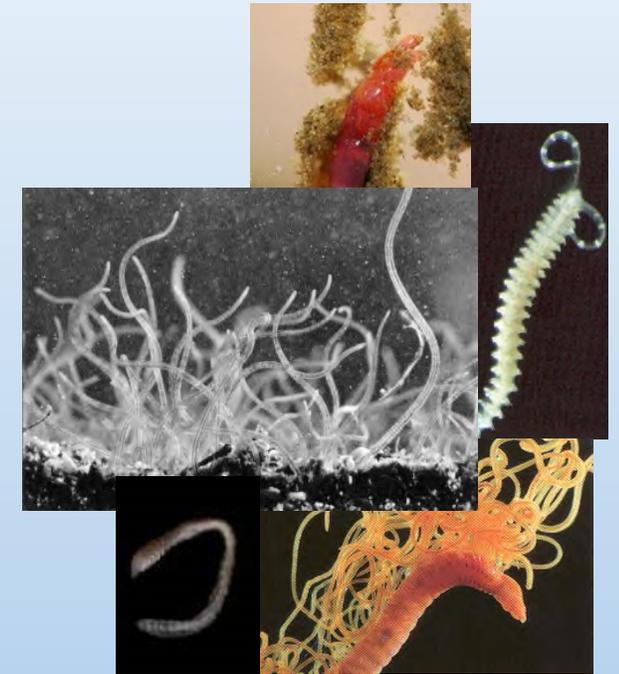
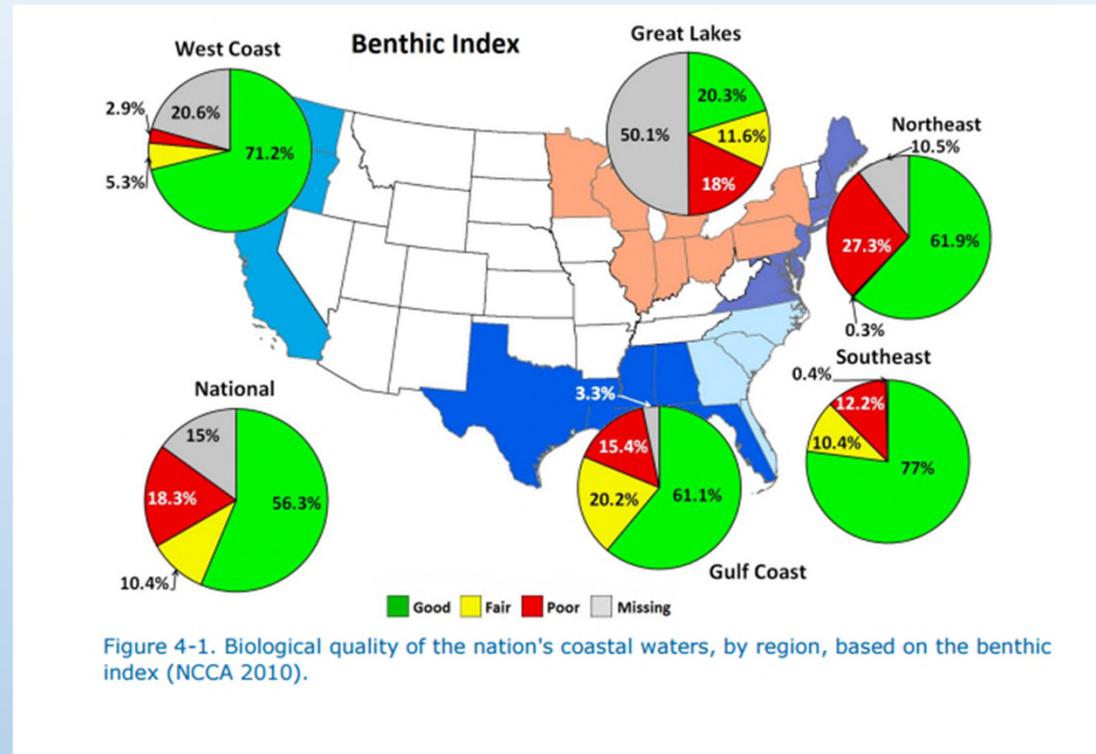
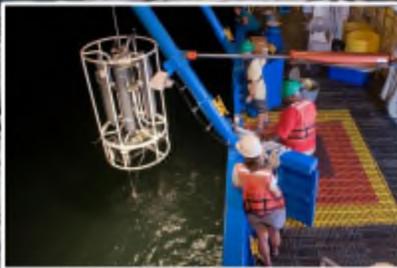


# Development and refinement of benthic indices to assess coastal waters for U.S. EPA's National Coastal Condition Assessment.



M.C. Pelletier<sup>1</sup>, D.J. Gillett<sup>2</sup>, A. Hamilton<sup>3</sup>, T. Grayson<sup>4</sup>, V. Hansen<sup>1</sup>, E.W. Leppo<sup>3</sup>, S.B. Weisberg<sup>2</sup>, A. Borja<sup>5</sup>, T. Angradi<sup>1</sup>, B. Hinchey-Malloy<sup>6</sup>, L.E. Burlakova<sup>7</sup>, H. Sullivan<sup>4</sup>  
<sup>1</sup>U.S. EPA Office of Research and Development, <sup>2</sup>SCCWRP, <sup>3</sup>TetraTech, <sup>4</sup>U.S. EPA Office of Water, <sup>5</sup>AZTI – Tecnalia, <sup>6</sup>U.S. EPA Great Lakes National Program Office, <sup>7</sup>Buffalo State

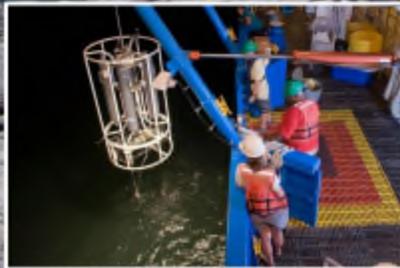
# National Coastal Condition Assessment 2010



## 2010 National Coastal Condition Assessment

- Extension of previous EPA marine and estuarine monitoring programs from 1990-2006
- Included near-shore Great Lakes monitoring for the first time
- Biological monitoring based on benthic invertebrates, assessed using benthic indices
  - Estuaries – 5 different indices for different parts of the country
  - Great Lakes – Oligochaete Trophic Index used to assess deep portions of Great Lakes by SOLEC

# National Coastal Condition Assessment 2010



This talk:

Improvements to benthic assessment for  
2015 & 2020 National Coastal Condition  
Assessment

## Estuaries

Development of single index for use in all  
continental US estuaries

## Great Lakes

Initial efforts to assess/improve the benthic  
index for nearshore Great Lakes waters

