Appendix A: Virtual Seminars Schedule and Presentations

Seminar I - Thursday, Nov 3, 2022	Presenter
Introduction: Investigating Food Webs: State of Knowledge and Investigative Approaches	Tom Stewart
A fisheries management perspective on food web information needs	Randy Claramunt
Applications of ecological tracers in the Great Lakes	Aaron Fisk et al.
The adaptive capacity of lake food webs	Bailey McMeans
Break	
Phosphorus to fish	Marten Koops
Integration of water quality objectives and fisheries objectives	Tim Johnson
Integrated Ecosystem Assessment	Mike Fraker
Adjourn	

Seminar II- Thursday, Nov 10, 2022	Presenter
Opening remarks	Tom Stewart
Understanding material and energy flow in aquatic ecosystems using linear inverse modeling	Dick van Oevelen
LIM approach to Ecopath mass balance and hypotheses testing	Tom Stewart
Modelling chance and necessity in natural systems	Benjamin Planque
Break	
Applications of Ecopath/Ecosim in the Great Lakes	Ed Rutherford
Atlantis: Potential Great Lakes applications	Doran Mason
Ecosystem behaviors	Bob Ulanowicz
Wrap-up and next steps	Tom Stewart

Investigating Food Webs: State of Knowledge and Investigative Approaches

Tom Stewart

Brian Weidel, USGS

Dick van Oevelen, Royal Netherlands Institute for Sea
Research

Investigating Food Webs: Sate of Knowledge and Investigative Approaches

Acknowledgements

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Great Lakes Fishery Commission
Cooperative Institute for Great Lakes Research

Steering Committee and Advisors

Nicholas Boucher, Aaron Fisk, Roger Knight, Doran Mason, Kevin McCann, Bailey McMeans, Lars Rudstam, Ed Rutherford, Heidi Swanson

Workshop Hosts

Cornell Biological Field Station

Investigating Food Webs: Sate of Knowledge and Investigative Approaches

What's this all about?

nvestigating Food Webs: Sate of Knowledge and Investigative Approaches





Objectives

- 1) Review and share the current state of food web investigative methods,
- 2) Determine food web-scale fisheries management information needs and possible investigative approaches, and
- 3) Develop collaborative study designs and proposals for potential funding addressing food web knowledge gaps relevant to fisheries management information needs.

Seminars Nov 3 & 10

Workshop Nov 14-16

Investigating Food Webs: Sate of Knowledge and Investigative Approaches

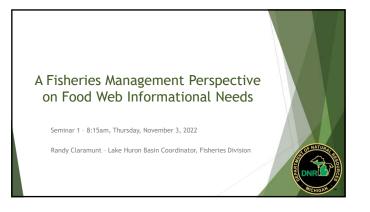
House keeping

- Presentation timer
- Presentation recordings
- Questions during seminar please raise your hand
- Workshop registration link is in the Zoom chat

To request recorded talks nboucher@glfc.org

Any other questions tomstewart54321@gmail.com

Investigating Food Webs: Sate of Knowledge and Investigative Approaches



Elevator Pitch for Food Webs

- ► The pitch 'theme': make smart decisions.
- ▶ Credibility booster: my background.
- \blacktriangleright Short and sweet pitch: GLs Δ + complex.
- ► What's the twist: food web information too complicated or theoretical for management.
- ▶ Call to action: What is needed or not
- One-liner: Hey look, without an understanding of food web dynamics, managing GLs fisheries is like driving at high speeds, in the dark, with no headlights!



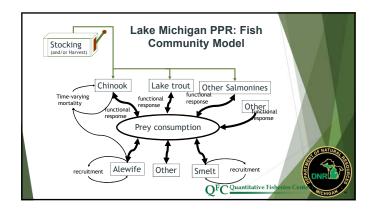
Make Smart Decisions

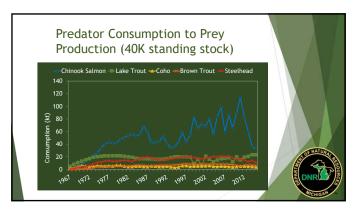
- Cost of Actions: a move from single species/agency management to fish community decisions in a shared management framework has exponentially increased the cost of said actions.
- ► Timeline: the diversity of stakeholder input has substantially increased the duration for complex decisions and management actions.
- Context: each decision / action is linked with or should be referred to previous actions and could include a range of data review, literature, and reports.
- ► Economics: management agencies with jurisdiction often lack the financial resources to adequately address the aforementioned challenges.
- Information needed from investigation of food webs are as follows:
 - ▶ Compensation
 - ► Stability and resiliency
 - ▶ Species interactions, prey preferences and habitat linkages
 - Productivity and trophic efficiency
 - ▶ Predictability (with or without management actions)

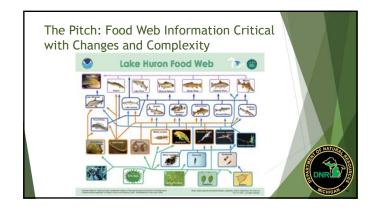


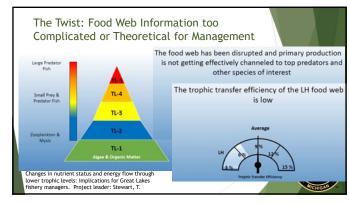
- ► Lake Michigan Research Biologist with a focus on stock assessment and predator-prey dynamics.
- ▶ Worked as a manager for both State and Tribal Fisheries Agencies.
- ▶ Predator-Prev dynamics substantial changes in
 - the GLs and observed: ▶ Single-species management to fish community
 - ▶ Need to balance productivity with mortality rates
 - ▶ Invasive species impacts
 - ▶ Advancing science and models
- Continued challenges

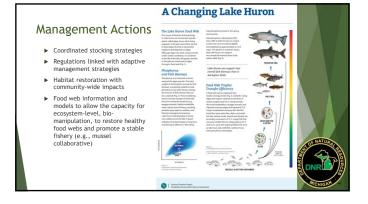


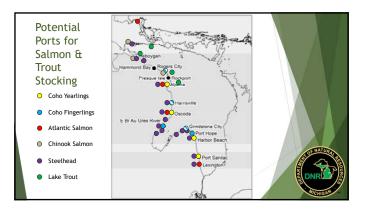












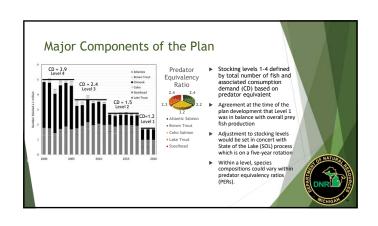


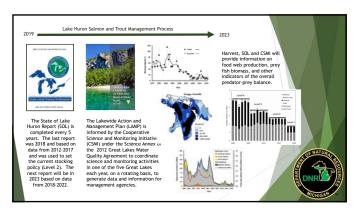
A sustainable and diverse salmon and trout fishery that maximizes the lakes production potential to provide exceptional fisheries for communities across Lake Huron.

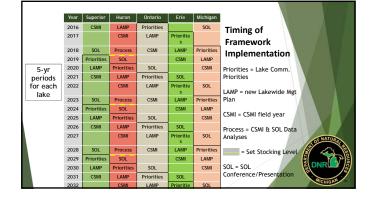


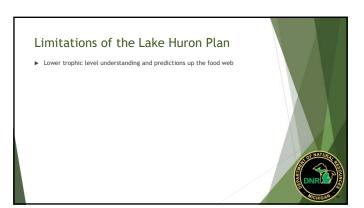
Species, predator equivalency ratios (PER), and factors considered in the stocking strategy for Lake Huron.

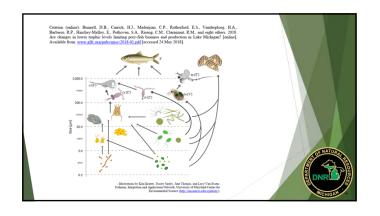
Species	Predator Equivalency Ratio (PER)	Catchability	Cost- effectiveness	Feeding Ecology	Movement- Straying	Wild Recruitment Potential	Social-Economical Benefits
Atlantic Salmon	2.4	Medium	Not available	Highly Diverse	Medium	Low	High
Brown Trout	2.2	Low	Low	Moderately Diverse	Low	Low	Medium-Low
Chinook Salmon	1.0	High	High	Pelagic Prey Only	High	High	High
Coho	3.2	Medium	Medium	Moderately Diverse	Medium	Medium	Medium-High
Lake Trout	2.3	High	Medium	Benthic Prey	Low	High	Medium-Low
Steelhead	2.4	Medium	Medium	Moderately Diverse	Medium	Medium	Medium-High



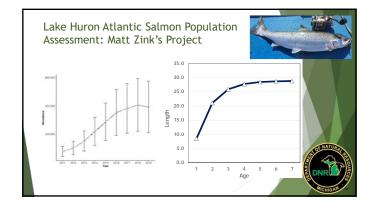




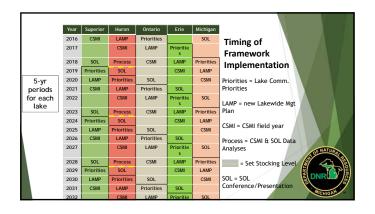


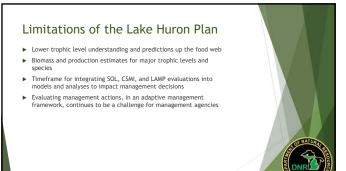


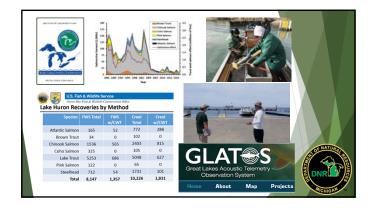
Limitations of the Lake Huron Plan ► Lower trophic level understanding and predictions up the food web ► Biomass and production estimates for major trophic levels and species

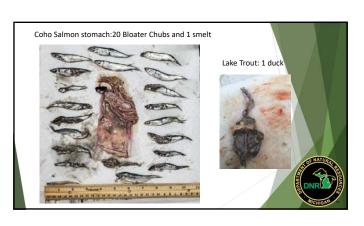


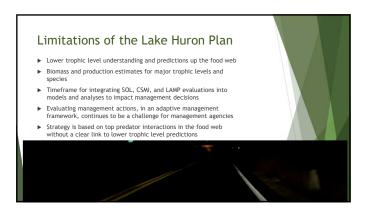








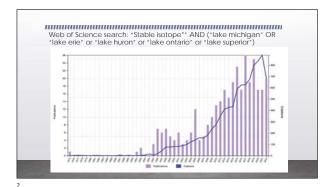






Investigating Food Webs: State of Knowledge and Investigative Approaches

Aaron Fisk
School of the Environment, University of Windsor

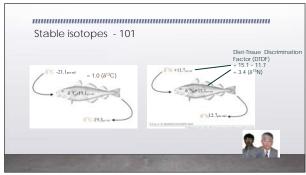


Stable isotopes - 101

- Most elements exist as multiple stable isotopes 1 or more additional neutron (same # of protons and electrons)
 - Isotopes of an element form the same chemical bonds but extra mass of neutron influences bonding strength (i.e., stronger bonds – longer to break) and movement in the environment (i.e., slower <u>movements</u>)
- Stable isotopes of the same element (e.g., ^{15}N and ^{14}N) have different abiotic and biotic process kinetics resulting in *predictable* changes in the relative amounts of the stable isotopes between pools in the environment (e.g., $^{515}N_{predator} > ^{515}N_{prey}$)

$$\begin{split} \delta^{15}N &= \left[\left(^{15}N \,/\, \left(^{15}N + ^{14}N\right)\right)_{sample} \,/\, \left(^{15}N \,/\, \left(^{15}N + ^{14}N\right)\right)_{standard}\right) - 1\right] \times 1000 \\ &- parts per thousands or per mil - symbol \% \end{split}$$

3





Nitrogen (8¹⁵N) and Carbon (8¹³C)

• 8¹⁵N tracer of relative trophic position/level (DTDF = 3.4 in FW) but also a tracer of nitrogenous inputs (e.g., fertilizers)

• 8¹⁹C tracer of carbon source/habitat use but also tracer of trophic level

• %C, %N and C:N indicators of lipid and chitin levels

Tetagic Pathway Benthic Pathway**

-15

Sulfur (8³⁴S)

• Tracer of marine vs freshwater / terrestrial resource use

• Processes and distribution in freshwater ecosystems poorly understood – but does provide insights into trophic pathways likely linked to sediment processes

• DTDFs – 0

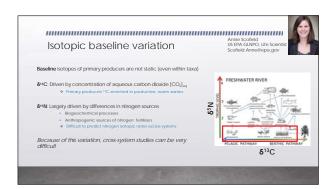
Oxygen (δ^{18} O) and Hydrogen (δ^{2} H)

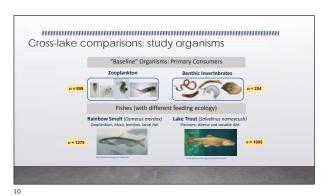
• δ^{18} O and δ^{2} H used for studies of migratory origin

• Large differences between δ^{2} H between terrestrial and aquatic systems (allochthonous vs autochthonous)

