

Gulf of Mexico Fishery  
Management  
Council  
SSC Meeting



*Southeast  
Fisheries  
Science  
Center*



**RESTORE**  
SCIENCE PROGRAM



# Part I Ecosystem Modeling the US Gulf of Mexico

Skyler R. Sagarese, Holden E. Harris, Igal Berenshtein,  
Matthew V. Lauretta, Amy Schueller, David D. Chagaris

UNIVERSITY  
OF MIAMI  
ROSENSTIEL  
SCHOOL of MARINE &  
ATMOSPHERIC SCIENCE



# U.S. Gulf of Mexico Ecosystem Model // Presentation Overview

## Part I - USGWEM Ecopath and Ecosim

- Diet and food web ([Sagarese et al. 2016](#))
- Fitting to time series ([NOAA technical memorandum](#))
- Gulf menhaden Ecological Reference Points ([Berenshtein et al. 2023](#))

## Part II - USGWEM Ecospace (*in dev.*)

- Data synthesis (GitHub: [SEFSC/IEA/GWEM/DataSynth](#))
- Validation and calibration
- Next steps



# Part I

- Need for ecosystem modeling
- Brief history of ecosystem modeling in the Gulf of Mexico (GOM)
- U.S. Gulfwide Ecopath Model
- U.S. Gulfwide Ecopath with Ecosim Model development via Restore Science Act Funding
- Developing Ecological Reference Points for Gulf menhaden
- Ecospace and future applications



*Some of the many features of the Gulf of Mexico.*

Image:

<http://ecosystems.noaa.gov/WhereIsEBMBeingUsed/GulfofMexico.aspx>

# Modeling challenges in the Gulf of Mexico

- Diversity

- > 1,100 fishes
- > 3,500 invertebrates
- > 30 marine mammals

- [www.fishbase.org](http://www.fishbase.org)

- [www.sealifebase.org](http://www.sealifebase.org)

- Fisheries



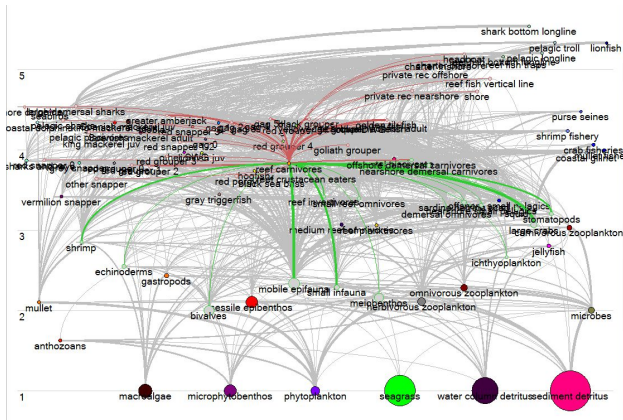
- Lack of detailed food habits data (Simons et al 2013) Image: <https://flowergarden.noaa.gov/>

- Sampling difficulties (e.g., reef species)
- Predator-prey dynamics (e.g., grouper predators)
- Large bias in available data from Eastern GOM (FWRI database)



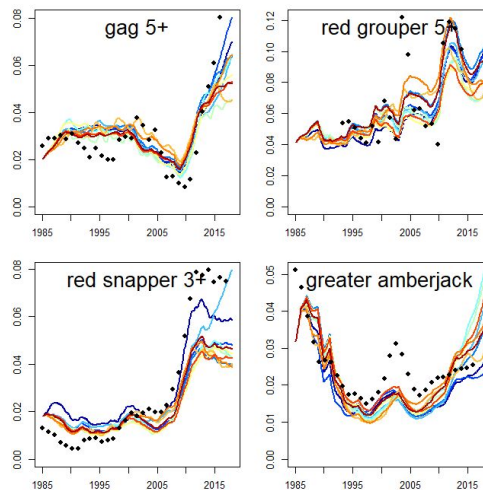
# Software // Ecopath with Ecosim and Ecospace

[www.ecopath.org](http://www.ecopath.org)



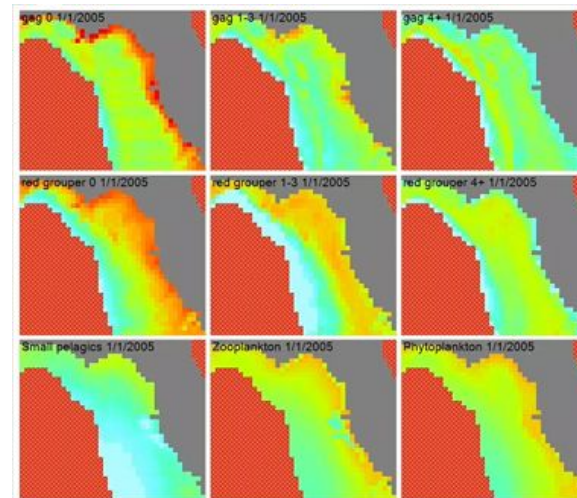
## Ecopath

- Static snapshot of the ecosystem
- Input: biomass, mortality, consumption, diet, and fishery removals
- Requires mass balance
- Starting point for dynamic simulations



## Ecosim

- Time dynamic simulations
- Environmental forcing
- Parameter estimation & time series calibration
- Future projection scenarios
- Policy analysis and tradeoffs



## Ecospace

- Spatially explicit simulations
- Input: dispersal rates, habitat maps, habitat preferences, fishing areas, MPAs, port locations
- Spatial-temporal drivers
- Red tide mortality

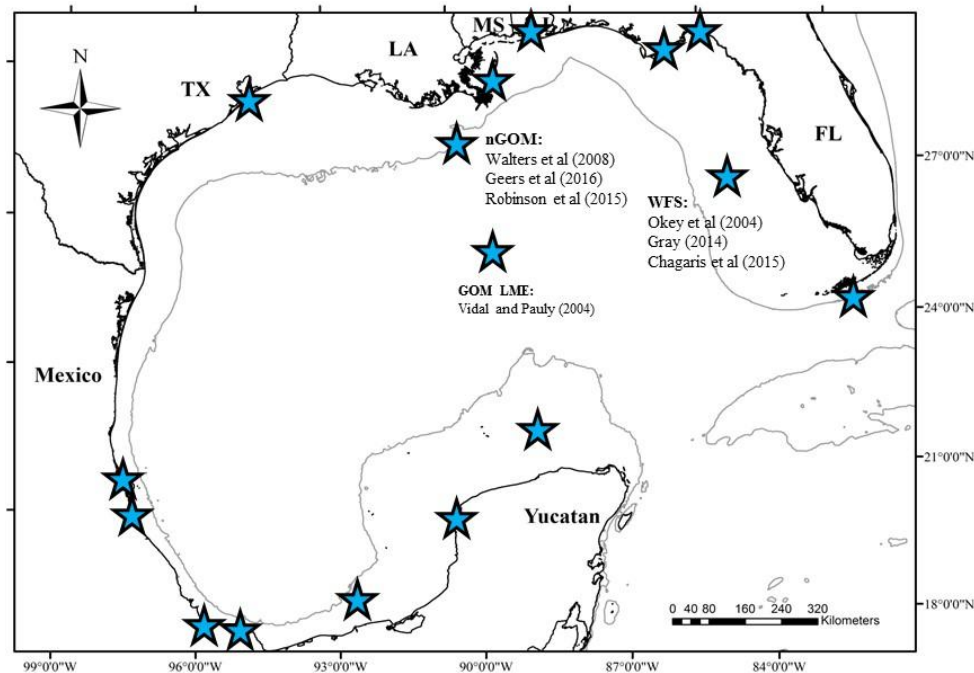
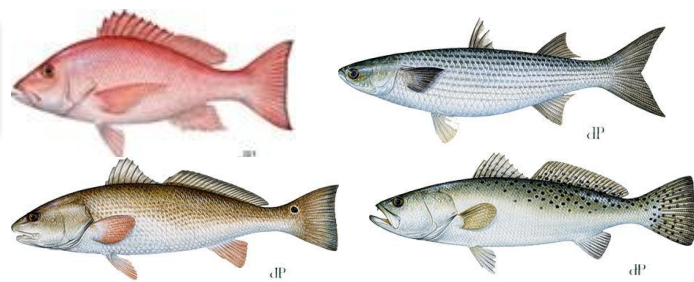
# Coastal GOM // Walters et al. (2008)

BULLETIN OF MARINE SCIENCE, 83(1): 251–271, 2008

## AN ECOSIM MODEL FOR EXPLORING GULF OF MEXICO ECOSYSTEM MANAGEMENT OPTIONS: IMPLICATIONS OF INCLUDING MULTISTANZA LIFE-HISTORY MODELS FOR POLICY PREDICTIONS

*Carl Walters, Steven J. D. Martell,  
Villy Christensen, and Behzad Mahmoudi*

- Focused on shrimp bycatch, red tide
- Functional groups tied closely to Florida
  - Multi-stanzas to represent different life stages
- Demonstrative exercise of the capabilities of EwE, fit to time series
- Geers et al. (2016) modified model to focus on Gulf menhaden



Images: <http://myfwc.com/research/saltwater/fish/>;  
[http://www.gsmfc.org/profiles/Gulf\\_menhaden/Gulf%20Menhaden.php](http://www.gsmfc.org/profiles/Gulf_menhaden/Gulf%20Menhaden.php)

# Utility of Ecosystem Models

- Both the Walters et al. (2008) and Geers et al. (2016) models considered as potential tools for informing damages to the Gulf of Mexico ecosystem following the Deep-water Horizon Oil spill, but...



## Concerns raised over:

- Diet matrix defining species interactions
  - Previously based heavily on expert opinion
  - Limited information on pelagic food-web and higher trophic level interactions
- Lack of fishery discards

Images: [https://en.wikipedia.org/wiki/Deepwater\\_Horizon](https://en.wikipedia.org/wiki/Deepwater_Horizon)

# The Gulf-wide Ecosystem Model // Ecopath

- Builds upon previous models and attempted to alleviate concerns
  - Focus on federally and internationally managed species on spatial scale matching management
  - Include a statistically-derived, more comprehensive definitions of species interactions
  - Model bycatch removals from the menhaden reduction fishery and large-scale fisheries



Contents lists available at [ScienceDirect](#)

Ecological Modelling

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



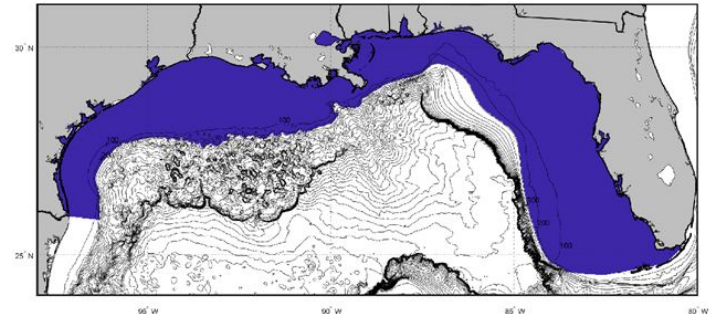
Progress towards a next-generation fisheries ecosystem model for the northern Gulf of Mexico



Skyler R. Sagarese<sup>a,\*</sup>, Matthew V. Laretta<sup>b</sup>, John F. Walter III<sup>b</sup>

<sup>a</sup> Cooperative Institute for Marine and Atmospheric Studies, Rosenstiel School of Marine and Atmospheric Science, University of Miami, 4600 Rickenbacker Causeway, Miami, FL 33149, USA

<sup>b</sup> Southeast Fisheries Science Center, National Marine Fisheries Service, 75 Virginia Beach Drive, Miami, FL 33149, USA



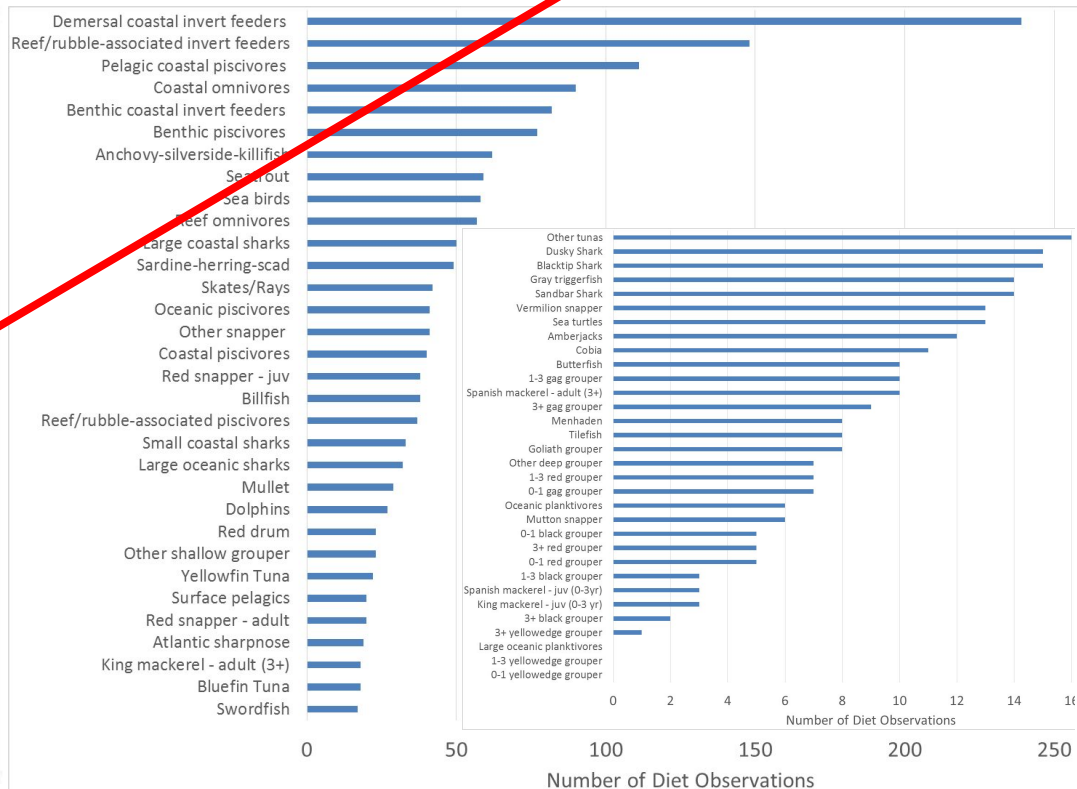


# GWEM Ecopath // Literature Review // Diet Matrix

## DATA

- **Sources:**
  - GOMEXSI (<http://gomexsi.tamucc.edu/>)
  - Journal articles
  - Technical documents
  - Theses
- **Diet composition (%W, %V):**
  - %FO only: converted into relative weight composition (Stobberup et al. 2009)
- 1 observation = 1 diet study\*
  - \*regions or length-classes separated out

- Compiled over 1,906 diet observations



# Diet Matrix // Meta Analysis

## DATA

- Sources:**
  - GOMEXSI (<http://gomexsi.tamucc.edu/>)
  - Journal articles
  - Technical documents
  - Theses
- Diet composition (%W, %V):**
  - %FO only: converted into relative weight composition (Stobberup et al. 2009)
- 1 observation = 1 diet study\*
  - \*regions or length-classes separated out

## PROCESSING

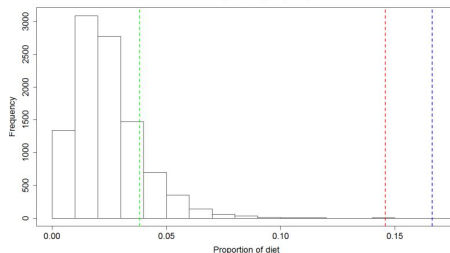
- Subsample:**
  - Randomly select 10 observations
  - Average diet proportions
    - weighted by area of study, method reported (i.e., %W vs %FO), and sample size
  - Re-normalize proportions
- Bootstrap:**
  - 10,000 samples with replacement

## FITTING

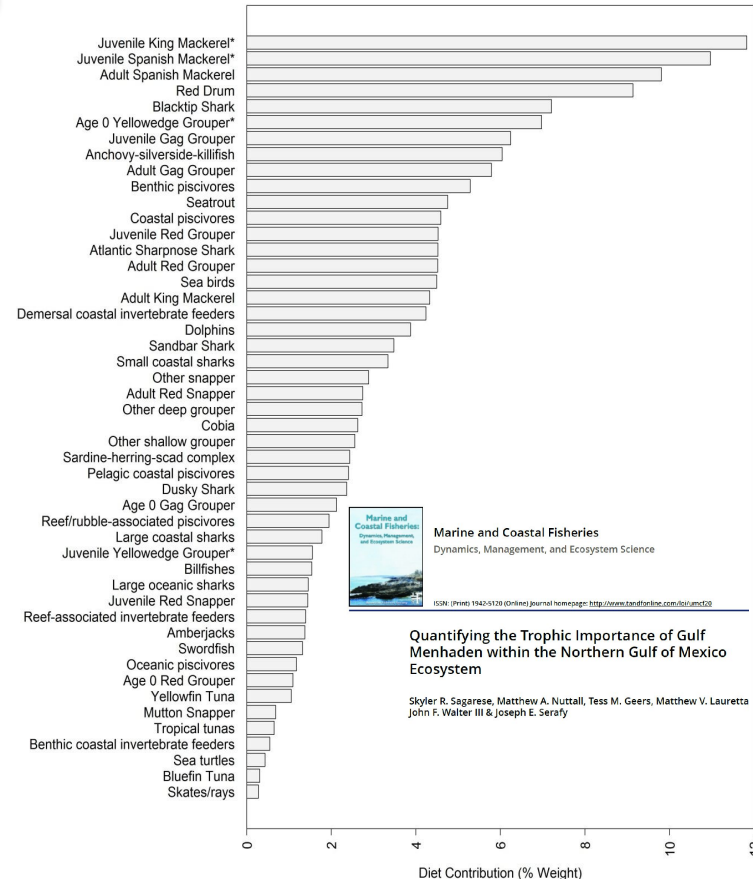
- Dirichlet distribution**
  - Fit to bootstrapped data in R
  - Obtain MLE estimates for each prey group

## EXAMPLE

Predator: Dolphin, Prey: Cephalopod



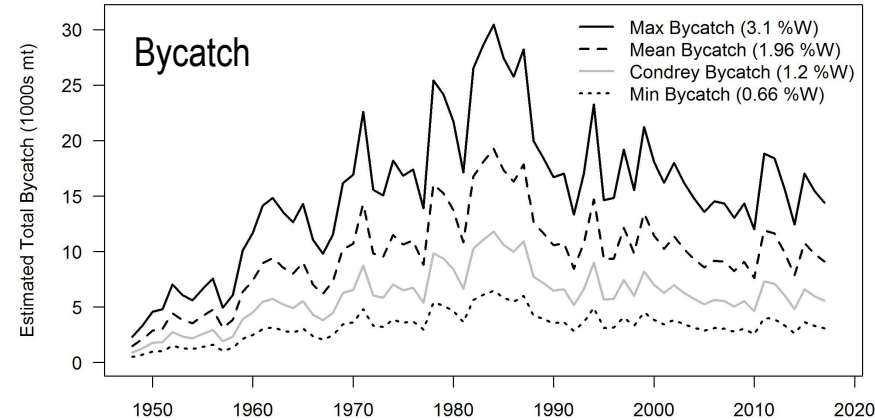
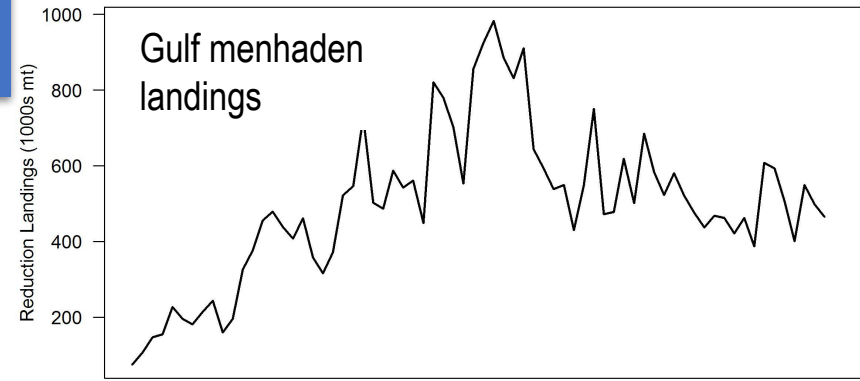
\* = limited sample size



Approach described in depth in Ainsworth et al (2010)

# Bycatch // Menhaden purse seine

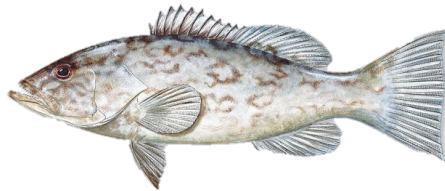
- Species composition and proportion of retained bycatch for the menhaden purse seine fishery were used to infer bycatch
  - Guillory and Hutton (1982)
  - de Silva and Condrey (1997)
  - de Silva et al. (2001)
- Dead discards (i.e., retained landings) were allocated based on percent by weight in the bycatch (2.35%, Guillory and Hutton 1982)



Additional details provided in SEDAR49-DW-04  
<https://sedarweb.org/documents/sedar-49-dw-04-review-of-bycatch-in-the-gulf-menhaden-fishery-with-implications-for-the-stock-assessment-of-red-drum/>

# Ecosystem Modeling for GoM Fisheries Management

- Funded by NOAA RESTORE FF0-2017, decision-support tool priority (PI: Chagaris)
- Goal: Integrate information on ecosystem stressors and predator-prey interactions into the assessment and management of fisheries in the Gulf of Mexico



Gag Grouper  
*Mycteroperca microlepis*



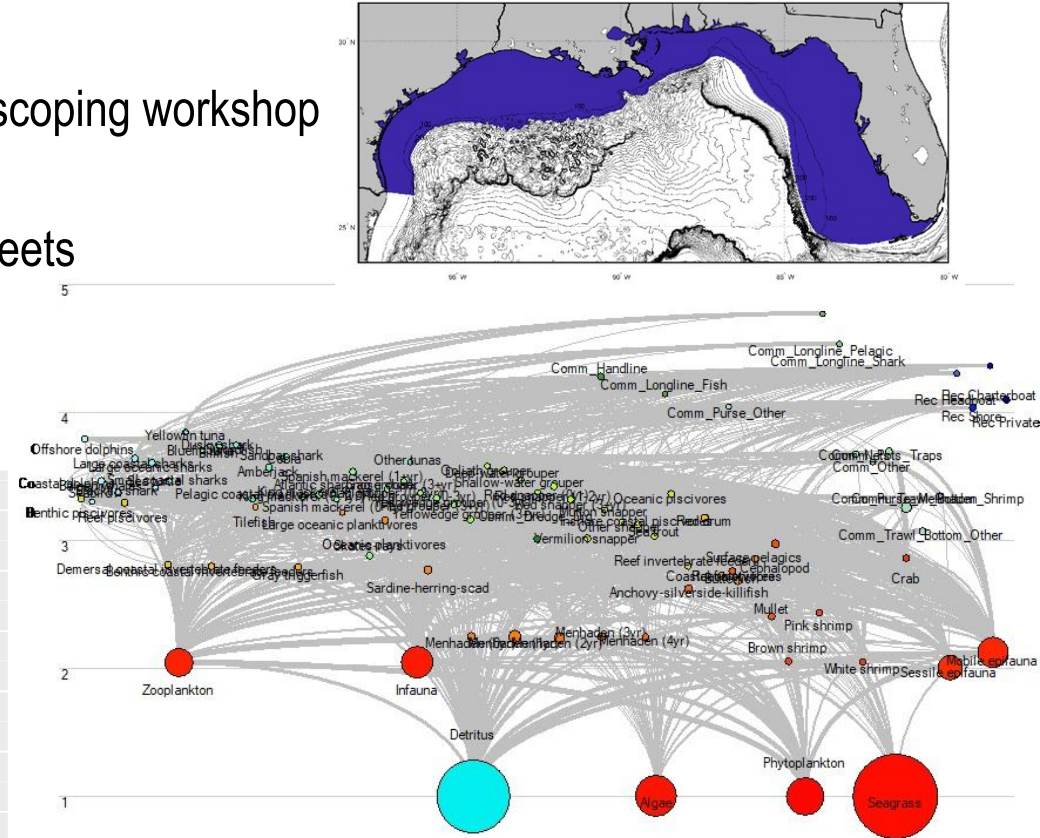
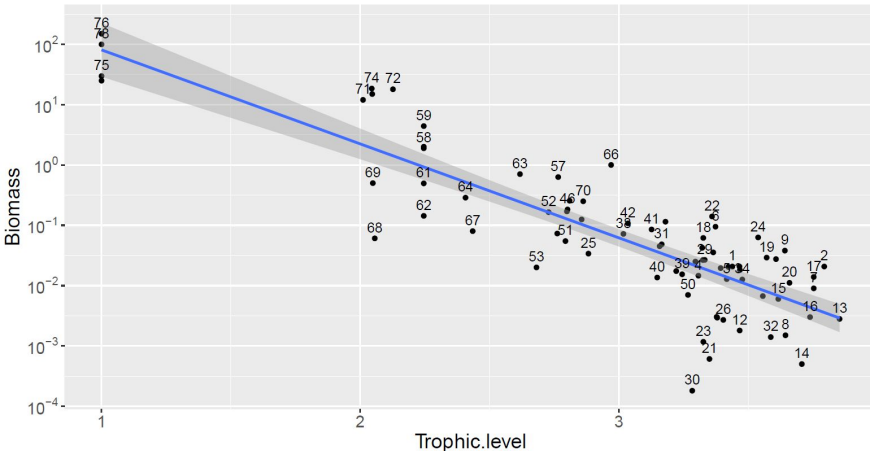
Gulf Menhaden  
*Brevoortia patronus*





