

The Chesapeake Bay Ecological Prediction System

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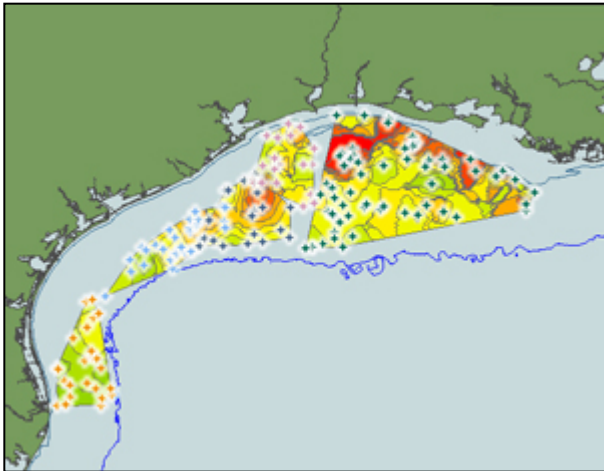
National Oceanic & Atmospheric Administration



Acknowledgements

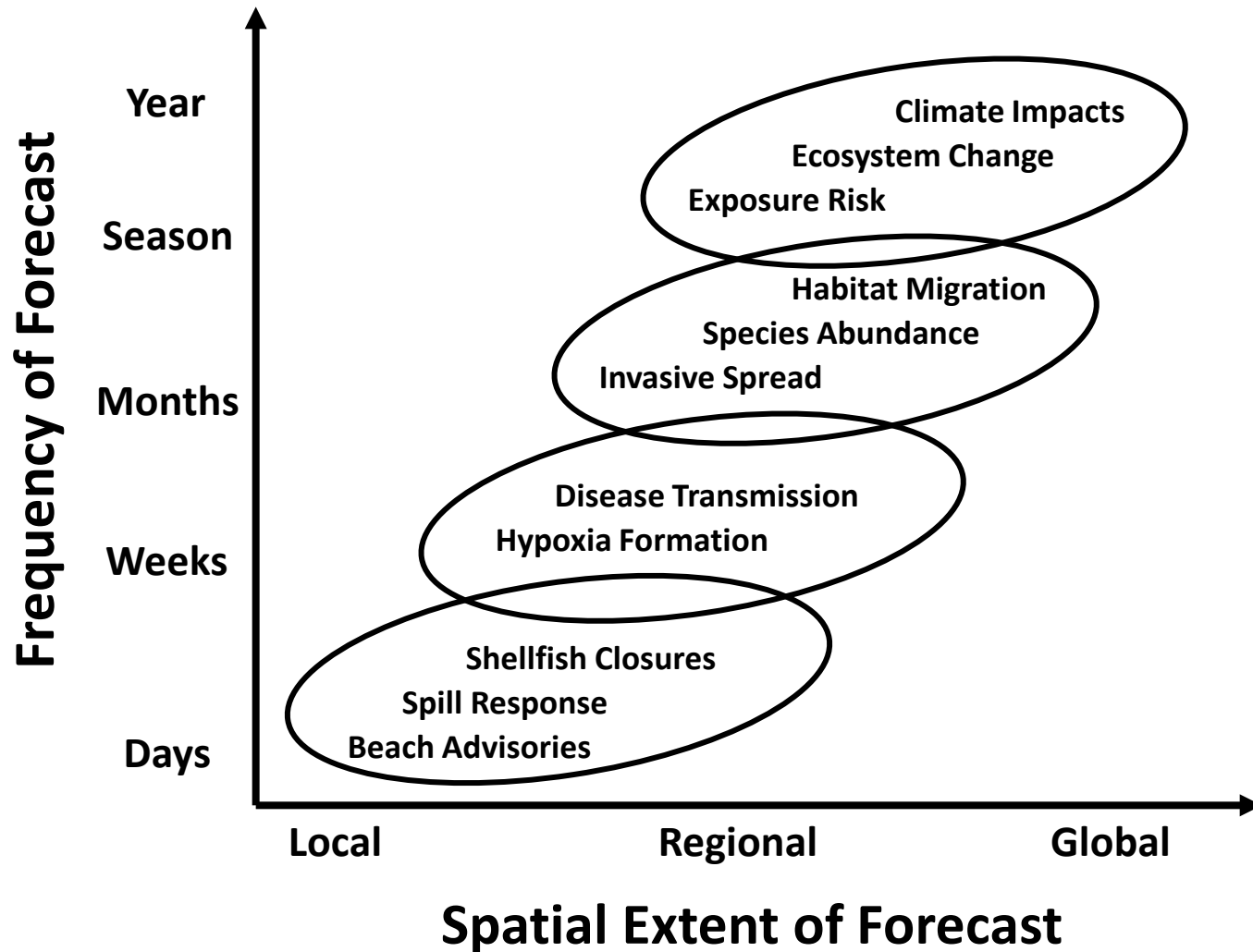
- Colleagues: Raleigh Hood, Mary Beth Decker, Wen Long, John Jacobs, Ragu Murtugudde, M. Bala Prasad, Jennifer Purcell, Douglas Ramers, Catherine Wazniak, Jerry Wiggert, Bob Wood, Stephanie Uz, and Jiangtao Xu
- Funding: NOAA EcoFore, MERHAB, MD SeaGrant

Ecological Forecasting



- Predicts the likely changes in ecosystems and their components in response to alteration in the environment
- Helps people, coastal managers and scientists make better decisions
- Extremely challenging: requires integration of physical, chemical, biological, economic, and social factors
- Feasible only recently with improvements in ecosystem understanding, observing systems, modeling, computing, and telecommunications

Ecological Forecasting: Time & Space Scales



Motivation of Ecological Forecasts in Chesapeake Bay

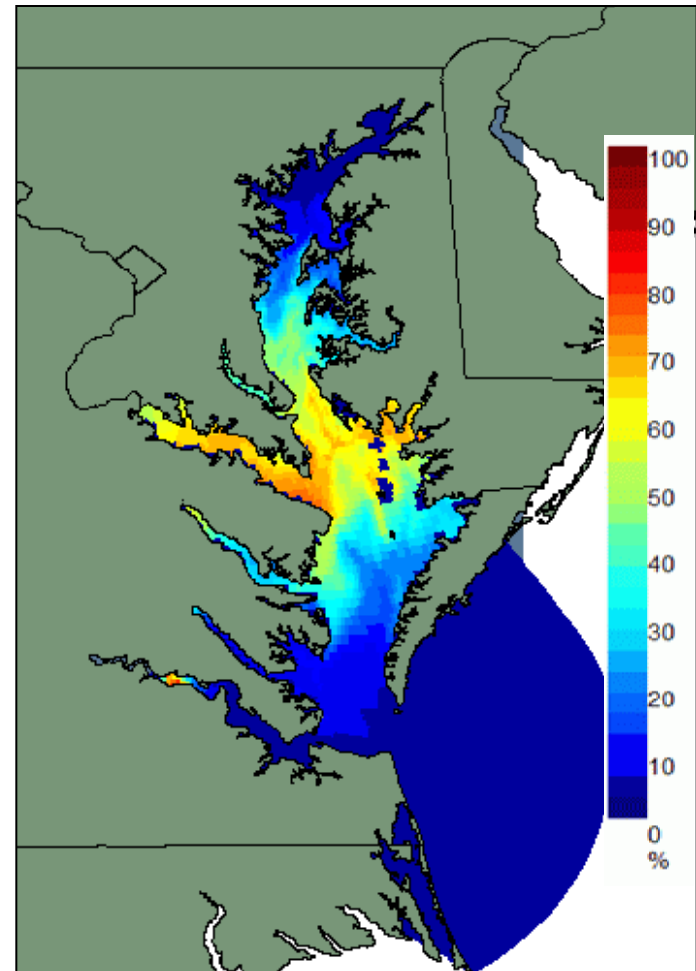
- Chesapeake Bay represents an extremely valuable regional resource
- Noxious conditions and organisms afflict the Chesapeake Bay, adversely affecting aquatic and human health, and local economies
- Predicting the timing and location of these conditions and events will improve monitoring capabilities and aid in mitigating their effects



SeaWiFS true-color image of Mid-Atlantic Region
from April 12, 1998.