• δ^{2} H may serve as a trophic marker

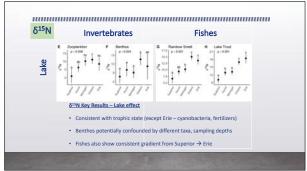
• Relative coupling of δ^{18} O and δ^{2} H may contribute insights on diet and physiology

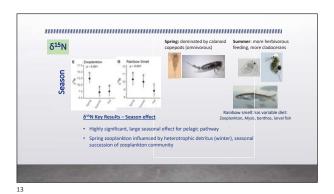


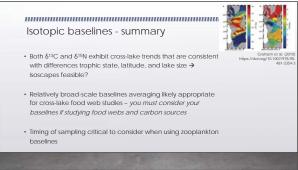


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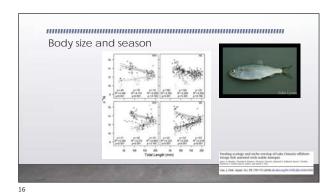


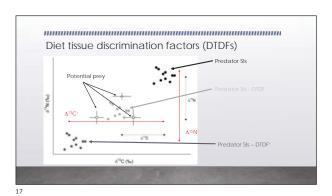


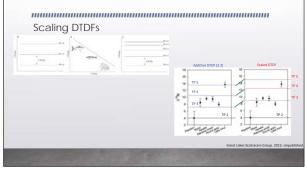


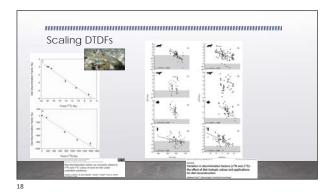
Fish movements and isotopes

Movement/feeding across different isotope regions complicates trophic position, carbon source and diet analysis
Opportunity - to quantify relative importance of different regions to diet







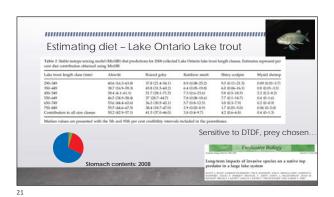


Scaling DTDFs

Scaled DTDFs were generally constant with SCA and were higher and greater range/variability than constant DTDF

Scaled DTDFs not as influenced by baseline species chosen nor carbon sources

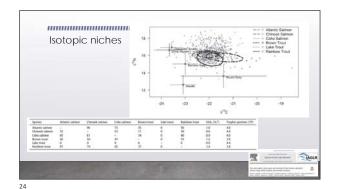
Scaled suggested a more complex trophic structuring of the Detroit River food webs



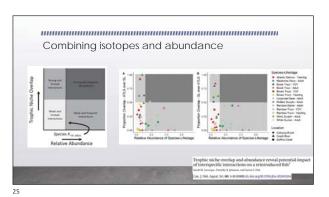
Estimating diet for lake Ontario salmonids 9.61-(9.69, 6.67) 6.10-(9.09, 6.15) 9.65-(9.02, 9.08-9.07-(9.00, 107) Sensitive to DTDF, prey chosen.

Estimating carbon sources through time – Lake Ontario Lake 81% offshore 1994-98 Sensitive to DTDF, prey 93% nearshore chosen.

23



8.12 (0.00, 8.22) 8.09 (0.00, 8.00) 8.09 (0.00, 8.02) 8.00 (0.00, 8.07) 8.00 (0.00, 8.01)



Other things to consider with stable isotopes in

food web ecology

• Lipids have with more negative δ¹³C because they are synthesized by the organism, need to extract or correct using modes when C:N > -3.5

- Exoskeletons have more negative $\delta^{13}C$ (zooplankton, benthic invertebrate), need to acid treat when C:N > ~ 3.5
- to acid treat when C:N > 3.5

 Tissue turnover times effect how long isotopes reflect feeding, in general:
 muscle (6-12 months) > liver (1-2 months) > blood plasma (2-3 days)
- Challenges to measuring stable isotopes in small organisms (algae, zooplankton)
- Compound specific stable isotope analysis (fatty acids and amino acids) can provide very specific information – analytically challenging and expensive

Moving forward with stable isotopes in food web ecology

- Use of $\delta^{34}\text{S},\,\delta^{18}\text{O}$ and $\delta^{2}\text{H},$ developing isoscapes
- Methods for analyzing smaller samples
- Need a better understanding of drivers of stable isotope dynamics in lower trophic levels
- More lab studies of DTDFs and turnover rates
- Combining stable isotopes with acoustic telemetry and biomass

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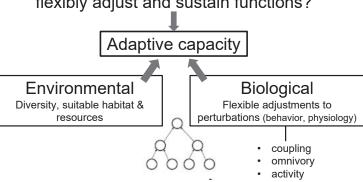
Acknowledgements

- Tim Johnson, Gord Paterson, Scott Rush, Doug Haffner, Scott Colborne, Sarah Larocque, James Mumby, Don Uzarski, Annie Scofield, Tom Stewart
- Funding: NSERC, CRC, GLFC, OMNRF

The adaptive capacity of lake food webs Bailey McMeans¹, Kevin McCann²

¹University of Toronto, ²University of Guelph

What structures within complex systems flexibly adjust and sustain functions?



Food webs serve key functions

- energy is transferred through trophic interactions, creates structure
- are not static
- dynamic structures <u>sustain functions</u> in the face of spatial and temporal variation

flexible **behavior**

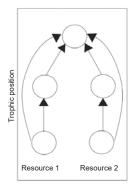
behavior & adaptive capacity

Dynamic food web structures,



Credit: T. Stewart et al. 2018

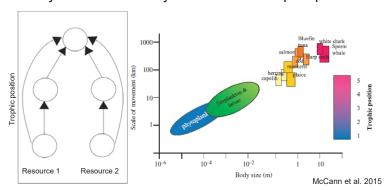
Simple but useful module within real food webs



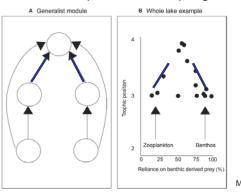
Network of interactions underpinned by common and widespread behaviors.

Highlights the importance of diverse habitats/resources and flexible behaviors.

Body size and mobility increase with trophic position

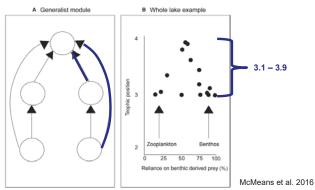


Real food webs are replete with coupling



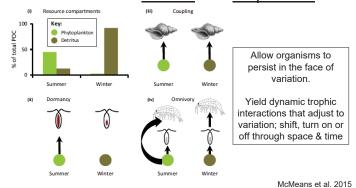
McMeans et al. 2016

Real food webs are replete with coupling and omnivory

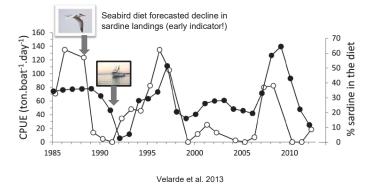


All organisms vary in their activity Multi-Annual Seasonal Junar Jun

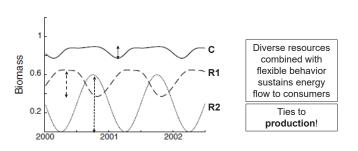
1. Behaviors are <u>flexible</u> and <u>responsive</u>



2. Behaviors are rapid



3. Flexible behaviors are important for stability



Takimoto et al. 2002

Empirical case study



Tyler Tunney, Matthew Guzzo, Emma Bloomfield

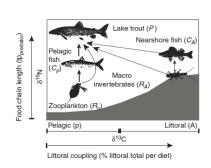


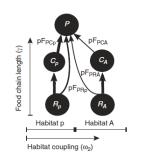




Liset Cruz-Font, Jake Vander Zanden, Paul Blanchfield, Mike Rennie, and many others!

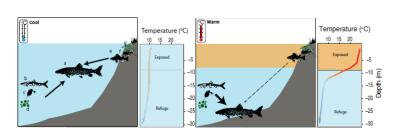
Lake trout food web





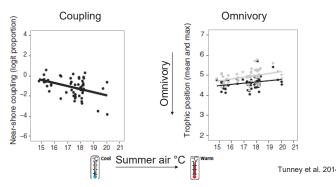
Tunney et al. 2012

Spatial temperature gradient



Tunney et al. 2014

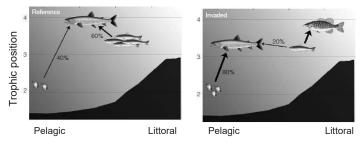
Coupling and omnivory increase in cooler lakes



Coupling and omnivory increase in smaller lakes

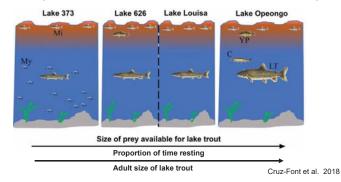
Coupling Omnivory 2.5 1.0 (%) 1.0 (

Coupling decreases in lakes with invasive competitors

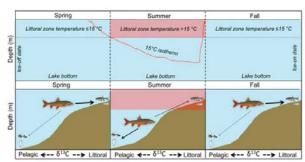


Vander Zanden et al. 1999

Activity increases in lakes with smaller prey



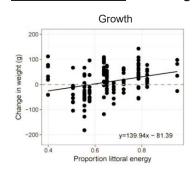
Seasonal temperature gradient



Guzzo et al. 2017

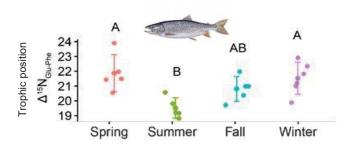
Coupling increases when cooler seasons are long

Coupling Proportion littoral energy logit(y)=0.08x - 2.51 Spring length (d)



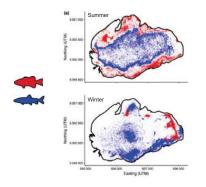
Guzzo et al. 2017

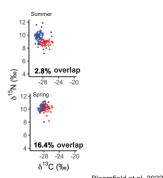
Omnivory increases in <u>warmer seasons</u>



Shipley et al. under review

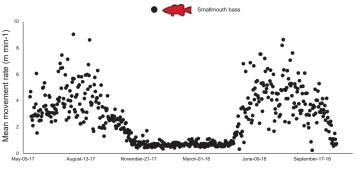
Overlap with competitors increases in cooler seasons





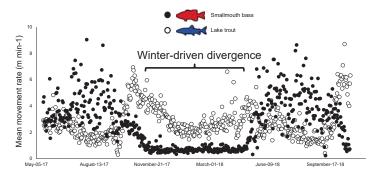
Bloomfield et al. 2022

Competitors decrease activity during cooler seasons



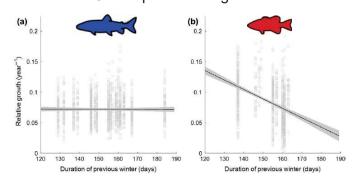
McMeans et al. 2020

Competitors decrease activity during <u>cooler seasons</u>



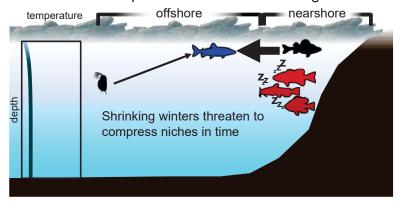
McMeans et al. 2020

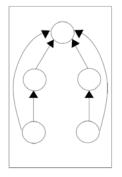
Consequences for growth



McMeans et al. 2020

Consequences for climate warming





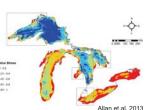
Summary

Common behaviors structure dynamic food webs:

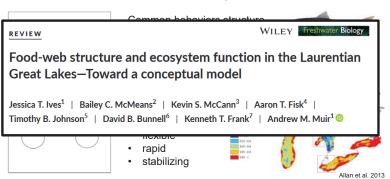
- coupling
- omnivory
- activity

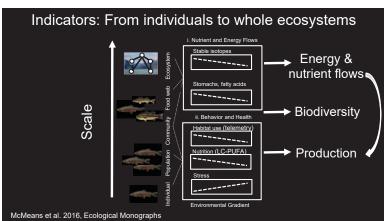
Contribute to adaptive capacity:

- flexible
- rapid
- stabilizing



Summary





Workshop: Towards Multi-scale Management of Great Lakes Ecosystems in a Changing World

Participants:

Great Lakes managers and scientists

Goals

Synthesize efforts, apply biotracers to management? Incorporate relatively simple food web modules into models?