# U.S. Gulf-wide Ecosystem Model // Ecopath

- Ecopath snapshot year: 1980
- Added additional age structure following scoping workshop
- 78 functional groups
- 12 commercial fleets and 4 recreational fleets
- Pre-Bal diagnostics (Link 2010)
- Best practices (Heymans et al., 2016)



# U.S. Gulf-wide Ecosystem Model // Ecosim

- Ecosim: 1980-2016
  - 160 input time series: biomass (B), catch (C), fishing mortality (F), fishing effort (E)
  - Data sources: SEDAR, SEAMAP, ICCAT, NOAA landings
  - Nutrient forcing: total Mississippi-Atchafalaya River Basin Loads
  - Fishing forcing: effort and mortality
  - Manual and automated calibration



NOAA Technical Memorandum NMFS-SEFSC-751  
doi:10.25923/zj8t-e656

TECHNICAL DOCUMENTATION OF A US GULF OF MEXICO ECOPATH WITH  
ECOSIM MODEL

BY

IGAL BERENSSTEIN, SKYLER R. SAGARESE, MATTHEW V. LAURETTA, MATTHEW  
A. NUTTALL, AND DAVID D. CHAGARIS

U.S. DEPARTMENT OF COMMERCE  
National Oceanic and Atmospheric Administration  
National Marine Fisheries Service  
Southeast Fisheries Science Center  
75 Virginia Beach Drive  
Miami, Florida 33149

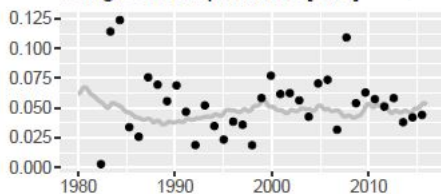
May 2021

# USGWEM // Ecosim // Biomass time series fits

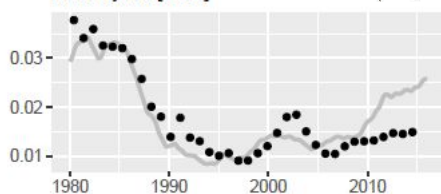
- Observed data
- **\*Assessment output**
- Ecosim prediction

[Sums of Squares]

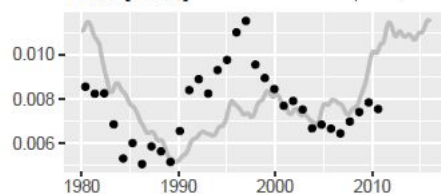
Pelagic coastal piscivores [8.79] 




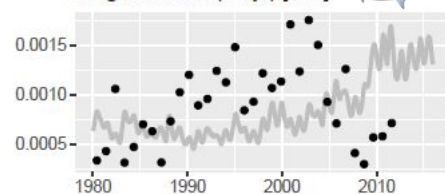
Amberjack [20.6] \*




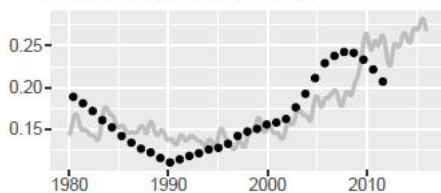
Cobia [14.14] \*




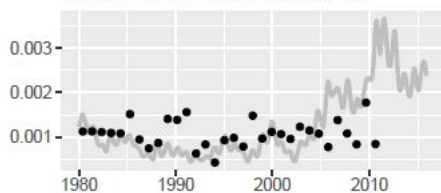
King mackerel (0-1yr) [158] \* 




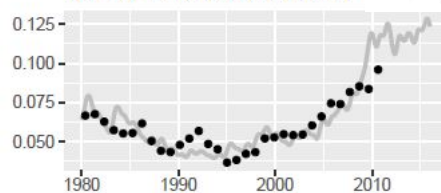
King mackerel (1+yr) [11.76] \* 




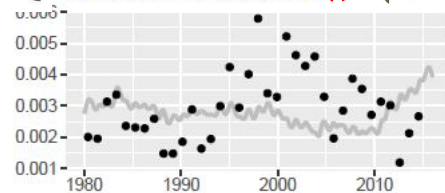
Spanish mackerel (0-1yr) [38.49] \* 




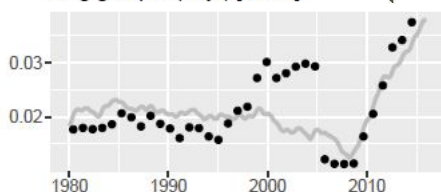
Spanish mackerel (1+yr) [5.86] \* 




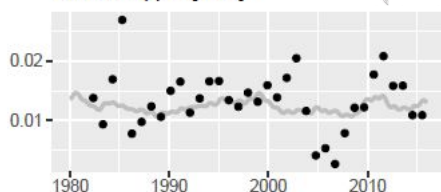
Gag grouper (0-3yr) [99.55] \* 



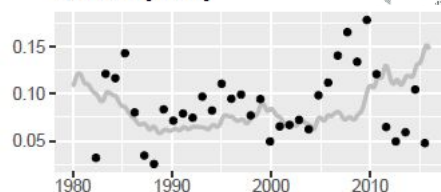
Gag grouper (3+yr) [48.66] \* 




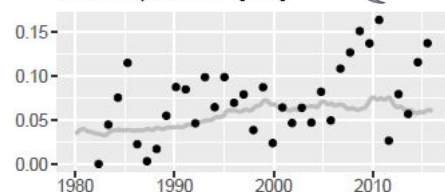
Other snapper [4.27] 



Sea trout [52.29] 



Oceanic piscivores [106] 

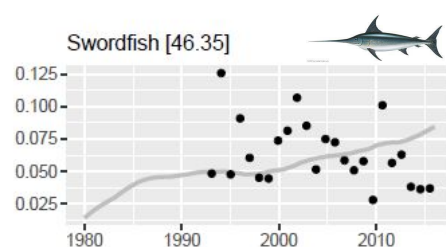
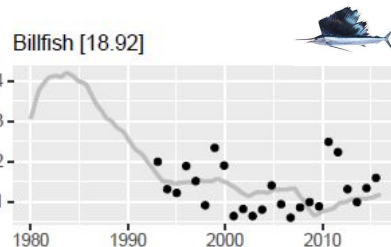
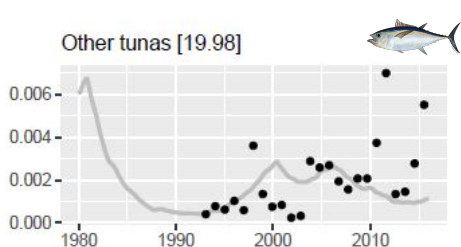
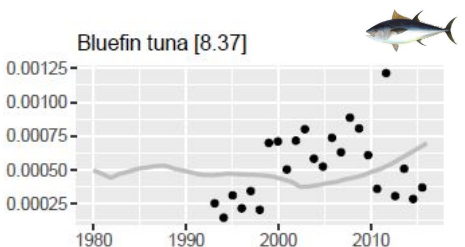
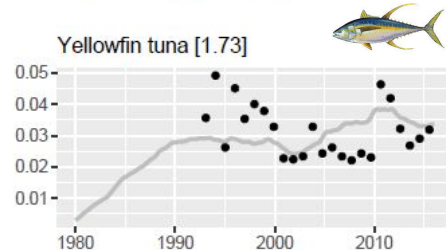
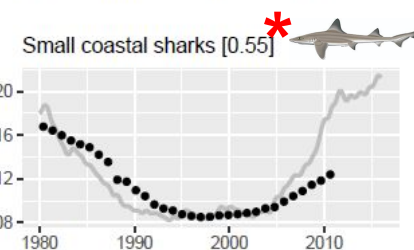
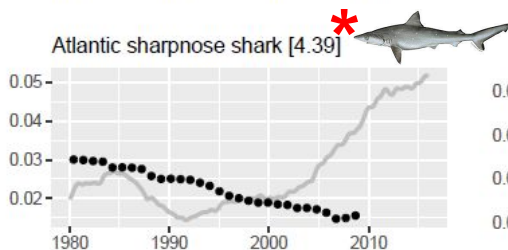
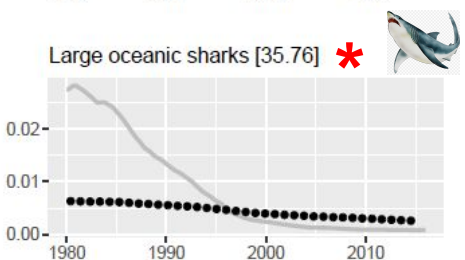
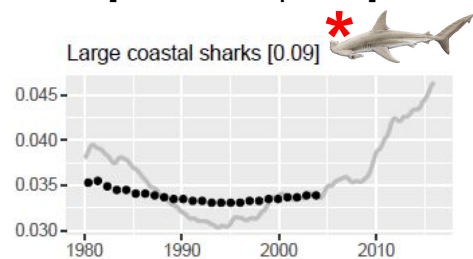
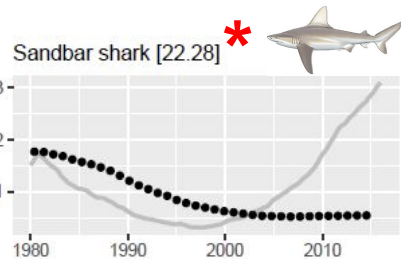
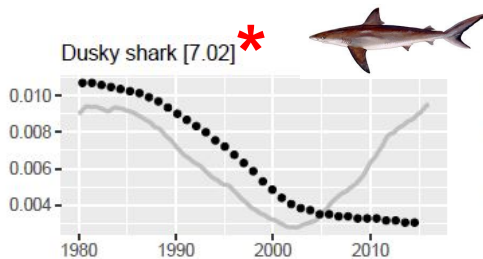
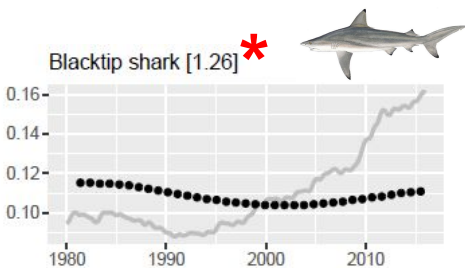


year

# USGWEM // Ecosim // Biomass time series fits

- Observed data
  - \*Assessment output
- Ecosim prediction

[Sums of Squares]

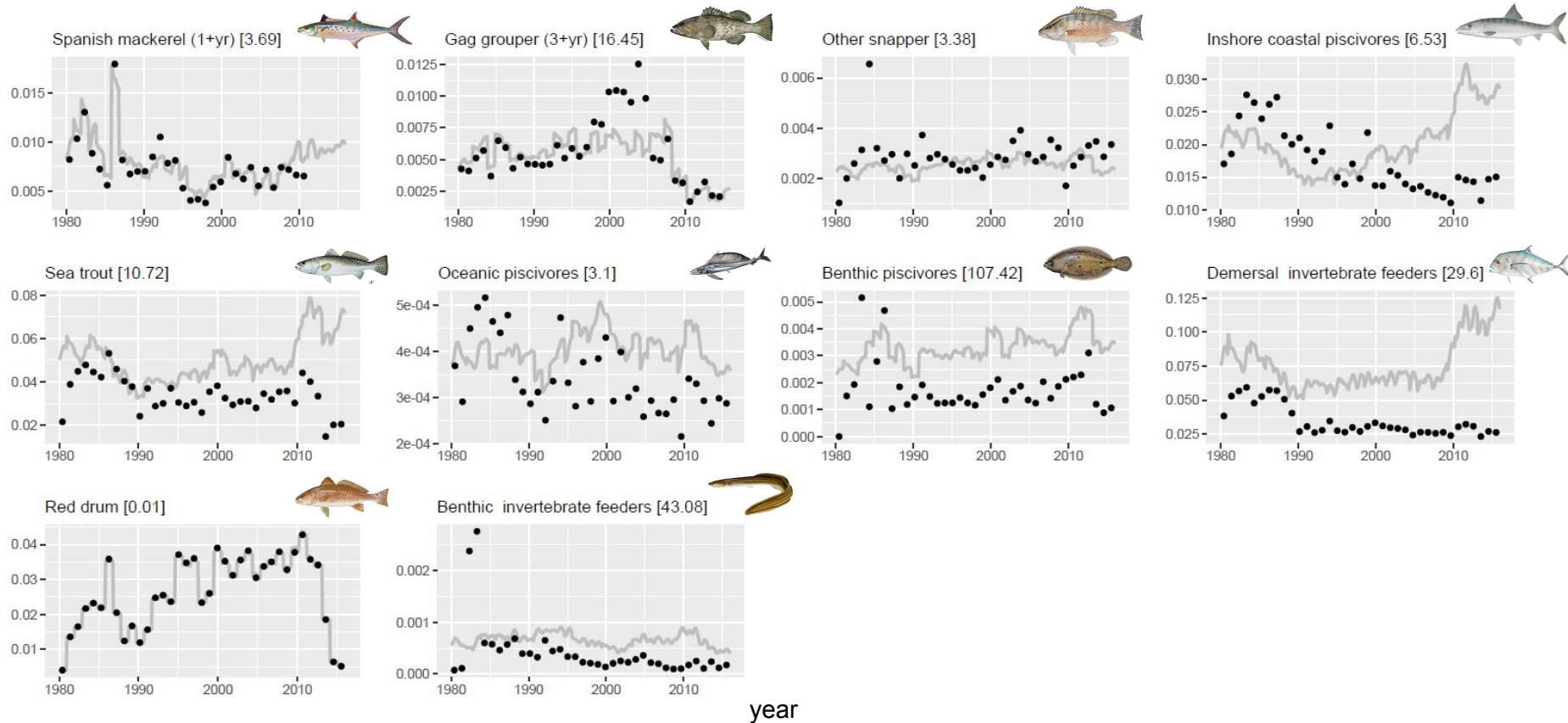


year



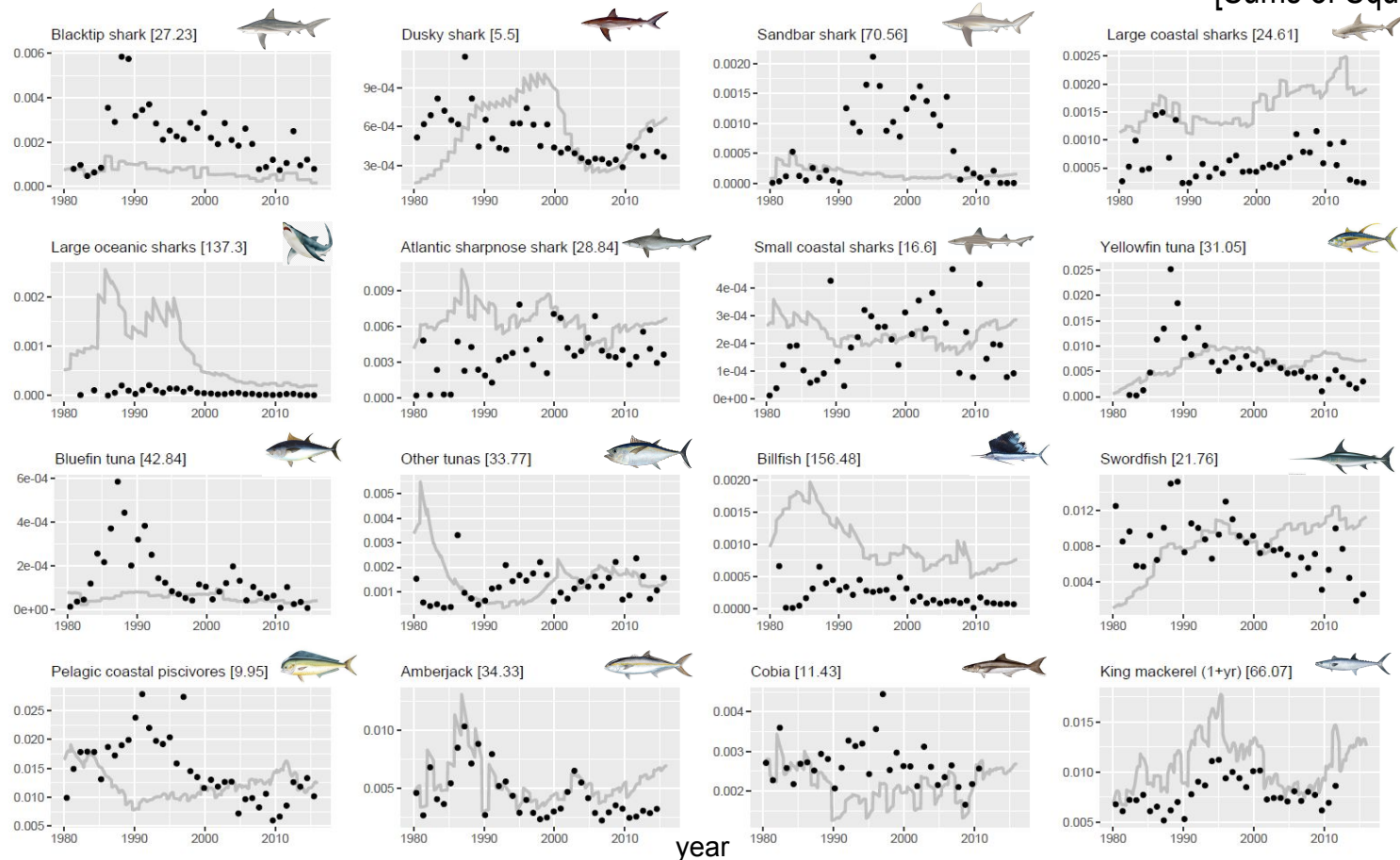
# USGWEM // Ecosim // Catch time series fits

- Observed data
- Ecosim prediction  
[Sums of Squares]



# USGWEM // Ecosim // Catch time series fits

- Observed data
  - Ecosim prediction
- [Sums of Squares]



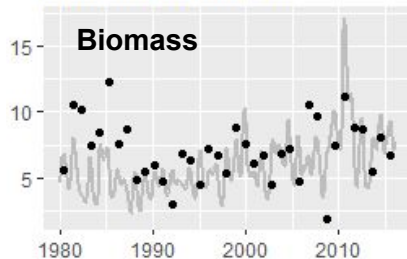
# Biomass time series fits // Gulf menhaden

- Oscillatory behavior largely driven by Mississippi River outflow

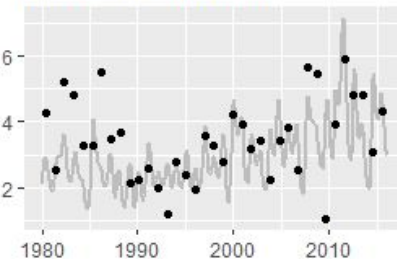


- Observed data
  - \*Assessment output
- Ecosim prediction
  - [Sums of Squares]

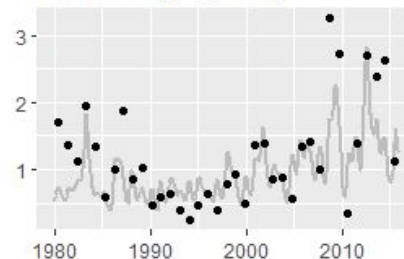
Menhaden (1yr) [81.3] \*



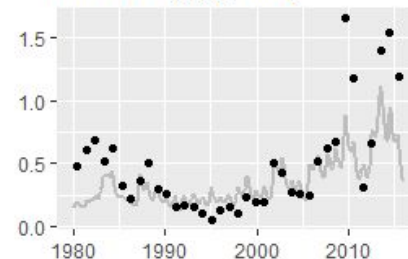
Menhaden (2yr) [68.19] \*



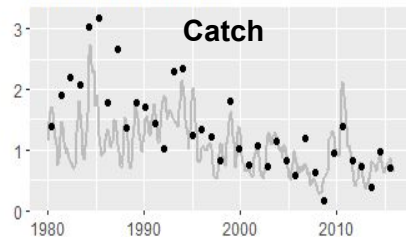
Menhaden (3yr) [113.2] \*



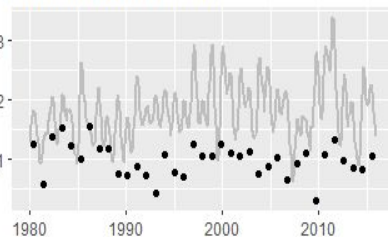
Menhaden (4yr) [152.6] \*



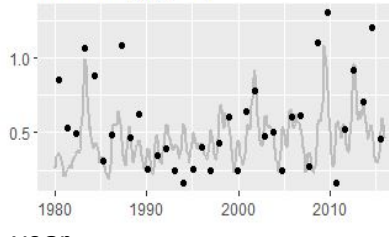
Menhaden (1yr) [34.2]



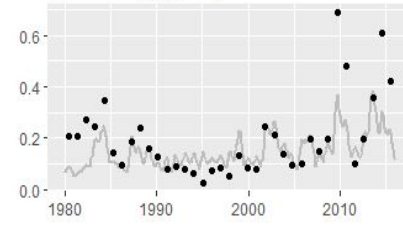
Menhaden (2yr) [146.51]



Menhaden (3yr) [47.7]



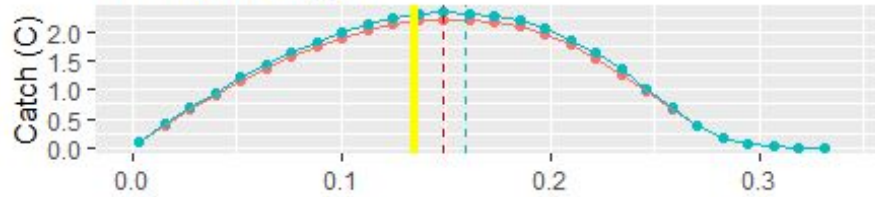
Menhaden (4yr) [65.5]