# Coastal Indices used for National Coastal Assessment and NCCA 2010 Reporting



Region/ Province	Data Source	Statistical Method	Component Metrics	Index Condition Scale		
				Good	Fair	Poor
Northeast/ Acadian	NCA 2000-2001	Logistic Regression Analysis	Diversity (Shannon $H'$ ) Pollution Tolerant Taxa Proportion Capitellids	> 5	4 – 5	< 4
Northeast/ Virginian	EMAP 1990-1993	Discriminant Analysis	Diversity (Gleason $D$ ) Abundance Tubificids Abundance Spionids	> 0	n/a	≤ 0
Southeast/ Carolinian	EMAP 1993-1994	Cluster Analysis	Abundance Species Richness Dominance Pollution Sensitive Taxa	> 2.5	2 – 2.5	< 2
Gulf/ Louisianan	EMAP 1991-1992	Discriminant Analysis	Diversity (Shannon $H'$ ) Abundance Tubificids Proportion Capitellids Proportion Bivalves Proportion Amphipods	> 5	3 – 5	< 3

- Benthic Indices developed and calibrated separately for each Region
- Concerns about cross-region comparability



West Coast Salinity-adjusted expected number of species

Region	Good	Fair	Poor
West	Observed species richness is more than 90% of the lower 95% confidence interval of expected species richness for a specific salinity.	Observed species richness is between 75% and 90% of the lower 95% confidence interval of expected species richness for a specific salinity.	Observed species richness is less than 75% of the lower 95% confidence interval of expected species richness for a specific salinity.

# Need for New National Benthic Index for Application to US Estuaries

Multivariate (discriminant analysis, logistic regression)

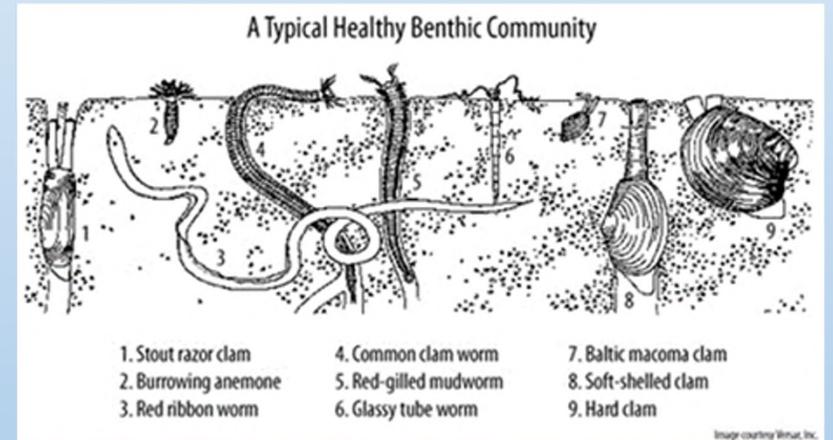
- Need reference and impaired sites
- Select from multiple potential metrics
- Combine metrics to develop index

B-IBI (multimetric)

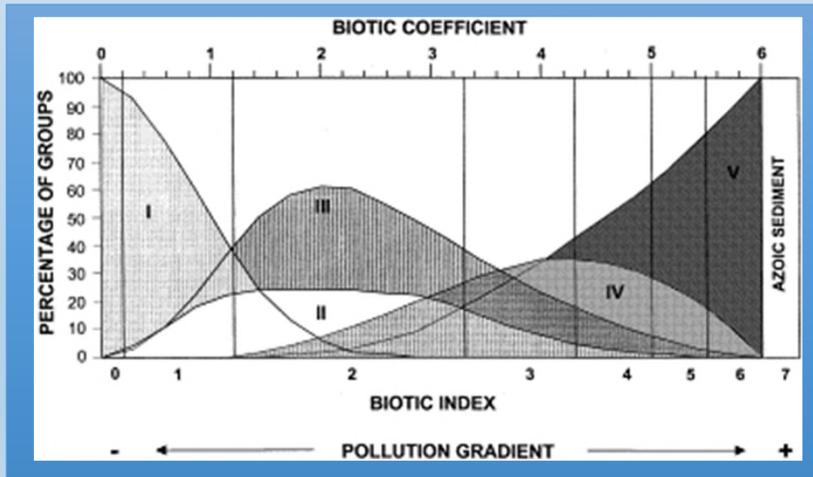
- Need reference and impaired sites
- Select from multiple potential metrics
- Combine metrics to develop index

AMBI (abundance-weighted tolerance index)

- Do Not Need reference and impaired sites
- Need to categorize tolerance values of benthic taxa



# What is AMBI (AZTI Marine Biotic Index)?



$BI =$

$$0 * \% EG I + 1.5 * \% EG II + 3 * \% EG III + 4.5 * \% EG IV + 6 * \% EG V$$

Range =

0 (unimpacted) to 6 (heavily impacted) or 7 (azoic)

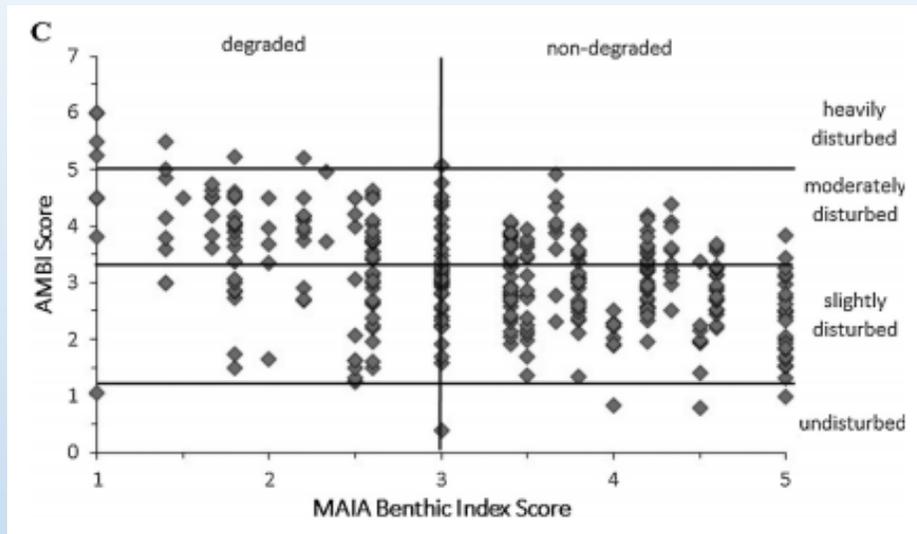
- Previous Case Studies in FL, SoCal, Chesapeake Bay, Northwest