Interested?

bailey.mcmeans@utoronto.ca ksmccann@uoguelph.ca





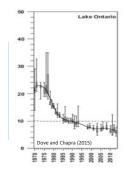




Core Project Team Affiliations



Rationale



- Eutrophication was a major issue when the original Great Lakes Water Quality Agreement (GLWQA) was singed.
- Efforts to manage TP have been generally successful
- TP targets have been met or exceeded in many
- Most of the Great Lakes are now experiencing an oligotrophication

Rationale

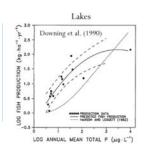


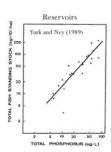
- In some locations, eutrophication and algal blooms have again been an issue
- Especially Lake Erie and some other nearshore areas
- The renewed GLWQA lowered TP targets for Lake Erie; committed to reviewing TP targets for the other lakes
- It is well established that fish production is related to ecosystem productivity
- In theory, these relationships can help to inform management of the consequences of TP reductions on fish and fisheries

Examples of Published TP Relationships

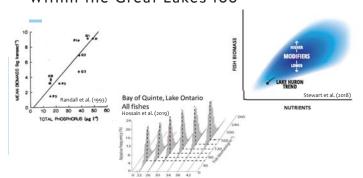
Dependent variable	Independent variable	Γ²	Location (N)	Source	
Fish and primary					
Fish yield	Phytoplankton standing stock	0.84	Natural Lakes, Northern-hemispheres (19)	Oglesby (1977)	
Sport fish yield	Chlorophyll a	0.83	U.S. lakes & reservoirs (25)	Jones & Hoyer (1982)	
Fish and phosphorus					
Fish yield	Total Phosphorus	0.84	North-temperate lakes (43)	Hanson & Leggett (1982)	
Fish standing stock	Total Phosphorus	0.84	Southern Appalachian reservoirs (21)	Ney et. al., (1990)	
Phytoplankton and phosphorus					
Phytoplankton	Phosphorus	0.97	North America (30)	Dillon and Rigler (1974)	
Phytoplankton	Phosphorus	0.88	Worldwide [38°S-75°N] (81)	Schindler (1978)	

Fish Production and Biomass as a Function of Total Phosphorus





Within the Great Lakes too

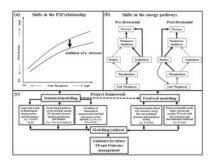


Objectives

- Examine the phosphorus to fishes (P2F) relationship across the Great Lakes basin.
- Relate nutrient reduction scenarios to fish biomass and harvest levels.
- Compare the P2F relationships to examine how energy dynamics may have changed.
- Evaluate how current and future TP reductions may affect sustainable biomass and yields.
- Project sustainable yields given TP, food web structures and environmental conditions.

Project Framework

- Compile data.
- Statistical analyses to examine the P2F relationships.
- Ecopath with Ecosim models to explore scenarios.
- Modelling synthesis.



P2F Fish Data Contributors



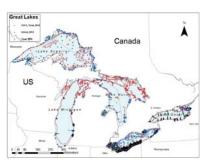


P2F Phosphorus Data Contributors

Agency	Contributor
ECCC	Alice Dove
ECCC	Violeta Richardson
USEPA	Anne Scofield
USEPA	Elizabeth Hinchey
USEPA	Eric Osantowski
USEPA	Kenneth Klewin
USEPA	Santina Wortman

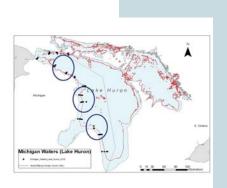
Locations Of Compiled Fish Data

- 13 partner agencies:
 CBFS, ECCC, InDNR, INHS, MDNR, MnDNR, NYSDEC, ODNR, OMNRF, PFBC, USEPA, USGS, WDNR
- 33 collaborators
- 22 gill net surveys
- 17 trawl surveys
- TP data from EPA, GLAHF, and ECCC



Statistical Models

- Gillnet survey data from Michigan waters (MDNR)
- · Three nearshore sites:
- · Hammond Bay
- Thunder Bay
- Port Austin
- TP data from USEPA and ECCC
- Three simple regression models



Regression Models

$$\begin{split} & \textbf{Model 1 (M1): simple P2F model} \\ & \text{Ln[Fish B]}_t = \alpha_{\text{O[R]}} + \alpha_{\text{S}} \text{std}[\text{TP}]_t \\ & \text{error} \sim G(0.001, 0.001) \\ & \text{k} = 1 \text{ for ECCC and 2 for USEPA TP data} \end{split}$$

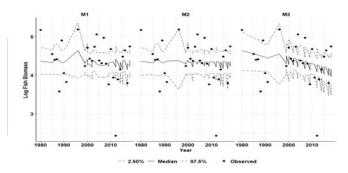
Model 2 (M2): test the effect of TP data source on P2F relationship

$$\begin{split} & \text{Ln[Fish B]}_t = \alpha_{0[k]} + \alpha_{1[k]} \text{std[TP]}_t \\ & \text{error} \sim \text{G(0.001, 0.001)} \\ & \text{k} = 1 \text{ for ECCC and 2 for USEPA TP data} \end{split}$$

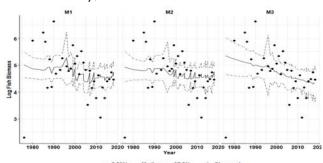
Model 3 (M3): test the effect of time on the prediction of fish biomass

$$\begin{split} &\text{Ln[Fish B]}_t = \alpha_{0[k]} + \alpha_1 \text{std}[\text{TP}]_t + \alpha_2 (t - t_0) \\ &\text{error} \sim \text{G}(0.001, 0.001) \\ &\text{k} = 1 \text{ for ECCC and 2 for USEPA TP data} \end{split}$$

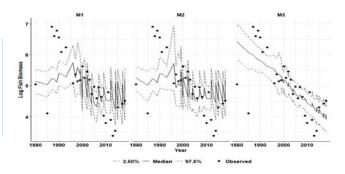
Preliminary Results: Hammond Bay, Nearshore Lake Huron



Preliminary Results: Thunder Bay, Nearshore Lake Huron



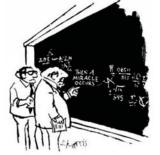
Preliminary Results: Port Austin, Nearshore Lake Huron



Preliminary Results: Nearshore Lake Huron

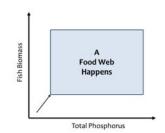
- Source of the TP data (EPA vs. ECCC) is not determining the P2F relationship
- Weak relationships between fish biomass and TP
- TP alone cannot explain the declines in fish biomass
- The time parameter suggests additional drivers of fish biomass need to be explored





"I think you should be more explicit here in step two."

Ecosystem Models

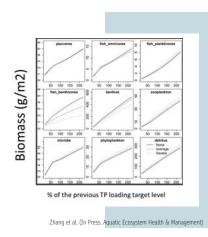


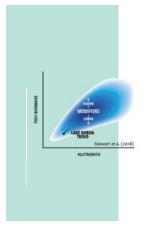
Ecopath with Ecosim (EwE) Models

Lake	Trophic status	Dominant predators	Ecopath year	Ecosim years	Model Status
Lake Michigan (LM)	Oligotrophic	Salmonids	1994-1998	1994-2015	Rutherford et al. (2021)
Lake Huron (LH)	Oligotrophic	Salmonids	1981-1984	1984-2006	Kao et al. (2016)
Lake Erie (LE)	Mesotrophic	Percids, Trout	1999-2001	1999-2010	Zhang et al. (2016)
Lake Ontario (LO)	Mesotrophic	Percids, Salmonids	2003-2008	2005-2018	Developed
Muskegon Lake, LM (ML)	Eutrotrophic	Percids	2001-2006	2003-2010	Developed
Saginaw Bay, LH (SB)	Eutrotrophic	Percids	1988-1990	1990-2010	Kao et al. (2014)
Western Lake Erie (WLE)	Eutrotrophic	Percids	1999-2001	1999-2016	Developed
Central Lake Erie (CLE)	Mesotrophic	Percids	1996-1998	1996-2020	Zhang et al. (in press)

Central Basin, Lake Erie

- Changing nutrients affects biomass of all model food web groups
- Some groups respond more strongly to nutrient changes than other groups
- Hypoxia has negative effects on some groups (e.g. benthos and benthivorous fishes) but had positive effects on some groups (e.g. plankton and planktivorous fishes).
- Nutrients have greater food web effects than hypoxia





Next Steps

- Complete statistical analyses across Great
 Lakes sites
- Add stressors as potential drivers
- Statistical comparisons across sites
- Models synthesis

Acknowledgements

- All the Great Lakes agencies who have invested in collecting both the fisheries and water quality data
- All the data contributors who have engaged in providing data and working with us on the data
- Great Lakes Fishery Commission for funding



Ministry of Natural Resources and Forestry

Managing multiple objectives

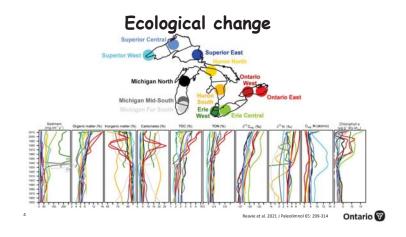
Tim Johnson Glenora Fisheries Station



Ontario 🗑

Allan et al. (2013) PNAS 110:32-377





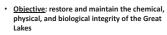
Great Lakes Governance

- · Two countries (Canada and U.S.)
- · 2 provinces & 8 states
- · Dozens of First Nations and tribes
- · Thousands of local governments















- Objective: promote science and establish working relationships among the players to improve and perpetuate the fishery and combat sea lampreys
- Principles: consensus; accountability; information sharing; and ecosystem-based management





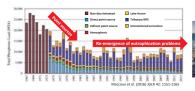








- Objective: restore and maintain the chemical, physical, and biological integrity of the Great Lakes
- Princip program understa em; eliminate Objective relations working ve and
- Principles: consensus; accountability; information sharing; and ecosystem-based management













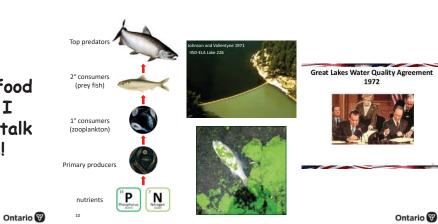
- Revised GLWQA (2012) revisit P loadings and targets
- Concern by fishery managers P reduction would further impact fisheries

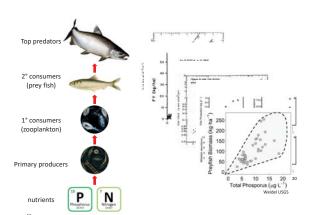


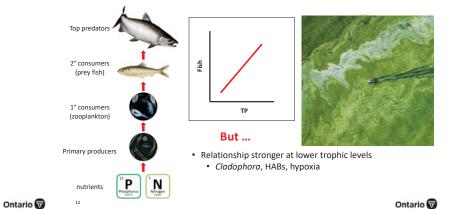


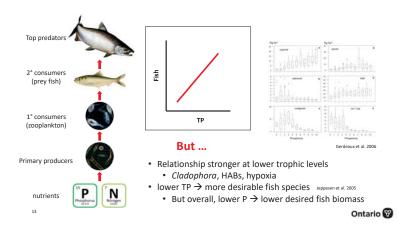


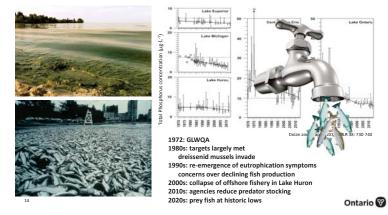
Wait, this is a food web workshop - I should probably talk about food webs!

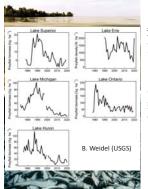


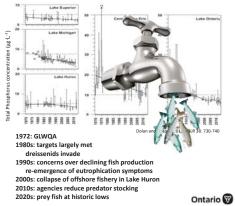


















Renewed GLWQA Update P targets Develop action plans

<u>By 2016</u> *Lake Erie* Review and revise P concentration and loads

<u>By 2018</u> Lake Erie Develop plans

Review and concentration and loads









- 1. Literature review on how P reductions influence upper food web
- 2. Update basin wide time series used by Bunnell et al. (2014) to investigate relationship among nutrients, lower food web, and fish production
- 3. Review models to link P to upper food web and fishery production

Recommendation Form cooperative monitoring and modelling committee to:

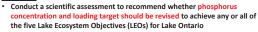
- 1. Review ecosystem forecasting science
- 2. Identify critical data and foster exchange to support integrative decision support 3. Evaluate trade-offs and benefits between nutrient management and fisheries
- 4. Promote joint decision making
- 5. Improve public communication w.r.t. consistency and clarity

"Major reductions in upstream nutrient loading to Lake Erie together with stakeholder concerns about increased coastal, nuisance Cladophora growth and declining salmonid biomass present a 'nearly-perfect storm' for both water quality and fisheries managers."

"It is incumbent on the SAB to advise the IJC of this risk that may result from taking a narrow view of the coastal Cladophora issue while neglecting possible impacts on pelagic food webs and ecosystem productivity."







· Hypoxia; nuisance algae; healthy algal community nearshore; cyanobacteria; oligotrophy and healthy ecosystem offshore

<u>Main recommendation:</u> "... cannot recommend a revision of the Lake Ontario phosphorus concentration objective or phosphorus loading target at this time."

- Maintain offshore water quality monitoring
 Improve understanding of inputs, distribution and fate of phosphorus
- Advance understanding of extent, drivers, and processes contributing to nuisance benthic algae in
- Establish modelling working group to inventory, review, and evaluate models, including nearshore Cladophora dynamics, food webs, and linkages between them
- Continue interdisciplinary collaboration between water quality, food web and fisheries researchers

Are you listening or just waiting to talk?

"... greater cooperation between water quality and fisheries agencies will be essential to maintaining a healthy and valuable fishery in the lakes."