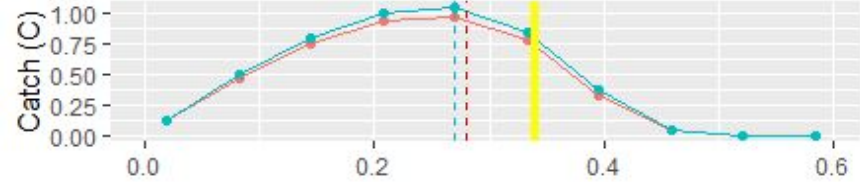
year

# Ecosim // $F_{MSY}$ analysis

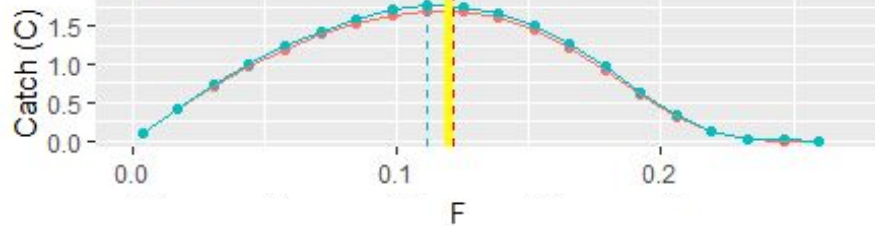
Vermilion snapper



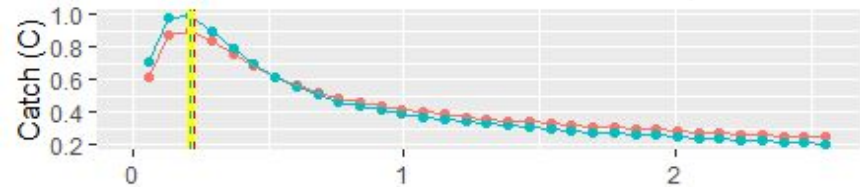
Cobia



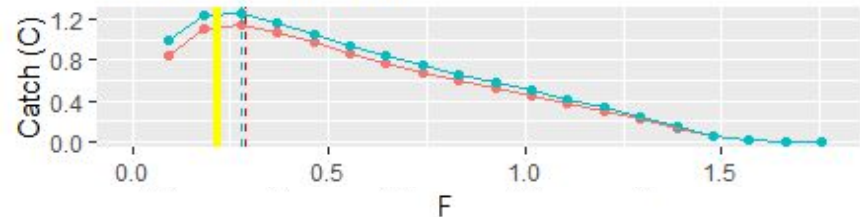
Tilefish



Amberjack



Red grouper (3+yr)



## FMSY Estimate

- Full compensatory (groups' biomasses change in response to the change in the target group)
- Stationary (other groups biomass fixed)
- Assessment



# Ecosim // $F_{MSY}$ analysis

$F_{MSY}$  estimated from the Gulfwide EwE model compared to single-species stock assessment estimates or proxies for key Gulf menhaden predators

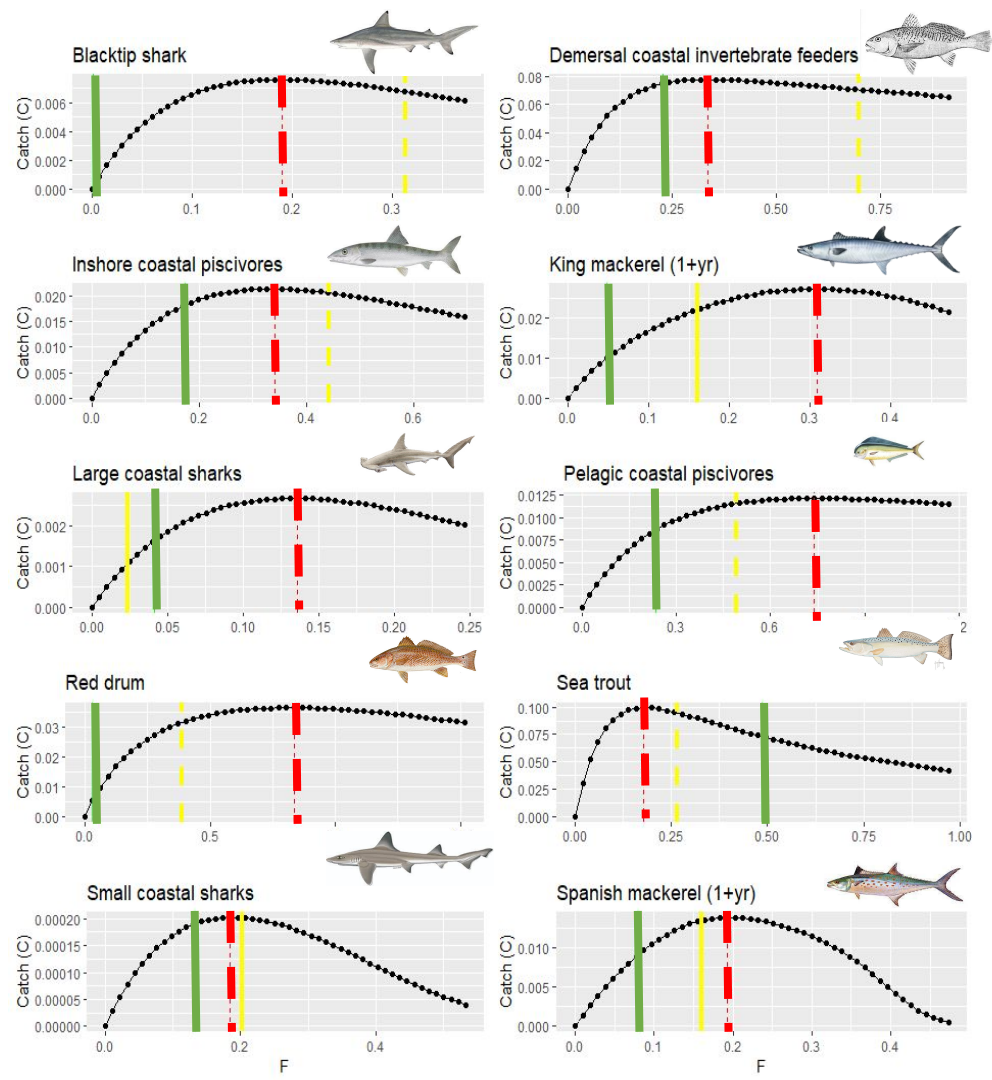
- Compensatory equilibrium analysis

-- EwE  $F_{MSY}$  estimates

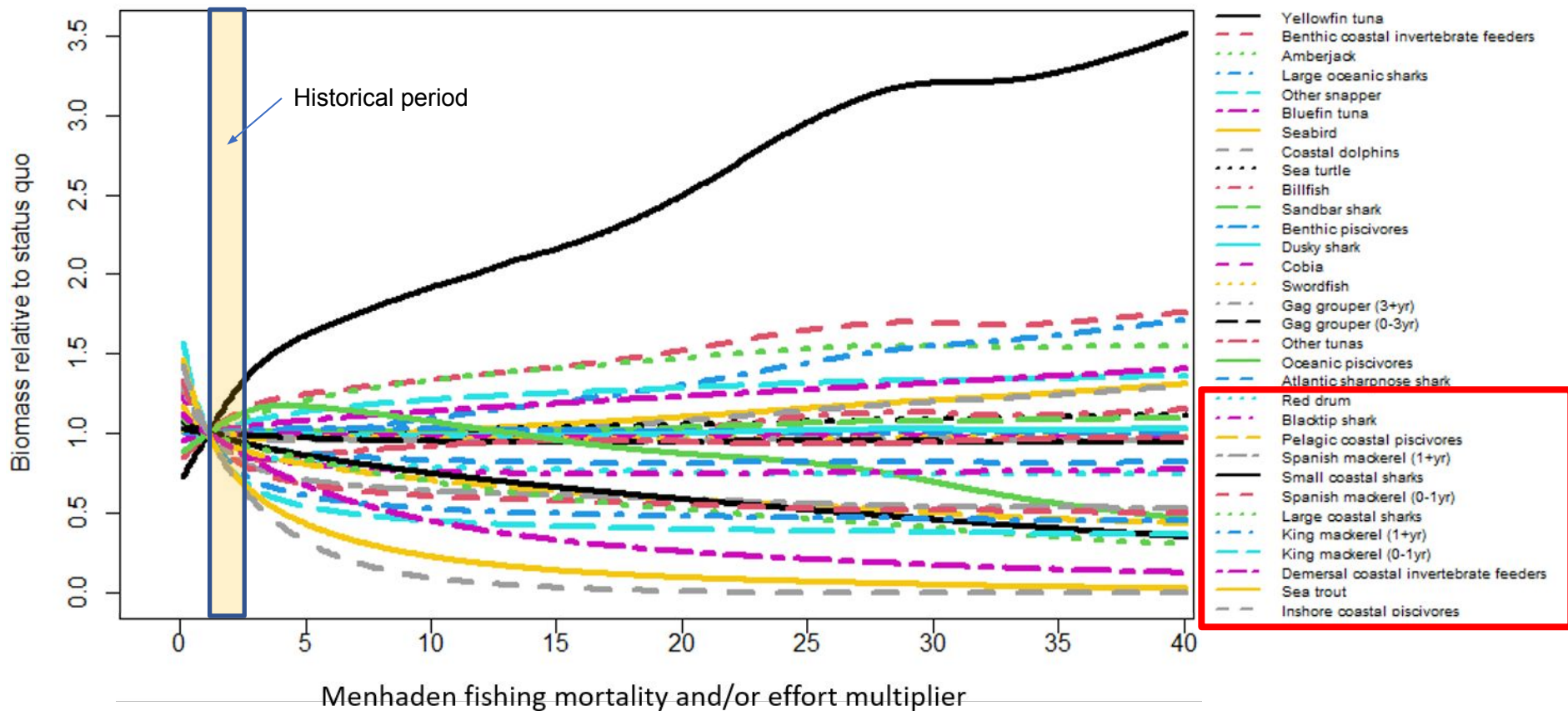
—  $F_{current}$  (2016)

— Assessment  $F_{MSY}$

-- Proxy ( $F_{MSY} = M$ )



# Gulf menhaden // Effect of F & E on predators



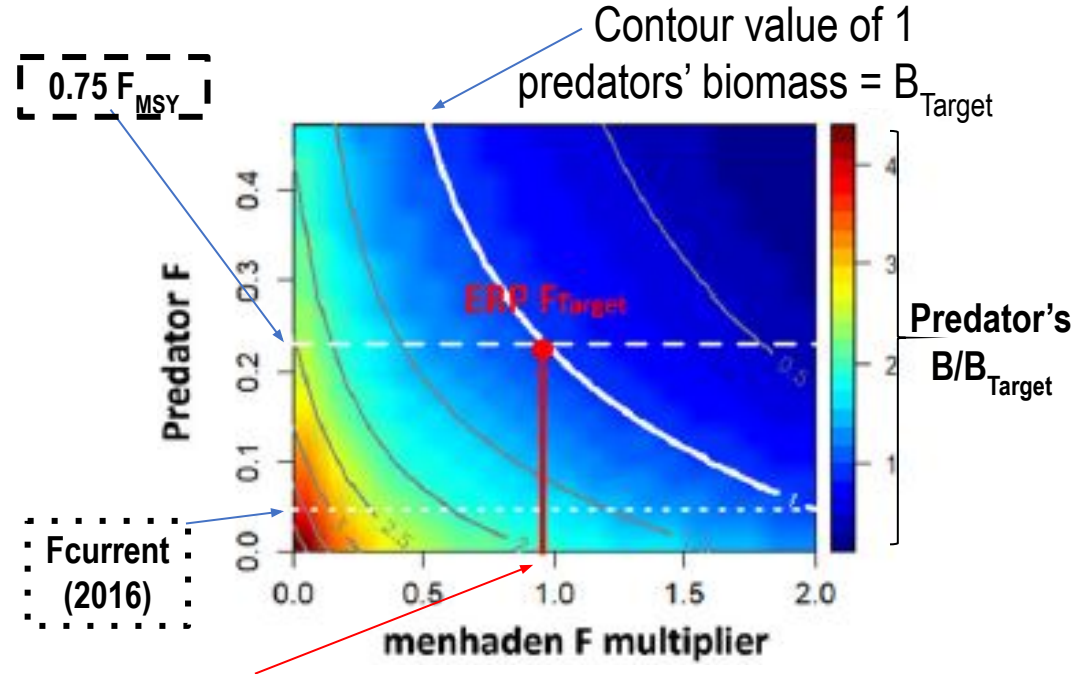
# Ecological Reference Points

- Indicator approach adapted from ERPs for Atlantic menhaden (SEDAR 69)

- Trade-off plot (color map)**

- Ratios of scenario's biomass relative to the target biomass ( $B_{\text{Target}}$ ) for menhaden predators as a function of variation in fishing mortalities for Gulf menhaden and their predators

Berenshtein I, Sagarese SR, Lauretta MV, Schueller AM and Chagaris DD (2023) Identifying tradeoffs and reference points in support of ecosystem approaches to managing Gulf of Mexico menhaden. Front. Mar. Sci. 9:935324. doi: [10.3389/fmars.2022.935324](https://doi.org/10.3389/fmars.2022.935324)

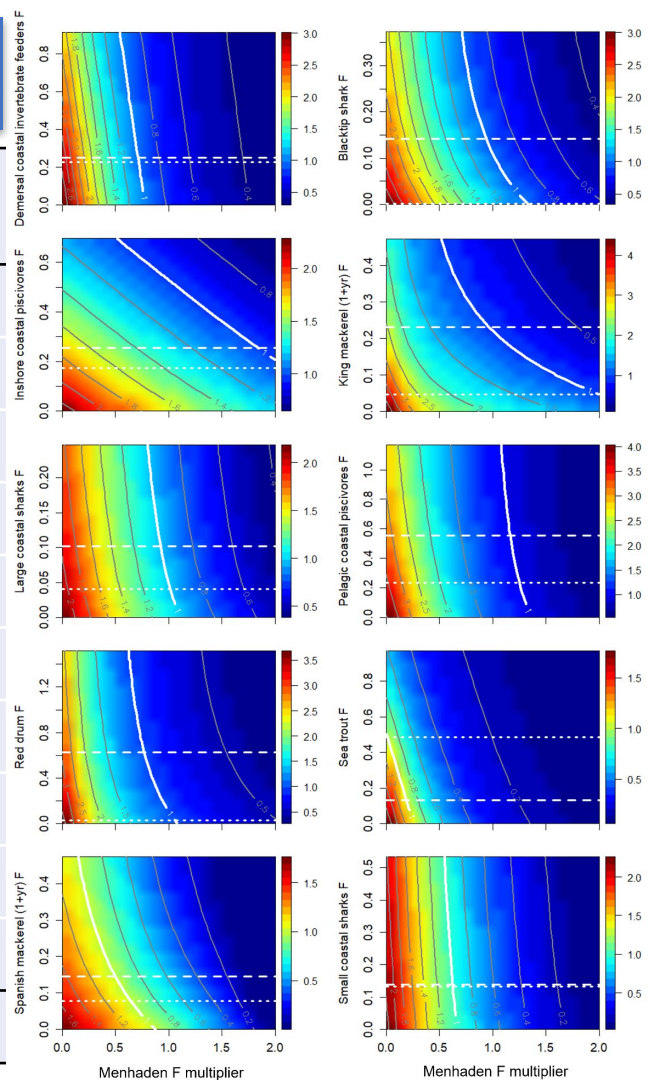


**ERP  $F_{\text{Target}}$**  = the multiplier of menhaden F that results in the  $B_{\text{Target}}$  of a given menhaden predator

- x-value of the intersect between the "1" contour and the horizontal dashed line (predator  $F_{\text{target}}$ )

# Ecological Reference Points

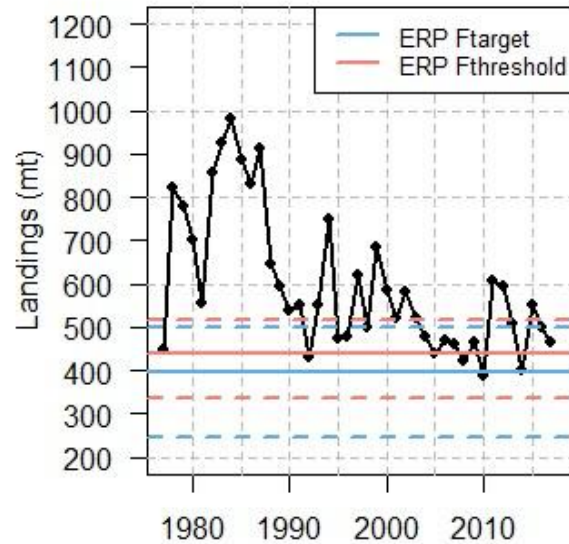
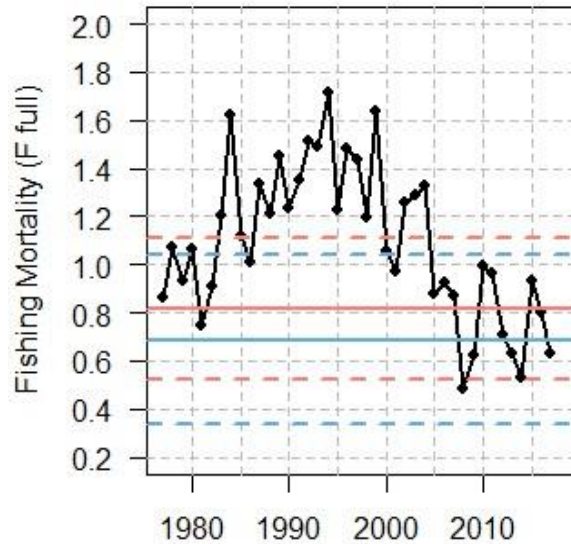
Group name	$F_{MSY}$	$B_{MSY}$	$F_{target}$	$B_{target}$	ERP $F_{target}$	ERP $F_{threshold}$
Blacktip shark	0.19	0.04	0.143	0.051	0.939	<b>1.232</b>
Demersal coastal invertebrate feeders	0.336	0.231	0.252	0.292	0.712	0.918
Inshore coastal piscivores	0.341	0.063	0.256	0.081	<b>1.837</b>	<b>&gt;2.0</b>
King mackerel (1+yr)	0.308	0.089	0.231	0.113	0.964	<b>1.286</b>
Large coastal sharks	0.136	0.02	0.102	0.026	0.924	<b>1.252</b>
Pelagic coastal piscivores	0.739	0.016	0.554	0.022	<b>1.164</b>	<b>1.559</b>
Red drum	0.836	0.044	0.627	0.058	0.770	<b>1.048</b>
Sea trout	0.178	0.557	0.134	0.699	0.189	0.299
Small coastal sharks	0.185	0.001	0.139	0.001	0.624	0.870
Spanish mackerel (1+yr)	0.193	0.072	0.145	0.09	0.500	0.738
<b>Mean across all 10 groups</b>					<b>0.86</b>	<b>1.02</b>





# ERPs // Gulf menhaden projections

- Gulf menhaden projections were run in Monte Carlo bootstrap mode using the 2019 Beaufort Assessment Model and the above-defined ERP target and threshold

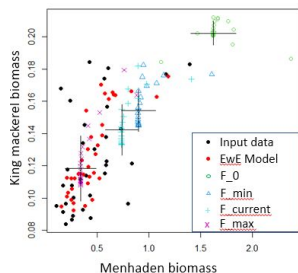


Observed landings since 2003 generally within the range of the projected equilibrium

- Horizontal solid lines = mean ERP **Ftarget** and **Fthreshold** averaged over all 10 predator species
- Horizontal dashed lines =  $\pm 1$  standard deviations

# Additional Analyses

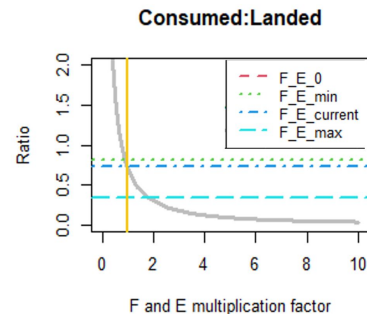
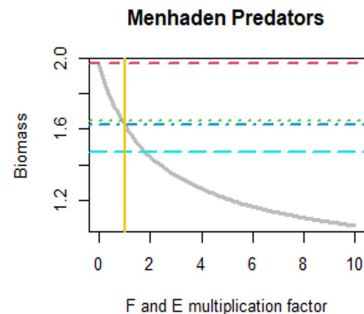
## • Ecological sensitivity scenarios



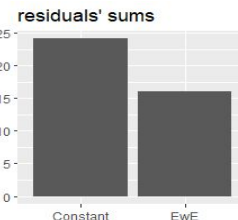
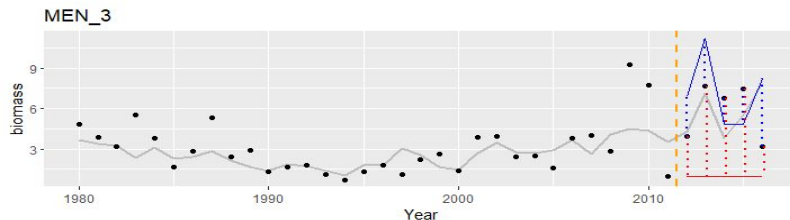
### Diagnostics

- Input data – realistic relationship
- Model recreates observed pattern
- Menhaden F Scenarios:
  - F\_0 (zero)
  - F\_min (historical minimum)
  - F\_current (2016)
  - F\_max (historical maximum)

## • Ecological indicators



## • Forecasting ability of the Ecosim model



- Provided a time series of M that was incorporated into a sensitivity run during the 2021 GDAR 03 Gulf Menhaden Stock Assessment Update

# USGWEM EwE // Key Takeaways

- The US Gulf-wide EwE model serves as a tool that could be used to address a number of ecological questions
- Gulf menhaden analysis demonstrated how  $B_{\text{Target}}$  could be achieved for a given predatory group, by modifying menhaden and/or the group's fishing pressure, and based on this relationship, ERPs were established
  - [https://igalberenshtein.shinyapps.io/r\\_shiny\\_app\\_menhadenf\\_4\\_pub/](https://igalberenshtein.shinyapps.io/r_shiny_app_menhadenf_4_pub/)
- Data and modeling limitations discussed in tech memo & Berenshtein et al. (2023)

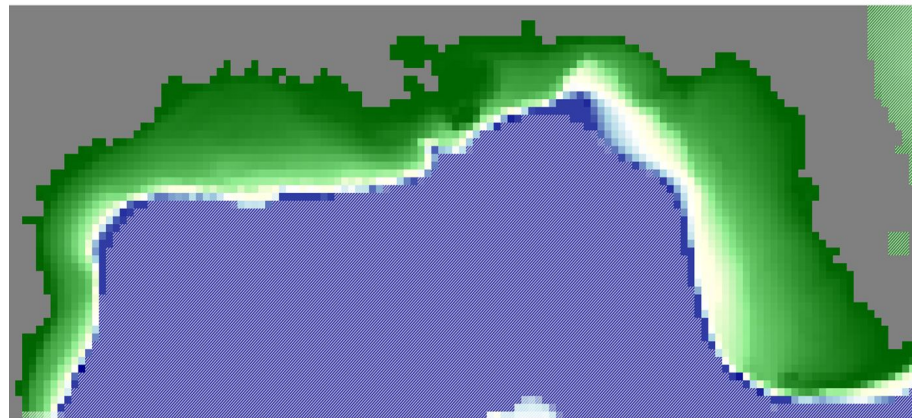
Table 26. Summary of data needs and considerations for applying the U.S. Gulf-wide Ecopath with Ecosim model for each functional group. Usability score (Score) includes: (1) model could be readily modified within a typical model development-review cycle; (2) model needs additional data and a typical model development-review cycle; (3) extensive data needed (e.g., long-time series) or the model is not feasible. Number of diet observations (i.e., studies, see Figure S1.1 for details) and number of stomachs feeding into the diet matrix are shown, as well as time series currently included in the model.