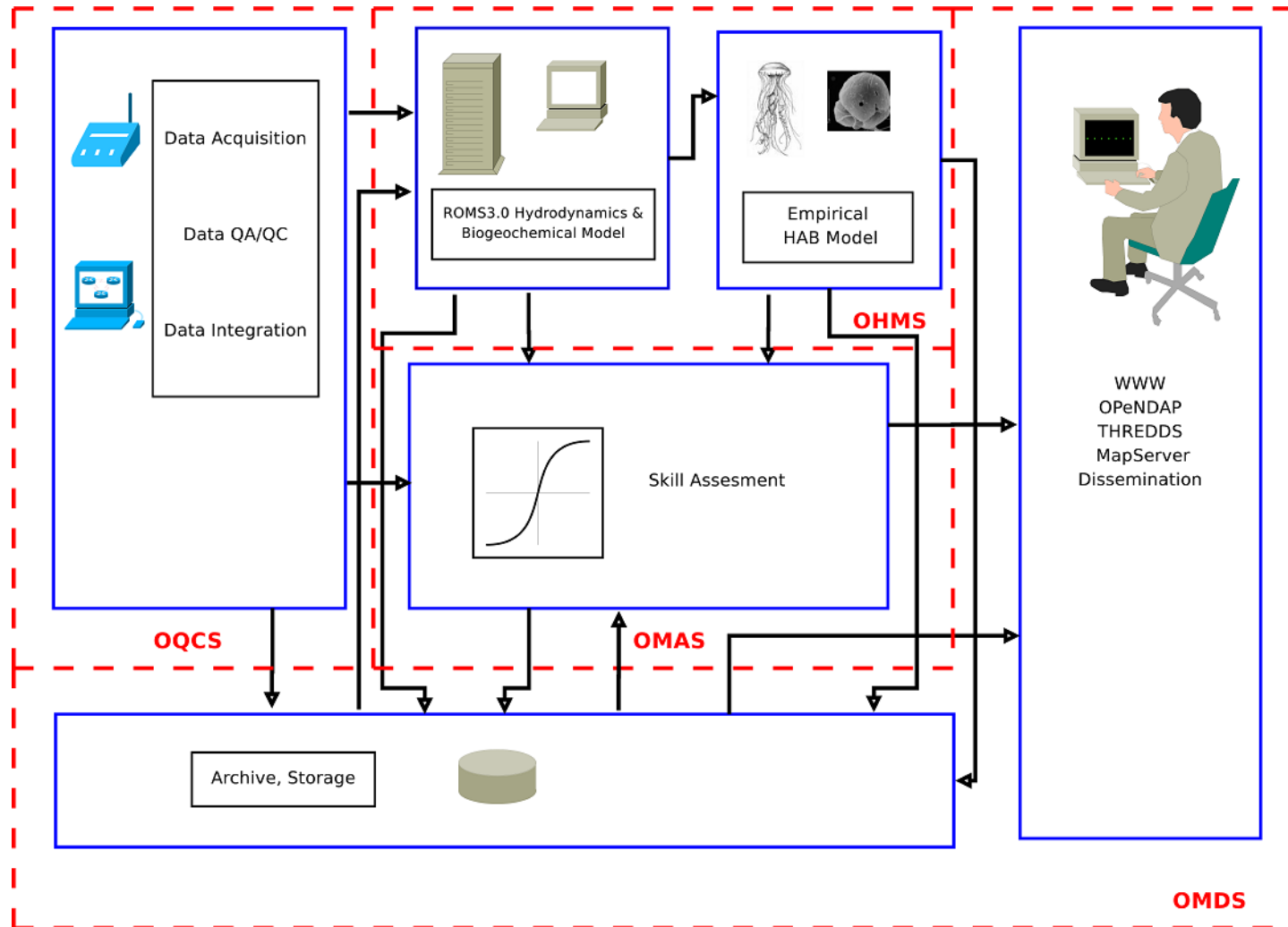
List of Forecasts

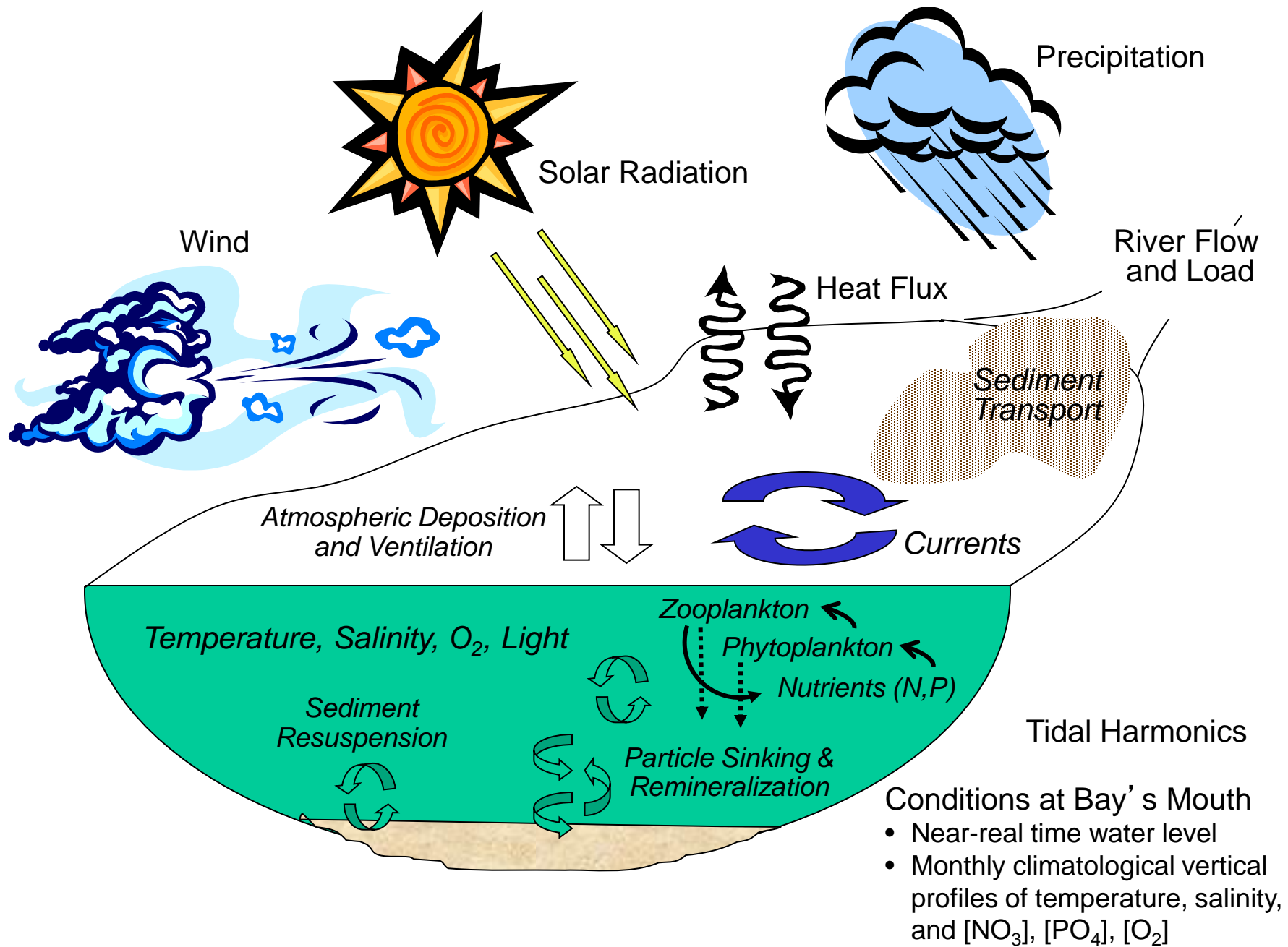
- Physical
 - *Temperature*
 - *Salinity*
 - *Current velocity*
 - *Sea Surface Height*
- Biogeochemical
 - *Nutrient concentrations*
 - *Phytoplankton, Zooplankton*
 - *Dissolved oxygen concentrations*
- Organismal
 - *Sea Nettles (Chrysaora quinquecirrha)*
 - *Harmful algal blooms*
 - *Water-borne pathogens*



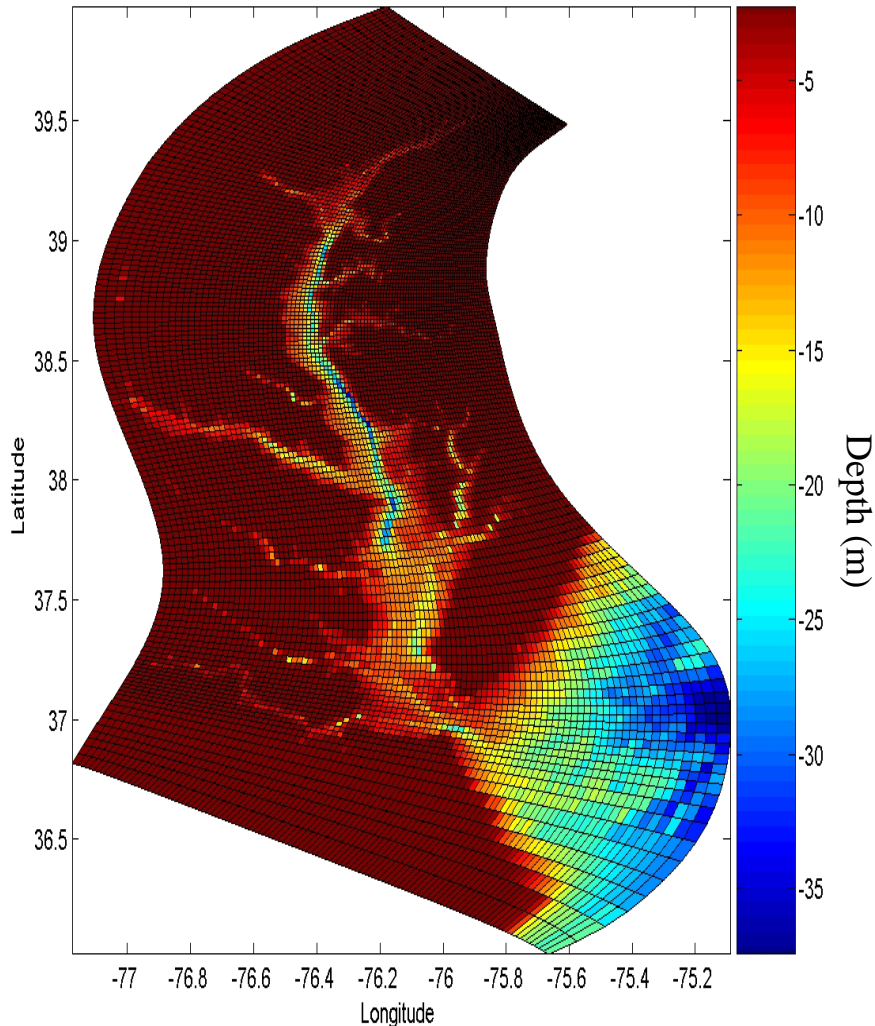
Example of nowcast: Likelihood of encountering sea nettles, *Chrysaora quinquecirrha*, in Chesapeake Bay on August 17, 2007.

Chesapeake Bay Ecological Prediction System (CBEPS)