# AMBI – Adaptation to US estuaries

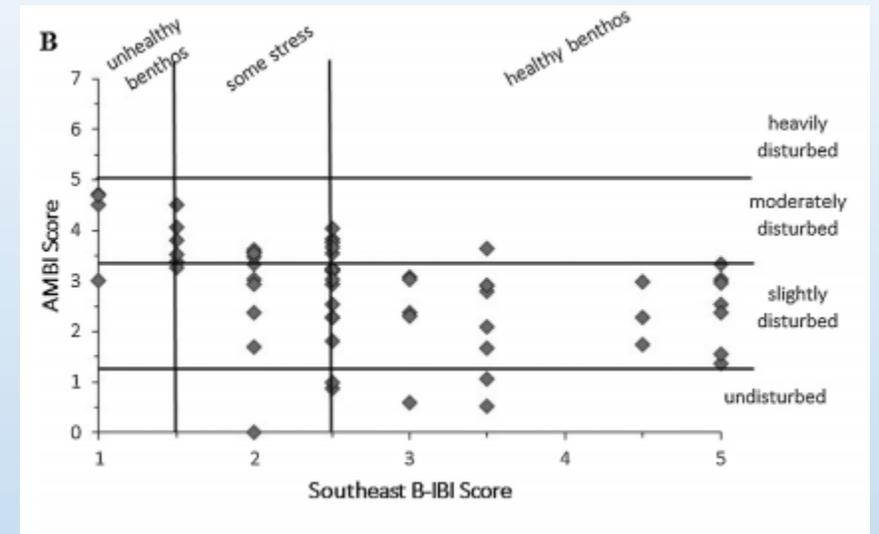
- 3 Day workshop in Cosa Mesa, CA (Sept 2011)
  - Decision to explore development of AMBI for US estuaries
  - Decision to compare ‘new’ index to existing local indices from around the country
  - NCCA species categorized by EG group (tolerance value)
- Creation of working group with larger group meetings at CERF (2011-2015)
  - 3 regional Datasets assembled – compare local index to AMBI
  - Workshop EG list augmented with existing European EG list
  - Published results in 2015