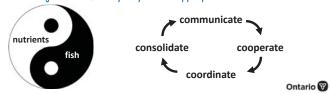
Hecky and DePinto 2020



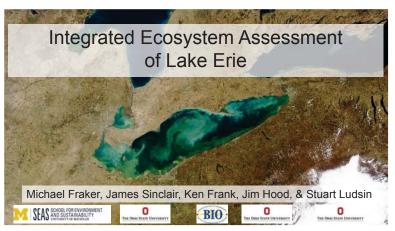


Conclusions

- Change in loading → change in biomass at all trophic levels
 Interacts with other stressors (AIS, climate, population growth) to affect biomass, structure and trophic efficiency
- · Water quality and fish groups are talking
- Scientifically based decision support tools are common ground
- Lakes are dynamic and complex
 - Goals and objectives of today may not be appropriate tomorrow









Ecosystem-based management

An approach that considers interactions among the biological, physical, chemical, and human components of the ecosystem when developing management strategies

Integrated Ecosystem Assessment

The IEA process:

- gives a holistic context to trends in the overall ecosystem and indicators of ecosystem state (e.g., GLFC Fish Community Objectives indicators, LAMP and SOLEC indicators)
- · better explains why trends occur and how they are linked
- sets up management strategy evaluation by giving a strong quantitative basis for model development
- helps guide management and restoration decision-making by identifying possible tradeoffs between ecosystem goals and helps managers avoid surprises from management actions due to ecosystem interconnectedness

Assessing drivers of ecosystem change: the role of temporal perspective



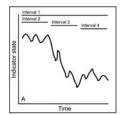
Collaborators

Jim Hood, Stu Ludsin, James Sinclair

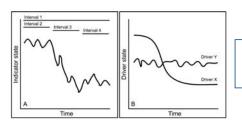
Ken Frank



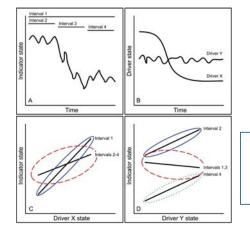




Pattern of an indicator time series can depend on length of time interval and dates covered



Pattern of driver variables may also vary (slow vs. fast drivers)



Relationships between drivers and ecosystem indicators may depend on chosen perspective of analysis

Western Lake Erie watershed is ~70% agricultural











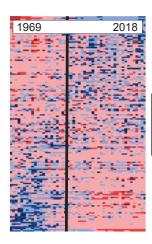










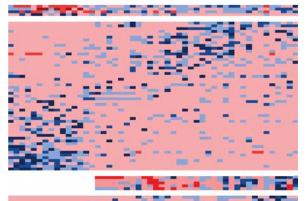


Increasing: soybean agriculture, charter boat fishing licenses, AMO, surface temperature, water clarity, white perch, round goby, lesser scaup, cormorants

Red = below long-term mean
Blue = above long-term mean

Darker shades are larger anomalies

Decreasing: ice cover, bottom oxygen, gizzard shad, yellow perch, carp, pelagic:benthic ratio of forage fish





Anthropogenic drivers (particularly soybean ag) are best long-term predictors, but physicochemical drivers become better in shorter intervals

	1969-2018	<u>1984-2018</u>	1999-2018
Bird abundances	Agriculture (0.13)	Meteorology (0.11)	Meteorology (0.08)
	River Discharge (0.09)	Lake Trophic (0.08)	Fishing (0.05)
	Meteorology (0.07)	Agriculture (0.08)	River Discharge (0.01)
Fish abundances	Agriculture (0.15)	Meteorology (0.12)	Lake Trophic (0.16)
	Fishing (0.15)	Lake Trophic (0.07)	Meteorology (0.08)
	Meteorology (0.03)	Agriculture (0.05)	Fishing (0.07)
Fish-derived ecosystem indicators	Fishing (0.16)	Meteorology (0.07)	Lake Trophic (0.17)
	Agriculture (0.15)	Agriculture (0.07)	Meteorology (0.09)
	Meteorology (0.03)	Lake Trophic (0.05)	Fishing (0.02)
Zooplankton abundances	NA	Meteorology (0.08) Agriculture (0.06) Lake Trophic (0.05)	Lake Chemistry (0.15) Meteorology (0.08) Fishing (0.07)

Fishes

Zooplankton

Birds

Anthropogenic drivers (particularly soybean ag) are best long-term predictors, but physicochemical drivers become better in shorter intervals

1969-2018 1984-2018 1999-2018 Agriculture (0.13) Meteorology (0.11) Meteorology (0.08) Bird abundances River Discharge (0.09) Lake Trophic (0.08) Fishing (0.05) Meteorology (0.07) River Discharge (0.01) Agriculture (0.08) Agriculture (0.15) Meteorology (0.12) Lake Trophic (0.16) Fish abundances Fishing (0.15) Lake Trophic (0.07) Meteorology (0.08) Meteorology (0.03) Fishing (0.07) Agriculture (0.05) Fishing (0.16) Meteorology (0.07) Lake Trophic (0.17) Fish-derived Agriculture (0.15) Meteorology (0.09) Agriculture (0.07) ecosystem indicators Meteorology (0.03) Lake Trophic (0.05) Fishing (0.02) Lake Chemistry (0.15) Meteorology (0.08) Zooplankton abundances NA Agriculture (0.06) Meteorology (0.08) Lake Trophic (0.05) Fishing (0.07)

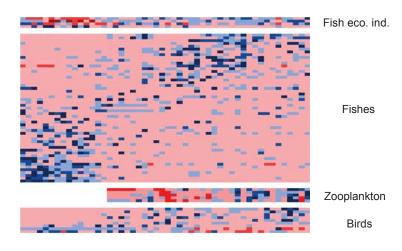
Anthropogenic drivers (particularly soybean ag) are best long-term predictors, but physicochemical drivers become better in shorter intervals

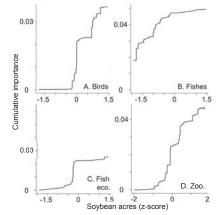
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Zooplankton abundances	NA	Meteorology (0.08) Agriculture (0.06) Lake Trophic (0.05)	Lake Chemistry (0.15) Meteorology (0.08) Fishing (0.07)

Slower and faster drivers co-occur





Soybean agriculture influences all three taxa, but at different thresholds

50-year mean = 3.2 million acres (standard deviation = 0.6)

Management implications:

- 1. Need to be clear about long-term vs. short-term planning
- 2. Greater uncertainty may exist if only short time series available
- Following a tiered approach (multiple time intervals and indicator groups) will provide the most complete understanding of the key drivers of patterns in ecosystem state and their most relevant operative time scales

Future work:

- 1. Modeling of focal species (walleye, yellow perch, whitefish) using structural equation modeling and fuzzy cognitive mapping
- 2. Effects of spatial scale (e.g., Heim et al. 2021)
- 3. Comparisons of driver-response relationships across all lakes

Tradeoffs between HABs, hypoxia, and fisheries related to nutrient loading targets



Collaborators

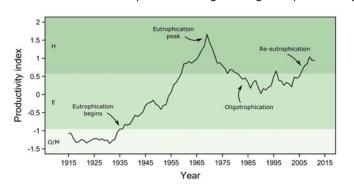
Jim Hood, Stu Ludsin, James Sinclair

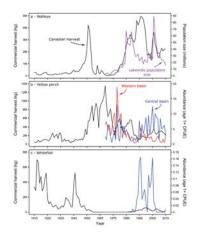
Euan Reavie





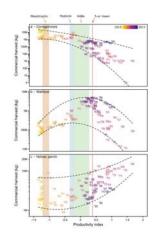
Western Lake Erie has experienced large changes in productivity





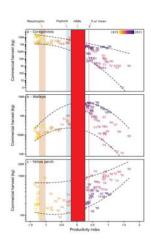
We have harvest data back to 1800s, but fishery-independent surveys only back to 1960s

However, both time series are reasonably well correlated during recent decades



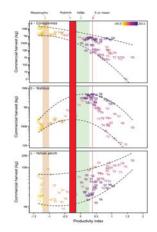
Analysis suggests potential tradeoffs with nutrient loading targets:

Walleye harvest is optimized under somewhat eutrophic conditions, yellow perch under hypereutrophic, and whitefish under mesotrophic



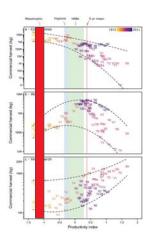
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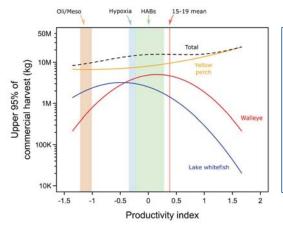
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Analysis suggests potential tradeoffs with nutrient loading targets:

Walleye harvest is optimized under somewhat eutrophic conditions, yellow perch under hypereutrophic, and whitefish under mesotrophic



Total harvest of the three species together is more stable across a range of trophic conditions (although it does decline with lower lake productivity)



Investigating Food Webs: State of Knowledge and Investigative Approaches

Tom Stewart

Brian Weidel, USGS

Dick van Oevelen, Royal Netherlands Institute for Sea
Research

Investigating Food Webs: Sate of Knowledge and Investigative Approaches

Acknowledgements

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Great Lakes Fishery Commission
Cooperative Institute for Great Lakes Research

Steering Committee and Advisors

Nicholas Boucher, Aaron Fisk, Roger Knight, Doran Mason, Kevin McCann, Bailey McMeans, Lars Rudstam, Ed Rutherford, Heidi Swanson

Workshop Hosts

Cornell Biological Field Station

Investigating Food Webs: Sate of Knowledge and Investigative Approaches

What's this all about?

nvestigating Food Webs: Sate of Knowledge and Investigative Approaches





Objectives

- 1) Review and share the current state of food web investigative methods,
- 2) Determine food web-scale fisheries management information needs and possible investigative approaches, and
- 3) Develop collaborative study designs and proposals for potential funding addressing food web knowledge gaps relevant to fisheries management information needs.

Seminars Nov 3 & 10

Workshop Nov 14-16

Investigating Food Webs: Sate of Knowledge and Investigative Approaches

Webinar I (Nov 3rd)

- Management information needs
 - Need for simple food web metrics
 - Dealing with critical unknowns (e.g. Goby production)
 - Linking food web "re-building" to habitat restoration
 - Linking LTL dynamics to fish recruitment
 - Consumptive demand on prey fish remains important
 - Linking stock assessment models to food web models
 - Phosphorus & Fish policy and scientific challenge
- Novel Approaches
 - Continued development of isotope applications
 - Linking isotopes and sub-models (interaction modules) into analysis to develop new metrics
 - Integrated Ecosystem Assessment

Investigating Food Webs: Sate of Knowledge and Investigative Approaches

Webinar II (Today)

- Diverse set of speakers
 - More emphasis on methods and modelling
 - Management and policy relevant
 - Special welcome to colleagues from "across the pond"
 - Dick van Oevelen Royal Netherlands Institute for Sea Research
 - Benjamin Planque- Institute for Marine Research, Norway
 - Special welcome to Robert Ulanowicz, University of Florida
 - Pioneering leader in ecological theory and network analysis
 - AND--Some other guys! (Myself, Doran Mason, Ed Rutherford and colleagues)

Investigating Food Webs: Sate of Knowledge and Investigative Approaches

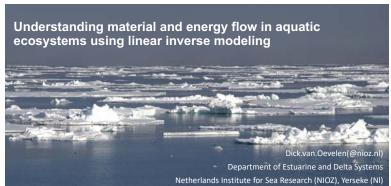
House keeping

- Presentation timer
- Presentation recordings
- Questions during seminar please raise your hand
- Workshop registration closed

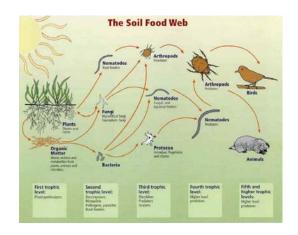
To request recorded talks nboucher@glfc.org

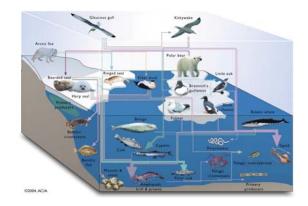
Any other questions tomstewart54321@gmail.com

Investigating Food Webs: Sate of Knowledge and Investigative Approaches







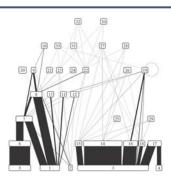




Connectance / binary food web, food web links 0 (absent) or 1 (present) (e.g. Pimm et al. Nature 1991)



Chesapeake Bay mesohaline ecosystem



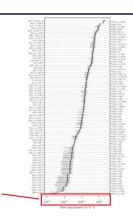
Banašek-Richter et al. Ecology 2009

Food webs are not binary



Cold-water coral ecosystem Van Oevelen et al. L&O 2009

Flow magnitudes range >6 orders of magnitude



From connectance to quantitative food webs

Is it important to account for the differences in the magnitude of food web flows? Yes! (at least in some cases...)

