



Meet Our SavioR

R is our **go-to tool** for besting the paradoxes of AIS detection and management (DM). R...

- Helps build, clean, and leverage huge new data sets.
- Magnifies rare data to guide detection.
- Collapses abundant data to guide management.
- Let's us share decision-making tools.
- Let's us "predict the future!"

Other platforms could do each thing R does for us; few (if any) could do *everything* R does for us.



R—The BuildeR

R has helped us build, maintain, and grow two *immense* data sets:

#1: Point-Intercept (PI) aquatic
plant survey database

- **Systematic** (unbiased) assessment of what plants live where in lakes.
- Collected by state agencies, counties, watershed districts, private companies, researchers, etc.
- No two data sets are alike! So no one has aggregated them...

A rake is thrown overboard at every site and retrieved. What gets stuck is what's recorded.





A grid is put on a lake. ALL points are sampled, no matter what, and ALL plants found are recorded, invasive or not.

R—The MagnifieR

Meet **PI Charter**, our R Shiny Hub World for our PI survey database!

- R Shiny enabled a portal for data set submission straight to our database!
- R Tidyverse enabled fast data cleaning and homogenizing—no humans needed!
- R JAGS allowed us to magnify these data via a Bayesian predictive model to guide early detection of starry stonewort.

Check it out at z.umn.edu/PICharter!

Hot colors = more likely to have starry stonewort



How Do I Use This Tab? On this tab, you can submit new Pl survey data files for incorporation into MAISRC's database! Not sure if what you have is a Pl survey? Consult the 'How is a Pl Survey Different From Other Kinds of Plant Surveys?' section on the 'Browse the Database' tab for more information! Below, on the left, the App will ask a *brief* series of questions. As you answer them, the App will eventually display--and maybe even reformat--your data on the right. The questions will help us ensure that your data are consistent with our database's structure and formatting requirements and will help the App

R—The BuildeR

R has helped us build, maintain, and grow two *immense* data sets: #2: The Minnesota Boater Movement Network database

- Inspectors survey boaters during inspections.
- These data are *immense*, complex, and contain inconsistencies—but have great potential for understanding boater movements!

Inspectors survey boats <u>and</u> their people.





R-The ShareR R *Tidyverse* helped us clean and process the data *way* faster and more easily! R *Quarto* allows us to document and share our process for greater transparency and reproducibility.

- R *igraph* turns these data into "network diagrams."
- R *Shiny* helps us share this info digestibly with stakeholders!

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R—The CollapseR

For each county, an optimized digest of which lakes to inspect in what order.

R also helps us "collapse" data when there's *too* much.

Meet **AIS Explorer**, our flagship R Shiny App for AIS D&M!

- R *OMPR* enables optimization models to find the *mathematically* best choices.
- R *Shiny* enables displaying optimized plans to stakeholders!

Check it out at www.aisexplorer.umn.edu!

	493 Insp	Inspection stations needed to inspect 100% of the ris boats in Cass County		
Rank	Name	DOW Number	Status	Percentage of the risky boats inspected 1
1	Leech	11020300	C 8 8	12.6%
2	Gull	11030500	0	20.6%
3	Cass	04003000	08	277%
4	Winnibigoshish	11014700	08	33.8%
5	Woman	11020100	0	377%
6	Ten Mile	11041300	0	411%
7	Thunder	11006200	8	44.5%
8	Little Winnibigoshish	31085000	0	47.7%
9	Washburn	11005900	8	50.8%
10	Roosevelt	11004300	8	53.7%
11	Pine Mountain	11041100	0	56%
12	Sytvan	11030400	0	58.3%
13	Pike Bay	11041500	-	60.4%

The model projects your intervention scenario would avert approximately 0.43 (0.75-0.32) new, boat-driven infestations over a five-year period.

R—The PredictoR

R even lets us "peer into our crystal ball!"

- Using R, we've made a simulation model that predicts which lakes will be infested in 5 years—unless we act! *Coming to AIS Explorer*!
- Another sim helps stakeholders understand the associated uncertainty.





Build a plan and simulate how life will be different 5 years from now under it!

My goals today

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- 2. Celebrate the majesty of R for us!
- 3. Provide a **whirlwind** tour how R has empowered us *over just the past year or so*.
- 4. Inspire and encourage you to take the plunge--tap R to fight for you in your sphere!

And if you'd appreciate an ally, reach out! Alex Bajcz -- <u>bajcz003@umn.edu</u>, or find me on LinkedIn!