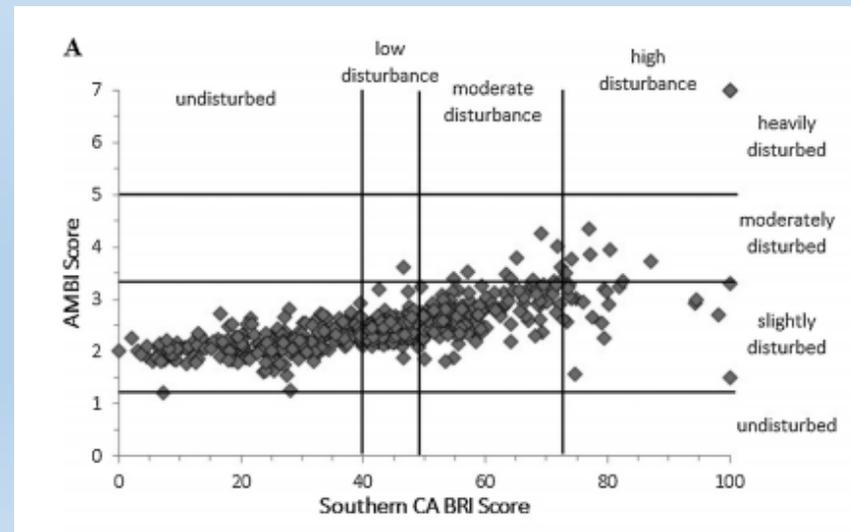
# AMBI – Adaptation to US estuaries



$R = 0.437, p < 0.0001$

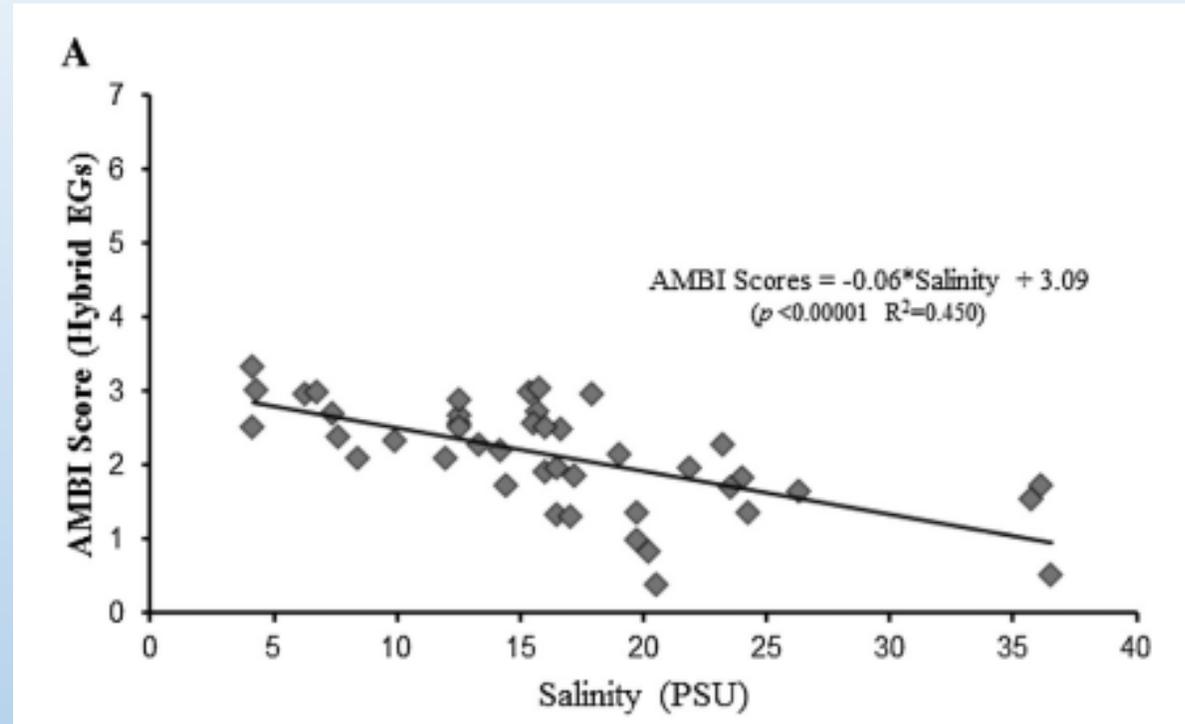


$R = 0.525, p < 0.0001$



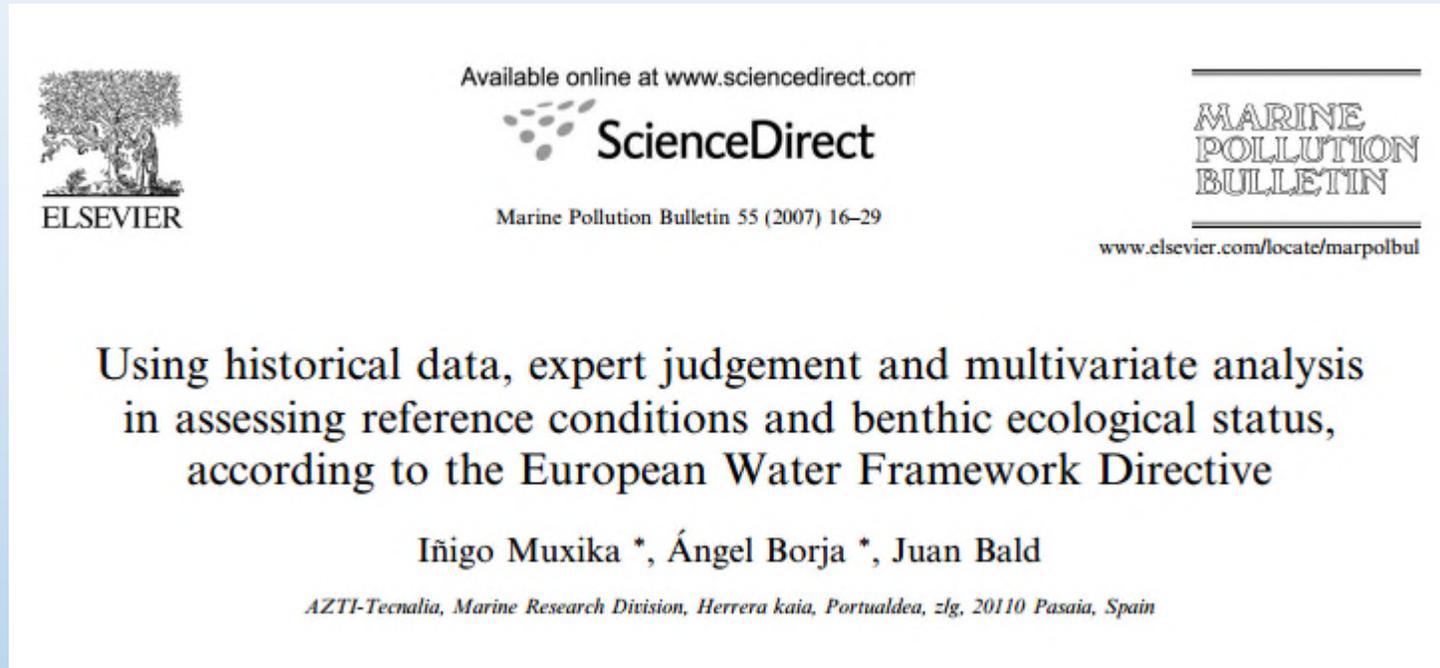
$R = 0.736, p < 0.0001$

# AMBI – Adaptation to US estuaries



- Strong salinity bias seen in unimpacted station in the southeast and mid-Atlantic (SoCal is primarily high salinity sites so a salinity bias would not be expected)

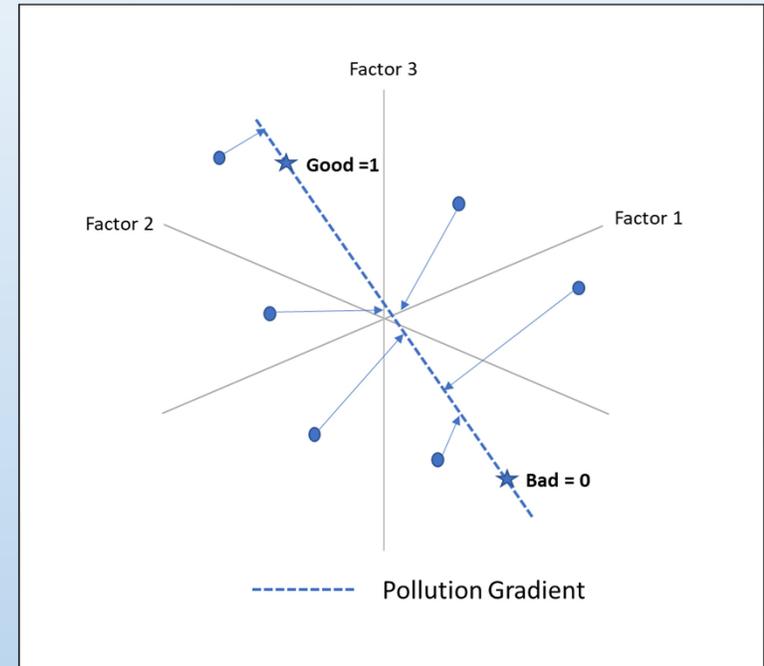
# Expansion of AMBI to M-AMBI (multivariate AMBI)



- Combines AMBI, diversity ( $H'$ ) and species richness into a new index
- Helps to deal with issues of salinity bias
- Allows discrimination of sites with low species richness

# How does M-AMBI work?

- Uses Factor Analysis to combine AMBI, diversity ( $H'$ ) and species richness into a new index
- Good and Bad endpoints ('reference' & 'impaired' derived for each metric)
  - Used to set up a pollution gradient
  - Station values from factor analysis projected onto the pollution gradient
- Classified by habitat (salinity and location (estuarine v. coastal))
- Range = 0 (Bad) to 1 (High)

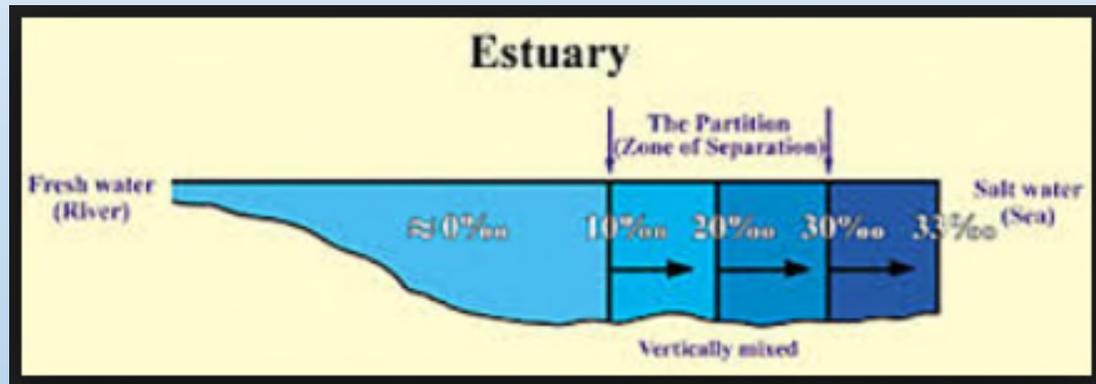


## Stretches

Oligo/mesohaline  
Polyhaline  
Euhaline (estuarine)  
Euhaline (coastal)