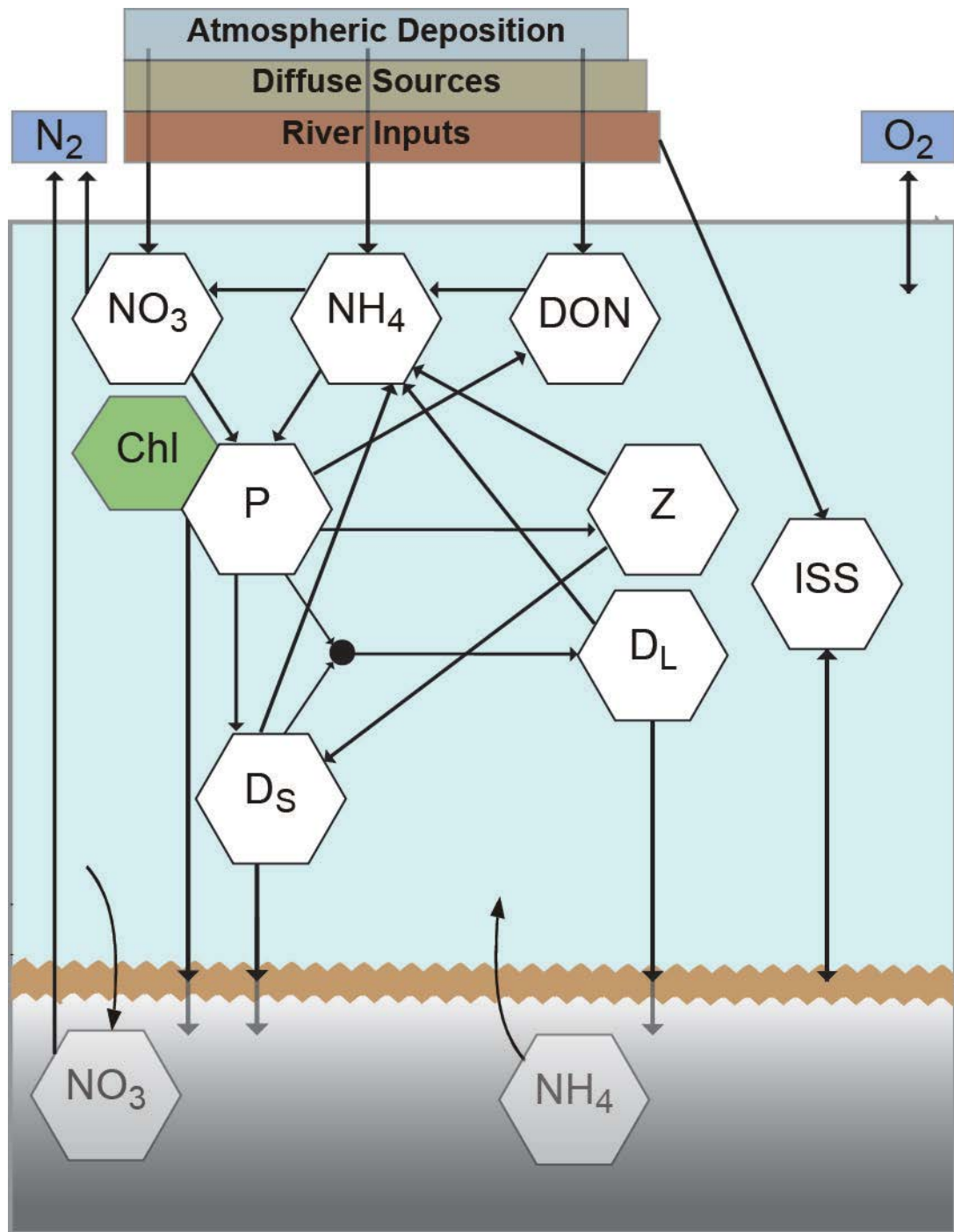


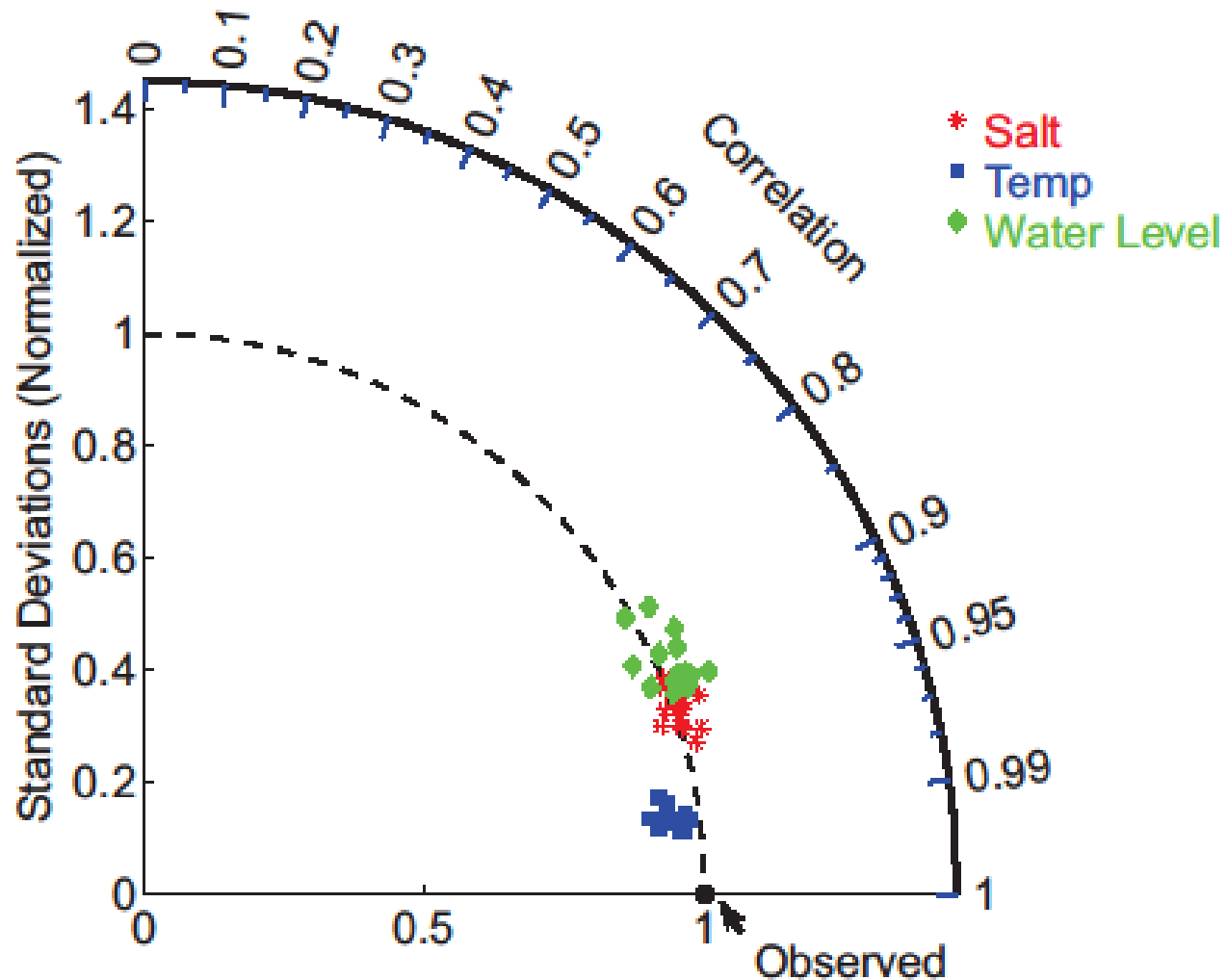
ChesROMS Hydrodynamic Model



- 3-D
- Sigma coordinate
- Coarse mesh
(100*150*20)
- Horizontal spatial
resolution (0.5 – 5 km)
- Validated w/15-year
hindcast

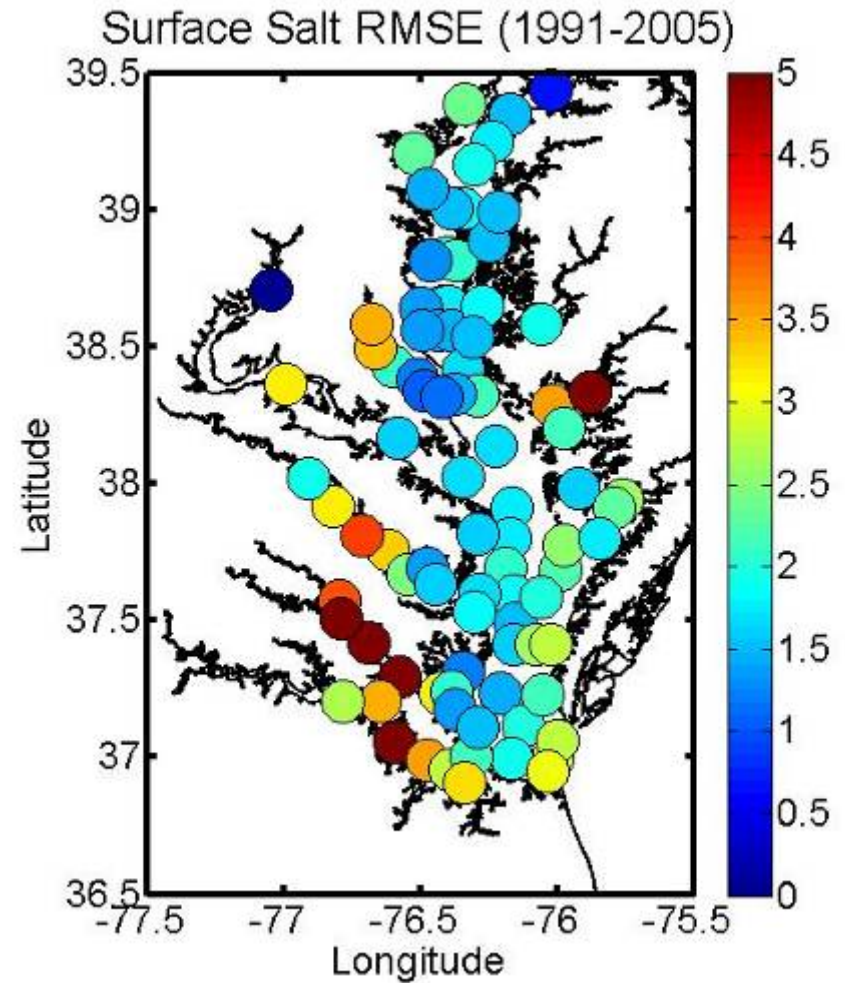
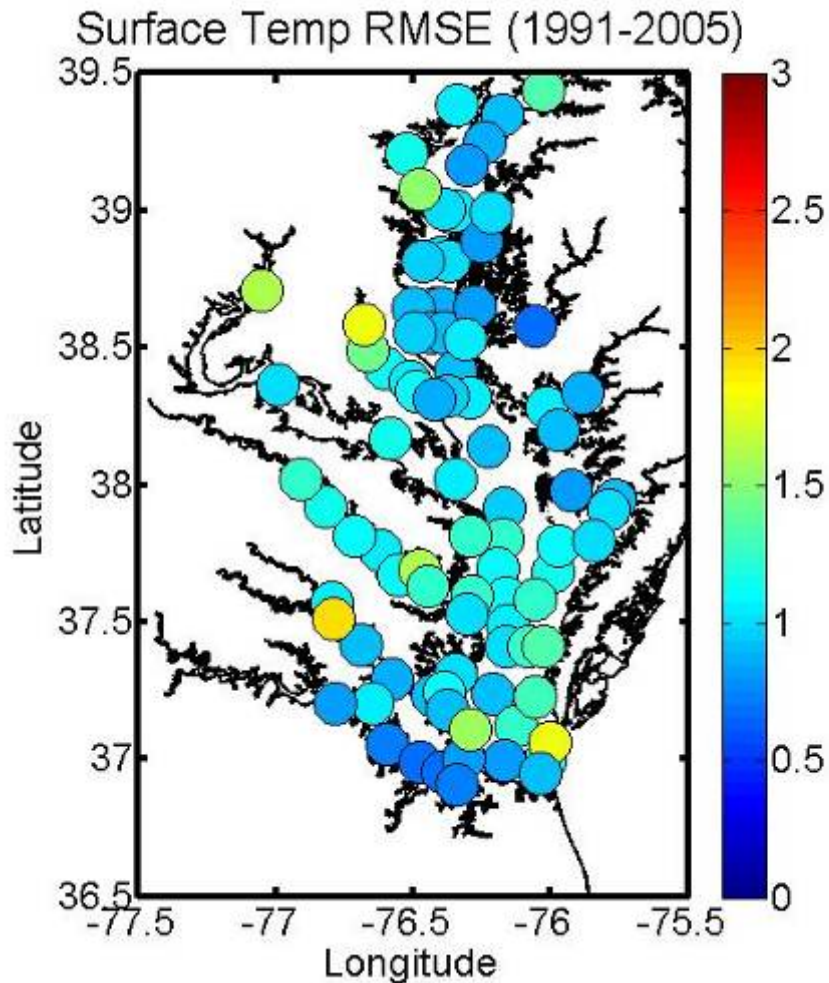
ChesROMS grid and bathymetry.

