## Prediction of starry stonewort invasion

 risks in Minnesota and Wisconsin based on lake level habitat suitability
## Where is starry stonewort going?

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# A comparison of eigenvalue decomposition and machine learning approaches for estimating distribution of a non-equilibrium species 

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## Collaborators

- Dan Larkin
- Wes Glisson (research fellow)
- Mike Verhoeven (grad. student)
- Carli Wagner (undergrad.)
- Luis Escobar
- Nick Phelps
- Megan Weber


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## Starry stonewort working group

- MN DNR
- University of Wisconsin
- Paul Skawiski
- Central Michigan University

- New York Botanical garden
- Ken Karol and Robin Sleith
- University of Geneva


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## Starry Stonewort



## Starry stonewort

Nitellopsis obtusa

- Green alga


Three domains of living organisms (Gogarten, Taiz et al. 2015)

## Starry stonewort

- Charophyte
- Closely related to stoneworts / muskgrasses native to Minnesota
- Ecologically important
- Water quality
- Habitat



Chara aspera

C. contraria


Nitella flexilis

## Invasion history

- Relatively new invader
- Quickly gaining ground
- Increasing concern for AIS management
- Currently in 9 lakes in MN


## Present distribution



## Ecological niche modeling

- Species distribution models
- Habitat suitability models
- Bioclimatic models
- MaxEnt


## Ecological niche modeling



Ecological niche modeling


## Ecological niche modeling



## Determining the niche

## Information from the current distribution



## Determining the niche

Information from the current distribution


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Information from the current distribution


## Challenges for predicting risk

- Is the current range all of the suitable habitat?
- Have you measured all the important factors?
- Biotic vs Abiotic constraints



## Environmental data



## Environmental data

## SCIENTIFIC REPRTS

OPEN Realized niche shift associated with the Eurasian charophyte Nitellopsis obtusa becoming invasive in North

## America

Luis E. Escobar ${ }^{1,2}$, Huijie Oiao ${ }^{3}$, Nicholas B Nitellopsis obtusa (starry stonewort) is a dioeci emerged as an aquatic invasive species in Nort
of its native range, but has spread rapilly in n of its native range, but has spread rapidy in no
interfere with recreation and may displace nati of $N$. obtusa, making it difficult to foreceast futu
investigated environmental variables associate investigated environmental variables associatt
data and remotely sensed environmental varia data, and remotely sensed environmental variz
distribution. We found that $N$. obtusa is exploit
which may help explain its invasiveness. While which may help explain its invasiveness. While there appears to have been a shift in its realize,
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estimates of the physiological tolerances of th Understanding how certain species experience gre
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invasive in Chile, Australia, and New Zealand ${ }^{\top}$. Ho
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Regardless of the underlying mechanisms, the hybridization, novel antelopathic weapons, and unc
Regardless of the underlying mechanisms, the
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May not be best strategy for aquatic species

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## Accepted: 14 June 201

 Published: 01 July 2016ized in their native range ${ }^{11}$. It is true that many ${ }^{1}$
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even rare species can potentially become dominan even rare species can potentially become dominan
conditions ${ }^{11,1516}$.
3 ONO
${ }^{1}$ Minnesota Aquatic Invasive Species Research Cent Popelation Medicicine, Colllege of Veterinary Menti
Laboratory of Animal Ecology and Conservation Biol China. ${ }^{4}$ Department of Fisheries, Wild life, and Cor USA. Correspondence and requests for materials sh

## Environmental data



## Environmental data



## Environmental data

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## Remote Sensing and Geospatial Analysis Laboratory



## Estimating the niche



## Estimating the niche



Dissolved
oxygen
Dissolved
oxygen
saturation
pH
Alkalinity
Hardness
Conductance
N
Ammonia

Nitrate
P
Ortho-P
Temp
Secchi depth
Chlorophyll A
Color
Salinity
S
Lead

## Estimating the niche



Nitrate
P
Ortho-P

Secchi depth Chlorophyll A Color
Salinity
S

## Estimating the niche



P
pH
Secchi depth Chlorophyll A

Conductance
N

## Estimating the niche







## Estimating the niche



Env 1

Env 2


Env 3


# Predicting suitable habitat 

- Three approaches
- Random forests
- Boosted regression trees
- Ecological niche factor analysis (presence only)


## Predicting suitable habitat

Low risk
High risk


## Predicting suitable habitat

Low risk
High risk

Random forest


Boosted regression tree


ENFA


## Expanding what we know

Boosted
Random forest Regression trees
ENFA


## Expanding what we know

- Field sampling $\longrightarrow$ Spatial interpolation



## Expanding what we know



## Expanding what we know

## Connectivity

Comnare nediction methods More samples



Expanding what we know Keep looking !


## Starry trek - How you can help

- Minnesota \& Wisconsin
- Statewide, coordinated shoreline searches

Saturday, Aug. 5 ${ }^{\text {th }}, 2017$


Photo: Dave Hansen

## Final thoughts

- "All models are wrong; some are useful"
- Models give the impression of precision
- More levels of extrapolation or interpolation mean more sources of error
- Use models for planning and prioritization


## Questions?



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