# M-AMBI – Adaptation to US estuaries

- Venice salinity classification to identify habitats



Habitat	Salinity (psu)	
Tidal Freshwater	<0.5	
Oligohaline	0.5	to <5
Mesohaline	5	to <18
Polyhaline	18	to <30
Euhaline	30	to <40
Hyperhaline	>=40	

- Needed to develop expectations for each habitat, but West Coast data used larger grab (0.1 m<sup>2</sup> grab vs 0.04 m<sup>2</sup> grab or equivalent) and sieve (1.0 mm vs. 0.5 mm)

Ecological Indicators 89 (2018) 818–827

Contents lists available at ScienceDirect

**Ecological Indicators**

journal homepage: [www.elsevier.com/locate/ecolind](http://www.elsevier.com/locate/ecolind)

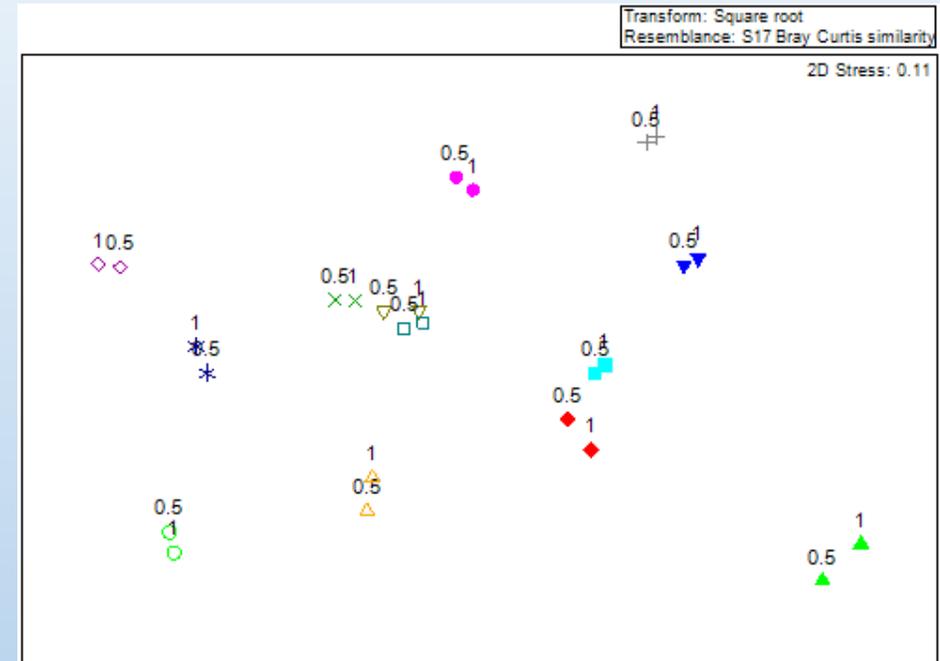
Research paper

Adaptation and application of multivariate AMBI (M-AMBI) in US coastal waters

Marguerite C. Pelletier<sup>a,\*</sup>, David J. Gillett<sup>b</sup>, Anna Hamilton<sup>c</sup>, Treda Grayson<sup>d</sup>, Virginia Hansen<sup>e</sup>, Erik W. Leppo<sup>c</sup>, Stephan B. Weisberg<sup>b</sup>, Angel Borja<sup>f</sup>

# M-AMBI Development – West Coast gear differences

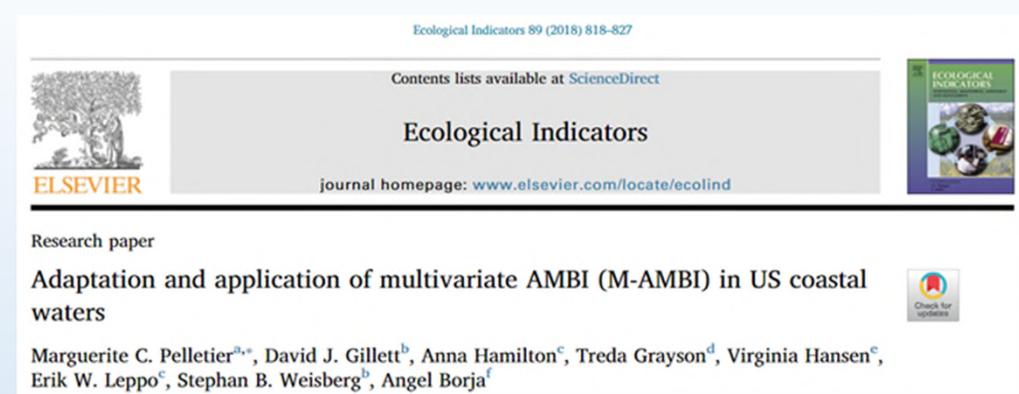
- Sieve size differences did not appear to be significant based on subset of stations using both sieves



- Grab size impacted the total number of species

95th percentile	Rest of US		WEST	
	S	H'	S	H'
Tidal Freshwater	16	2.06	15	1.80
Oligohaline	16	2.14	14	1.82
Mesohaline	26	2.50	17	1.97
Polyhaline	43	2.93	77	3.30
Euhaline	59	3.29	92	3.60