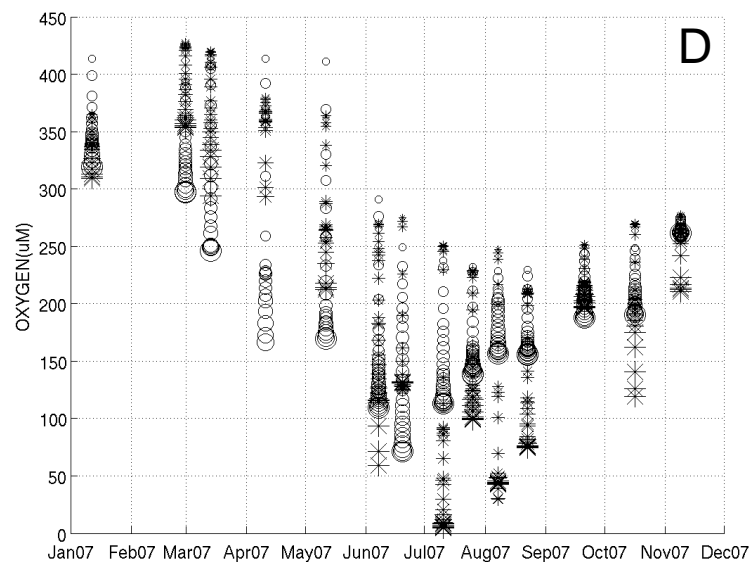
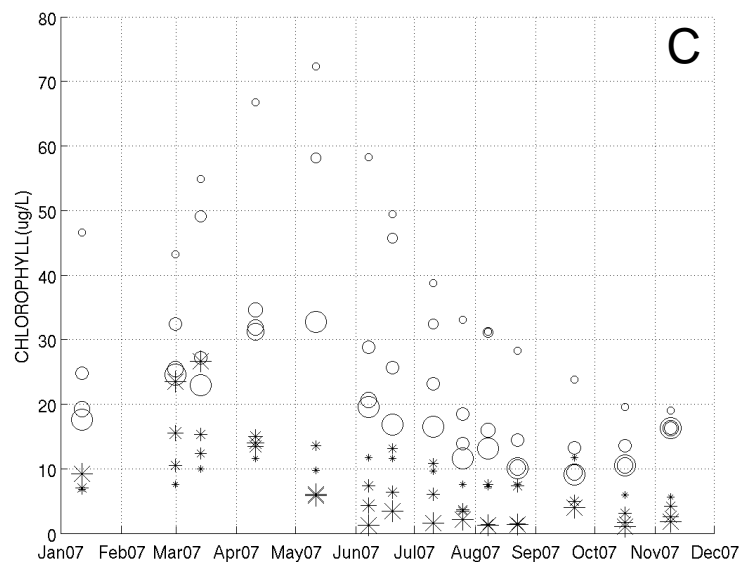
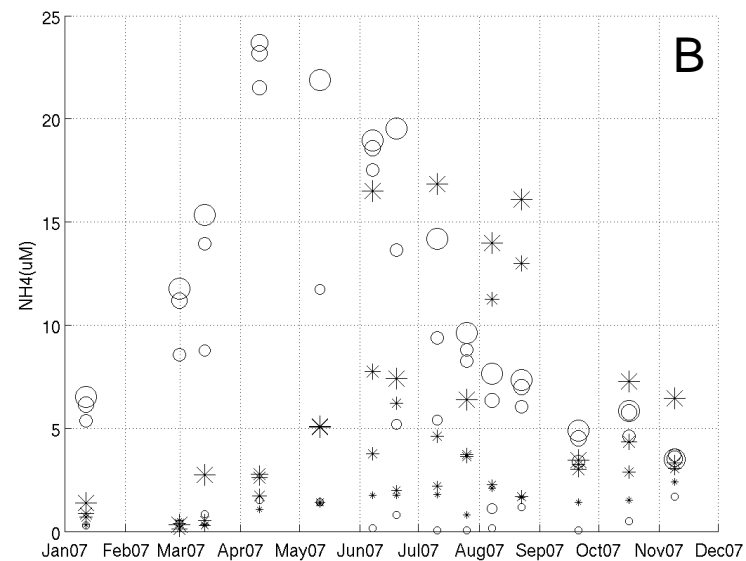
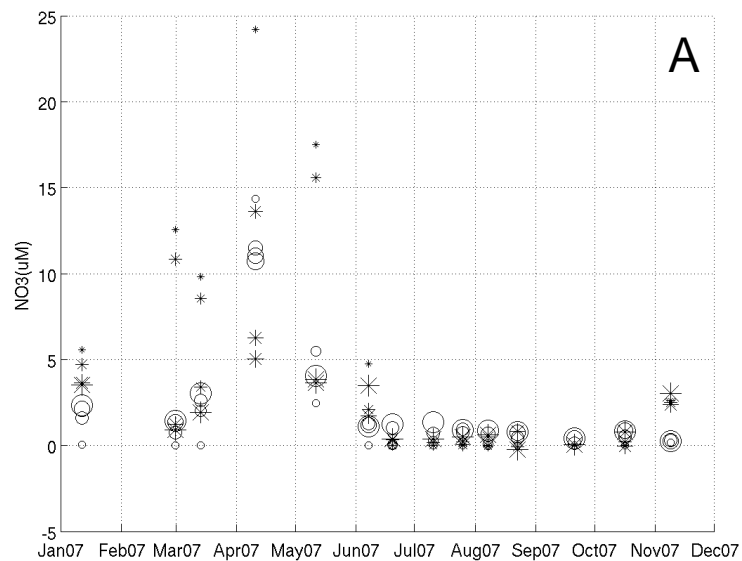




Taylor diagram of annual salinity, temperature, and water level predicted by ChesROMS over 15 years (1991 – 2005). From Xu et al., 2012.

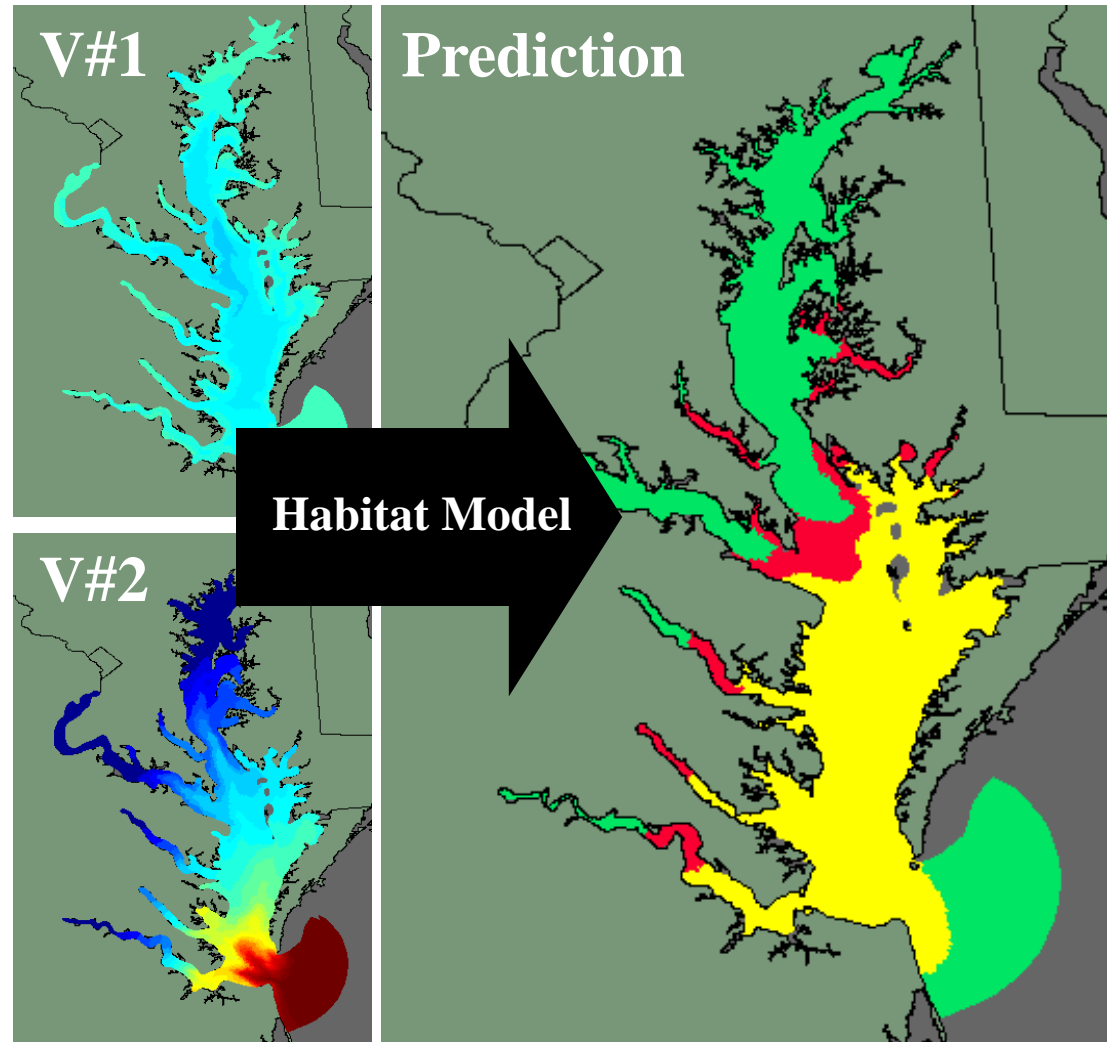
RMS Error of Surface SST & SSS





Empirical - Mechanistic Approach

Using **real-time and forecast data** acquired and derived from a variety of sources and techniques to **drive multi-variate empirical habitat models** that predict the abundance or presence of the target species



Motivation for Hybrid Approach

- Few existing methods work well and in near-real time
 - *Satellite remote sensing: low taxonomic resolution*
 - *In-situ probes: spatially and / or temporally limited*
 - *Mechanistic modeling: biology of most species not well known*

Statistical – Mechanistic Approach

- Based on concept of niche
 - *Identifies the geographic locations where ambient conditions coincide with the habitat of target organism*
- Feasible
 - *Many relevant variables can be acquired, estimated, or simulated in near-real time and forecast*
- Approach can be generalized to any location and any organism if sufficient habitat data and access to near-real time environmental data exists

Empirical Model Limitations

- Geographically and temporally limited
- Does not provide insights into the various processes

Habitat Models Used in CBEPS

Species	Model Type	Input Variables	Forecast	Accuracy (correct forecasts/n)	Reference
<i>Chrysaora quinquecirrha</i>	Logistic Regression	SST, SSS	Probability of Occurrence	87%	Decker et al., 2007
<i>Karlodinium veneficum</i>	Artificial Neural Network	SST, SSS, Month	Relative Abundance	84%	Brown et al., In prep.
<i>Microcystis aeruginosa</i>	Artificial Neural Network	SSS, DO, Chla, DIN, TN, NH4, TSS, Kd	Probability of Bloom Occurrence)	90%	Ramers et al., Unpublished
<i>Prorocentrum minimum</i>	Logistic Regression	Chla, NH4, TON, TSS, Month	Probability of Bloom Occurrence	88%	Ramers et al., Unpublished
<i>Vibrio cholerae</i>	Logistic Regression	SST, SSS	Probability of Occurrence	77%	Constantin de Magny et al., 2009; Louis et al., 2003
<i>Vibrio parahaemolyticus</i>	Logistic Regression	SST, CHL	Probability of Occurrence	82%	Jacobs et al., In prep.
<i>Vibrio vulnificus</i>	Logistic Regression	SST, SSS	Probability of Occurrence	93%	Jacobs et al., 2010; Jacobs et al., In prep.