- Weak trophic interactions in long trophic loops dampens the destabilizing effect of these long loops (De Ruiter et al. 1995; Neutel et al. 2002)
- Food webs are structured such that top predators couple energy channels that differ in productivity and turnover rate (Rooney et al. 2006)
- Decoupling of C (food quantity) and P (food quality) uptake by zooplankton fostered omnivory in a lake food web (Gaedke et al. 2002)
- Food web descriptors, e.g. trophic level/omnivory, are more robust when magnitude of link is considered (Banašek-Richter et al. 2004)

A major obstacle in making the transition from connectance to energetic food webs is the imbalance between food web complexity and data availability/uncertainty

Two solutions to address this imbalance:

Increase data availability

Optimize data integration using models

Increase data availability

Data used to resolve fluxes in food webs:

- Abundance / biomass
- Stoichiometric composition (C, N, P, +)
- Organism physiology, e.g. assimilation efficiency, growth efficiency
- Process rates, e.g. primary production, grazing, (community) respiration
- Feeding relations, e.g. gut contents, relation based on size and/or functional type, ¹³C/³⁴S stable isotopes, fatty acid composition, ¹³C/¹⁵N isotope tracer experiments
- Trophic level, e.g. ¹⁵N-bulk or ¹⁵N-amino acid stable isotopes
- ...

Linear inverse modelling

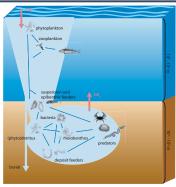
A major obstacle in making the transition from connectance to energetic food webs is the imbalance between food web complexity and data availability/uncertainty Two solutions to address this imbalance: Increase data availability Optimize data integration using models

prediction • parameters • initial conditions forward model data analysis • observations on variables in time/space intime/space inverse model • trajectories of variables in time/space • parameter values • initial conditions • unmeasured processes

Note that I refer to data analysis models, not conceptual models

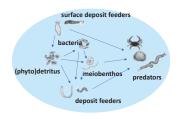
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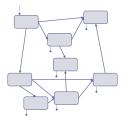
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Soetaert & Van Oevelen et al. Oceanogr. 2009

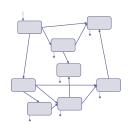
Mass balances couple the components in a food web

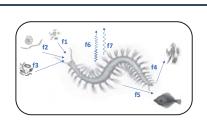




Soetaert & Van Oevelen et al. Oceanogr. 20

Mass balances





$$\frac{dB}{dt} = \underbrace{\text{f1+f2+f3}}_{\text{ingestion}} - \underbrace{\text{f4}}_{\text{defecation}} - \underbrace{\text{f5}}_{\text{predation}} - \underbrace{\text{f6}}_{\text{basal}} - \underbrace{\text{f7}}_{\text{growth}}$$

Soetaert & Van Oevelen et al. Oceanogr. 2009

Linear inverse modelling



$$\frac{dB}{dt} = \underbrace{\mathbf{f1} + \mathbf{f2} + \mathbf{f3}}_{\text{ingestion}} - \underbrace{\mathbf{f4}}_{\text{defecation}} - \underbrace{\mathbf{f5}}_{\text{predation}} - \underbrace{\mathbf{f6}}_{\text{basal respiration}} - \underbrace{\mathbf{f7}}_{\text{growth respiration}}$$

 $\begin{array}{c|c} \text{Linear} & \frac{dx_j}{dt} = \overline{\sum flow_{i \to j}} - \overline{\sum flow_{j \to k}} \\ \text{Steady-state} & 0 = \sum f(x, pl, \ldots) - \sum f(y, p2, \ldots) \\ \\ \text{Dynamic} & (\text{mechanistic} + \frac{dx}{dt} = \sum f(x, pl, t, \ldots) - \sum f(y, p2, t, \ldots) \\ \\ \text{Topical initial position} & \frac{dx}{dt} = \sum f(x, pl, t, \ldots) - \sum f(y, p2, t, \ldots) \\ \\ \text{Topical initial position} & \frac{dx}{dt} = \sum f(x, pl, t, \ldots) - \sum f(y, p2, t, \ldots) \\ \\ \text{Topical initial position} & \frac{dx}{dt} = \sum f(x, pl, t, \ldots) - \sum f(y, p2, t, \ldots) \\ \\ \text{Topical initial position} & \frac{dx}{dt} = \sum f(x, pl, t, t, \ldots) - \sum f(y, p2, t, t, \ldots) \\ \\ \text{Topical initial position} & \frac{dx}{dt} = \sum f(x, pl, t, t, \ldots) - \sum f(y, p2, t, t, \ldots) \\ \\ \text{Topical initial position} & \frac{dx}{dt} = \sum f(x, pl, t, t, \ldots) - \sum f(y, p2, t, t, \ldots) \\ \\ \text{Topical initial position} & \frac{dx}{dt} = \sum f(x, pl, t, t, \ldots) - \sum f(y, p2, t, t, \ldots) \\ \\ \text{Topical initial position} & \frac{dx}{dt} = \sum f(x, pl, t, t, \ldots) - \sum f(y, p2, t, t, \ldots) \\ \\ \text{Topical initial position} & \frac{dx}{dt} = \sum f(x, pl, t, t, t, \ldots) - \sum f(y, p2, t, t, \ldots) \\ \\ \text{Topical initial position} & \frac{dx}{dt} = \sum f(x, pl, t, t, t, \ldots) - \sum f(y, p2, t, t, \ldots) \\ \\ \text{Topical initial position} & \frac{dx}{dt} = \sum f(x, pl, t, t, t, \ldots) - \sum f(y, p2, t, t, \ldots) \\ \\ \text{Topical initial position} & \frac{dx}{dt} = \sum f(x, pl, t, t, \ldots) - \sum f(y, p2, t, t, \ldots) \\ \\ \text{Topical initial position} & \frac{dx}{dt} = \sum f(x, pl, t, t, \ldots) - \sum f(x, p2, t, t, \ldots) \\ \\ \text{Topical initial position} & \frac{dx}{dt} = \sum f(x, p2, t, t, t, \ldots) - \sum f(x, p2, t, t, \ldots) \\ \\ \text{Topical initial position} & \frac{dx}{dt} = \sum f(x, p2, t, t, \ldots) - \sum f(x, p2, t, t, \ldots) \\ \\ \text{Topical initial position} & \frac{dx}{dt} = \sum f(x, p2, t, t, \ldots) - \sum f(x, p2, t, t, \ldots) \\ \\ \text{Topical initial position} & \frac{dx}{dt} = \sum f(x, p2, t, t, \ldots) - \sum f(x, p2, t, t, \ldots) \\ \\ \text{Topical initial position} & \frac{dx}{dt} = \sum f(x, p2, t, t, \ldots) - \sum f(x, p2, t, t, \ldots) \\ \\ \text{Topical initial position} & \frac{dx}{dt} = \sum f(x, p2, t, t, \ldots) - \sum f(x, p2, t, \ldots) - \sum f$

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$$\frac{dB}{dt} = \underbrace{f1 + f2 + f3}_{\text{ingestion}} - \underbrace{f4}_{\text{defecation}} - \underbrace{f5}_{\text{predation}} - \underbrace{f6}_{\text{basal resolution resolution}} - \underbrace{f7}_{\text{growth}}$$

$$\frac{dx_{j}}{dt} = \overbrace{\sum flow_{i \to j}}^{input} - \overbrace{\sum flow_{j \to k}}^{output}$$

Dynamic

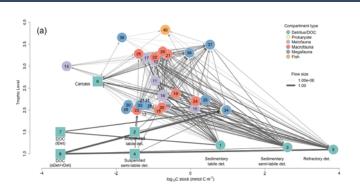
transient)

Steady-state (mechanistic)
$$0 = \sum f(x, pl,...) - \sum f(y, p2,...)$$
 Dynamic (mechanistic + transient)
$$\frac{dx}{dt} = \sum f(x, pl, t,...) - \sum f(y, p2, t,...)$$

$$\frac{dB}{dt} = \underbrace{\text{f1} + \text{f2} + \text{f3}}_{\text{ingestion}} - \underbrace{\text{f4}}_{\text{defecation}} - \underbrace{\text{f5}}_{\text{predation}} - \underbrace{\text{f6}}_{\text{basal respiration}} - \underbrace{\text{growth}}_{\text{respiration}}$$

Linear (snap-shot)	$\frac{dx_{j}}{dt} = \sum_{i} flow_{i \to j} - \sum_{j} flow_{j \to k}$			
Steady-state (mechanistic)	$0 = \sum f(x, p1,) - \sum f(y, p2,)$			data
Dynamic (mechanistic + transient)	$\frac{dx}{dt} = \sum f(x, p1, t, \dots) - \sum f(y, p2, t, \dots)$	reality	complexity	availability

A linear inverse model solves the food web flows



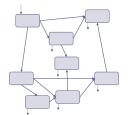
De Jonge et al. Prog Ocean 2020

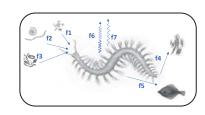
Linear inverse modelling

We often get the critique that (food web) processes are non-linear, so that a linear representation is therefore inherently problematic and unrealistic

Underlying processes are undoubtedly non-linear, however ...

a linear inverse model evaluates mass balances over a 'certain' time interval and while the underlying processes may be highly non-linear, the terms over that time interval are linear





$$\frac{dB}{dt} = \underbrace{f1 + f2 + f3}_{\text{ingestion}} - \underbrace{f4}_{\text{defecation}} - \underbrace{f5}_{\text{predation}} - \underbrace{f6}_{\text{basal respiration}} - \underbrace{f7}_{\text{respiration}}$$

Soetaert & Van Oevelen et al. Oceanogr. 2009

$\frac{dB}{dt} = \underbrace{\text{f1} + \text{f2} + \text{f3}}_{\text{ingestion}} - \underbrace{\text{f4}}_{\text{defecation}} - \underbrace{\text{f5}}_{\text{predation}} - \underbrace{\text{f6}}_{\text{basal respiration}} - \underbrace{\text{growth}}_{\text{respiration}}$



linear functions of the food web flows

numerical data

$$\begin{pmatrix}
1 & 1 & 1 & -1 & -1 & -1 & -1 & \cdots & n \\
\vdots & & & & & & n
\end{pmatrix} \cdot \begin{pmatrix}
f_1 \\
f_2 \\
f_3 \\
\vdots \\
f_n
\end{pmatrix} = \begin{pmatrix}
dB/_{dt} \\
\vdots \\
f_n
\end{pmatrix}$$
food web flows

We need to express data as a linear function of the food web flows

Data collected to resolve food webs:

$$\frac{dB}{dt} = \underbrace{\text{fl} + \text{f2} + \text{f3}}_{\text{ingestion}} - \underbrace{\text{f4}}_{\text{defecation}} - \underbrace{\text{f5}}_{\text{predation}} - \underbrace{\text{f6}}_{\text{respiration}} - \underbrace{\text{f7}}_{\text{respiration}}$$

- Abundance / biomass Biomass data are always used with other data, such as ingestion: $f1 + f2 + f2 = biomass \cdot \frac{Q}{R}$
- $\begin{tabular}{ll} \bf S to ichiometric composition \\ {\bf Stoichiometric coupling of fluxes in a food web:} & f1_N=f1_C \cdot \frac{N}{C_{prey}} \\ \end{tabular}$
- $\begin{array}{l} \blacksquare \ \ \text{Physiology} \\ \text{Ratio of fluxes, e.g. assimilation efficiency:} \ \ \frac{f1+f2+f3}{f1+f2+f3} \frac{f4}{f3} + (\text{NE} \ AE) \cdot f2 + (1-AE) \cdot f3 f4 = 0 \\ \end{array}$
- Process rates:

Single or combination of fluxes: f6 + f7 = respiration

• Feeding relations: Stable isotopes:

 $\delta^{13}C \ \left(\underline{\delta^{1}} \frac{f_{2}^{1} \cdot \delta^{13} S_{1}}{f_{1}^{1}} - \frac{\delta^{13} S_{2}}{f_{1}^{2}} + \frac{f_{1}^{2} f_{1}^{1} \frac{\delta^{13} S_{2}}{f_{2}^{2}} + \frac{\delta^{13} S_{2}}{f_{1}^{2}} - \frac{\delta^{13} S_{2}}{f_{1}^{2}} \right) \cdot f2 + \left(\delta^{13} S_{1} - \delta^{13} C \right) \cdot f3 = 0$ ■ Trophic level: similar to feeding relations, but taking trophic level fractionation into account A LIM in matrix notation

- Topological food web
 - "who eats who"
 - primary sources
- export terms
- Flux measurements
 - biomass-based flows • flow measurements
- respiration Physiological constraints
 - growth rates
 - growth efficiency assimilation efficiency

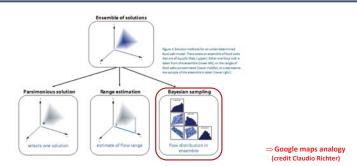
LIM equations

Mass balances and $\mathbf{A} \cdot \mathbf{x} = \mathbf{b}$ data equations

Constraints on and $G \cdot x \ge h$ relations between flows

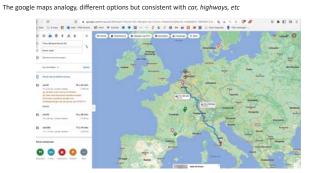


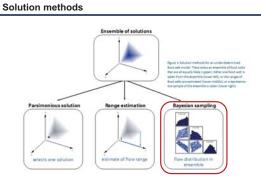
Network analysis



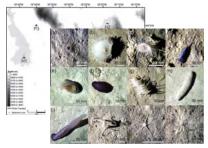
Soetaert and Van Oevelen (2009) Oceanography