# M-AMBI – Adaptation to US estuaries



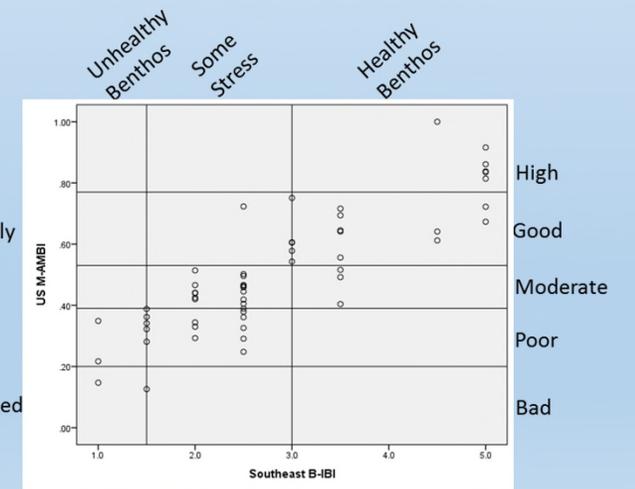
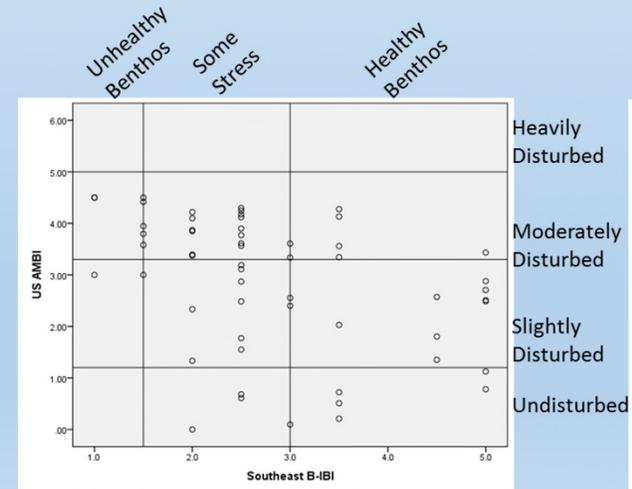
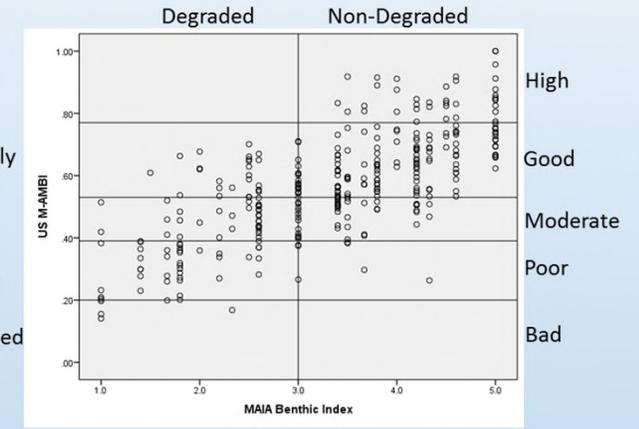
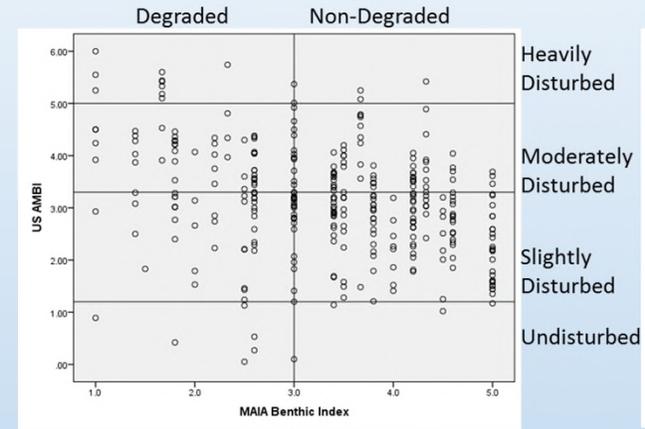
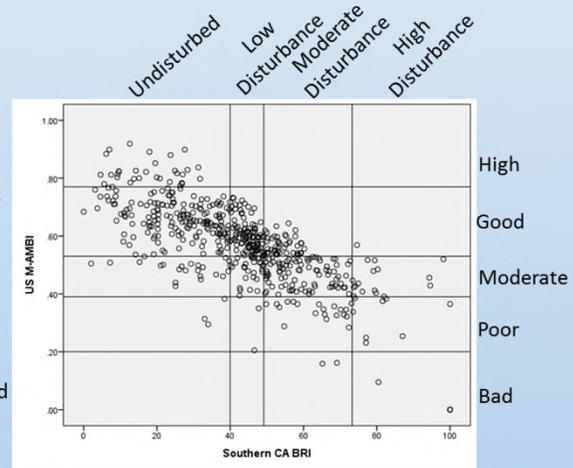
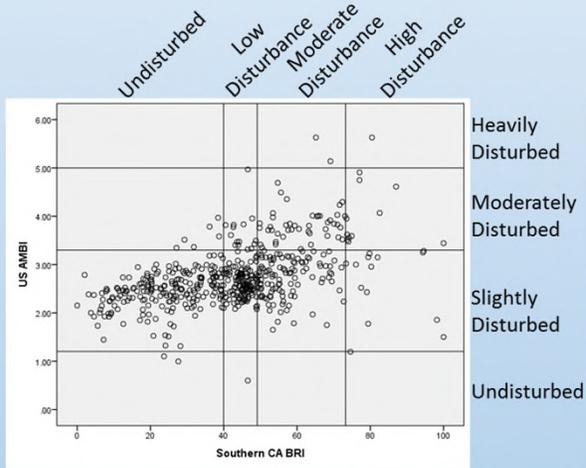
- Bad endpoints were same for all habitats ( $S$ ,  $H' = 0$ ,  $AMBI = 6$ )
- High endpoints based on 95<sup>th</sup> percentile of 1999-2006 NCA data, based on habitat
  - $TF$ ,  $O$ ,  $M$ ,  $H$  calculated for entire US
  - $P$  and  $E$  – calculated for West and Rest of US
- Using 1999-2006 NCA data, explored use of metrics other than  $S$  and  $H'$  (i.e., % oligochaetes for tidal freshwater habitat)
- Calculating M-AMBI for 3 validation datasets
  - Compare to local indices
  - Look at calibration accuracy vs. apriori Good/Bad sites
  - Look to see if salinity correlation has been reduced or eliminated

# M-AMBI – Good Classification Accuracy

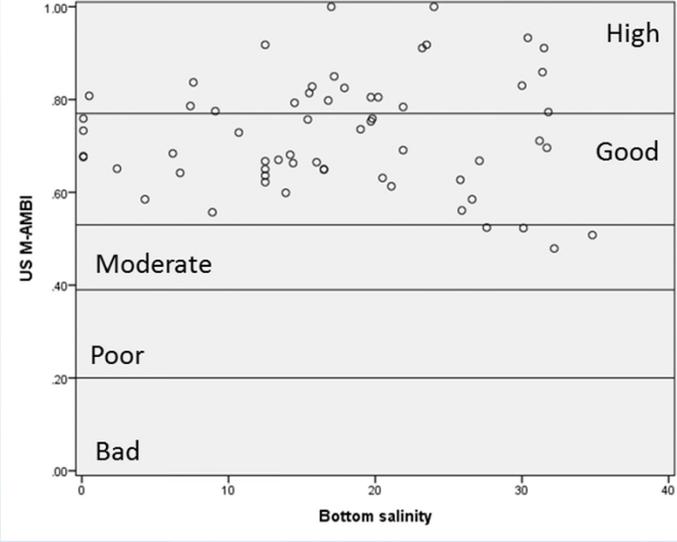
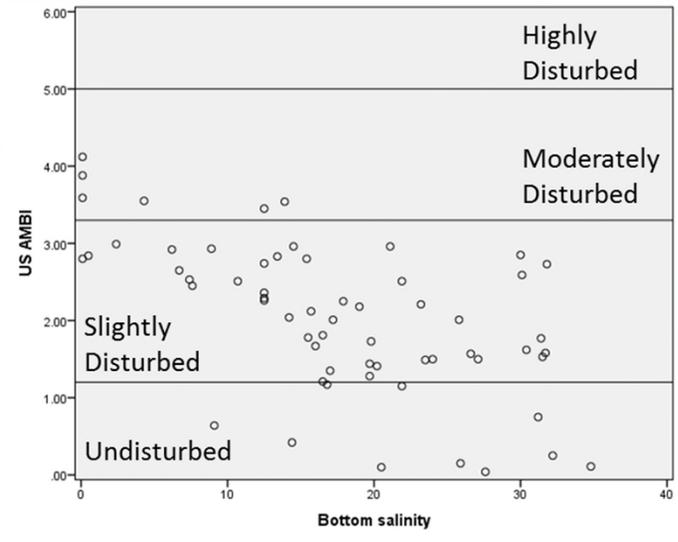
Dataset	M-AMBI vs Local Index			M-AMBI accuracy compared to a priori Good-Bad sites	
	Correlation (Spearman's $\rho$ )	Overall Classification		Overall Classification	
		Accuracy (%)	Kappa	Accuracy (%)	Kappa
Mid-Atlantic	$r = 0.715$ ( $p < 0.0005$ )	87.3	0.65 (substantial)	83.2	0.57 (moderate)
Southeast	$r = 0.829$ ( $p < 0.0005$ )	97.4	0.95 (almost perfect)	90.3	0.81 (almost perfect)
Southern California	$r = -0.746$ ( $p < 0.0005$ )	86.3	0.49 (moderate)	100	1.00 (almost perfect)