Functional Group	Notable species and importance	Diet observations (Stomachs)	Time series	Score	Data Needs and Considerations
Coastal dolphins	All protected under Marine Mammal Protection Act (MMPA)	27 (739)	-	3	Need species-specific biomass, incidental bycatch in fisheries, diet composition, and time series

# USGWEM // Next steps and Future Applications

## Modeling needs:

- Build in spatial component to better capture spatial dynamics, such as species overlap and bycatch
- Incorporate the effect of additional environmental drivers (e.g., temperature and hypoxia)
- Follow approach used for Atlantic Menhaden (SEDAR 69)
  - Develop alternative model configurations or models
    - MICE model focused on key predator groups
  - Technical review akin to a stock assessment review



## Ecospace

- Spatially explicit simulations
- Input: dispersal rates, habitat maps, habitat preferences, fishing areas, MPAs, port locations
- Spatial-temporal drivers



Gulf of Mexico Fishery  
Management  
Council  
SSC Meeting



*Southeast  
Fisheries  
Science  
Center*



## Part II Ecosystem modeling the US Gulf of Mexico: Ecospace

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Holden E. Harris, Skyler R. Sagarese, David D. Chagaris

UNIVERSITY  
OF MIAMI  
ROSENSTIEL  
SCHOOL of MARINE &  
ATMOSPHERIC SCIENCE



# USGWEM // Part II // Outline

## 1. Context

- a. Ecospace
- b. and its GoM applications

## 2. Data inputs

- a. Habitat
- b. Environmental drivers
- c. Preference functions
- d. Dispersal rates

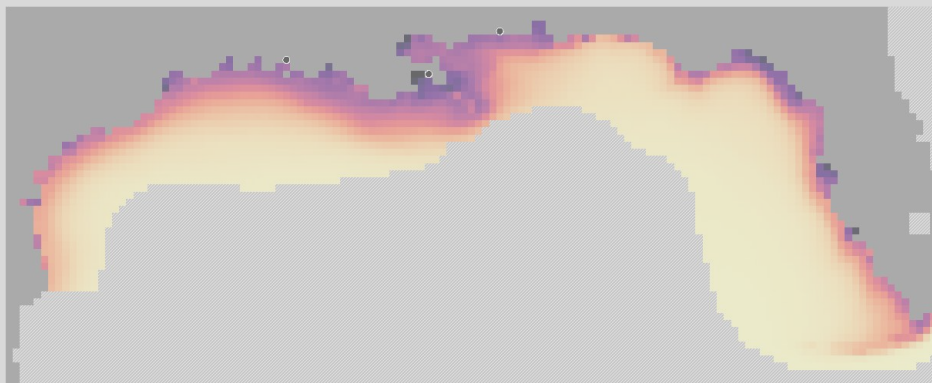
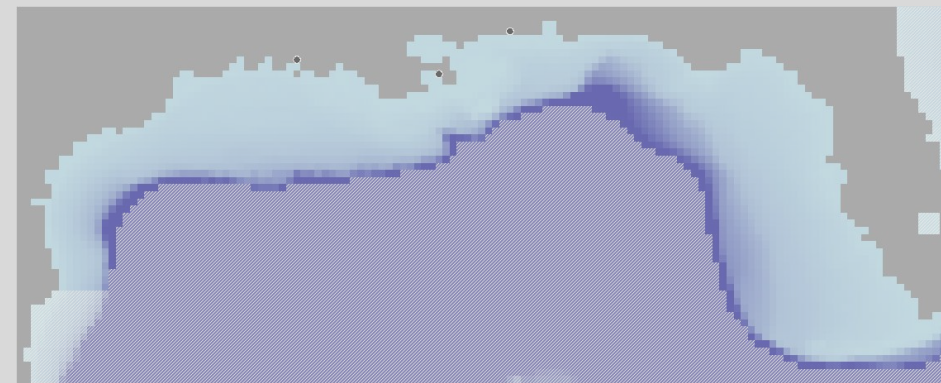
## 3. Next steps

- a. Fitting & calibration
- b. Qualitative & fisher input

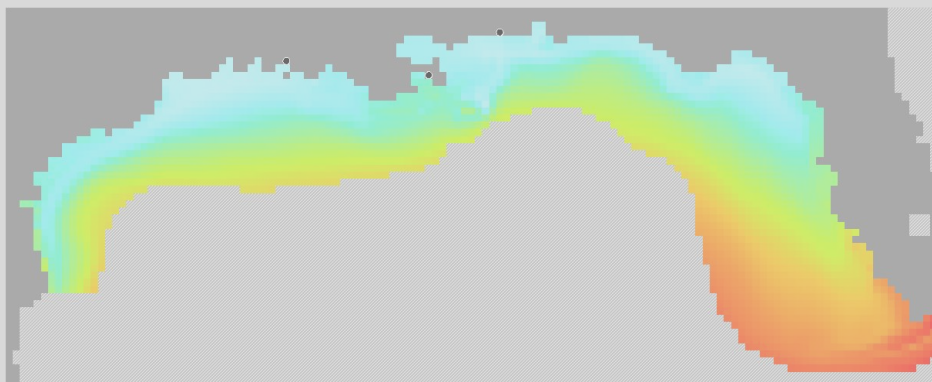
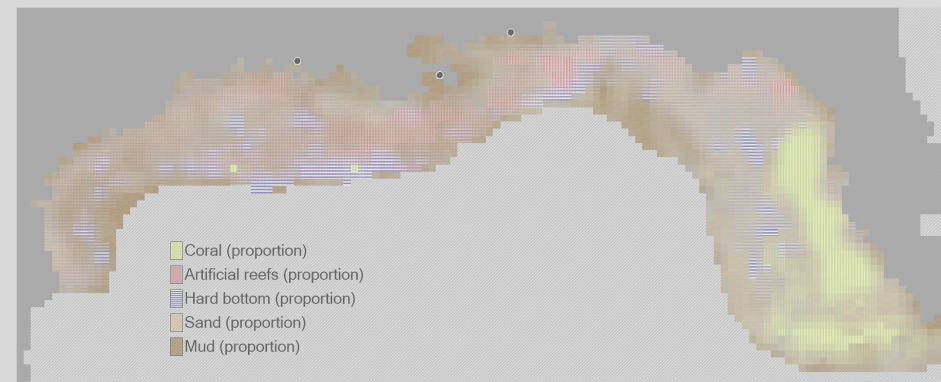
## 4. Potential research directions

- a. Operationalization into Fisheries Ecosys. Plans
- b. Spatial-temporal closures & bycatch reduction
- c. Climate change
- d. Structured habitat (O&G & wind)

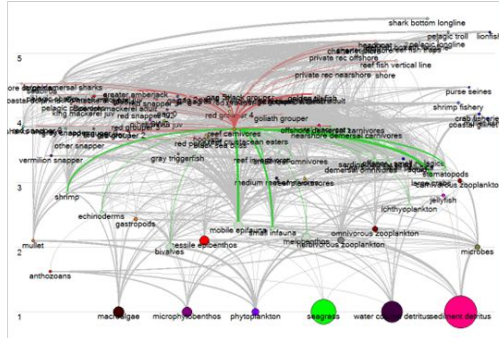




## Introduction // Ecospace

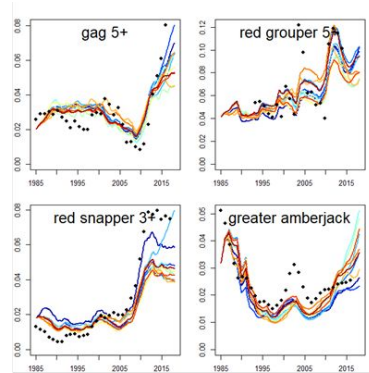


# Context // Ecospace



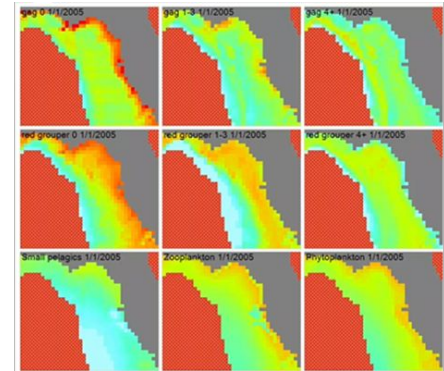
## Step 1 - Ecopath

Static mass-balanced model



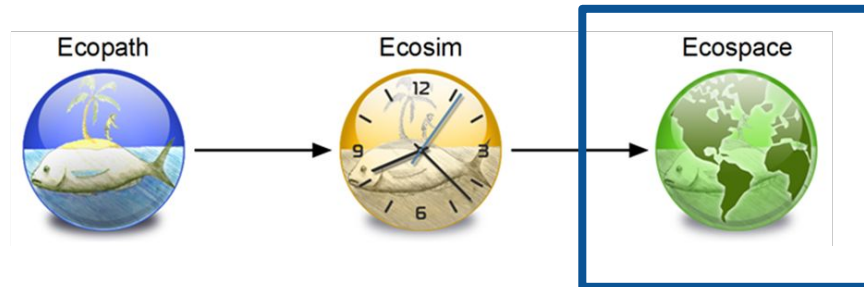
## Step 2 - Ecosim

Time dynamic simulations



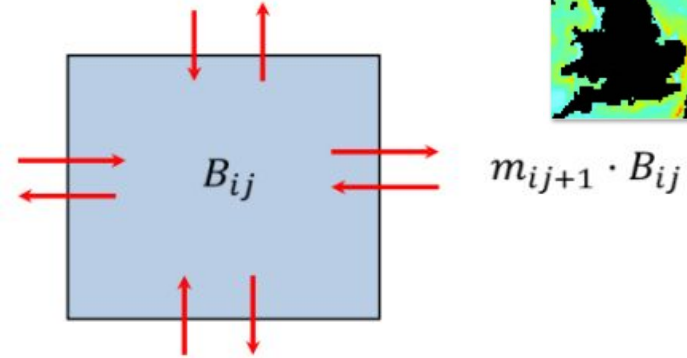
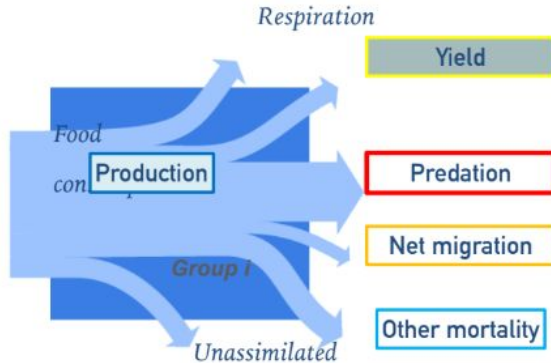
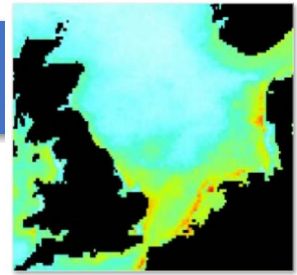
## Step 3 - Ecospace

Spatial-temporal dynamics



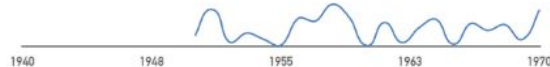


# Context // Ecospace



- For each cell, the immigration rate  $m_i$  is the sum of emigration flows from the four surrounding cells.
- Immigration / emigration rates based on
  - Abiotic factors: **habitat & environmental drivers** (e.g., temp. & salinity)
  - And biotic factors: to **feed** & avoid **predators**
- Fishing effort uses Ecopath base or predicts fleet dynamics with a gravity model

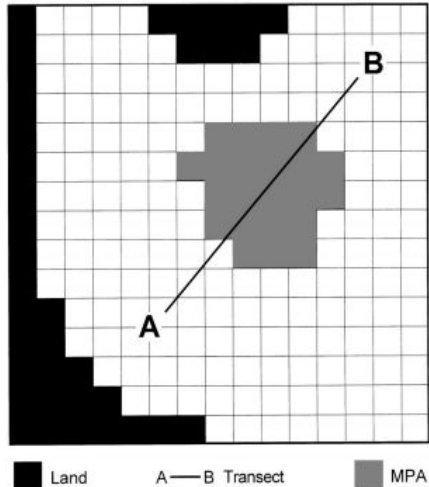
$$\frac{dB_i}{dt} = g_i \cdot \sum_{Pred\_j=1}^n c_{ji}(B_i, B_j) - \sum_{Pred\_j=1}^n c_{ij}(B_i, B_j) + I_i - (M0_i + F_i + e_i) \cdot B_i$$



# Context // Ecospace // MPAs

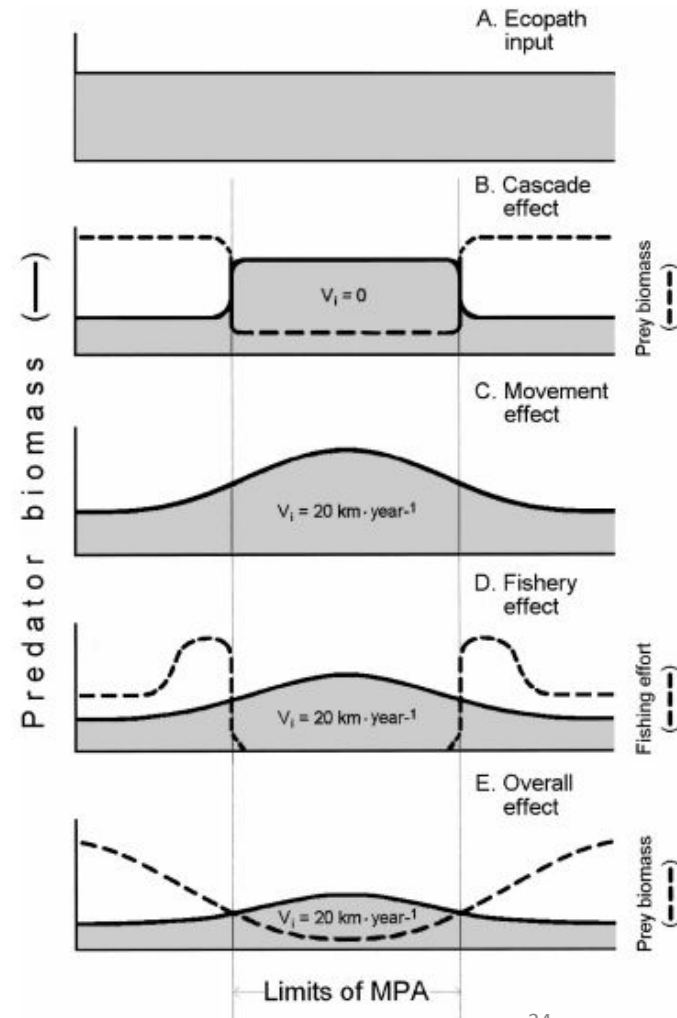
Illustrative (and first) Ecospace example:

[Walters et al. 1999](#) examining MPAs



Panels:

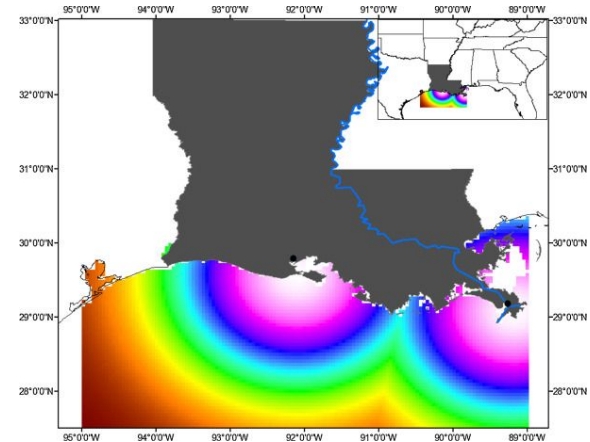
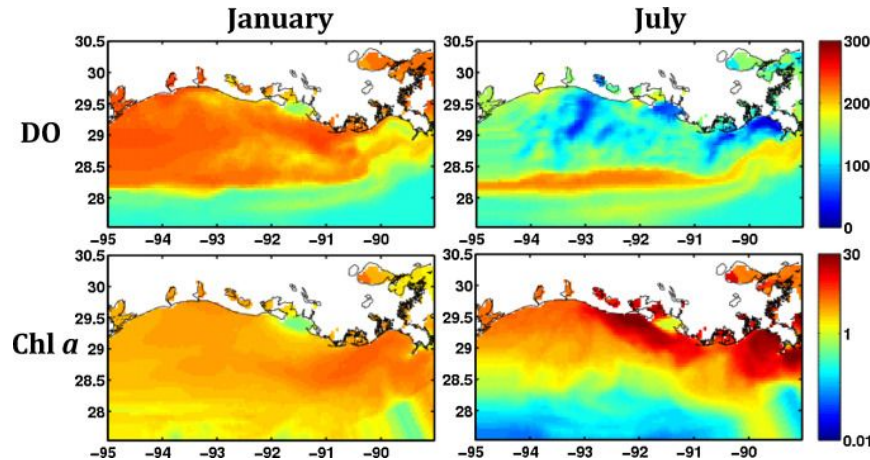
- A. Uniform EwE model
- B. No spatial mixing
- C. Mixing with higher biomass in MPA
- D. Aggregated fishing along MPA edges
- E. Overall effect



# Ecospace // GoM Applications // Hypoxia

## [de Mutsert et al., 2016](#): Effects of **hypoxia** on LA fisheries

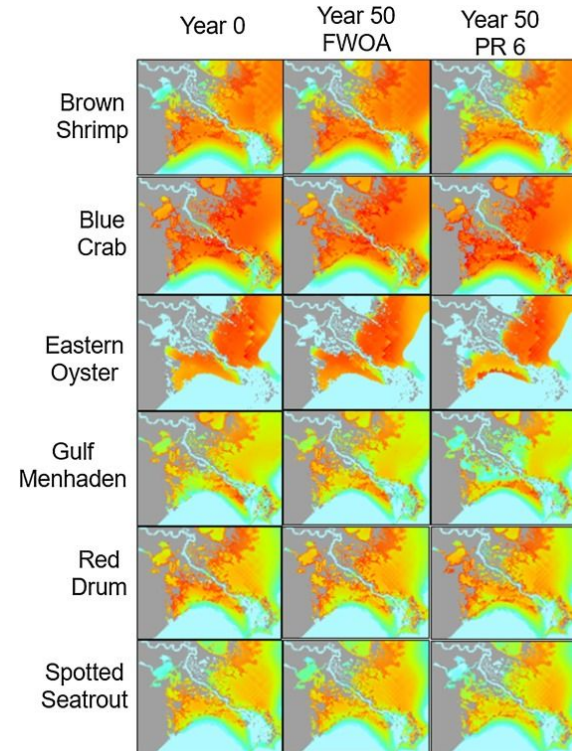
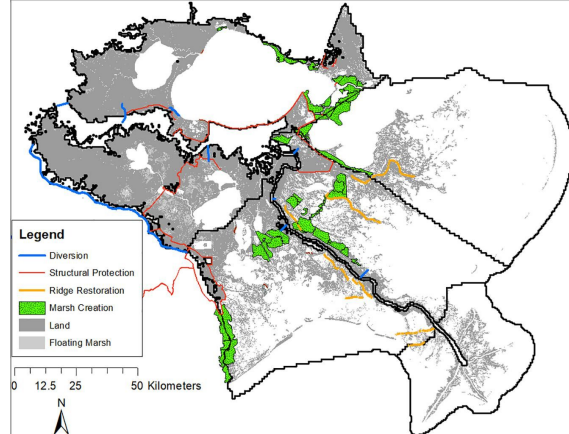
- Nekton able to move out of poor habitat conditions
- Primary production gains largely counteracted deleterious effects of hypoxia



# Ecospace // GoM Applications // Habitat Restoration

## [de Mutsert et al., 2017](#): Effects of Miss. Rvr. freshwater diversions on LA fisheries

- ID'd winners & losers from restoration efforts
- Results used by restoration authority to prioritize diversion projects





# Ecospace // GoM Applications // HAB

[Vilas et al. 2023](#): Effects of **red tide** on WFS gag grouper and fisheries

- Incorporated lethal and sublethal effects
- Produced time-series of episodic mortality that can inform stock assessment

**Ecosystem Modeling of Red Tide Impacts on West Florida Shelf Fisheries**

David Chagaris, Daniel Vilas, Joe Buszowski, and Zach Siders

In Collaboration with:  
Jeroen Steenbeek, Lisa Ailloud, Skyler Sagarese, Ted Switzer, Dylan Sinnickson, Kim de Mutsert, Matthew Lauretta, Carl Walters, Rob Ahrens

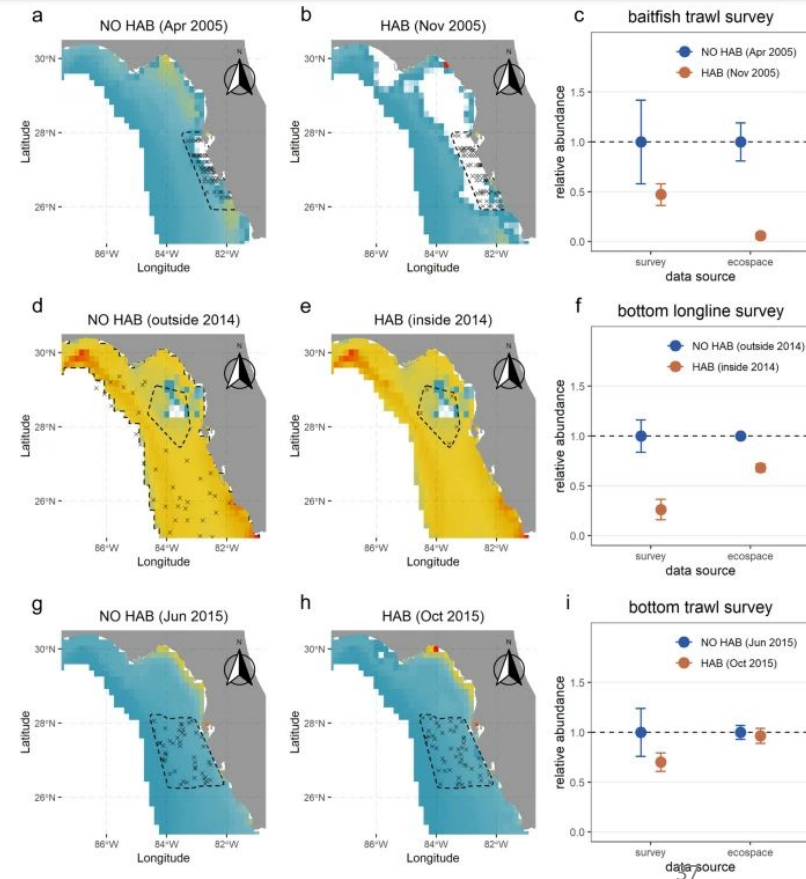
Gulf SSC Meeting, September 28, 2021

**UF**  
UNIVERSITY of  
FLORIDA  
NATURE COAST  
BIOLOGICAL STATION

**Ecopath**  
International  
Initiative

**NOAA**  
NATIONAL OCEANOGRAPHIC  
ADMINISTRATION  
U.S. DEPARTMENT OF COMMERCE

**RESTORE**  
SCIENCE PROGRAM



# Ecospace // GoM Applications // Climate Change

## Effects of **freshwater provisioning** on estuary ecosystems & fisheries

- Spatial-temporal changes in salinity, temperature, and nutrients
- Concurrent bottom-up & top-down drivers