Chla = chlorophyll-*a* concentration; SST = sea-surface temperature; SSS = sea-surface salinity; TON = total organic nitrogen; ISS = Inorganic suspended solids

K. veneficum Habitat Model

- Training / Testing Data Set (n= 151): MD DRN PP Live Count
- Habitat Model: Neural Network
- Input variables:
 - Sea-surface temperature
 - Sea-surface salinity
 - Month
- Output: Relative abundance (low (0 - 10 cells/ml), medium (10 - 2000 cells/ml), and high or "bloom" (> 2000 cells/ml))



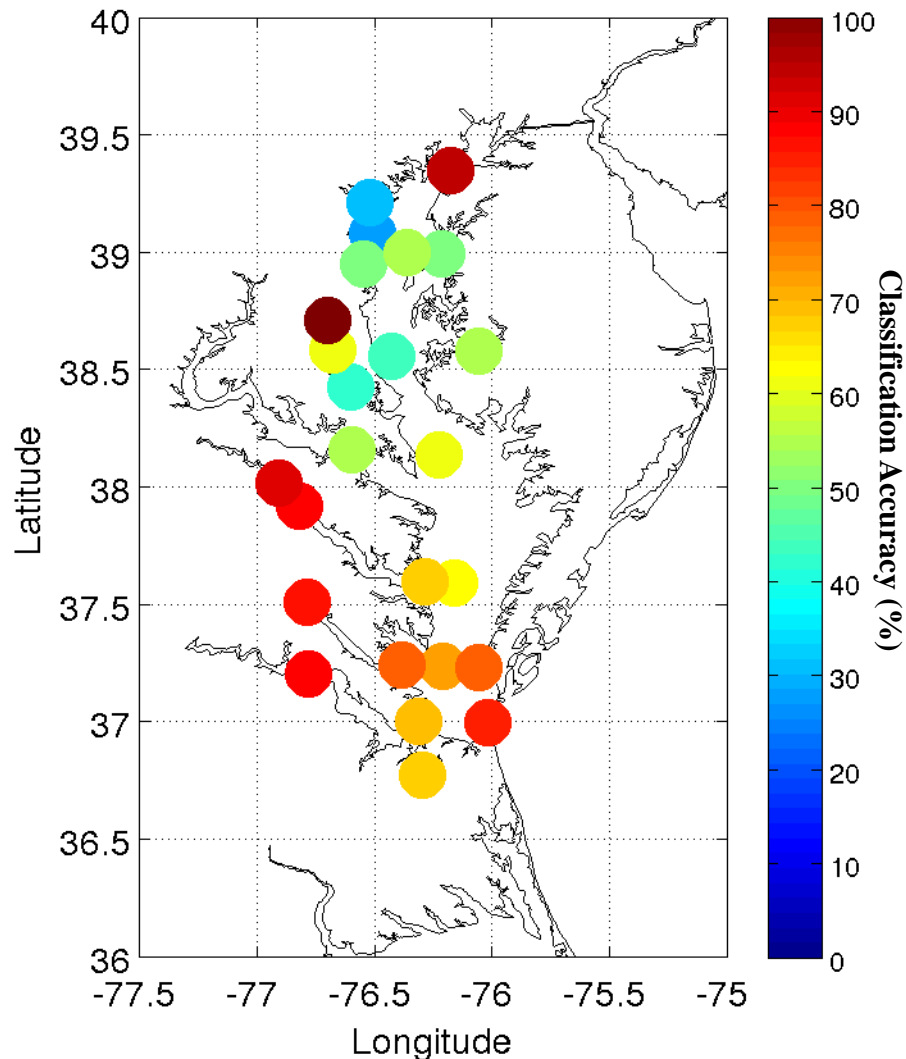
Photomicrograph of the
dinoflagellate
Karlodinium veneficum

Validation Confusion Matrix (n=81):

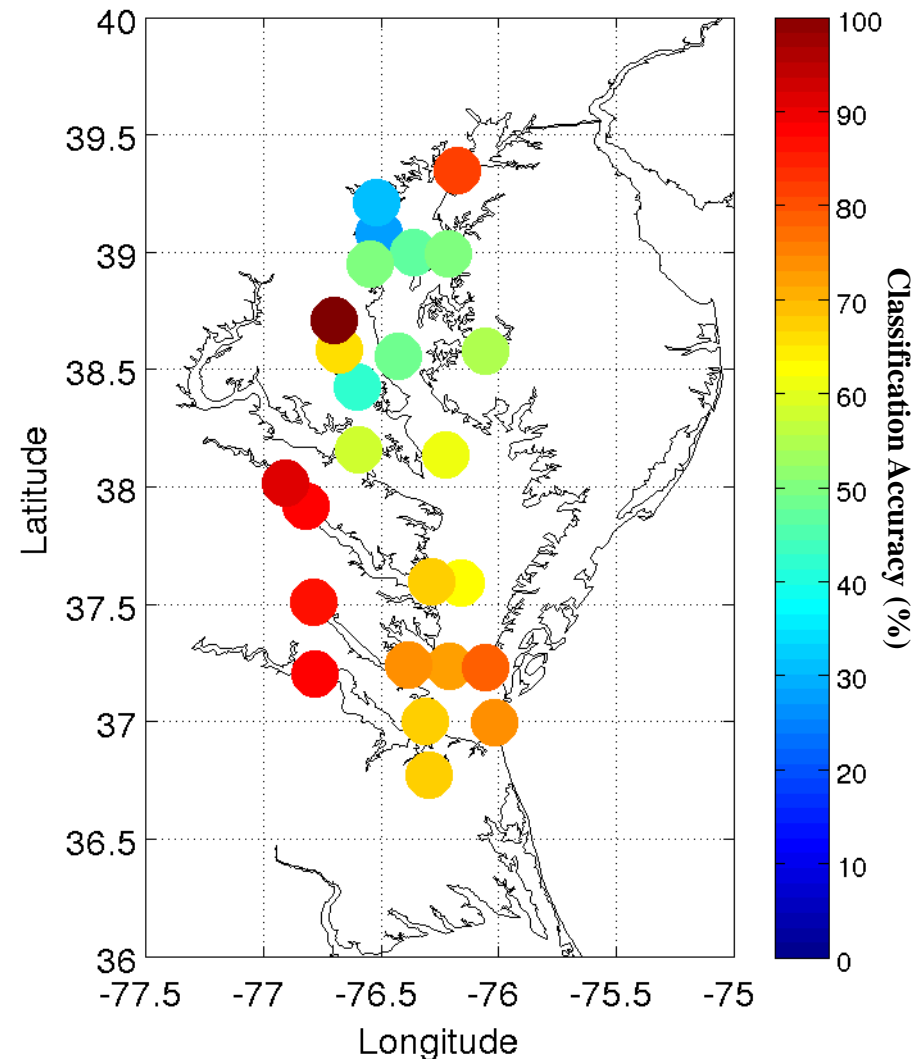
		Predicted		
		Low	Med.	High
Observed	Low	81%	28%	0%
	Med.	13%	61%	0%
	High	6%	11%	100%