Solution methods

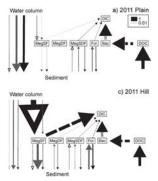




Soetaert and Van Oevelen (2009) Oceanography



Durden et al. 2015 (Prog Oceanogr)



Durden et al. 2017 (L&O)



Quantification of Barents Sea food web to determine food web transfer in phytoplankton- and microbial-dominated food webs

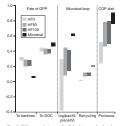
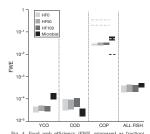


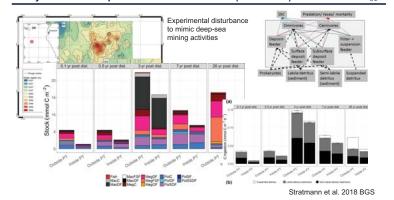
Fig. 1. Differences believes in the 3 herthereces is not webs ingress the present of the present of the present of the group term of the present of the present of the present group immary production (GPF) proportions sinking to the heathers (To bendros) and excreted as disorded organic cruel to the other present of the present of the present of the present of the ratio of better one of the present of the present sensing from recycling (Recycling) as quantified by Finn's cycling index. Englist extent—died to oppose properties of protocos in copaped date (Protocos). Bars represent 90% CI and were calculated by analyzing all Nr. Fasilest deathings of



rig. 4. Took we'l entirelistry (First, Expressed as Indiano calculated by dividing the production of young cod (YCO adult cod (COD), copposits (COP) and ill fish (ALL FISH) to the sum of net primary production and input of copepo biomass by the Atlantic current. Dashed line: FWEs for coppod production found by Bergiund et al. (2007). Bars represent 90% Cl and were calculated by analysing all N realise solutions of the food web model

De Laender et al. 2010 (MEPS)

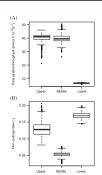
Analysis of LIMs: Experimental disturbance site (Peru Basin)



Analysis of LIMs

Food web differences between head, middle and lower section of the Nazare Canyon (Portugal)

LIM food webs analyzed with network indices



Van Oevelen et al. 2011 (DSRII)

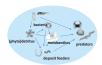
Recap 33



Food webs are not binary



Sampling the solution space generates many possible solutions (google/maps)



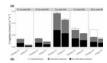
Mass balances in a food web **and** data represented in matrix equations to create a LIM



A LIM provides a single snapshot of a food web



Use R package *LIM* to setup and solve LIM www.rforscience.com



and can be used for time series as repeated LIMs and network analysis

LIM approach to Ecopath mass balance and hypotheses testing

Tom Stewart

Dick van Oevelen

Brian Weidel

Origins of the project

- A workshop in 2017, illustrated the potential for comparative studies of GL food webs to provide fisheries management insights
- The comparison of food metrics (e.g. TTE) relied primarily on Ecopath descriptions of food webs
- While valuable, I thought comparative analysis could benefit from methods that better accounted for uncertainty.
- Linear inverse model (LIM) approach to Ecopath is conceptually straightforward, and offers advantages for both comparative studies, and other applications, but had been rarely applied in Great Lakes context

Advantages of LIM

- Flexibility to deal with different types of data and sources of variation and uncertainty using stochastic computational methods
- Ability to generate objective ensembles of massbalance solutions and associated metrics to facilitate statistical treatment
- Allows for ancillary trophic information, (e.g. isotopes), stoichiometry, spatial flows (e.g. nearshore/offshore) to be reconciled with massbalance

Disadvantages of LIM

- Not as user-friendly or as easy to understand as Ecopath
- Requires coding knowledge and experience
- Requires explicit coding of all system flows and bioenergetics parameters, and constraints
- If observational variation is to be applied as a constraint, must be able to estimate

Objectives

- Demonstrate the potential for LIM in a Great Lakes context by,
 - Designing a LIM coding application that exploits existing Great Lakes Ecopath frameworks and available data sets
 - Basic data requirements are essentially the same
 - Consider how best to account for uncertainty across multiple model inputs
 - Apply LIM in a Great Lakes comparative analysis as "proof of concept"

Methods

- Applied method to existing Ecopath comparative analysis of Lake Ontario offshore food web (Stewart and Sprules, 2011)
 - Updated alewife and dreissenid mussel biomass
- Compared pre-dreissenid (1987-91) to postdreissenid establishment (2001-05)
- Included multiple sources of variation

Principle of mass balance in a food web

Species-group production is either consumed, exported (harvested), or dies and becomes detritus (recycled)

Species-Group Growth = dB/dt = ingestion – predation mortality – defecation – respiration

Food web flows, as defined over space and time, must balance all inputs and outputs under these constraints

Configuring Ecopath in LIM

Ecopath Parameters

LIM Parameters (this study)

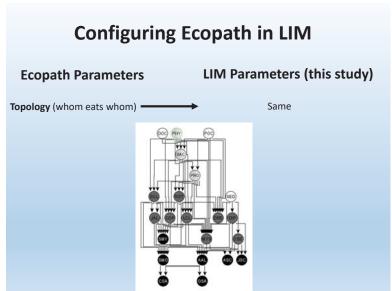
Topology (whom eats whom)

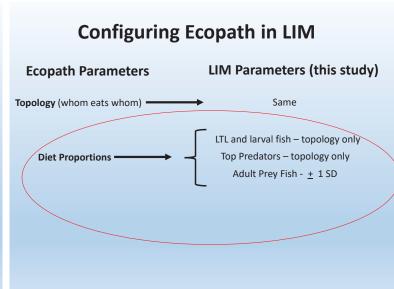
Diet Proportions

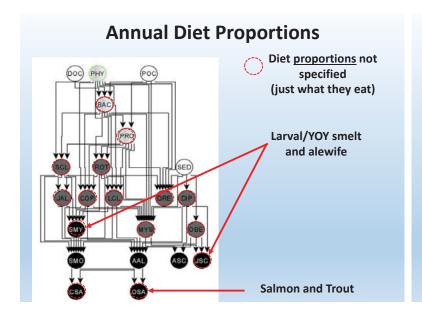
Production = P/B x Biomass

Consumption = Q/B x Biomass

Defecation (% of Consumption), **Respiration** (calculated as a residual component)

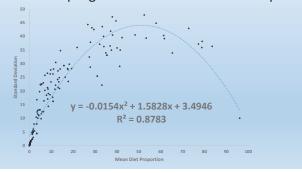






Variation in Adult Prey Fish Diet Proportions

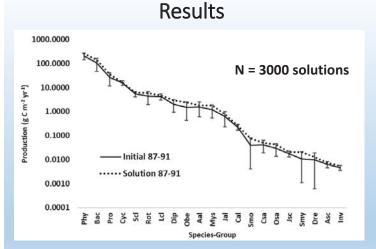
- Determined variance in adult Alewife, Rainbow Smelt and Sculpin diet proportions
- Used independent sets of samples over space and time to develop a generalized mean vs variance pot



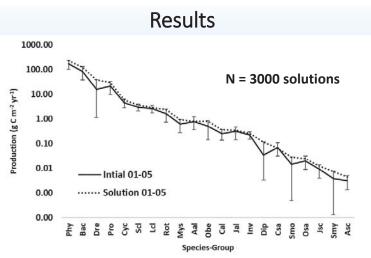
Configuring Ecopath in LIM LIM Parameters (this study) **Ecopath Parameters** Topology (whom eats whom) Same LTL and larval fish – (topology only) Top Predators – (topology only) Adult Prey Fish - + 1 SD Production = P/B x Biomass Production = $P/B \times (Biomass + 2SD)$ Consumption = Production Consumption = Q/B x Biomass x GE (range) Defecation (% of Consumption), Defecation and respiration Respiration (calculated as included in GE a residual component)

Achieving mass-balance

- General LIM
 - Numerous options (parsimonious, range, Monte Carlo randomization) while maintaining internal consistency and user-specified constraints
- This study (LIM)
 - Monte Carlo randomization with constraining bounds based on observed variation in species-specific diet proportions (or unconstrained topology), production, and consumption.



Initial and mean solution production values for 1987-91 time-period. Error bars are the initial production bounds. The species-groups for each time-period are plotted in order from highest solution production to lowest solution production.



Initial and mean solution production values for 2001-05 time-period. Error bars are the initial production bounds. The species-groups for each time-period are plotted in order from highest solution production to lowest solution production.

Hypotheses

 The overall food web trophic transfer efficiency declined between time periods due to a redirection of primary production and detritus to invasive mussels

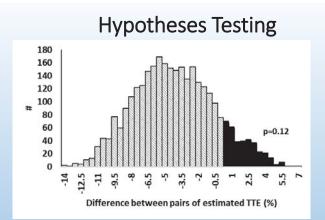
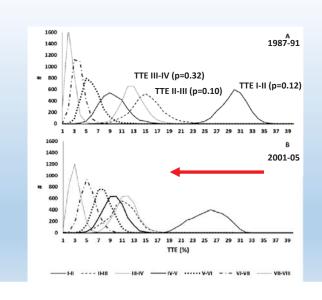
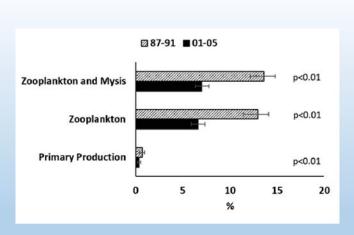


Illustration of how test-statistic probabilities are determined for the difference between trophic transfer efficiency from trophic levels I-II for 1987-1991 and 2001-2005. The dark bars are the number of times independent pairs of 3000 solutions had a difference ≥ 0 (m). The p-values is the probability that the difference ≥ 0 (null hypotheses) occurred by chance, calculated as m/3000.



Hypotheses

 The decline in nutrients, and increased flows of primary production (PP) to dreissenid mussels reduced PP flows to Alewife and production efficiency declined

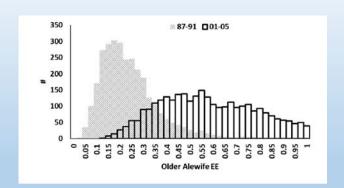


Median older alewife production efficiency for each timeperiod. Error bars are the 5% and 95% percentiles (N=3000).

Hypotheses

 The risk to alewife over-consumption and potential predator response (measured as Alewife ecotrophic efficiency), increased, due to system wide production declines and invasive species establishment

Shift in the risk of alewife over-consumption



Distribution of older alewife ecotrophic efficiency (EE) values for each time-period (N=3000).

Caveats

- This is a prototype model, and it needs some tweaking
 - GEs were too liberal for some species and need to use more system-specific or Great Lake specific data from existing bioenergetics models or a general literature review
 - Need to consider some constraints on un-bounded diet proportions
 - However, a cursory look indicated that un-bounded diets associated with solutions were very reasonable and realistic

Conclusions

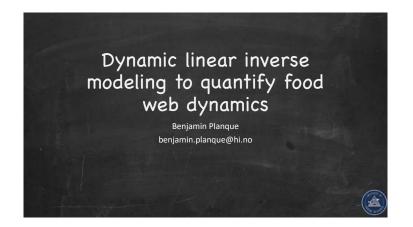
- Coding of Ecopath structures and data sets into LIM is relatively straightforward
- Flexible approaches for handling uncertainty
- Unknowns can be objectively dealt with (e.g. diet proportions, uncertain levels of production, bioenergetics)
- Results in objective ensemble of mass-balance solutions
 - Facilitates comparative statistical approaches (CI, probability distributions, hypotheses testing)
- User-defined metrics can be tabulated and evaluated

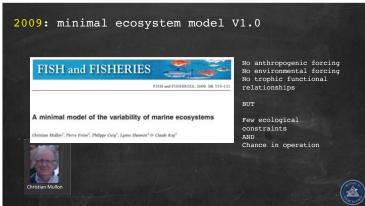
Acknowledgements

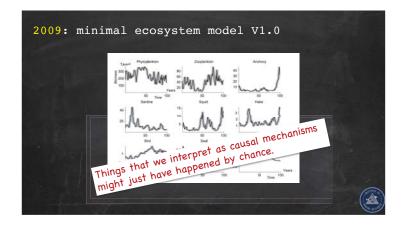
 GLFC provided funding for this project through the Technical Assistance for Fisheries Research Program (TARF)

Future Applications

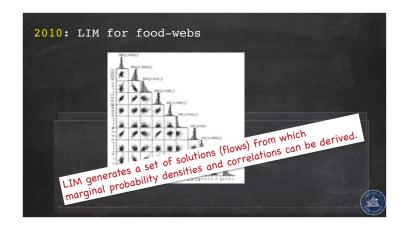
- Design comparative Great Lake food web studies that exploit existing Ecopath data sets (likely close to 20 such data sets available)
- Include isotope values flow process and constraints in models
 - i.e. diets must be consistent with measured isotope values
 - increases confidence in outcomes and decrease the need for difficult and variable diet studies
- Use it to explore hypothetical data deficient food webs
 - Explore pre-colonization endemic food web function
 - Predict the potential range of food web responses to introduced species
- Flexibility to link other sub-model constraints into the mass balance models
 - Nearshore to offshore exchanges
 - Nearshore mussel dynamics could be described as a separate model and linked to more general food web model
 - Bailey et al. predator-prey "interaction modules" or variations could be included as constraints