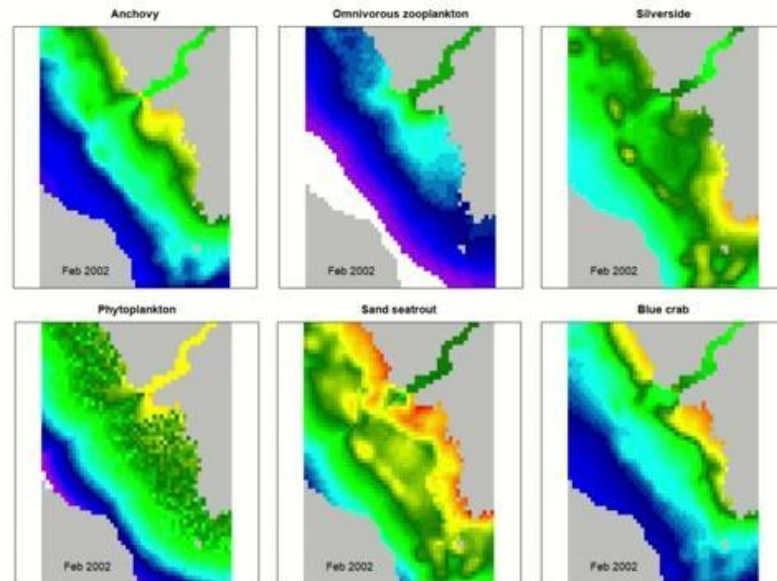
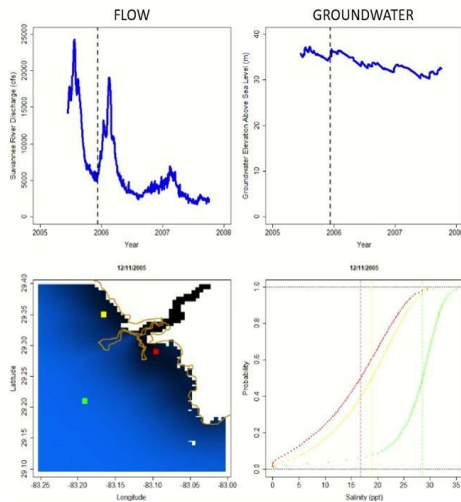
Changes in Land Use and Climate

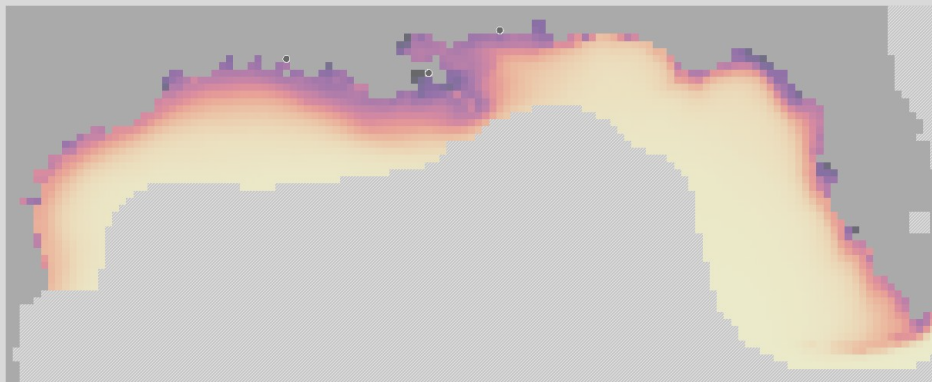
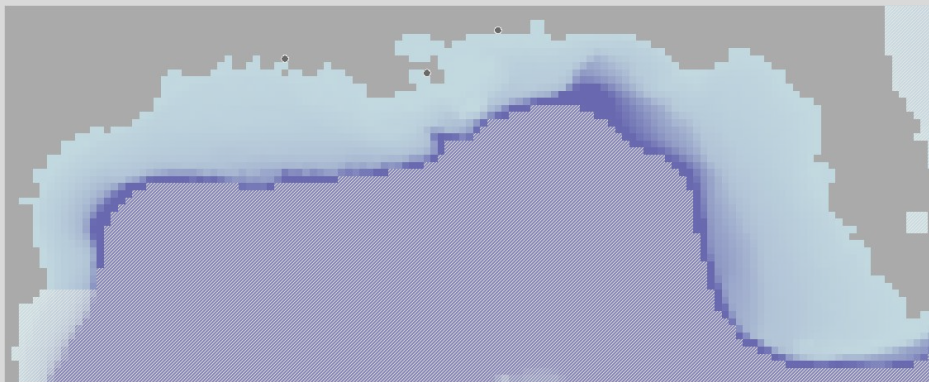


Coastal Ecosystems

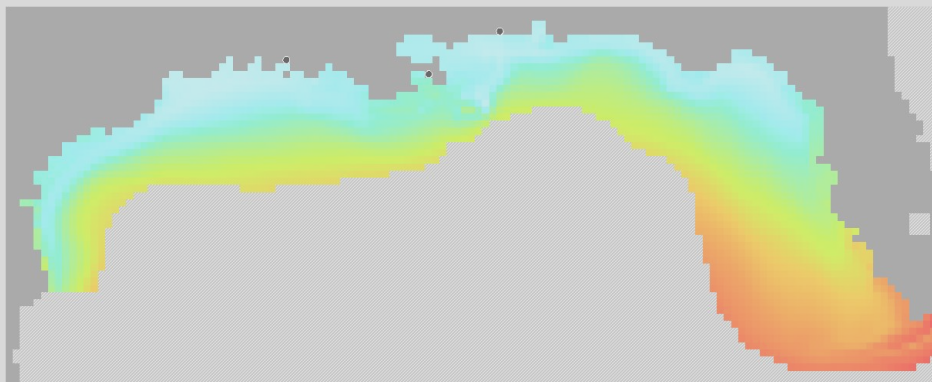
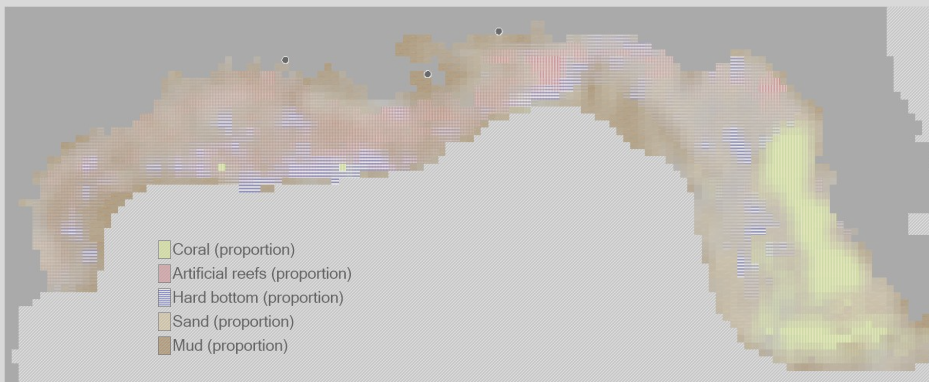


### Historic Salinity (2005-2007)





## USGWEM Ecospace // Data Inputs



# USGWEM Ecospace // Data inputs // Overview

## Maps

- Habitat maps (static)
  - Depth / basemap
  - Habitats
    - Hardbottom
    - Coral
    - Artificial reefs
    - Sand
    - Mud
- Environmental drivers (Spatial-temporal; ST)
  - Primary production
  - Salinity
  - Temperature (bottom, surface, avg.)

## Responses

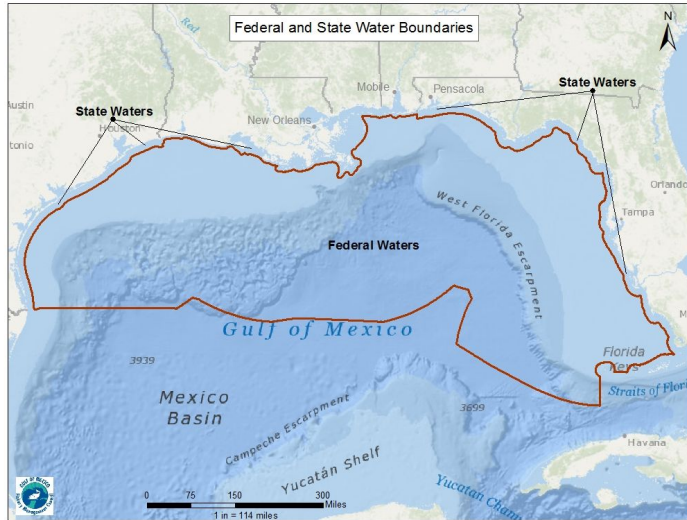
- Dispersal rates
- Preference functions



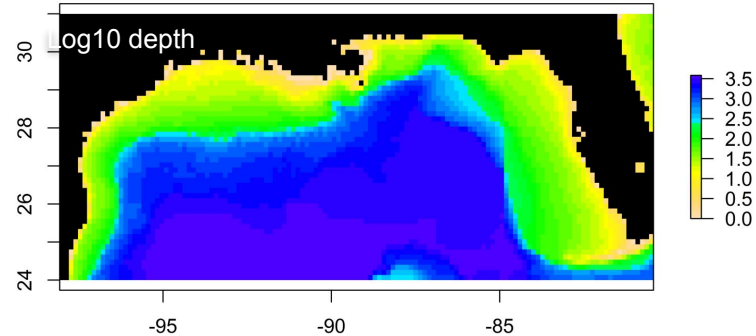


# USGWEM // Spatial extent and base map

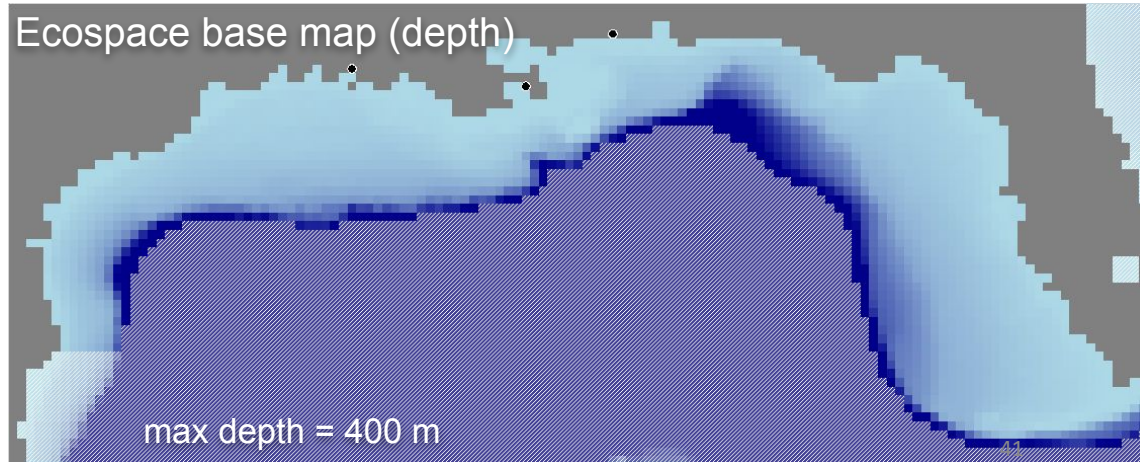
## GOM Managed LME



## Spans >3 orders of magnitudes



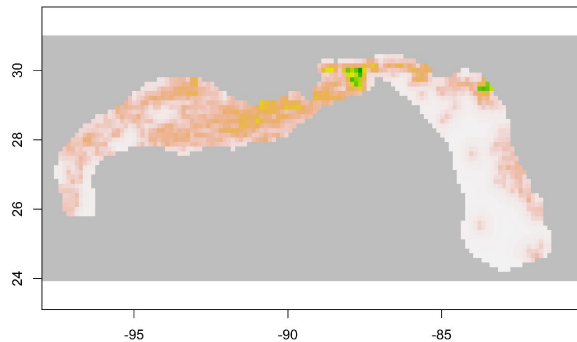
## Ecospace base map (depth)



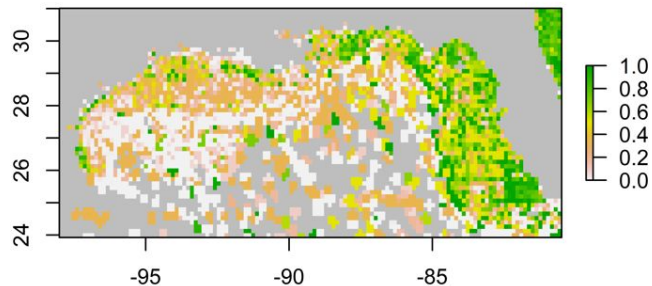
[GitHub repository for making basemap](#)

# Data Inputs // Habitat Maps

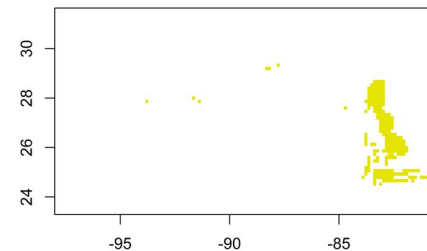
Artificial reefs (scaled inverse distance)



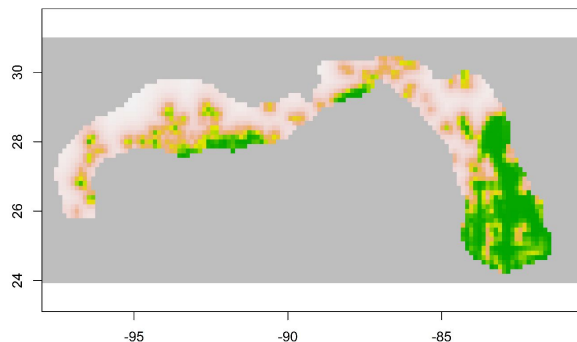
Sand



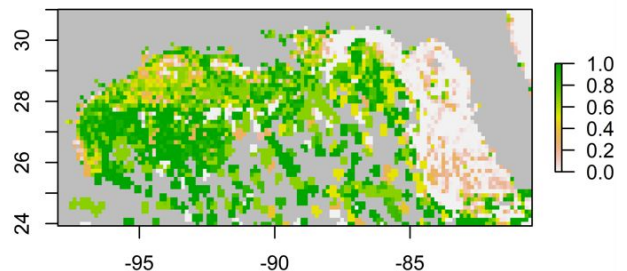
Coral habitat



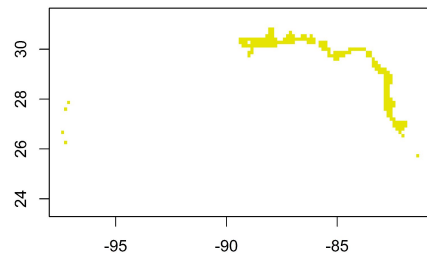
Hard bottom (scaled inverse distance squared)



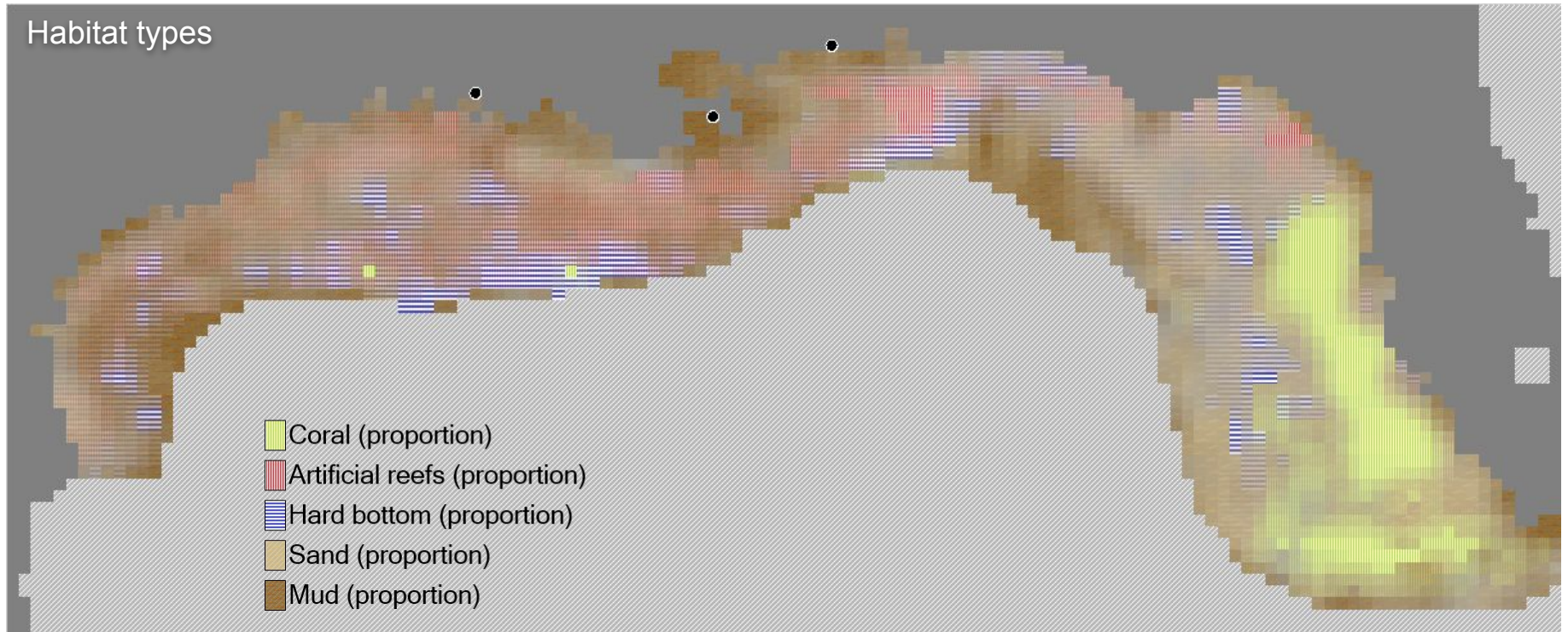
Mud



Seagrass habitat

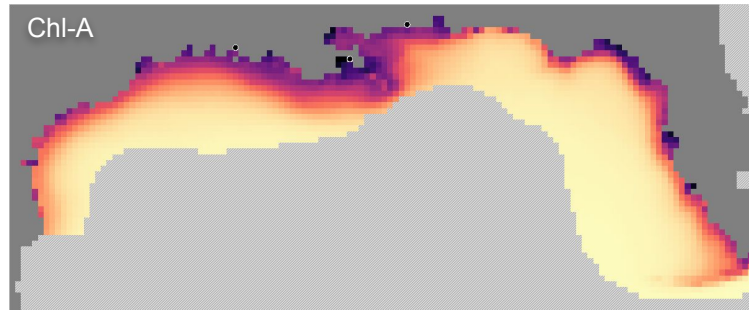
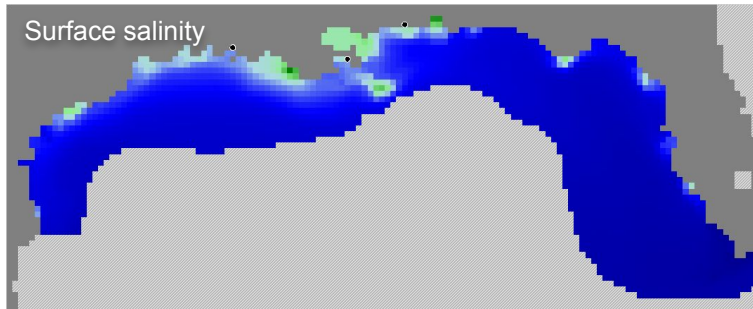
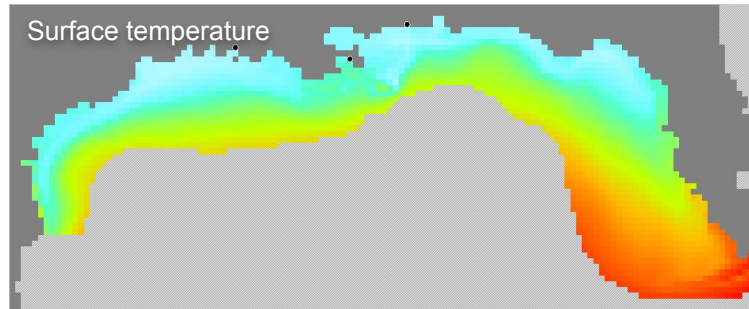
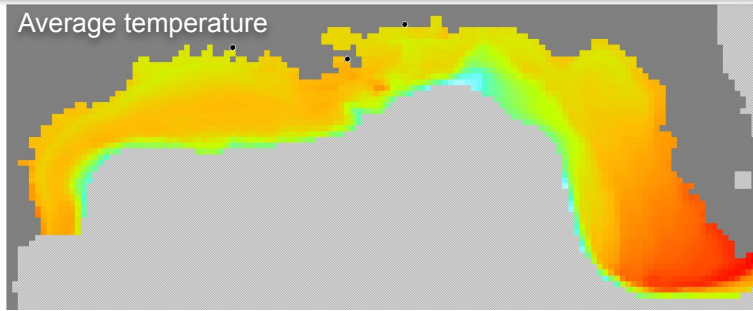
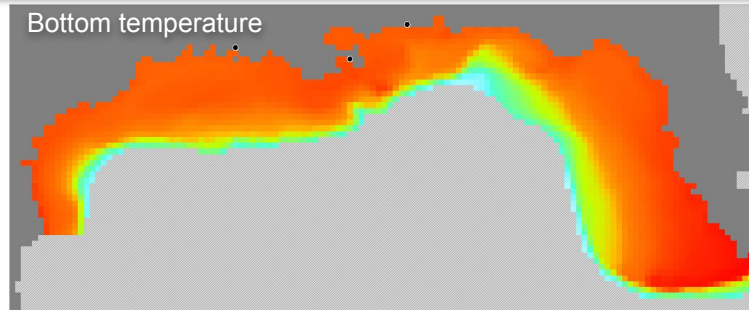