K. veneficum Forecasting Skill

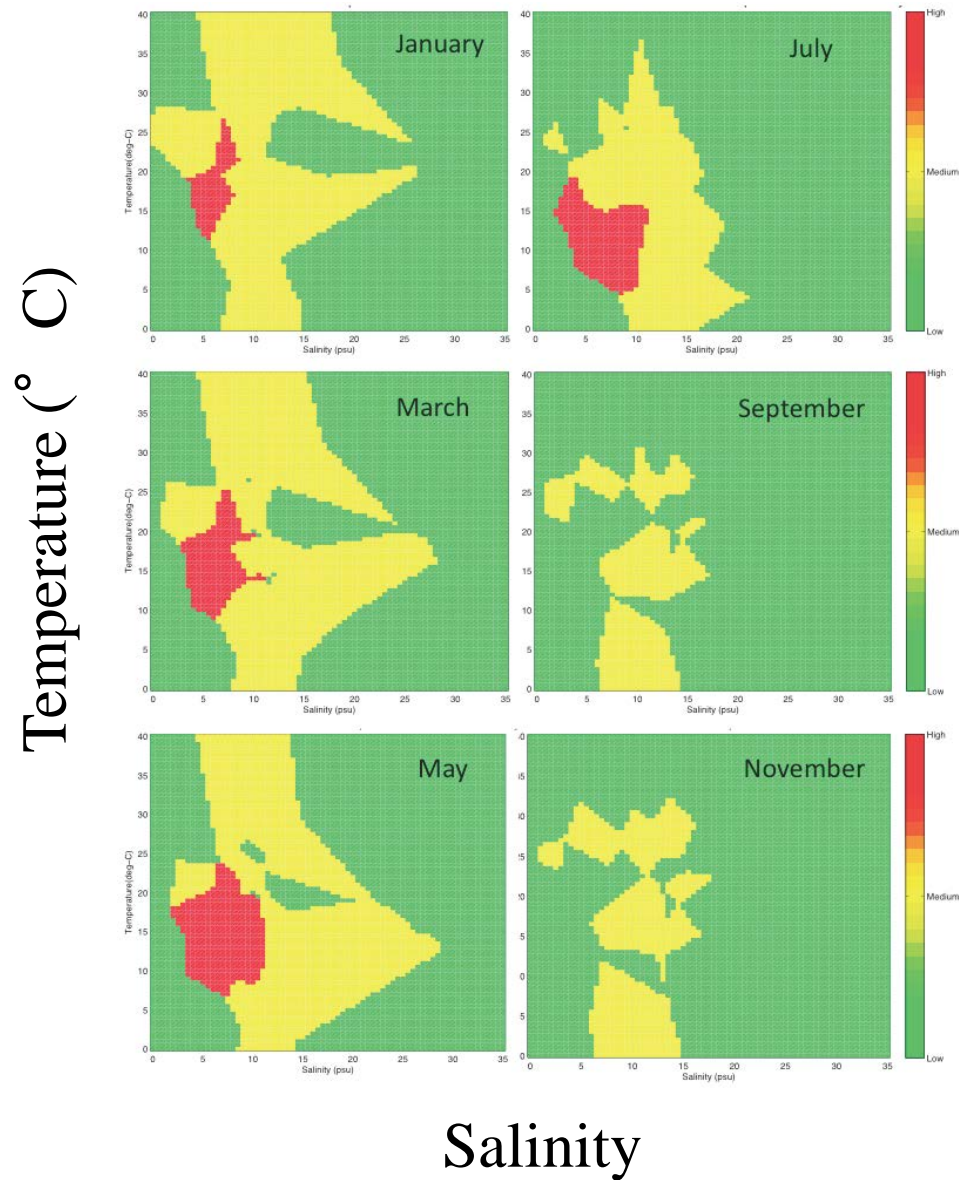
Nowcasts

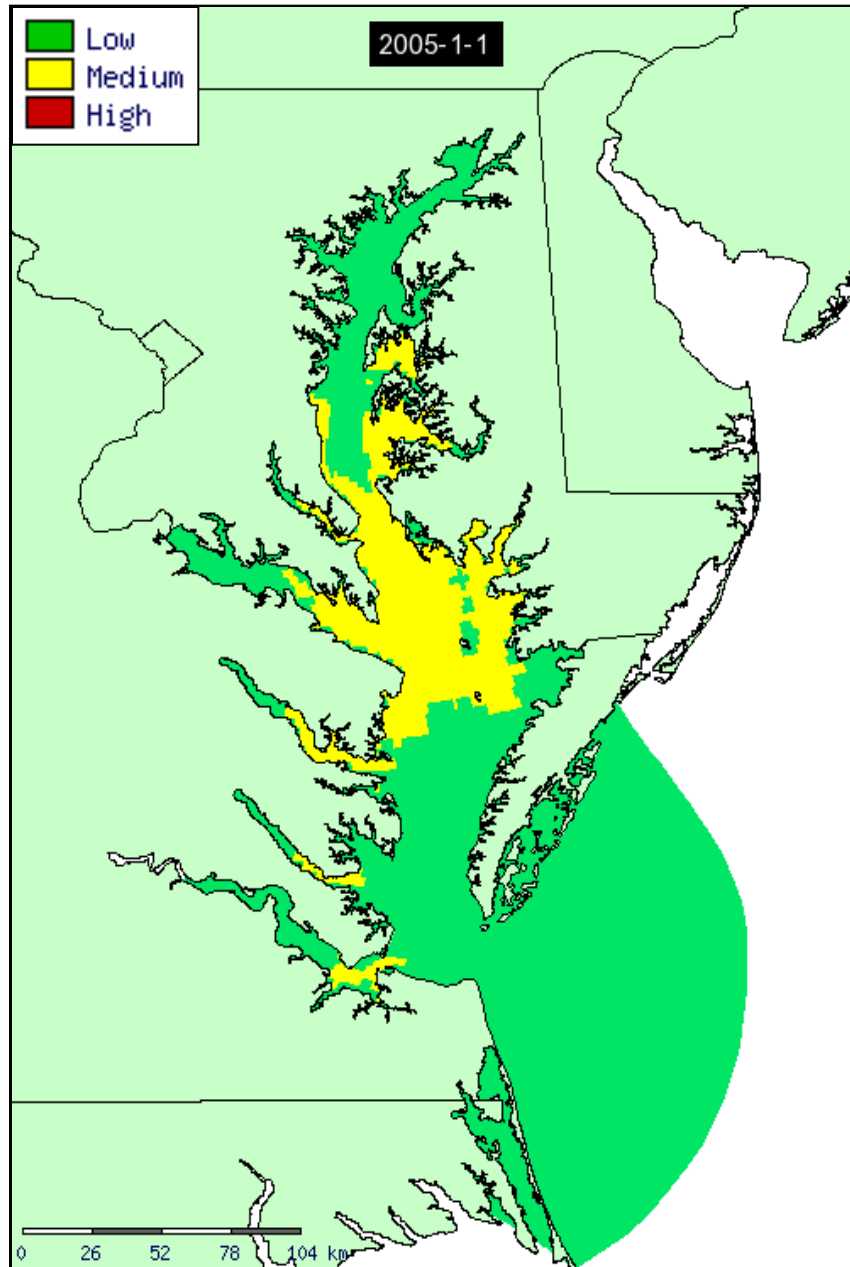


3-Day Forecasts



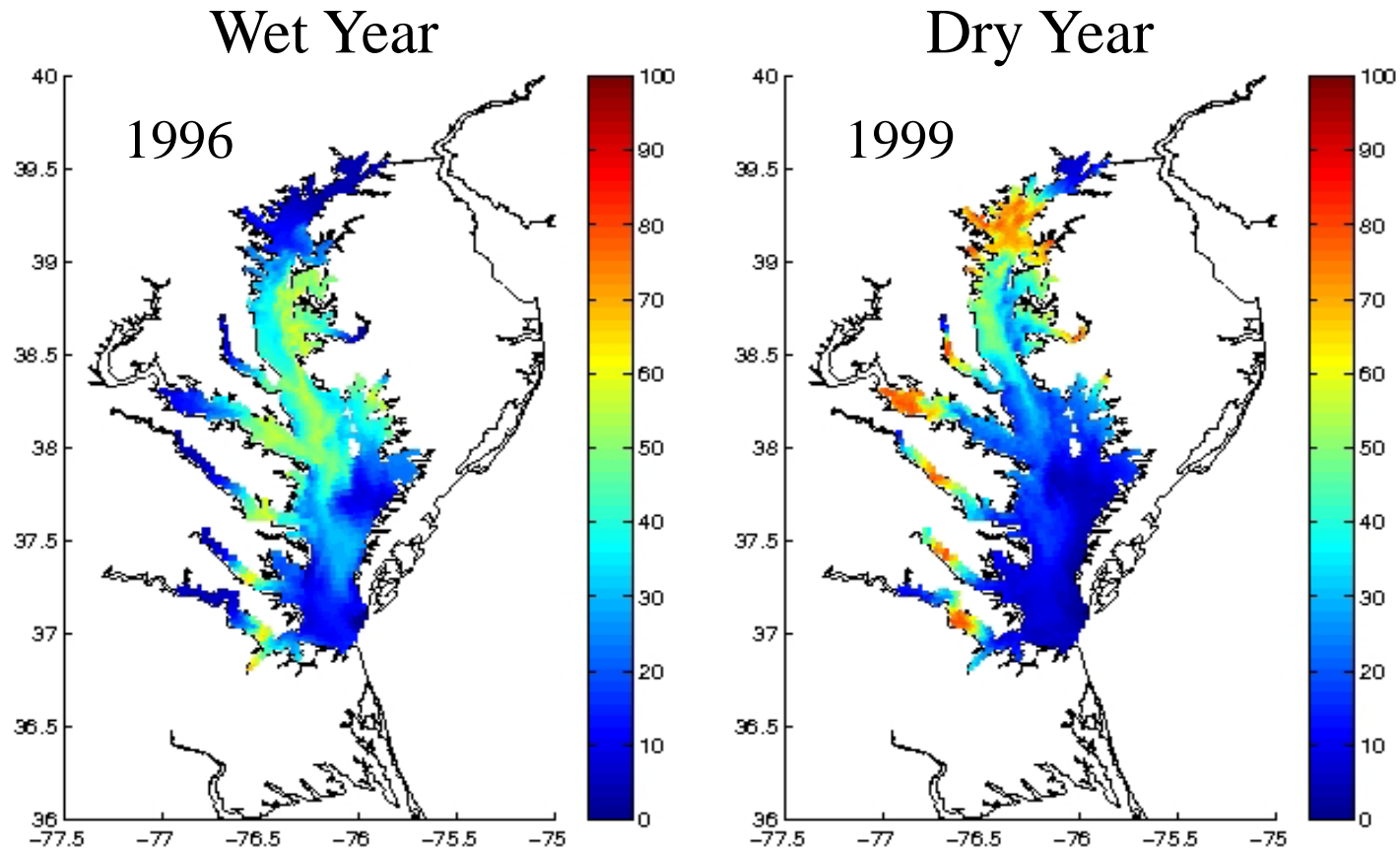
Predicted *K. veneficum* Relative Abundance





Animation of
predicted daily
Karlodinium
veneficum relative
abundance from
January 1 – December
31, 2005 using the
preliminary *K.*
veneficum habitat
model

Interannual Variability



Hindcasts depicting probability of occurrence of *V. vulnificus* in wet (1996) and dry (1999) years. Both figures represent conditions present on August 1st of each year. Color scale represents probability of occurrence with red high (100%) and blue low (0%).

Next Steps

- Continue validating forecasts
- Develop and incorporate mechanistic models to account for the relevant biotic and abiotic processes
- Predict toxicity and virulence associated with HABs and pathogens
- Assimilate additional sources of environmental information into prediction system
 - *Satellite imagery*
 - *In-situ observations*
- Include socio-economic models
- Transition vetted products to NOAA operations

Regional Earth System Model

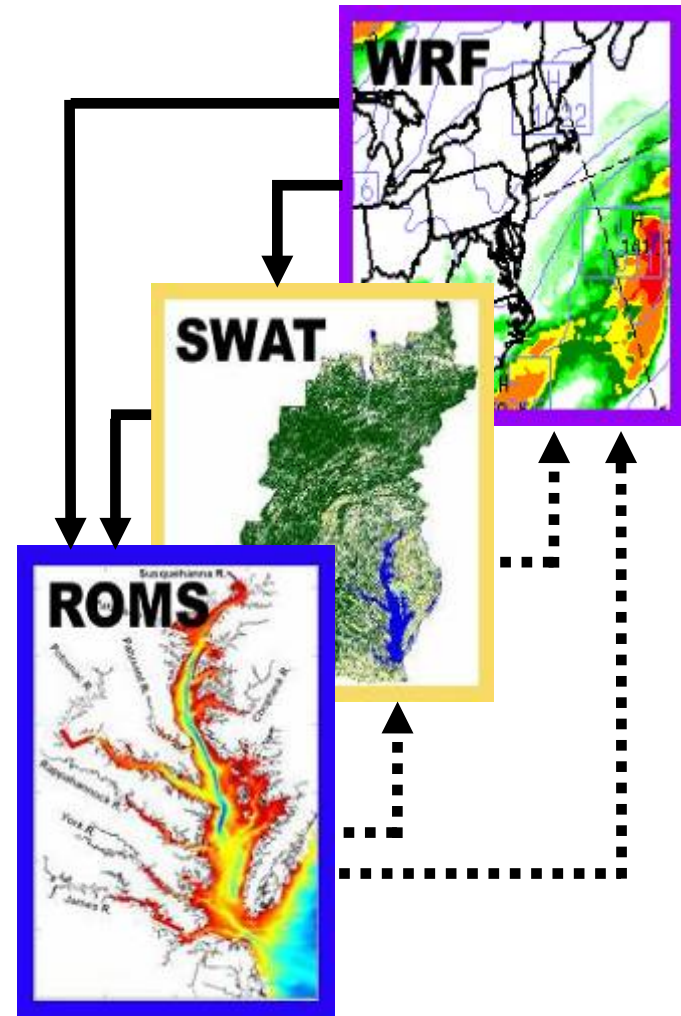
- **Objective**

- Revive and expand the Chesapeake Bay Forecast System -- a fully integrated, ecosystem model of the Chesapeake Bay and its watershed that assimilates *in-situ* and satellite-derived data by adapting and coupling existing models