- Good correspondence with local indices
- Good correspondence with a-priori Good/Bad sites from local datasets

# M-AMBI – Tighter correspondence to local indices



# M-AMBI – Removal of Salinity Bias seen with AMBI



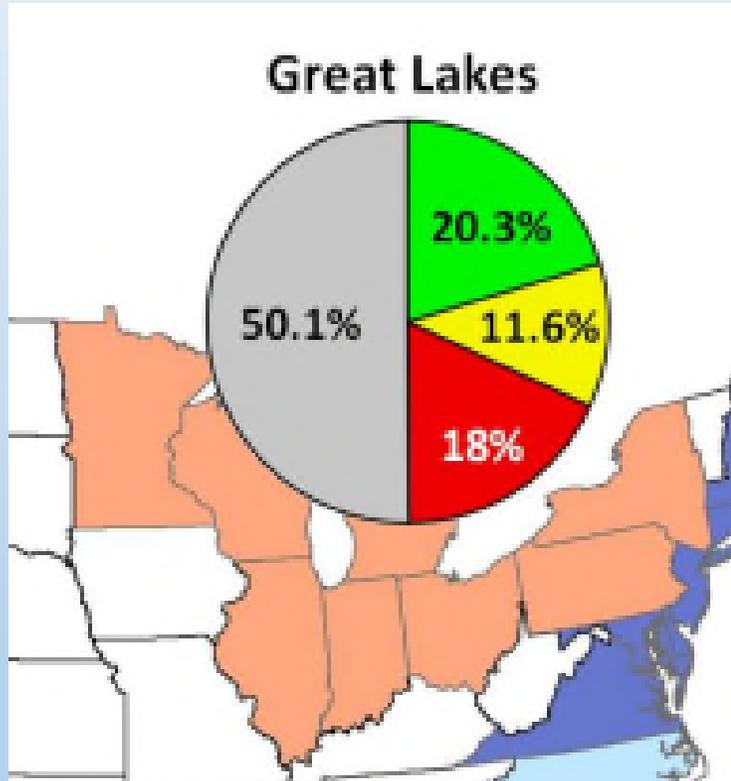
Dataset		US M-AMBI	US AMBI	MAIA BI
Mid-Atlantic	Spearman	0.043	<b>-0.613**</b>	0.017
	p-value	0.742	<b>&lt;0.0005</b>	0.933
	N	61	<b>61</b>	26
Southeast	Spearman	-0.410	<b>-0.900*</b>	-0.671
	p-value	0.493	<b>0.037</b>	0.215
	N	5	<b>5</b>	5
Southern California	Spearman	-0.008	-0.132	-0.156
	p-value	0.973	0.568	0.499
	N	21	21	21

\* Correlation is significant at the 0.05 level (2-tailed).

\*\* Correlation is significant at the 0.01 level (2-tailed).

Used M-AMBI to assess all estuaries in the 2015 NCCA Report

# Great Lakes Index assessment/improvement



$$OTI = c \times \frac{\frac{1}{2} \sum n_0 + \sum n_1 + 2 \sum n_2 + 3 \sum n_3}{\sum n_0 + \sum n_1 + \sum n_2 + \sum n_3}$$

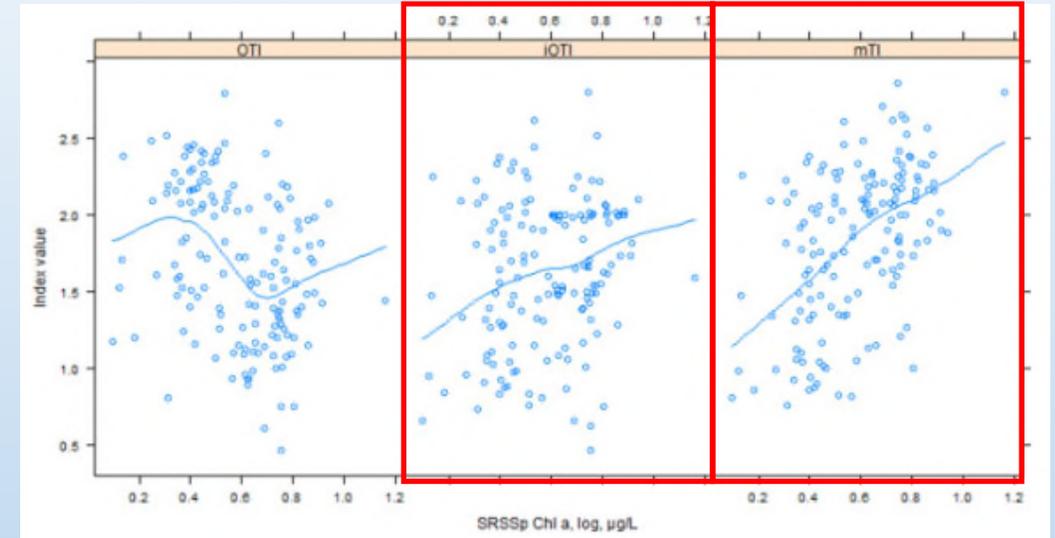
$c = 1$	if	$n > 3,600$
$c = 0.75$	if	$1,200 < n < 3,600$
$c = 0.5$	if	$400 < n < 1,200$
$c = 0.25$	if	$130 < n < 400$
$c = 0$	if	$n < 130$