# Data Inputs // Habitat Maps





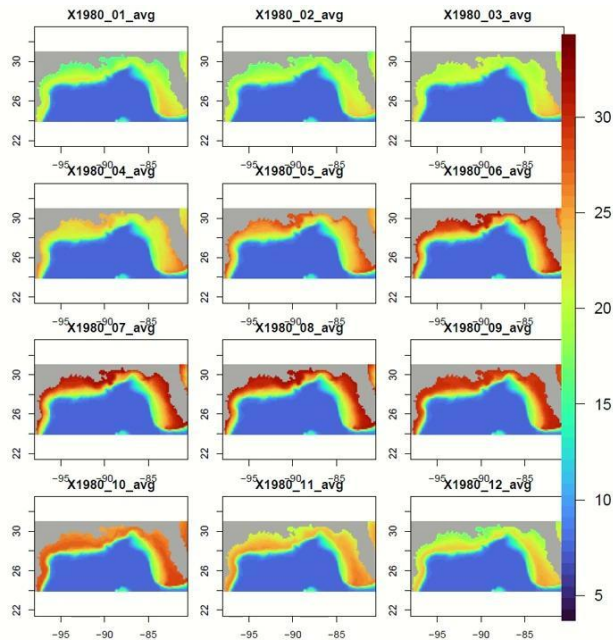
# Data Inputs // Environmental Drivers



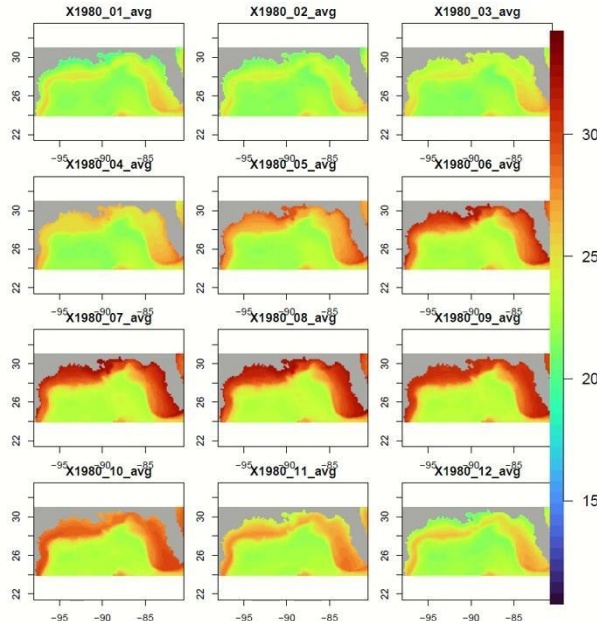


# Data Inputs // ST Env. Drivers // Temperature

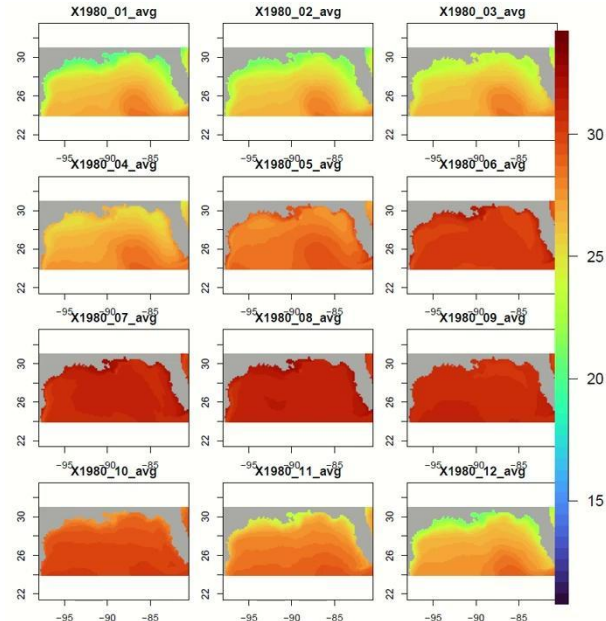
## Bottom



## Average



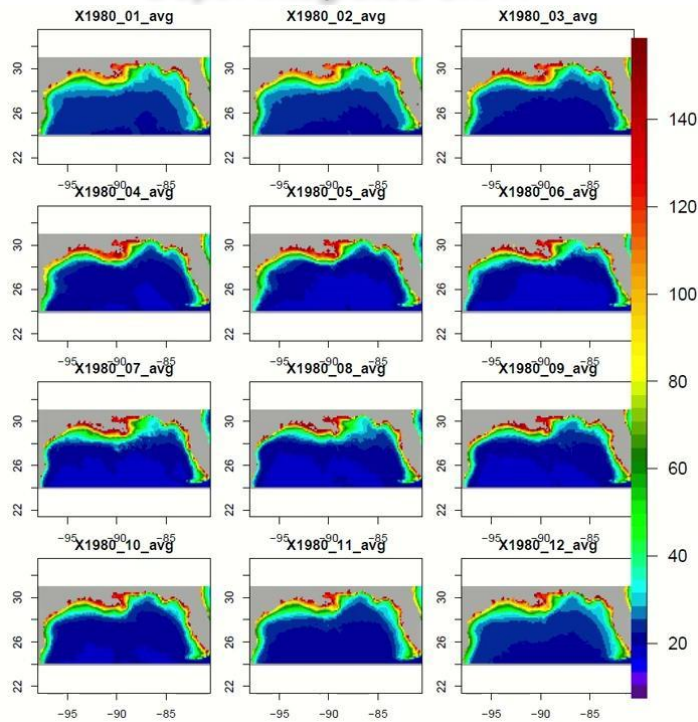
## Surface



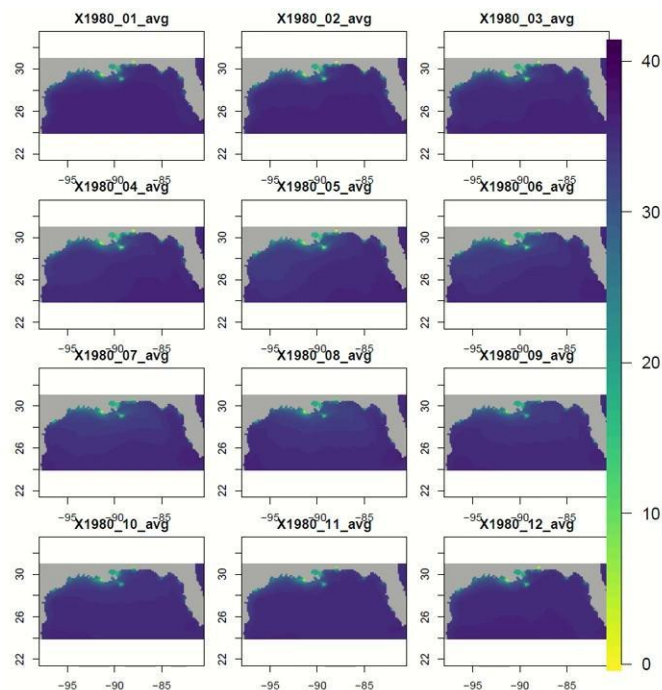
Monthly averages: 1980-1993 | Years of data available: 1993-2022  
Data source: HYCOM | [Files and PDF maps](#)

# Data Inputs // ST Env. Drivers // Prim. Prod. & Salinity

## Depth Integrated Chl-A



## Surface salinity



PP monthly averages: 1980-2003 | Years of data available: 2003-2020  
Data source: MODIS | [Files and PDF maps](#)

# Data inputs // Dispersal rates

Step 1: Query characteristics from FishBase R

Step 2: Estimate a relative swim speed (Simbilay 1990),

$$\log_{10}(S) = -0.828 + 0.61961 \log_{10}(L) + 0.3478 \log_{10}(A) + 0.7621(M)$$

where  $L$  represents a species' common length in mm,  $A$  is the aspect ratio of its caudal fin, and  $M$  is a binary swimming mode

Step 3: Scale to experimentally determined dispersal rates.

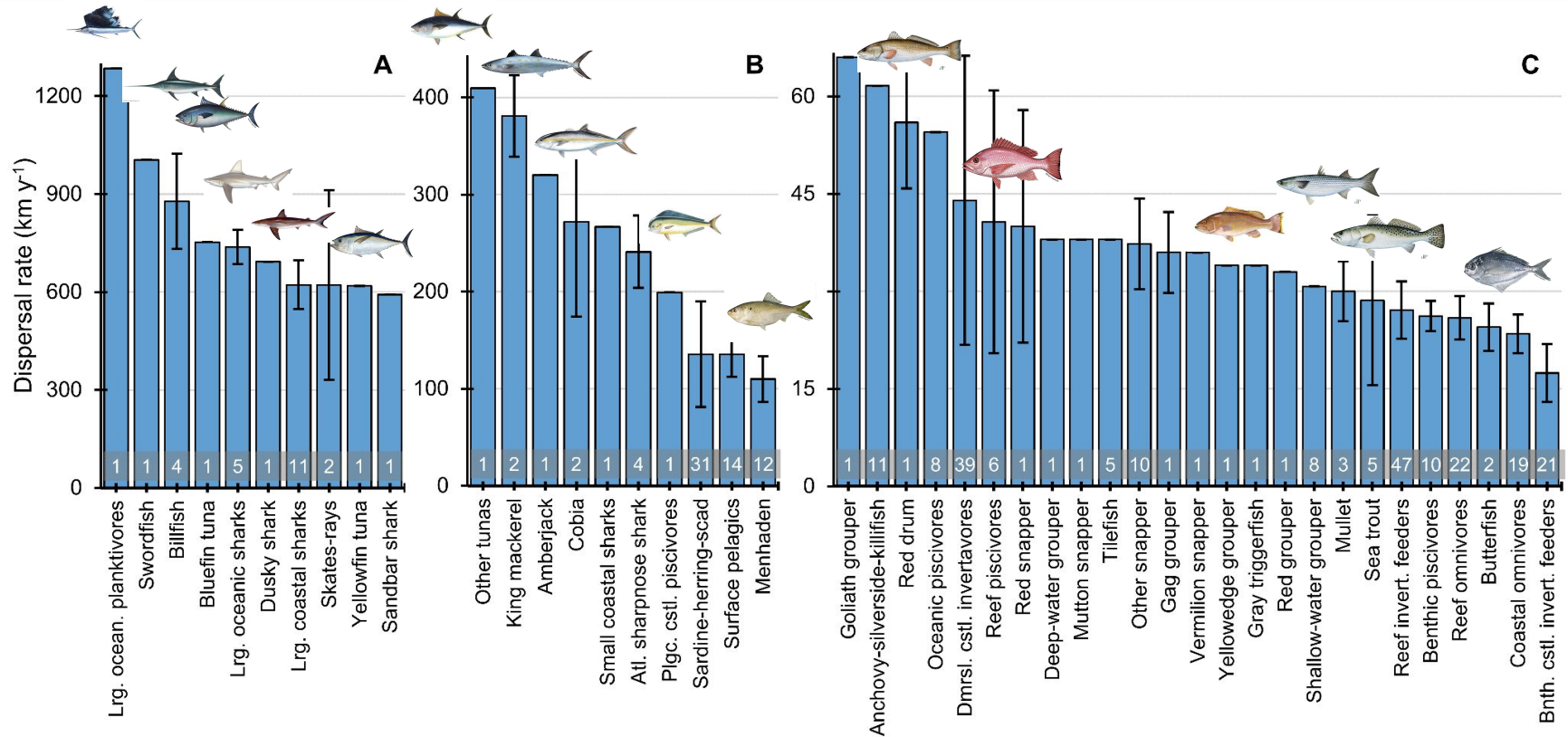
$$\begin{aligned} m_{red\ drum} &= 60.75 \text{ km y}^{-1} \text{ and} \\ m_{spotted\ seatrout} &= 28.26 \text{ km y}^{-1} \text{ (Nelson et al., 2021)} \end{aligned}$$

Step 4: Average and aggregate by functional group

Table 1 – Relative swim speed and scaled dispersal rates per species based on common length, tail aspect ratio, and swim mode.

Scientific name	Common name	EwE FG	FG num	Aspect ratio	Common length (cm)	Swim mode	Rel. swim speed km h <sup>-1</sup>	Scaled dispersal rate km y <sup>-1</sup>
<i>Cetorhinus maximus</i>	Basking shark	Lrg. oceanic planktivores	55	3.162	700	1	68.4	1285
<i>Makaira nigricans</i>	Blue marlin	Bilfish	16	7.829	290	1	54.3	1020
<i>Xipias gladius</i>	Swordfish	Swordfish	17	7.052	300	1	53.5	1005
<i>Isidophorus albicans</i>	Atlantic sailfish	Bilfish	16	5.502	240	1	49.7	934
<i>Isidophorus platypterus</i>	Indo-pacific sailfish	Bilfish	16	5.901	270	1	47.1	886
<i>Carcharhinus brevipinna</i>	Spinner shark	Large cstl. sharks	09	5.879	250	1	44.8	842
<i>Alopias vulpinus</i>	Common thresher shark	Large oceanic sharks	10	1.815	450	1	42.9	806
<i>Galeorhinus galeus</i>	Tiger shark	Large cstl. sharks	09	1.386	500	1	41.7	783
<i>Pristis pectinata</i>	Smalltooth sawfish	Skates-rays	25	1.116	550	1	41.0	770
<i>Alopias superciliosus</i>	Bigeye thresher	Large oceanic sharks	10	2.486	350	1	40.9	769
<i>Isurus paucus</i>	Shortfin mako shark	Large oceanic sharks	10	3.761	270	1	40.2	756
<i>Thunnus thynnus</i>	Bluefin tuna	Bluefin tuna	14	6.342	200	1	40.1	753
<i>Sphyrna tiburo</i>	Great hammerhead	Large cstl. sharks	09	1.954	370	1	38.2	719
<i>Prionace glauca</i>	Blue shark	Large oceanic sharks	10	2.084	335	1	37.5	704
<i>Sphyrna tiburo</i>	Scalloped hammerhead	Large cstl. sharks	09	1.808	360	1	37.3	701
<i>Carcharhinus obscurus</i>	Dusky shark	Dusky shark	07	3.360	250	1	36.9	693
<i>Tetrapodon lineatus</i>	Longbill spearfish	Bilfish	16	6.420	165	1	35.7	671
<i>Carcharhinus longimanus</i>	Oceanic whitetip shark	Large oceanic sharks	10	2.470	270	1	34.8	654
<i>Acanthocybium solandri</i>	Wahoo	Pelagic cstl. piscivores	18	5.619	170	1	34.7	653
<i>Carcharhinus altimus</i>	Bignose shark	Large cstl. sharks	09	2.520	250	1	33.4	627
<i>Thunnus albacares</i>	Yellowfin tuna	Yellowfin tuna	13	6.014	150	1	32.9	619
<i>Hexanchus griseus</i>	Squal shark	Large cstl. sharks	09	1.570	300	1	31.7	596
<i>Carcharhinus plumbeus</i>	Sandbar shark	Sandbar shark	08	3.178	200	1	31.5	592
<i>Carcharhinus falciformis</i>	Silly shark	Large cstl. sharks	09	2.026	250	1	30.9	582
<i>Carcharias taurus</i>	Sand tiger	Large cstl. sharks	09	1.614	250	1	28.6	537
<i>Carcharhinus leucas</i>	Bull shark	Large cstl. sharks	09	1.444	280	1	28.2	530
<i>Pristis pristis</i>	Largetooth sawfish	Skates-rays	25	1.116	250	1	25.1	473
<i>Carcharhinus signatus</i>	Night shark	Large cstl. sharks	09	1.638	200	1	25.0	470
<i>Neaportus brevirostris</i>	Lemon shark	Large cstl. sharks	09	1.073	240	1	24.2	455
<i>Katsuwonus pelamis</i>	Skipjack tuna	Other tunas	15	6.498	80	1	22.9	431
<i>Carcharhinus limbatus</i>	Blacktip shark	Blacktip shark	06	2.000	150	1	22.4	422
<i>Caranx lugubris</i>	Black jack	Pelagic cstl. piscivores	18	6.261	70	1	20.8	391
<i>Thunnus atlanticus</i>	Blackfin tuna	Bluefin tuna	15	5.808	72	1	20.6	388
<i>Seriola lalandi</i>	Almeco jack	Pelagic cstl. piscivores	18	3.744	90	1	20.3	382
<i>Scomberomorus cavalla</i>	King mackerel	King mackerel	21,22	5.815	70	1	20.3	381
<i>Seriola lalandi</i>	Amberjack, greater	Amberjack	19	2.832	100	1	19.7	370
<i>Euthynnus alletteratus</i>	Little tunny	Pelagic cstl. piscivores	18	4.184	80	1	19.7	369
<i>Trachurus falcatus</i>	Pennet	Pelagic cstl. piscivores	47	2.870	84	1	19.0	359
<i>Axius thazard</i>	Frigate mackerel	Pelagic cstl. piscivores	18	5.799	60	1	18.4	346
<i>Coryphaena hippurus</i>	Common dolphinfish	Pelagic cstl. piscivores	18	2.235	100	1	18.1	341
<i>Caranx hippos</i>	Crevalle jack	Pelagic cstl. piscivores	18	3.413	75	1	17.6	331
<i>Sardinia sarda</i>	Atlantic bonito	Pelagic cstl. piscivores	18	5.758	50	1	16.4	309
<i>Carcharias porosus</i>	Smalltail shark	Small cstl. sharks	12	2.014	90	1	16.4	309
<i>Caranx latus</i>	Horse-eye jack	Pelagic cstl. piscivores	18	3.534	60	1	15.5	291
<i>Caranx ruber</i>	Bar jack	Pelagic cstl. piscivores	18	4.228	50	1	14.7	277
<i>Sphyrna tiburo</i>	Shark, bonnethead	Small cstl. sharks	12	1.774	80	1	14.6	274
<i>Rachycentron canadum</i>	Cobia	Cobia	20	6.987	110	1	14.5	272

# Data inputs // Dispersal rates



Results consistent with “300-30-3 rule” for assigning dispersal rates to relatively fast-moving, slow-moving, and sedentary species, respectively (e.g., [Zeller and Reinert, 2004](#), [Chen et al., 2009](#), [Piroddi et al., 2011](#), [Fouzai et al., 2012](#))



# Data Inputs // Preference Functions

Step 1: Query environmental preferences from FishBase R ([code here](#))

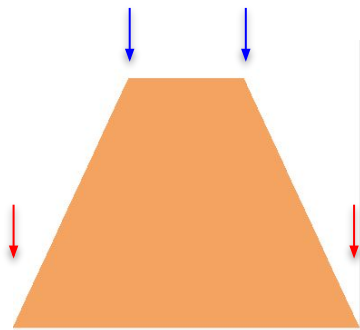
Step 2: Aggregate by species and functional group ([table](#))

Step 3: Review and adjust based on expert opinion ([table](#))

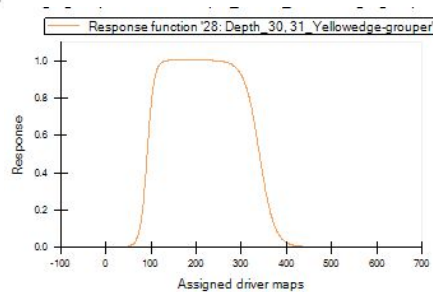
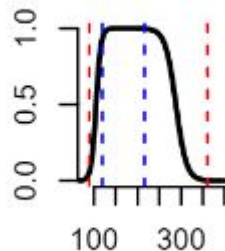
Step 4: Make double logistic function ([code](#))

Four Parameters: **minimum**, **preferred minimum (10th %ile)**, **preferred maximum (90th %ile)**, **maximum**

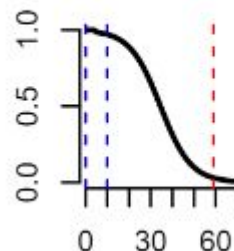
Three environmental drivers: depth (static), temperature (monthly), salinity (monthly)



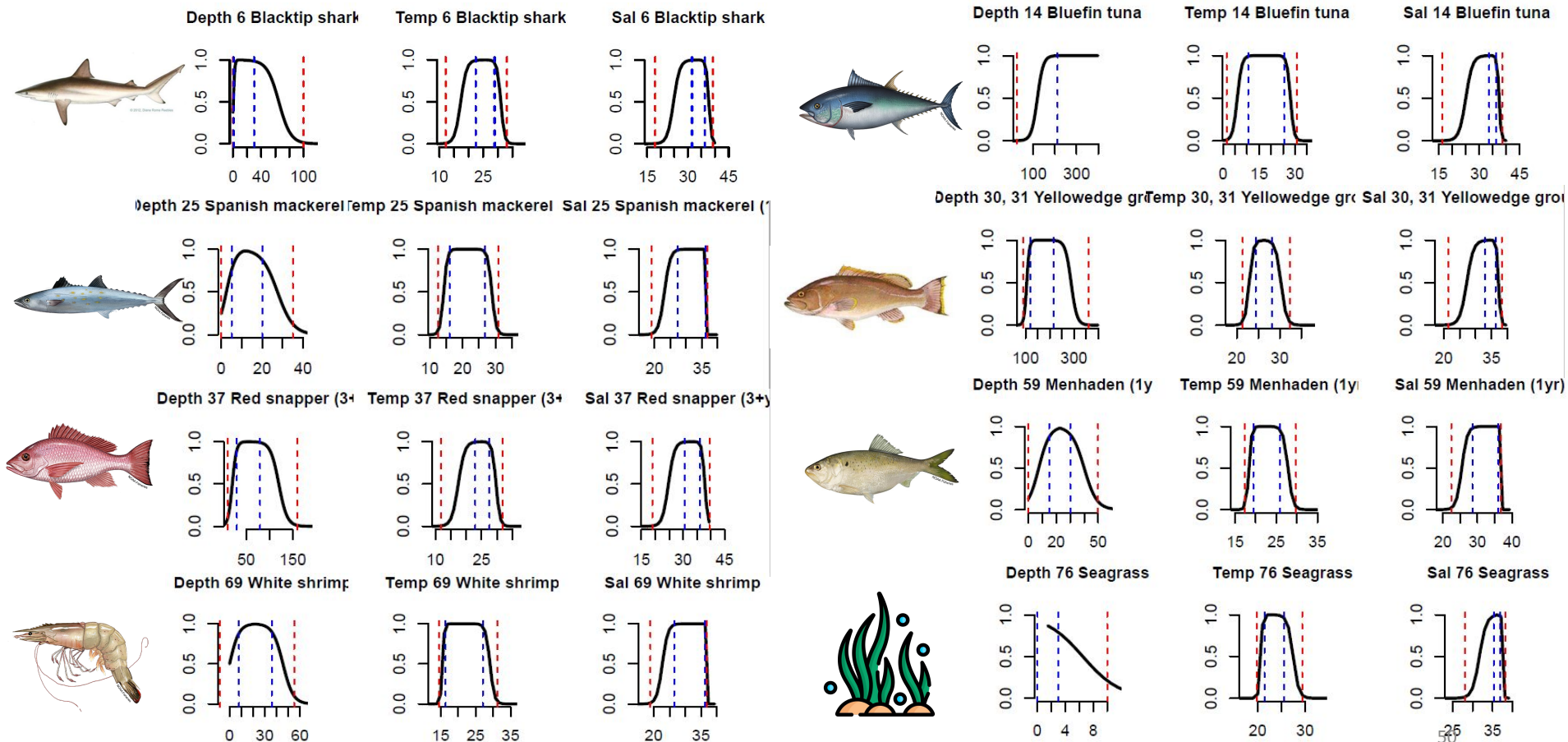
Depth 30, 31 Yellowedge gr

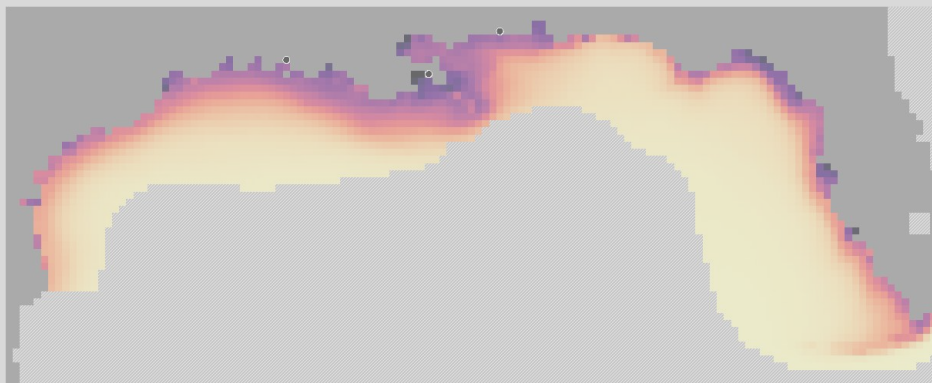
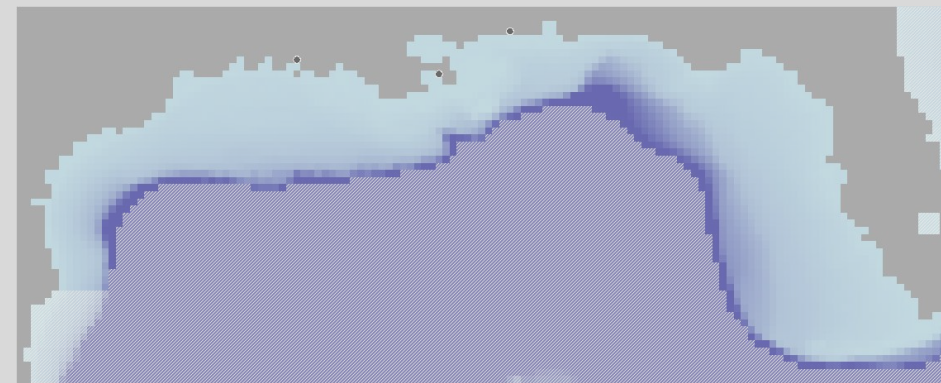