- **CBFS System Components**

- **Air:** Atmosphere - Weather Research and Forecasting (**WRF**) Model
- **Land:** Land - Soil and Water Assessment Tool (**SWAT**)
- **Coastal Ocean:** Regional Ocean Modeling System (**CBEPS**)

- **Partners:** UM System, NASA, NOAA



Backup Slides

Forecast Availability

Predictions are generated daily and are available as data files and on the World Wide Web



Sea nettle forecasts: <http://chesapeakebay.noaa.gov/remote-sensing-for-coastal-management/forecasting-sea-nettles>

Forecasts Generated by the CBEPS

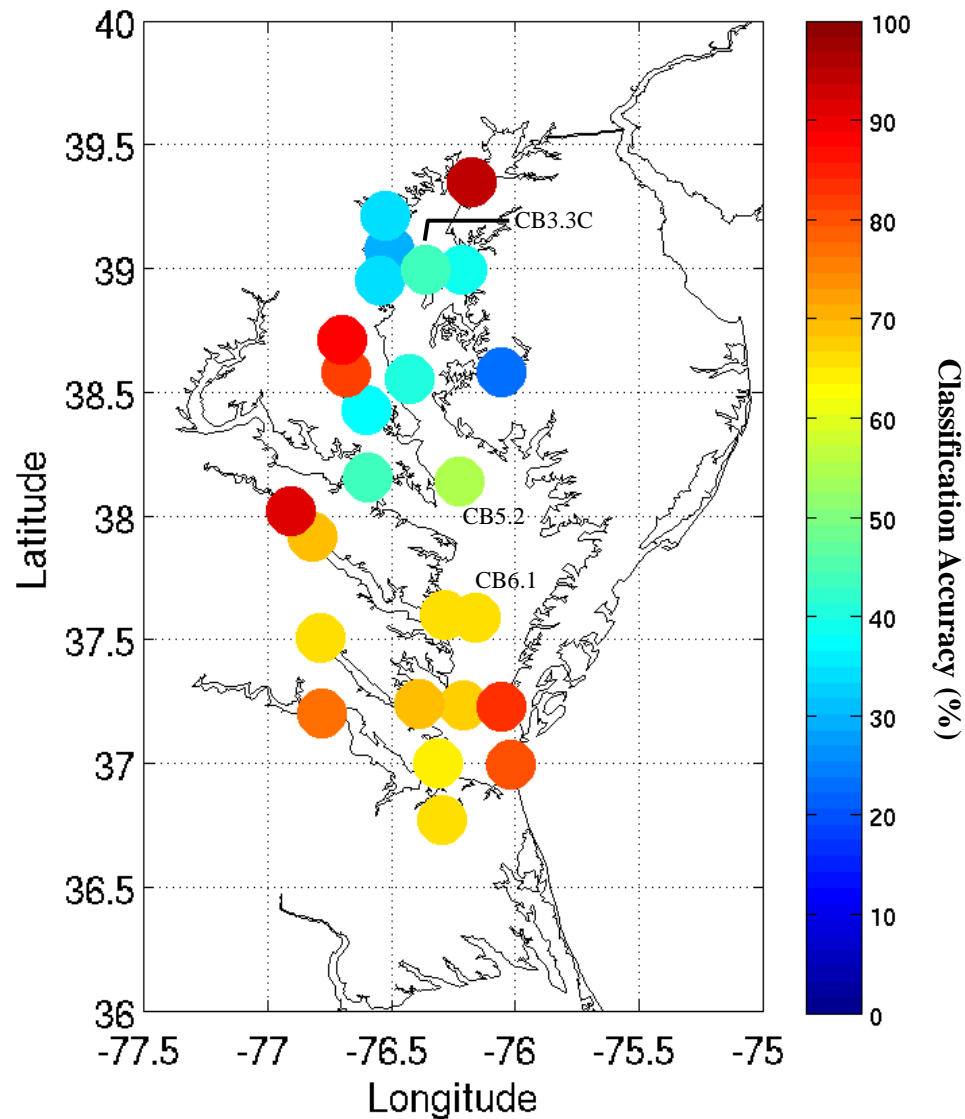
Physical	Biogeochemical	Organismal
Temperature	Nitrate (NO ₃)	<i>Chrysaora quinquecirrha</i> (scyphomedusa)
Salinity	Ammonia (NH ₄)	<i>Karlodinium veneficum</i> (dinoflagellate)
Water density	Dissolved Organic Nitrogen	<i>Prorocentrum minimum</i> (dinoflagellate)
Current velocity (u, v, w)	Chlorophyll	<i>Microcystis aeruginosa</i> (cyanobacteria)
Sea Surface Height (tidal and non-tidal water level)	Inorganic suspended sediments	<i>Vibrio cholerae</i> (bacteria)
Turbulent eddy viscosity	Detritus (small and large component)	<i>Vibrio parahaemolyticus</i> (bacteria)
Turbulent kinetic energy	Dissolved Oxygen	<i>Vibrio vulnificus</i> (bacteria)
Diffuse attenuation coefficient	Phytoplankton	
	Zooplankton	

K. veneficum Habitat Model

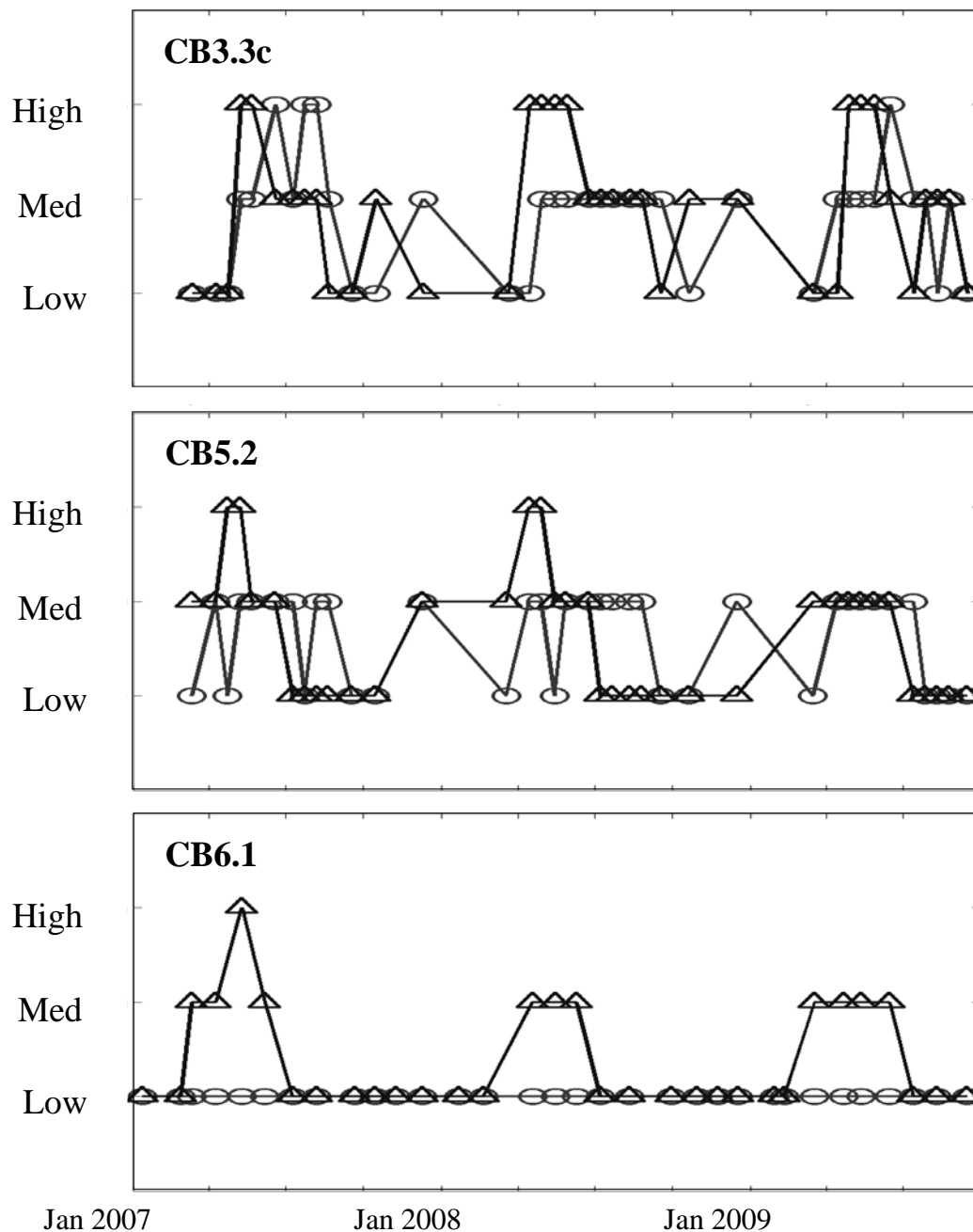
Development

- o Neural Network (NN) employs **sea surface temperature, salinity** and **month** to predict the relative abundance of *K. micrum* at low, medium and high or “bloom” concentrations
- o NN trained with samples (n = 151) of *in-situ K. micrum* abundance and various environmental variables
- o A test data set (n = 81) was extracted from the available data to assess the model’s performance

K. Veneficum Nowcasting Accuracy

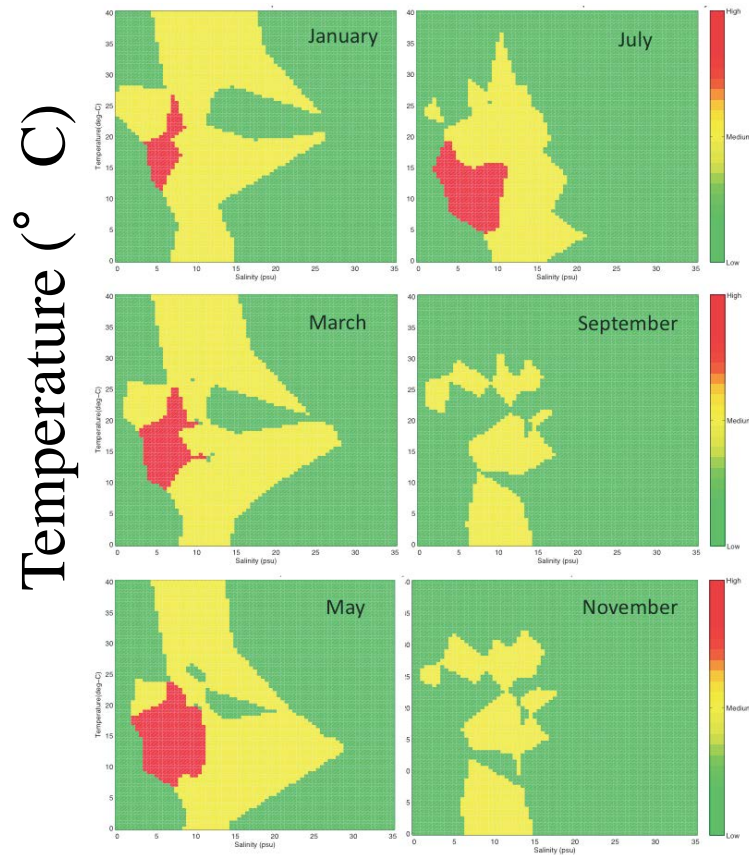


Relative Abundance of *K. veneficum*



Comparison of observed (○) and nowcast (△) *K. veneficum* relative abundance at Chesapeake Bay stations in the upper (CB3.3c), mid (CB5.2), and lower (CB6.1) Chesapeake Bay from January 2007 to October 2009.