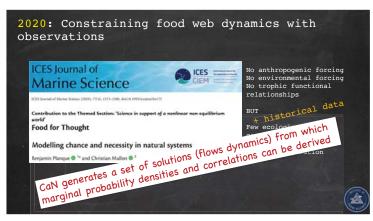


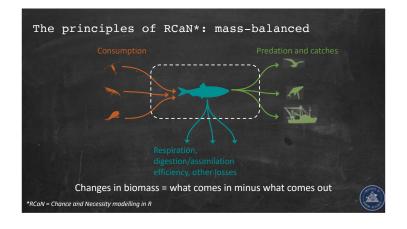


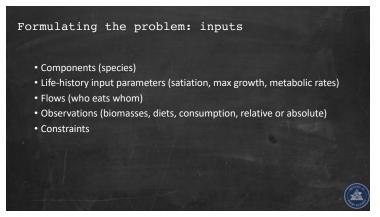


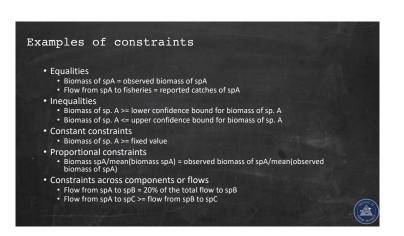






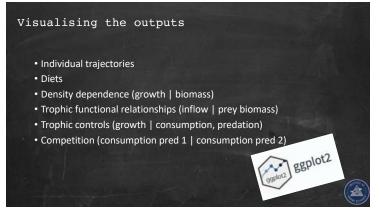


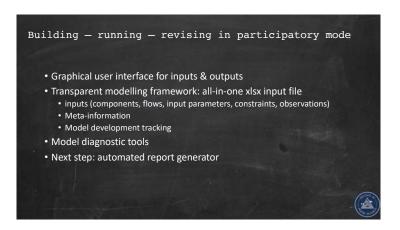


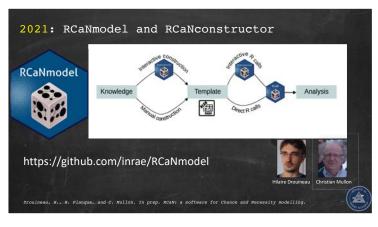


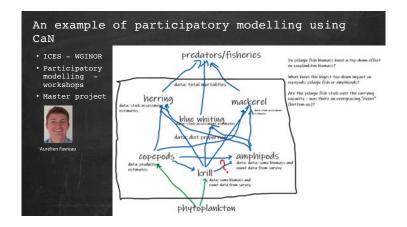


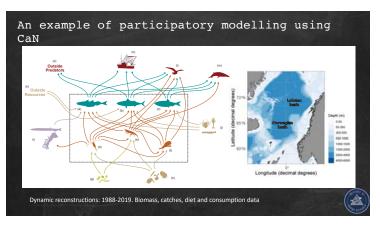


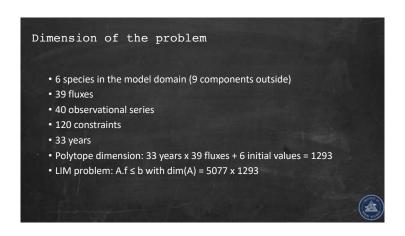


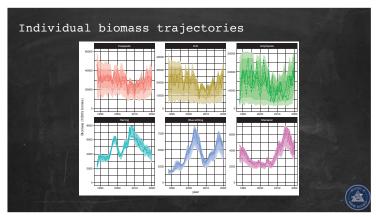


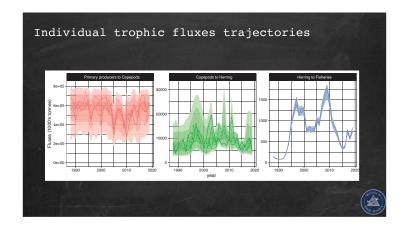


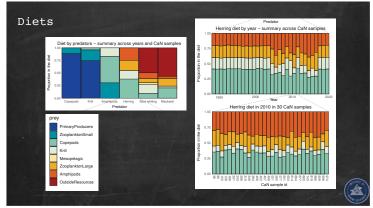


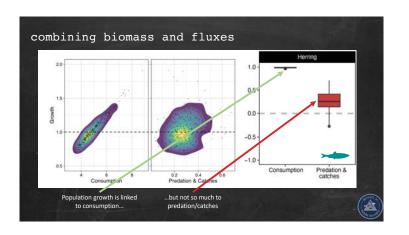


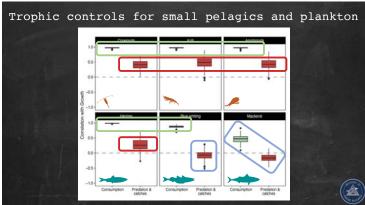


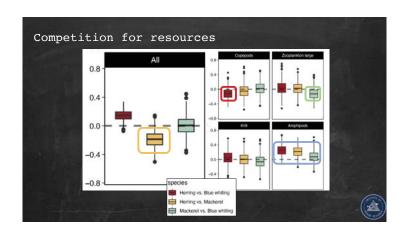




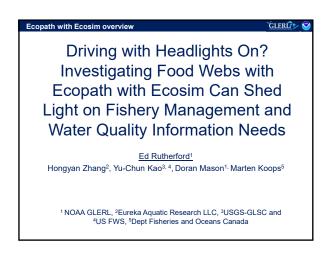


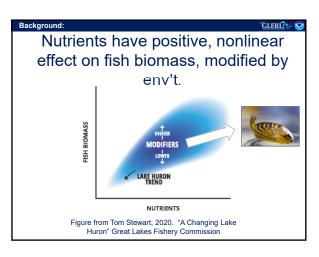


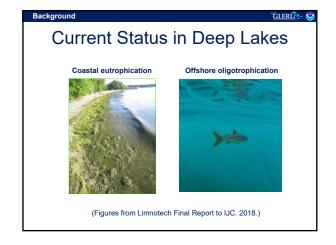












Why Use Ecopath with Ecosim (EwE)? • EwE is freely available as download. Used worldwide to evaluate stressor and harvest effects on fisheries. Over 35 years of development. • EwE can be used to address managers' concerns (eg. increased water quality vs productive fisheries; balance prey production with predator demand; evaluate stocking strategy for predators; restore habitats and native species to promote stable fisheries). • Scenario simulations can inform hypotheses on how food webs may respond differently to drivers (invasive species, fisheries, eutrophication, etc.) over time and space. • Can incorporate uncertainty and be linked to other models (hydrodynamics, economics, land use)

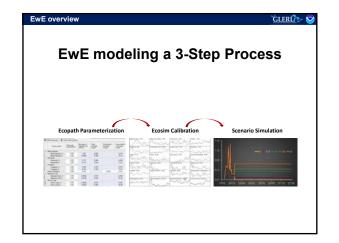
Drawbacks: requires lots of data and some time to construct balanced food web

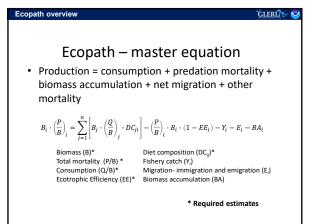


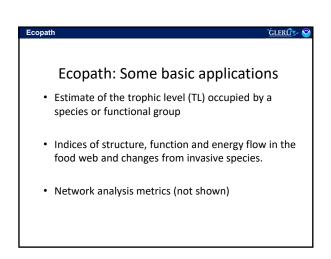


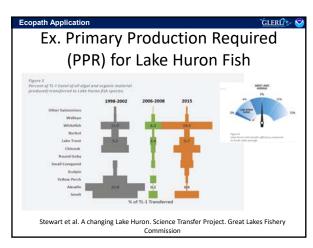
Ecopath with Ecosim- components

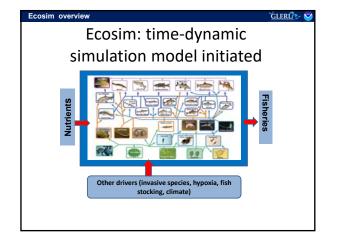
- Ecopath a static, <u>mass-balanced snapshot</u> of the food web and interactions, with biomass groups consisting of a single species, or species groups representing ecological guilds.;
- Ecosim a <u>time dynamic simulation</u> module for policy exploration;
- Ecospace a spatial and temporal dynamic module primarily designed for exploring impact and placement of protected areas

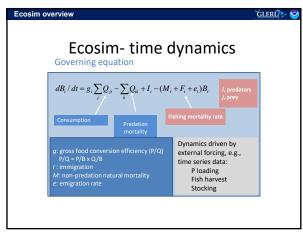


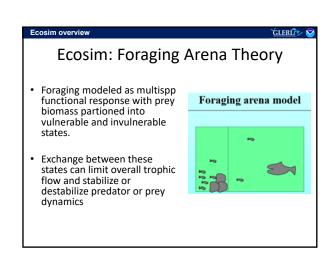




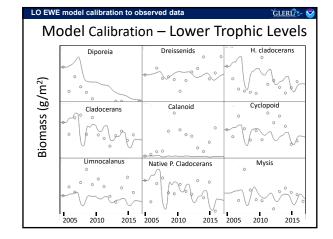


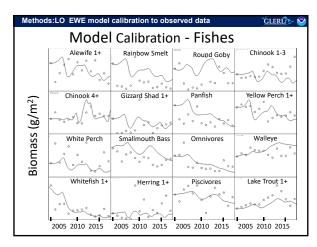


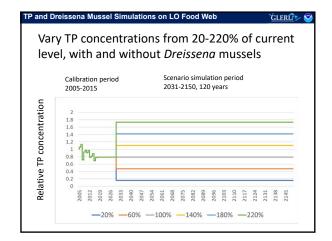


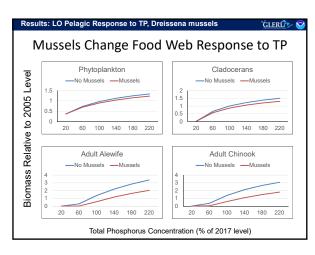


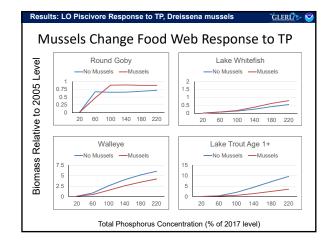


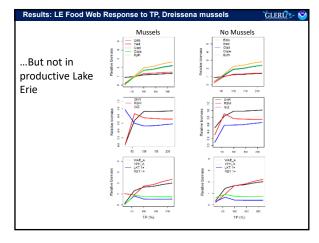


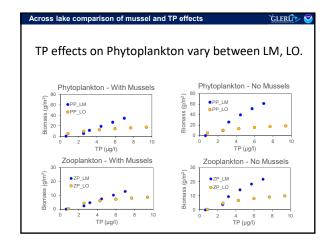


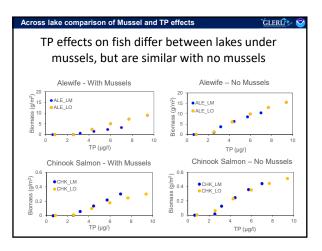


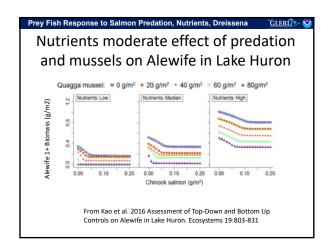




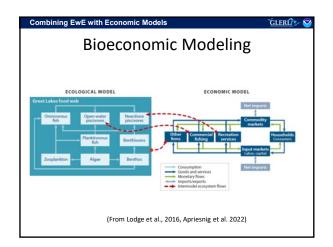


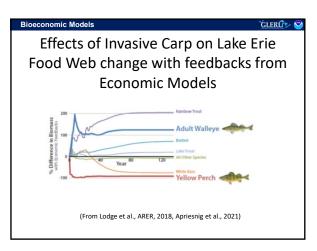




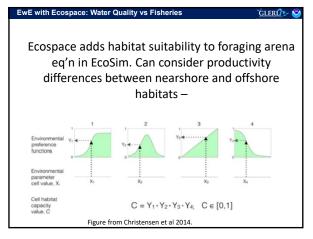












EwE in Coupled Model Framework

Inputs:

- Downscaled current and past climate (precip; temperature)
- · Landscape nutrient delivery model (point & non-point nutrient inputs)
- Hydrodynamic and water quality model (currents, nutrients, primary production, light, etc)
- Food web model (spatially explicit biomass, movement, predator-prey dynamics and harvest potential for model groups/species)

Outputs:

 Assessment and implications of nutrient variability for water quality and fisheries in nearshore and offshore habitats

Discussion and Conclusions

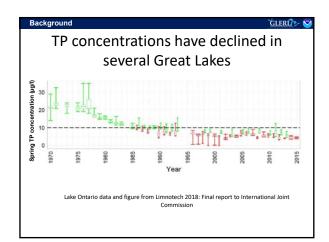
GLERL

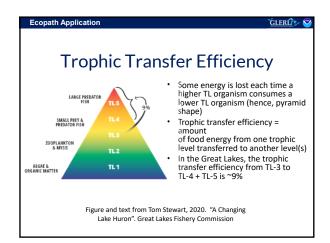
Conclusions

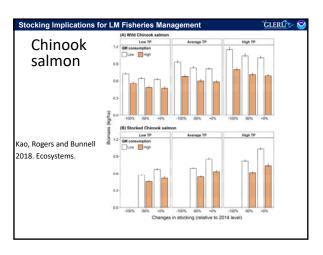
- EwE models are useful, and can be used to inform managers how current and future stressors will affect fisheries production, and allow evaluation of potential actions.
- Continued dialogue and meetings between Managers and Modelers will improve these models and their usefulness to Managers. In turn, managers may grow to trust the information provided by the models
- · Let's keep the lights on!