>50% of samples classified as 'missing'

- ~1/2 were sites that were not sampled – hard bottom or grab 'blowout'
- ~1/2 were sites with no oligochaetes, or unclassified oligochaetes

# Potential modification of Great Lakes Index

- 3 day workshop in Chicago, IL (Oct 2016)
  - L. Burlakova presented 2 new indices;
    - Improved OTI (iOTI)
    - Modified TI (mTI)
  - Group classified additional taxa into OTI groups
  - Group agreed to assemble additional validation datasets



For deeper water sites in Lake Erie, both the iOTI and mTI had a better relationship with surface remote-sensed chlorophyll than the 'original' OTI

Journal of Great Lakes Research 44 (2017) 618–628



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Development of new indices of Great Lakes water quality based on profundal benthic communities

Lyubov E. Burlakova <sup>a,\*</sup>, Katya E. Kovalenko <sup>b</sup>, Kurt L. Schmude <sup>c</sup>, Richard P. Barbiero <sup>d</sup>, Alexander Y. Karatayev <sup>a</sup>, Barry M. Lesht <sup>d,e</sup>

<sup>a</sup> Great Lakes Center, Buffalo State College, 1300 Elmwood Ave., Buffalo, NY 14222, USA

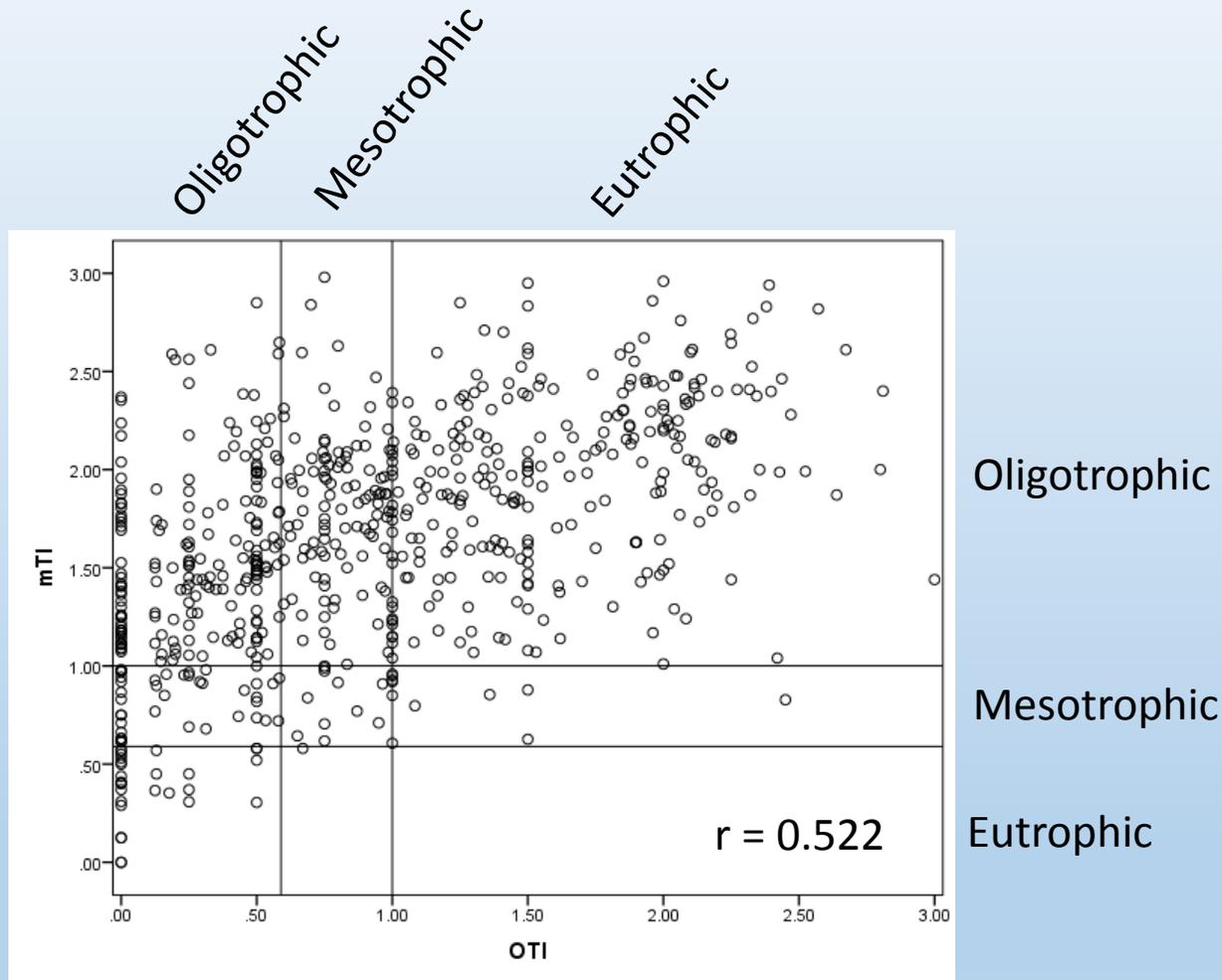
<sup>b</sup> Natural Resources Research Institute, University of Minnesota-Duluth, 5013 Miller Trunk Highway, Duluth, MN 55811, USA

<sup>c</sup> Department of Natural Sciences, Lake Superior Research Institute, University of Wisconsin-Superior, 801 N. 28<sup>th</sup> St., Superior, WI 54880-5400, USA

<sup>d</sup> CSRA, 1359 W. Elmdale Ave. Suite 2, Chicago, IL 60660, USA

<sup>e</sup> Department of Earth and Environmental Sciences, University of Illinois at Chicago, 845 W. Taylor St., Chicago, IL 60607, USA

# Potential modification of Great Lakes Index



2010 & 2015 NCCA data

- mTI tends towards more eutrophic conditions than the OTI
  - mTI - used coding from the Burlakova paper + additional coding from workshop
  - OTI calculated using original coding
- For most sites, most taxa are not coded
  - OTI median: 20% of abundance consists of coded oligochaetes
  - mTI median: 30% of taxa abundance coded

# Next steps for Great Lakes Index



- Code additional taxa
- Relate OTI and mTI to stressor gradients to validate indices
- May need to develop new index that uses OTI (or mTI) as a metric within a larger index to encompass more of the benthic community

# Summary

- M-AMBI is multivariate AMBI
  - Accounts for naturally structuring parameters (e.g., salinity)
  - Improves index performance by adding additional metrics
  - Good correspondence with well-calibrated local indices
  - Appropriate for assessing condition at a national scale
- Great Lakes Index
  - Additional validation needed to relate index to stressors
  - May need to add additional metrics to mTI to better characterize the benthic community