Demersal coastal invertel

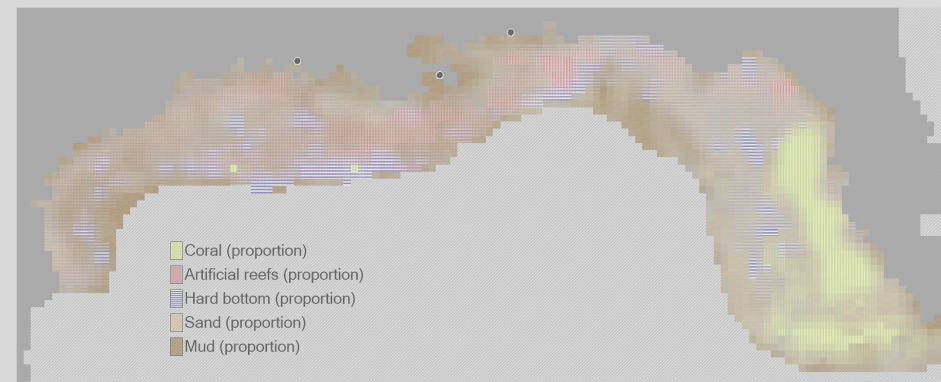


# Data Inputs // Preference Functions // A few examples

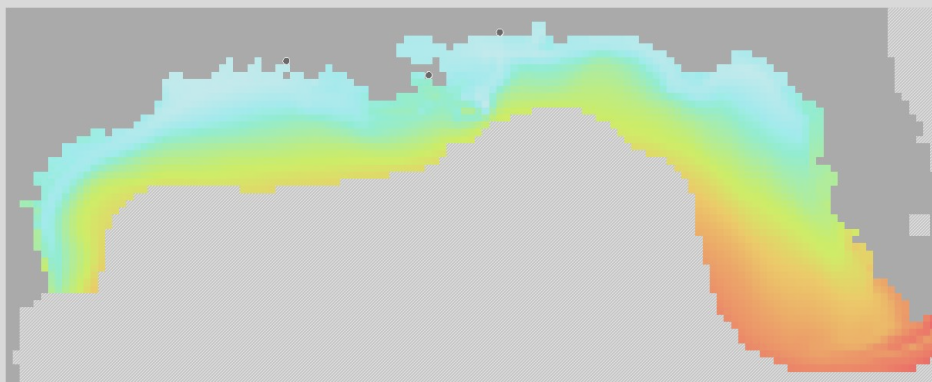




## USGWEM Ecospace // Next Steps

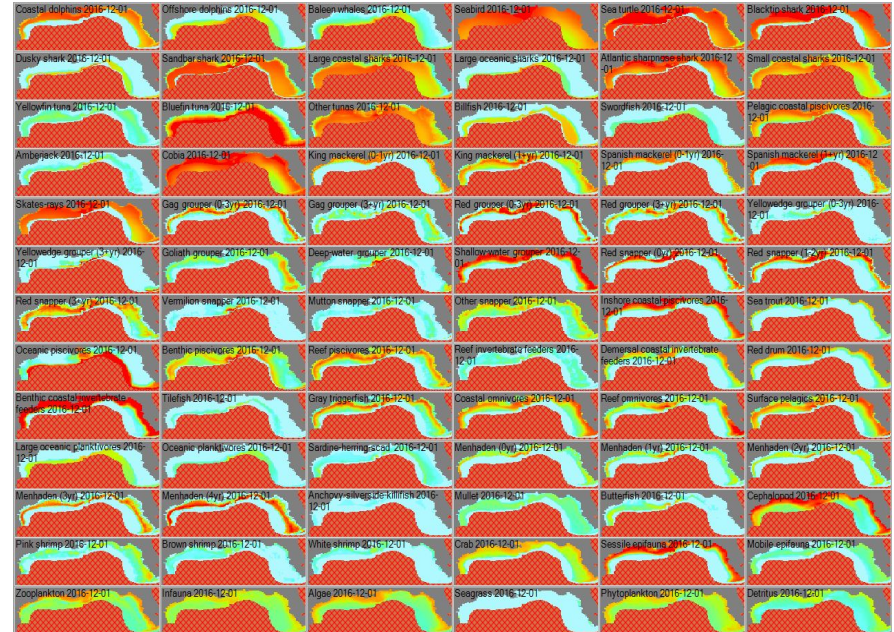
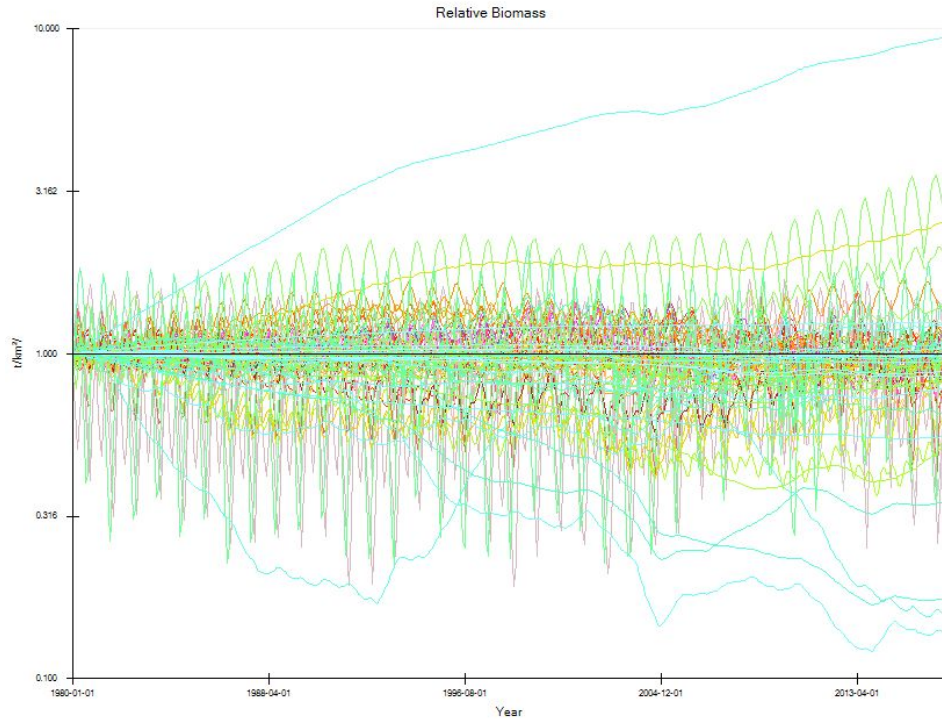


- Coral (proportion)
- Artificial reefs (proportion)
- Hard bottom (proportion)
- Sand (proportion)
- Mud (proportion)





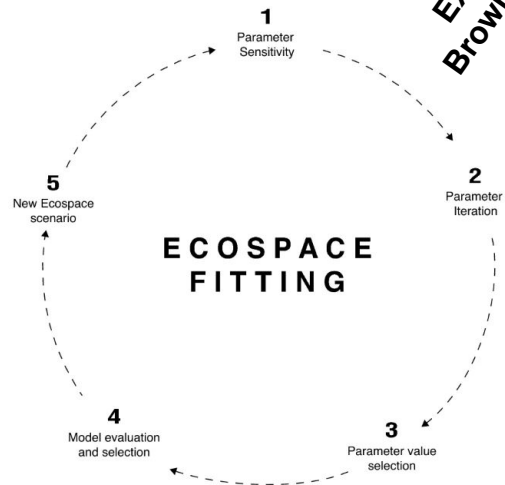
# USGWEM Ecospace // 10,000 foot view





# Next steps // Calibration and fitting

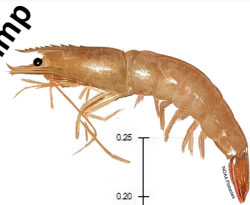
Schematic: [Vilas et al. 2023](#)



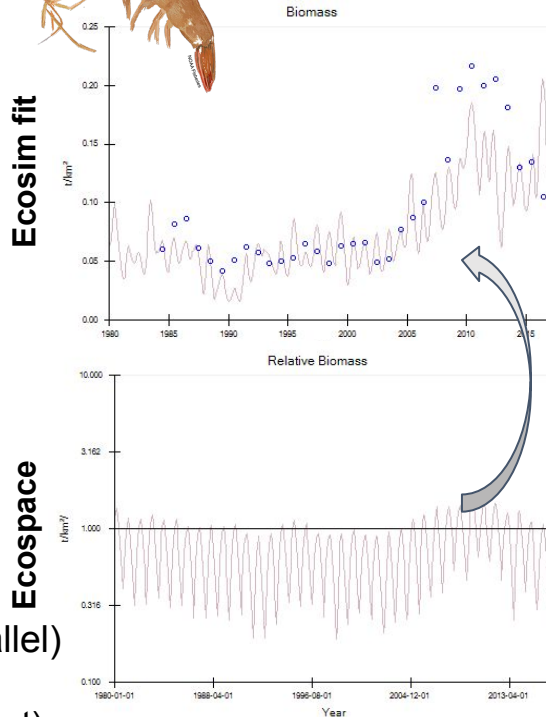
Calibration via iterative parameter adjustments (simulations run in parallel)

- Ecosim times series (done)
- Empirical time series (not done yet)
- Spatial surveys (not done yet)

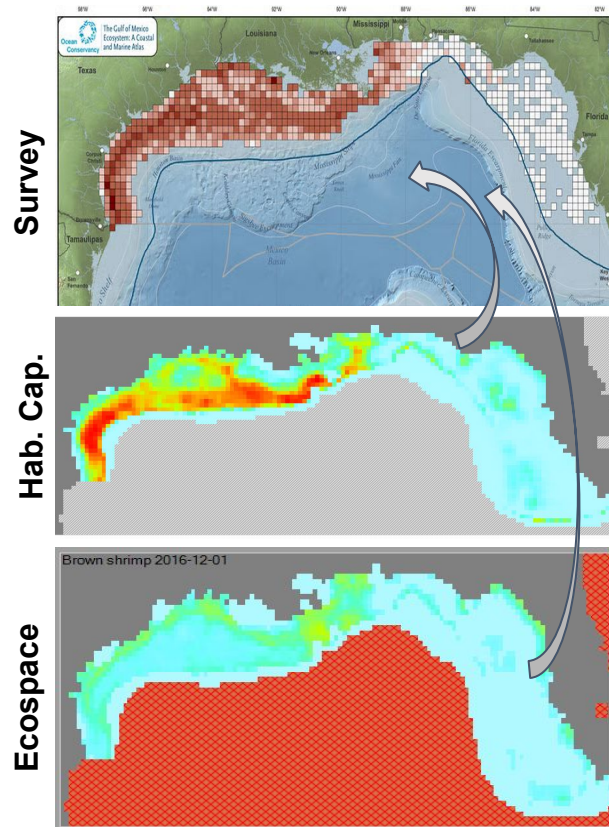
Example:  
Brown shrimp



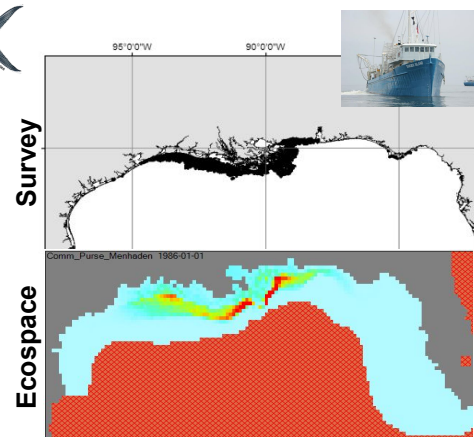
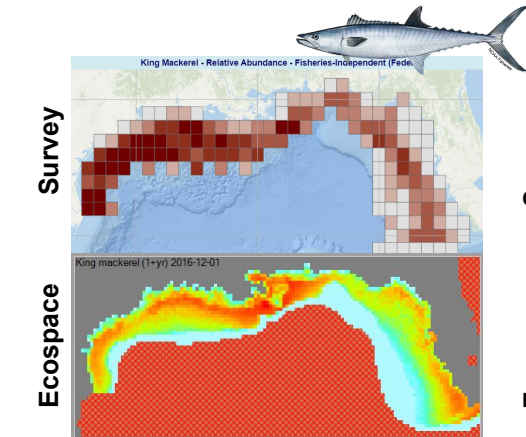
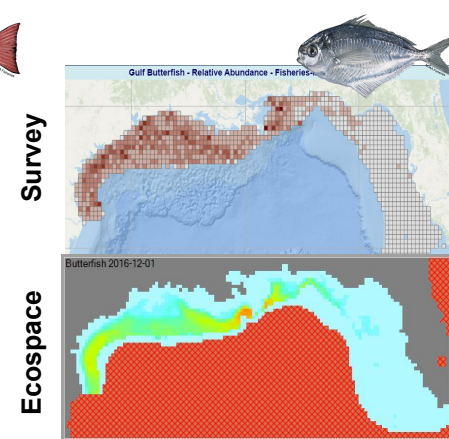
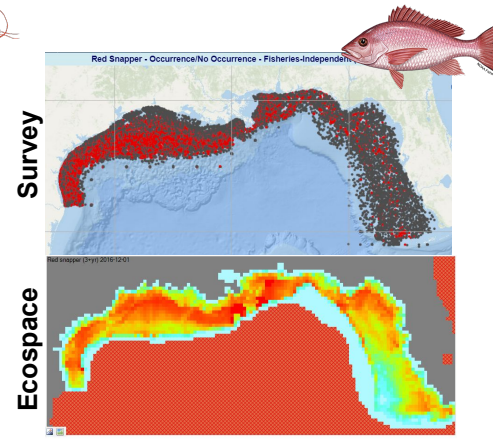
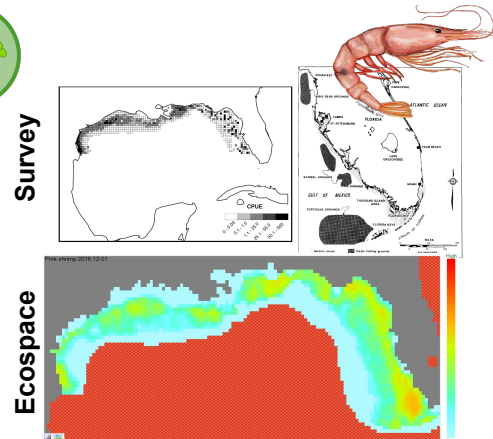
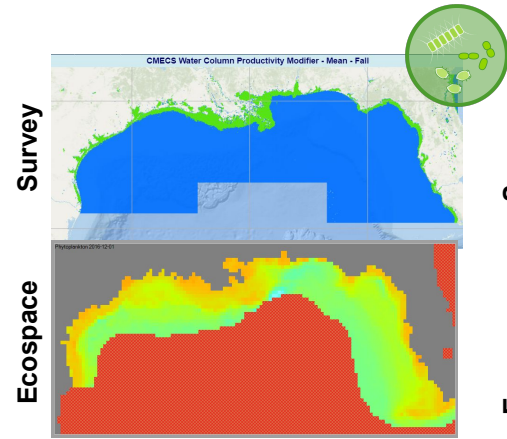
Fit to time...



...and space...?



# Next steps // Validation, calibration, and fitting



## Data sources for validation, calibration, and fitting:

- NOAA NCEI [GoM Data Atlas](https://www.noaa.gov/data-atlas)
- MODIS (e.g., CMEC PP)
- SEAMAP surveys (e.g., reef fish, long lines; <https://www.gsmfc.org/seamap-gomrs.php>)
- Literature (e.g., Pink shrimp [Drexler & Ainsworth 2013](#), Pelagic longline [Baum et al. 2003](#))
- SEDAR (<https://sedarweb.org/>)

# Next steps // Incorporate expert and stakeholder knowledge



ICES Journal of  
Marine Science

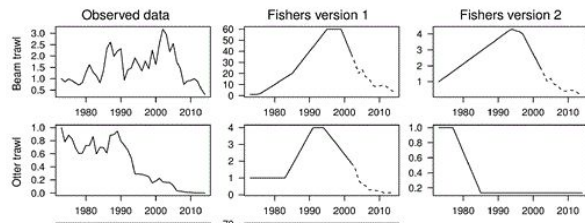


ICES Journal of Marine Science (2019), 76(4), 897–912. doi:10.1093/icesjms/fsz003

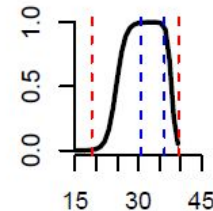
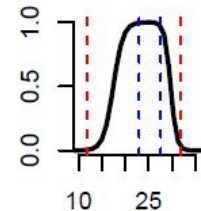
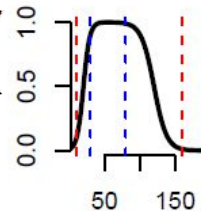
## Original Article

### Fishers' knowledge improves the accuracy of food web model predictions

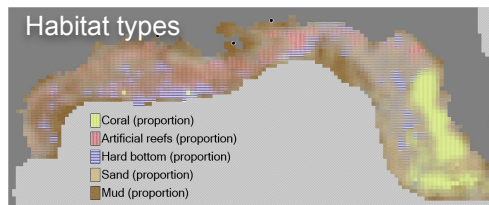
Jacob W. Bentley<sup>1\*</sup>, Natalia Serpetti<sup>1</sup>, Clive Fox<sup>1</sup>, Johanna J. Heymans<sup>1,2</sup>, and David G. Reid<sup>3</sup>



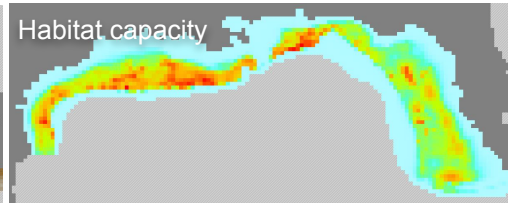
Depth 37 Red snapper (3+ Temp 37 Red snapper (3+ Sal 37 Red snapper (3+



Habitat types



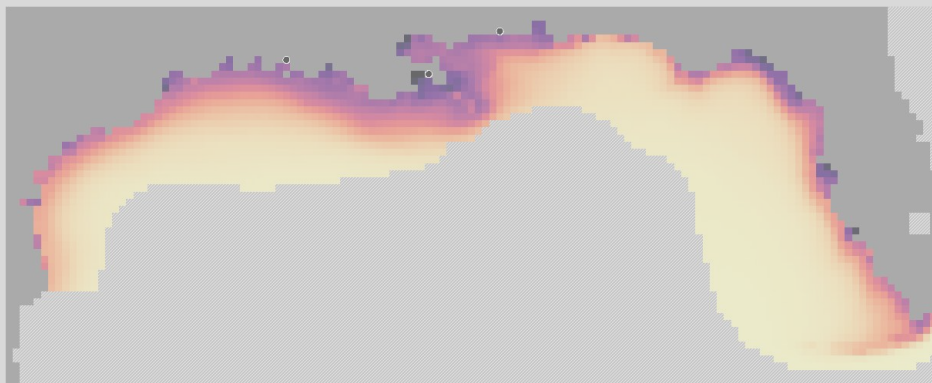
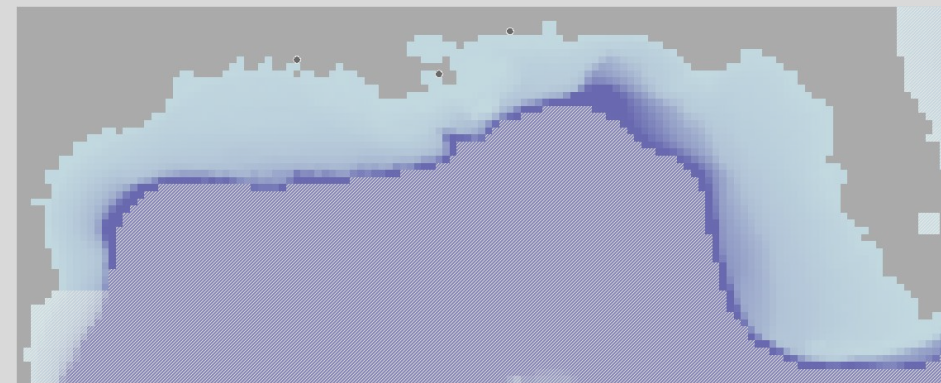
Habitat capacity



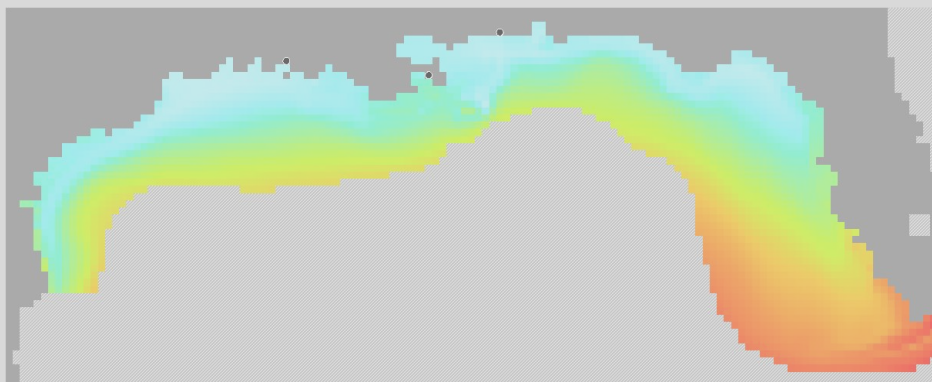
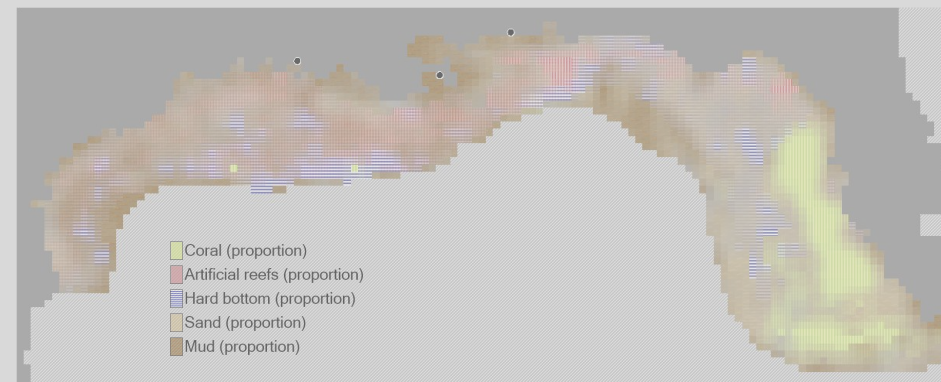
Incorporate scientific and fisher qualitative knowledge to...

- Review and validate results & trends
- Update parameters for preference functions
  - Depth, temperature, salinity
  - Habitat use, spatial validation





## USGWEM // Direction & Operationalization





# The Way to Operationalization

“The requisite conditions for enhanced operational use of EwE to support and inform resource management decisions exists, and these models can contribute to both strategic and tactical management decisions.”