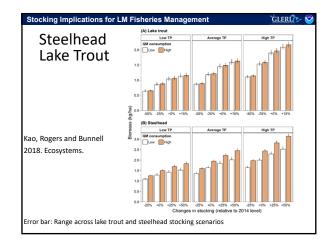
Acknowledgments

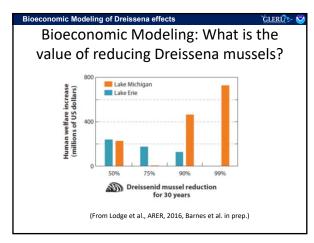
- Funding agencies: GLFC, EPA GLRI, NOAA GLERL
- Scientists and managers at Provincial, State, and Federal Agencies and universities who collaborated and shared data











Atlantis Ecosystem Modeling Framework

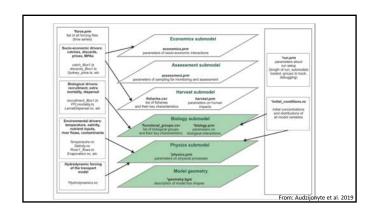
Nick Boucher, David Cannon, Ayume Fujisake-Manome, Beth Fulton, Bec Gordon, Haoguo Hu, James Kessler, Doran Mason, Edward Rutherford, Jia Wang, Hongyan Zhang

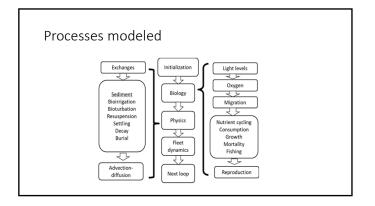
Outline

- What is the Atlantis Ecosystem Modeling Framework?
- How has it been used?
- How can it be used in the Great Lakes?

What is it?

- Deterministic, dynamic, 3-dimensional, end-to-end model integrating physics, geochemistry, biology, fisheries management and assessment, and economics
- Modular by design
- Developed by scientists at the Australian Commonwealth Scientific and Industrial Research Organization (CSIRO)



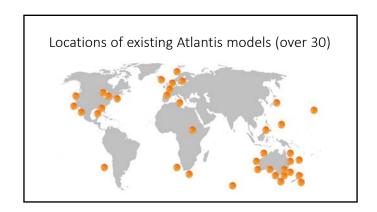


Highlights of some of the features

- Passively advected tracers (e.g., nutrients, oxygen) and actively moving tracers (e.g., organism movement driven by habitat quality, foraging, season, and/or density
- Primary producers dependent upon nutrients, light, space, and species specific growth rate and mortality
- Predation is determined by predator-prey interaction matrix modified by prey availability (e.g., spatial overlap, refuge, spawning time, gapelimitations) in a functional response model
- Maintenance costs modeled (e.g., respiration)
- Early life stages not modeled, however use stock-recruitment relationships

Some challenges

- Very complex model
- Data needs are intense
- Requires a multi-disciplinary team of people
- Traditional sensitivity analysis is not possible
- Parameter uncertainty is often evaluated by bounding the particular parameters



Applications

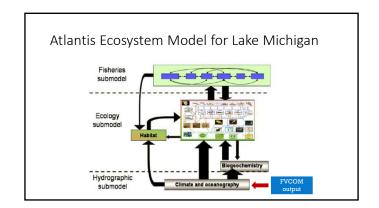
- Best used for scenario-based evaluations of competing forcing factors and simulating what-if scenarios
- Ecosystem-based applications have included:
 - · Fisheries assessment and management
 - Assessment of ecosystem indicators
 - Evaluation of marine protected areas
 - Effects of anthropogenic stressors:

Climate change Ocean acidification Invasive species Eutrophication Fishing pressure Oil spills

Lake Michigan example

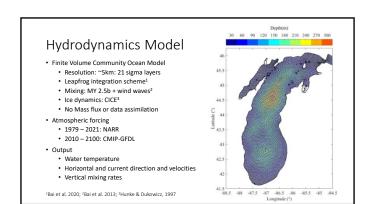
Objectives

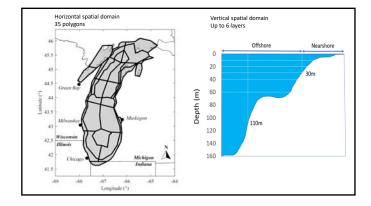
- Use outputs of Lake Michigan FVCOM hydrodynamics model, and nutrient inputs from GLERL watershed hydrology model, as inputs to the Atlantis Ecosystem Model;
- Calibrate the Atlantis Ecosystem Model for the 1994-2020 period;
- Use linked model system to simulate food web and fisheries response to nutrients, invasive mussels, and climate (cold vs warm years).
- Funding for this projects: Great Lakes Restoration Imitative, Great Lakes Fishery Trust

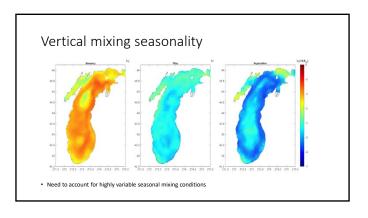


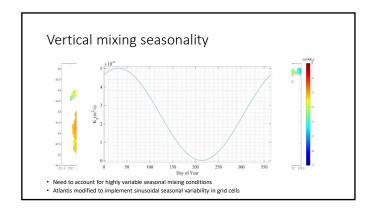
Data sources

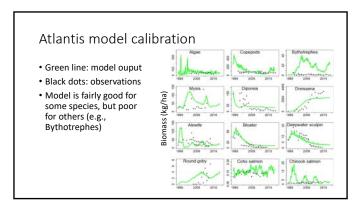
- Nutrients (GLERL, EEGLE, EPA, GLNPO)
- Plankton, benthos, and fish: Ecopath for LM (Rutherford et al. 2021, USGS fish surveys, EPA GLNPO)
- Water temperature and hydrodynamics (FVCOM)
- Solar radiation (Field observations)
- Fish stocking data (Great Lakes fish stocking database)
- Fisheries catch data (commercial and recreational)

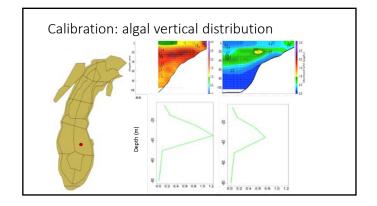


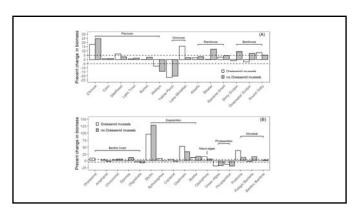


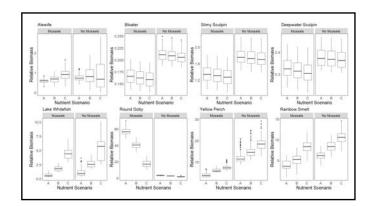












Summary

- Atlantis Ecosystem Modeling Framework is best used as scenario-based exercise to evaluate competing factors or the effects of multiple factors (e.g., climate change, fisheries harvest, invasive species, nutrients, etc)
- Integrates physics, chemistry, biology, fisheries management and assessment, and economics in a 3-dimensional spatial domain
- The approach is extremely complex an is likely best used in the context of a multiple modeling approach using models of different complexities
- The application of this approach is best used in concert with discussions and interactions with managers to define needs and competing issues managers need to contend with.

Ecosystem Behaviors: A World beyond the Scope of Physics

Robert E. Ulanowicz

Arthur R. Marshall Laboratory University of Florida

Center for Environmental Science University of Maryland

> Investigating Food Webs November 10, 2022

Ultimate physical reductionism is impossible in a heterogeneous world.

"All causality originates from below, and there is nothing 'down there' but the laws of physics."

Nobel physicists Murray Gell-mann, Stephen Weinberg and David Gross (Kauffman 2008)

Physics is all about <u>objects</u> moving according to universal laws.

Whitehead & Russell (1913)

Principia Mathematica

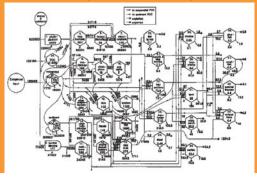
=> Logic of laws of physics limited to operations upon sets of homogeneous entities

Ecology is all about <u>relationships</u> among <u>heterogeneous</u> processes.

Walter Elsasser (1981) A Form of Logic Suitable for Biology

=> Any laws of biology cannot resemble the laws of physics.

Relationships are every bit as important as objects.
Chesapeake mesohaline ecosystem flow network
(Baird & Ulanowicz 1989)



Configurations of process relationships constitute the causal foundations of life.

Tiezzi's dead deer:

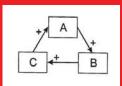
Same mass, molecules, genes, bound energy

So what's missing?

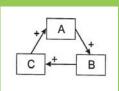
Answer: Configuration of processes

Causality can be distributed over more than just two entities.

Example: Autocatalysis



Autocatalysis can exert endogenous facultative selection within a system.



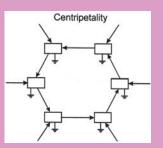
Can select beneficial changes in the performance of individual nodes.

Autocatalysis can be self-stabilizing and possibly served as memory before the advent of genetic coding.

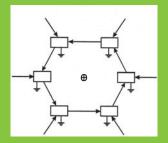
The obverse action of autocatalytic selection is to mitigate changes that diminish contingent disturbances. This effect tends to stabilize current configurations.

In higher systems and organisms the effect can also be described as "healing".

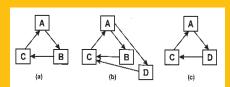
Autocatalysis induces <u>centripetality</u> – the keystone to subjective tendencies.



Centripetality also creates a virtual center for the autocatalytic loop.



Can also select advantageous replacements of nodes and/or pathways, i.e., drive competition.



Competition is secondary to and derivative of mutual benefaction.

Competition occurs when two or more centripetal configurations overlap in a field of mutually necessary resources.

Without centripetal action (driven by mutuality) at some lower level, competition cannot occur.

Autocatalysis imparts a temporal but nondeterminate direction to system evolution.

Autocatalysis always selects changes that benefit the process of autocatalysis.

There is no fixed goal (endpoint) of the direction, which changes with each new incorporated contingency. (It is not teleology)

Autocatalysis is semi-autonomous of universal laws (indeterminate), but not random in behavior

"Indeterminate, but non-random" sounds like an oxymoron, but simple examples exist. An analogy of such self-organization was given by John Wheeler in terms of an inventive parlor game.

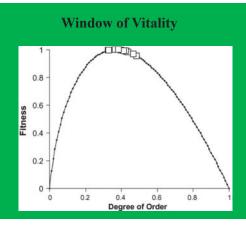
Information Theory can be invoked to quantify the degree of coherent action in a network as distinct from disordered events.

If T_{ii} represents the effect of node i upon node j, then

 $a = \{\Sigma_{i,j}T_{ij}log([T_{ij}\Sigma_{k,l}T_{kl}]/[\Sigma_kT_{kj}][\Sigma_lT_{ij}])\}/\{-\Sigma_{p,q}T_{pq}log(T_{pq}/\Sigma_{k,l}T_{kl})\} \geq 0$ measures the degree of coherent action in the system, while

$$\begin{split} & \emptyset = \{\Sigma_{i,j}T_{ij}log([T_{ij}^2]/[\Sigma_kT_{kj}][\Sigma_lT_{il}])\}/\{\Sigma_{p,q}T_{pq}log(T_{pq}/\Sigma_{k,l}T_{kl})\} \geq 0, \\ & quantifies disorder among processes. \end{split}$$

Since $a + \omega = 1$, the measures are strictly complementary (i.e., in dialectical opposition).



Indirect mutualism is the crux of <u>agency</u> – a dynamic that does not simply respond to events, but can make qualitatively new things happen.

Simple physical forces are driven by external constraints. They can express no choice among what interacts with or guides them. Agency, however, selects from among the myriad of impinging actions those which benefit its own actions. It thereby "creates" its own new dynamics.

Eugene P. Odum (Science Magazine 1977)

"To achieve a truly holistic or ecosystematic approach, not only ecology, but other disciplines in the natural, social and political sciences as well must emerge to new hitherto unrecognized and unrestricted levels of thinking and action."