ORIGINAL ARTICLE | [Full Access](#)

It is past time to use ecosystem models tactically to support ecosystem-based fisheries management: Case studies using Ecopath with Ecosim in an operational management context

J. Kevin Craig  Jason S. Link

First published: 27 February 2023 | <https://doi.org/10.1111/faf.12733>

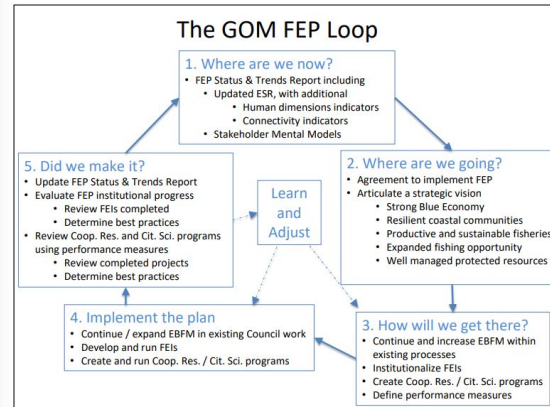
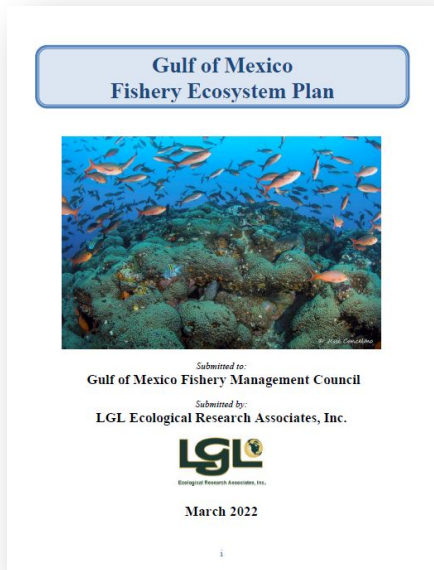
## Criteria to operationalize EwE:

1. Well-defined **objective(s)** for managing fisheries ecosystem issues
2. Clear **trade-offs**
3. A **mgmt process** that can respond to these trade-offs
4. Model(s):
  - a. That are accessible, well-documented, and follow best practices
  - b. Ideally, multiple models to assess structural uncertainties
5. Coproduction:
  - a. Early engagement from scientists, stakeholders, and managers (here we are!)
  - b. Collaborative, iterative model development
6. A rigorous **review** process

# Operationalization // Fisheries Ecosystem Plans



“FEPs require models of the ecosystem for stakeholders and managers to visualize and make predictions about how fishery ecosystems function . . . Perhaps the most effective use of Mathematical Ecosystem Models is within a hypothesis-testing framework” (pg. 29, [GoM Fisheries Ecosystem Plan](#))



## The GOM FEP Loop

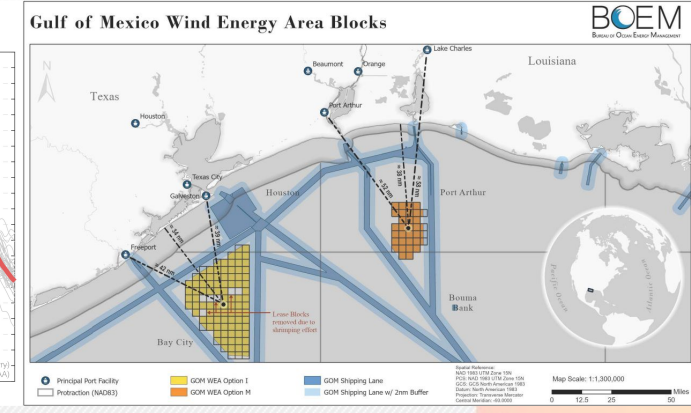
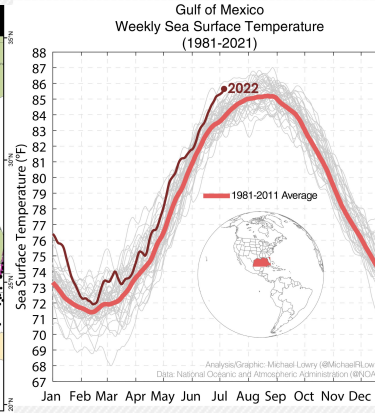
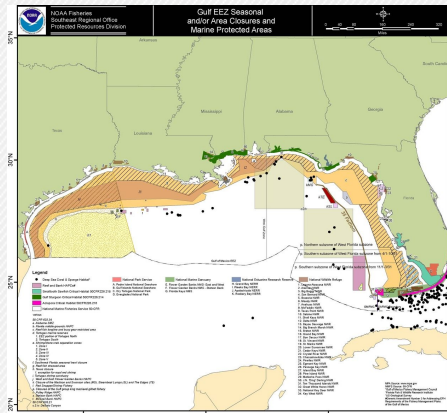
1. Where are we now?
2. Where are we going?
3. How do we get there?
4. Implement the plan
5. Did we make it?  
(Again, where are we now?)

**Learn & adjust**

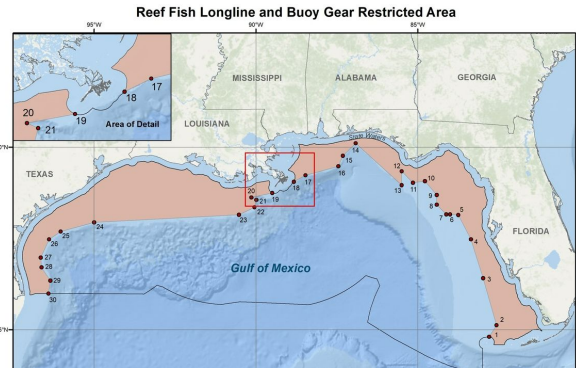
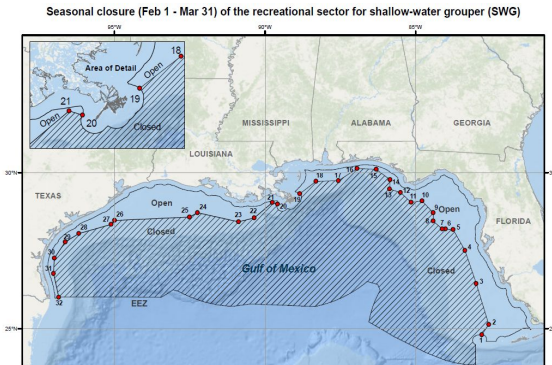
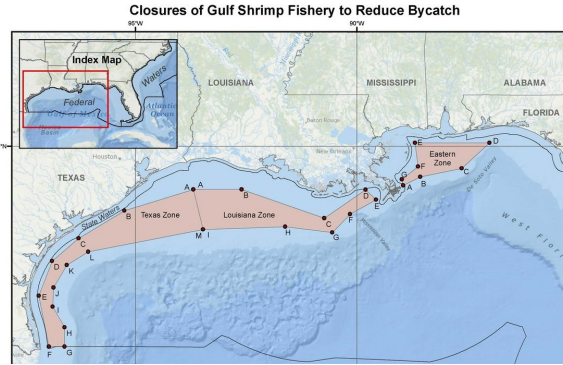
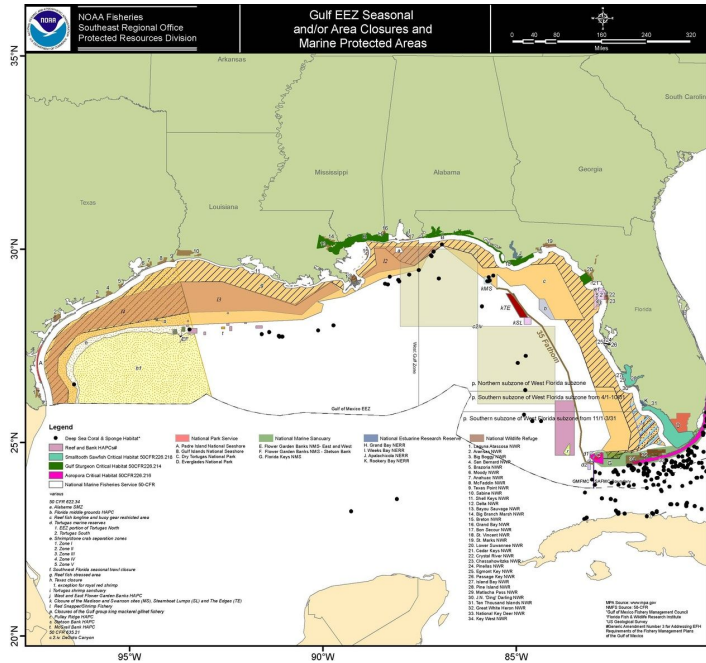
**Assess trade-offs  
Simulate mgmt actions  
Inform MSE**

# Potential research directions // Fisheries Ecosystem Issues

1. Spatial-temporal fishery closures and bycatch reduction
2. Climate change and environmental stressors
3. Changing artificial habitat: oil and gas (O&G) decommissioning & offshore wind farm (OWF) development



# Potential research directions // Spatial-temporal fishery closures & bycatch reduction



**Fisheries ecosystem issue (EAF) Multi-species, overlapping fisheries with bycatch and discard mortality**

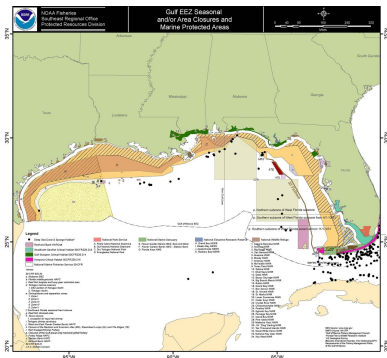


### Key trade-offs

- ↑Reef fish discard mortality vs. recreational fishing effort
- Temporal fishing access vs. out-of-season catches and high-grading
- Commercial net fishing (e.g., by Menhaden purse seine and shrimp trawls) vs. potential future yield of bycatch of juvenile finfishes

### Potential decision support from the USGWEM

- ID times and places where bycatch interactions should be monitored or limited
- Setting opening/closing dates for fisheries
- Optimize time and place for annual spawning season closures



### Strengths

- ST management a traditionally capacity of Ecospace and could inform mgmt decisions

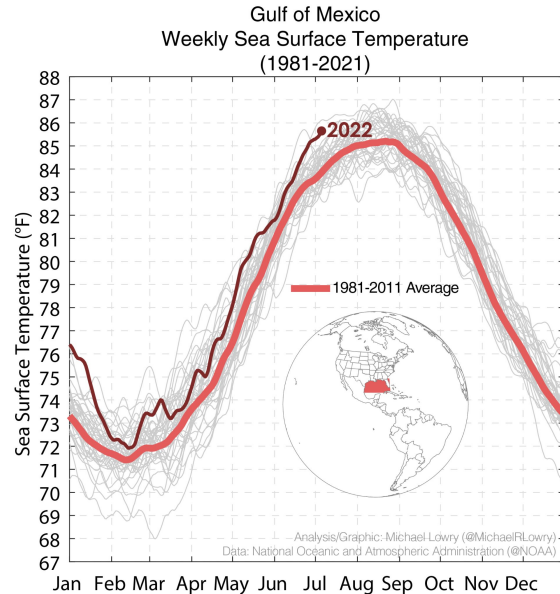
### Weaknesses/Caveats

- Gulf-wide scale model may be too large/complex for some applications
- Model of finer scale or lower complexity may be appropriate for specific deliverables, e.g., MICE model for Atlantic Menhaden generated inputs for SEDAR ([Chagaris et al. 2020](#), [Drew et al. 2021](#))

# Potential research directions // Climate change & environmental stressors

## Fisheries ecosystem issue (EBFM)

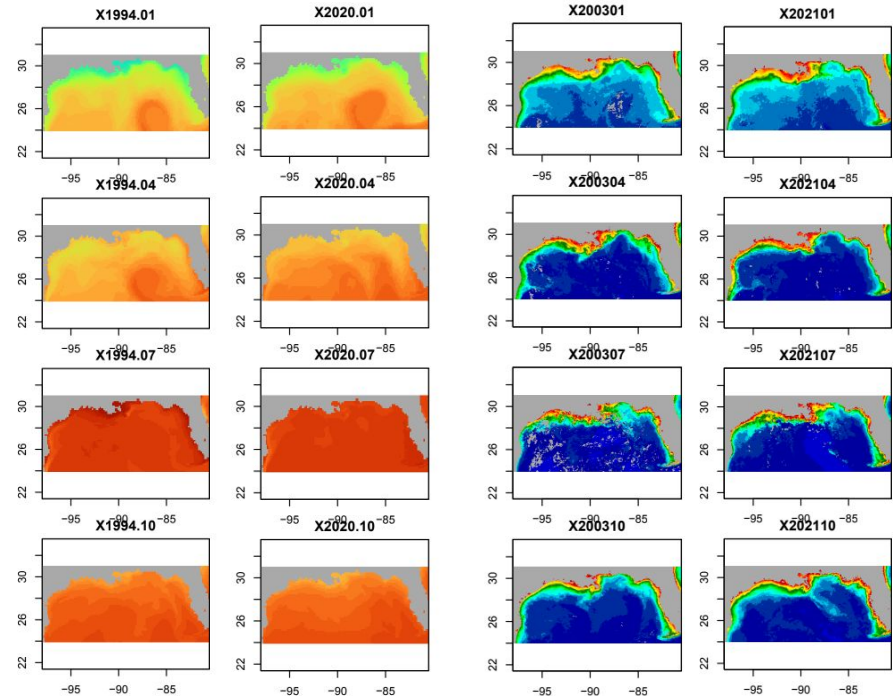
Wide-ranging and potentially compounding environmental stressors from climate change



Example USGWEM environmental driver maps:

SST

Prim. Prod.



# Potential research directions // Climate change & environmental stressors

## Key trade-offs

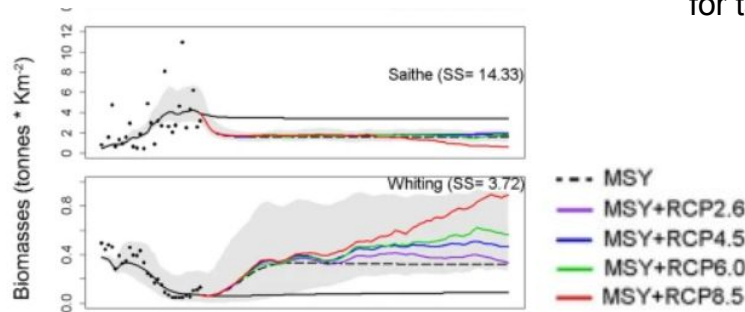
- ↓ Current yield vs. long-term precautionary management
- US agriculture production vs. Miss. Rvr. eutrophication
- Winners vs. losers from rising temperatures and changing environment

## Possible EBFM/EBM decision support from the USGWM

- Forecast winners & losers, e.g., [Serpetti et al. 2017](#)  
→ support precautionary management of impacted species
- Anticipate long-term responses to changing environment (e.g., warmer waters, changing Miss. Rvr. discharge)
- Model “tropicalization” food-web impacts from poleward range expansions of tropical species (e.g., mangroves, snook)

## Impact of ocean warming on sustainable fisheries management informs the Ecosystem Approach to Fisheries

N. Serpetti<sup>1</sup>, A. R. Baudron<sup>1</sup>, M. T. Burrows<sup>1</sup>, B. L. Payne<sup>1</sup>, P. Helaouët<sup>1</sup>, P. G. Fernandes<sup>2</sup> & J. J. Heymans<sup>1</sup>



## Strengths

- Gulf-wide spatial scale could consider climate change impacts
- Support for and from [NOAA Climate, Ecosystems, and Fisheries Initiative](#)



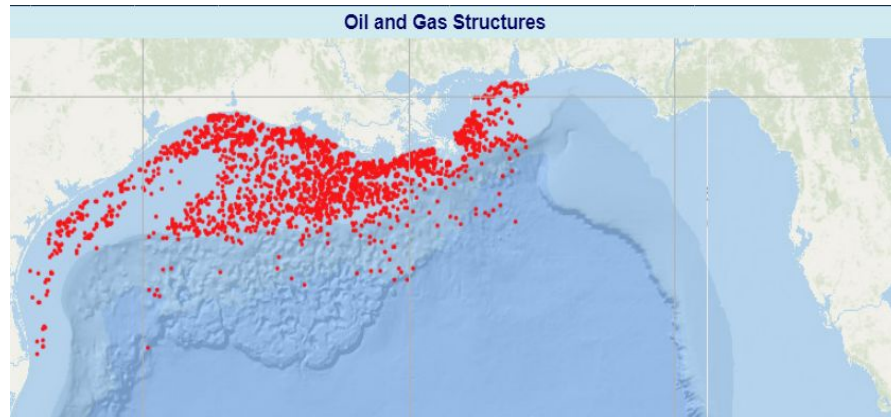
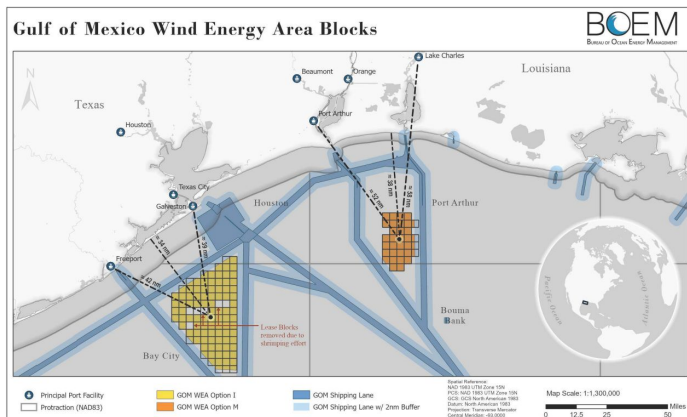
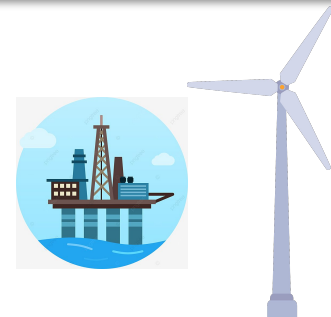
## Weaknesses/caveats

- Advice will be long-term and strategic
- Unclear how model results might be used for tactical decision making

# Potential research directions // Changing artificial habitat (OWF + O&G)

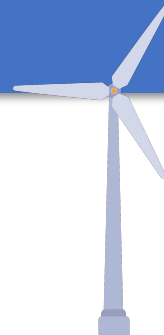
## Fisheries ecosystem issue (EBM)

Cross-sector management of fisheries, protected species, and habitat management facing forthcoming changes in O&G decommissioning and OWF infrastructure development





# Literature // Recent EwE applications for OWF development



Trade-offs between fisheries, offshore wind farms and marine protected areas in the southern North Sea – Winners, losers and effective spatial management

Miriam Püts<sup>a,\*</sup>, Alexander Kempf<sup>a</sup>, Christian Möllmann<sup>b</sup>, Marc Taylor<sup>a</sup>

<sup>a</sup> Thünen Institute of Sea Fisheries, Hohenlockse 31, 27572 Bremerhaven, Germany  
<sup>b</sup> Institute of Marine Economics and Fishery Science, Center for South-South Research and Sustainability (CSRS), Glinde Elbeufer 133, 22767 Hamburg, Germany



Spatial ecosystem modelling of marine renewable energy installations: Gauging the utility of Ecospace

Karen A. Alexander<sup>a,\*</sup>, Sophie A. Meyjes<sup>b</sup>, Johanna J. Heymans<sup>a</sup>

<sup>a</sup> Scottish Association for Marine Science, Scottish Marine Institute, Clon, Argyll PA37 7QA, United Kingdom  
<sup>b</sup> University of St Andrews, North St, St Andrews, Fife KY16 9AJ, United Kingdom



ICES Journal of Marine Science (2020), 77(3), 1238–1246. doi:10.1093/icesjms/fy125

Contribution to the Themed Section: 'Decommissioned offshore man-made installations'  
**Original Article**

**An ecosystem approach for studying the impact of offshore wind farms: a French case study**

Jean-Philippe Pezy<sup>a,\*</sup>, Aurore Raoux<sup>a</sup>, and Jean-Claude Dauvin<sup>a</sup>



Isotopic analyses, a good tool to validate models in the context of Marine Renewable Energy development and cumulative impacts

Aurore Raoux<sup>a,\*</sup>, Jean-Philippe Pezy<sup>b</sup>, Bruno Ernande<sup>c</sup>, Nathalie Niquil<sup>d</sup>, Jean-Claude Dauvin<sup>b</sup>, Karine Grangeré<sup>a</sup>

## JOURNAL ARTICLE

**Spatialized ecological network analysis for ecosystem-based management: effects of climate change, marine renewable energy, and fishing on ecosystem functioning in the Bay of Seine**

Quentin Nogues<sup>a,\*</sup>, Emma Airaingous, Pierre Bourdaud, Ghassen Halouani, Aurore Raoux, Éric Foucher, François Le Loc'h, Frédérique Loew-Turbout, Frida Ben Rais Lasram, Jean-Claude Dauvin, Nathalie Niquil

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Effects of established offshore wind farms on energy flow of coastal ecosystems: A case study of the Rudong offshore wind farms in China

Junjie Wang<sup>a,\*</sup>, Xingxi Zou<sup>b,c</sup>, Wenwen Yu<sup>b,c,d</sup>, Dongjia Zhang<sup>b,c</sup>, Tang Wang<sup>b,c</sup>



The environmental impact from an offshore windfarm: Challenge and evaluation methodology based on an ecosystem approach

Jean-Philippe Pezy<sup>a</sup>, Aurore Raoux<sup>a</sup>, Jean-Claude Dauvin<sup>a</sup>

**Modeling Small Scale Impacts of Multi-Purpose Platforms: An Ecosystem Approach**

Natalia Serpetti<sup>a,\*</sup>, Steven Benjamins<sup>a</sup>, Stevie Brain<sup>a</sup>, Maurizio Collu<sup>b</sup>, Bethany J. Harvey<sup>a</sup>, Johanna J. Heymans<sup>a</sup>, Adam D. Hughes<sup>a</sup>, Denise Risch<sup>c</sup>, Sophia Rosinski<sup>d</sup>, James J. Waggitt<sup>e</sup>, Ben Wilson<sup>e</sup>

## EwE assessments for OWF: The North Sea

- [Puts et al. 2023](#)
- [Serpetti et al. 2021](#)
- [Alexander et al. 2016](#)

## West Coast of France

- [Jean-Philippe et al. 2020](#)
- [Halouani et al. 2020](#)
- [Raoux et al. 2020](#)
- [Pezy et al. 2020](#)
- [Nogues et al. 2022](#)

## China

- [Wang et al. 2019](#)

## EwE can help assess trade-offs, e.g.,

Positive spillover effects predicted from OWF spatial closures in W. coast of France model (Halouani et al. 2020) but not W. coast of Scotland model (Alexander et al., 2016)

# Potential research directions // Changing artificial habitat (OWF + O&G)

## Key trade-offs

- Energy sector vs. fisheries sector vs. impacts to protected species (e.g., birds, mammals, turtles)
- Reef effect: ↑production vs. ↑catchability from aggregation; also habitat for invasive lionfish
- Exclusion effect: ↑fisheries spillover vs. ↓fishing access
- Hydrologic and primary production: hypoxia vs. up/downwelling and mixing ([Daewal et al. 2022](#))

## Possible EBM decision support from the USGWEM

- Consider designs of reef structures (e.g., scour aprons as reef structures)
- Inform experimental design and monitoring programs for artificial habitat
- Ultimately, help inform O&G decommissioning plans (e.g., “rigs to reefs”) and spatial management OWF

NCEAS proposal in review: [Ecosystem impacts of GoM wind energy development on fish and fisheries](#) (Harris, Sagarese, Chagaris, Klajbor)

## Strengths

- Ecosystem modeling can inform management and research design *before* wide-scale OWF construction and O&G decommissioning
- Stakeholder & management interest

## Weaknesses/Caveats

- No precedent for EwE in ecosystem impact assessments
- USGWEM spatial scale may be too large to assess artificial habitat in specific areas (e.g., OWF development in W. Gulf)



# USGWEM // Ultimate Goals

## An **operational model** that supports decision making

- EAF (e.g., spatial-temporal closures)
- EBFM (e.g., climate stressors in fisheries management), or
- EBM (e.g., artificial habitat effects from marine energy infrastructure)

## A **co-produced model** that's valuable for management

- Early and iterative direction and input
- Leverage collaboration with stakeholders for validation and guidance

## A **robust model** that withstands rigorous review, e.g.,

- ERPs for Atlantic Menhaden ([Anstead et al. 2021](#), [Chagaris et al. 2020](#), [Drew et al. 2021](#))
- NEFSC Groundfish Assessment and Review Meeting
- CIE review for California Current Atlantis model ([Kaplan & Marshall, 2016](#))



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## Gulf of Mexico Fishery Management Council SSC meeting

02 May 2023

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## Further info

[Sagarese et al. \(2016\)](#)  
[NOAA technical memorandum](#)  
[Berenshtein et al. \(2023\)](#)  
[SEFSC/IEA/GWEM/DataSynth](#)



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