Predicted *K. veneficum* Relative Abundance



Salinity

Fish kill in Middle River this November
(Bay Journal, Rona Kobell, 11/17/2015)



Enhance Ecological Forecasting

Ecological Event
Outlook



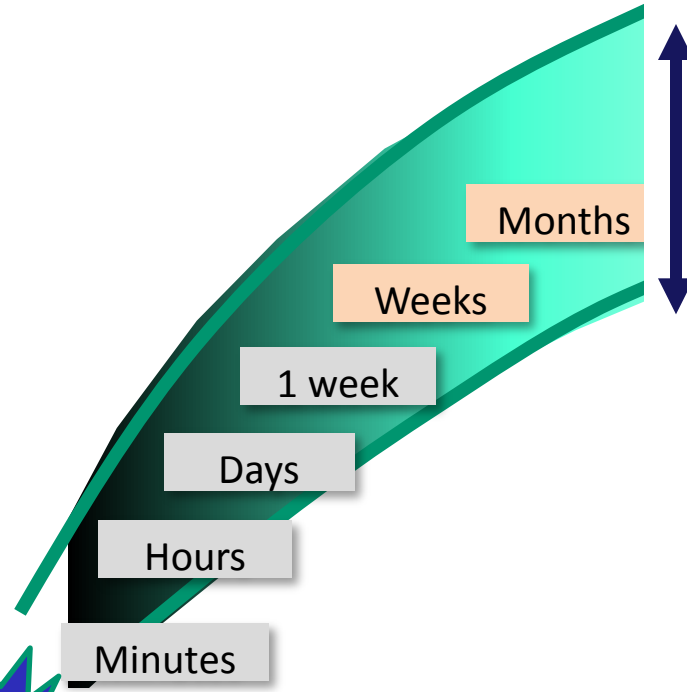
Ecological Event
Prediction



Ecological Event
Forecast



Ecological Event
Warning



Recovery

Planned Response

Preparation: Control, Mitigate



Water-Borne Pathogens

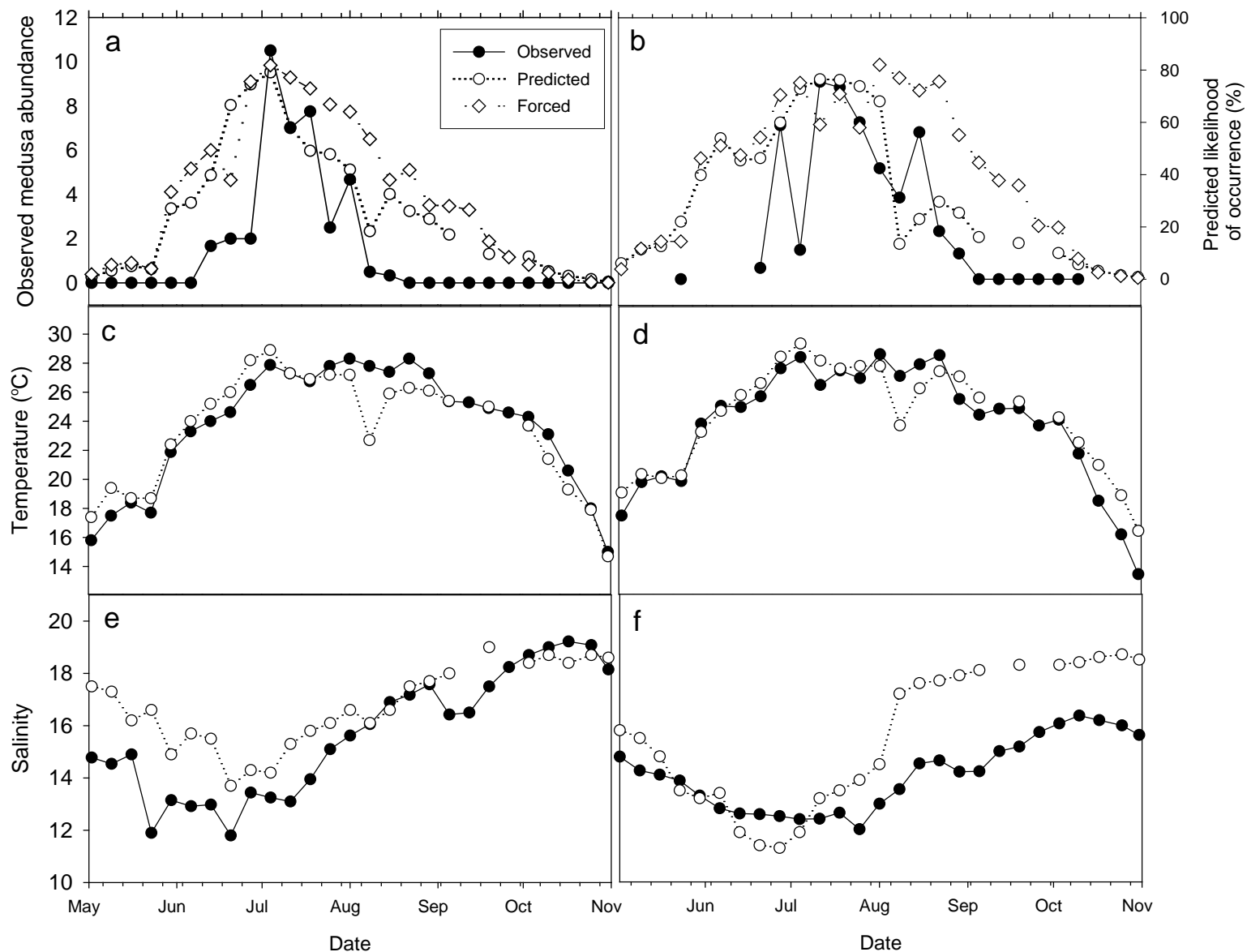
- Vibrio Forecasts
 - *V. cholerae*
 - *V. vulnificus*
 - *V. parahaemolyticus*



Courtesy of safeoysters.org

CBL

HPL



P. minimum Habitat Model

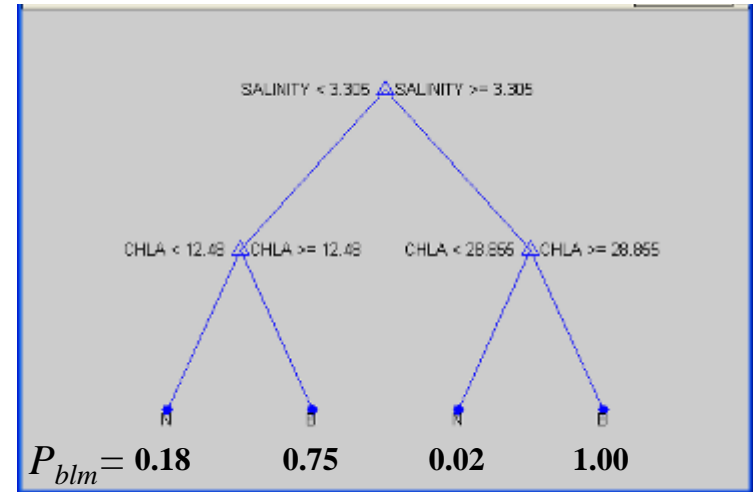
- Training / Testing Data Set: CBP (Morgan State)
- Habitat Model: Logistic Regression
- Input variables:
 - “APR-MAY”
 - Chl *a*
 - NH₄
 - TON
 - TSS
- Output: Probability of Bloom, where a bloom \equiv cell counts > 3,000 cells/ml

		Observed		
		Bloom	No Bloom	
Predicted	Bloom	16 7.1%	22 9.8% (False + 's)	42.1% 57.9%
	No Bloom	4 1.8% (False - 's)	182 81.3%	97.8% 2.2%
		80.0% POD*	89.2% 10.8%	88.4% Skill

* *Probability of Detection*

M. aeruginosa Habitat Model

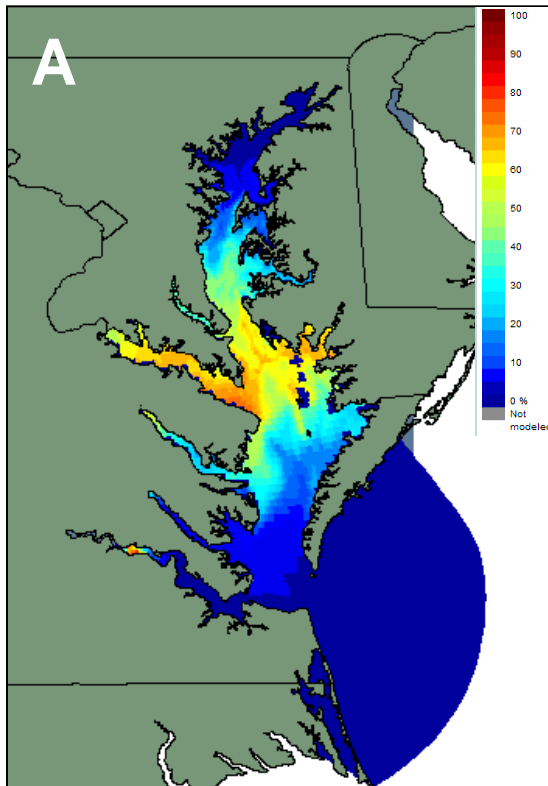
- Training / Testing
Data Set: MD DRN PP
Live Count
- Habitat Model:
Hierarchical Decision
Tree
- Input variables:
 - Chlorophyll concentration
 - Sea-surface salinity (SSS)
- Output: Probability of
Bloom (P_{blm}), where
bloom \equiv cell counts $>$
10,000 cells/ml



		Observed		
		Bloom	No Bloom	
Predicted	Bloom	202 11.6%	65 3.7%	75.7% 24.3%
	No Bloom	52 3.0%	1425 81.7%	96.5% 3.5%
		79.5% 20.5%	95.6% 4.4%	93.3% 6.7%
		1	2	Target Class

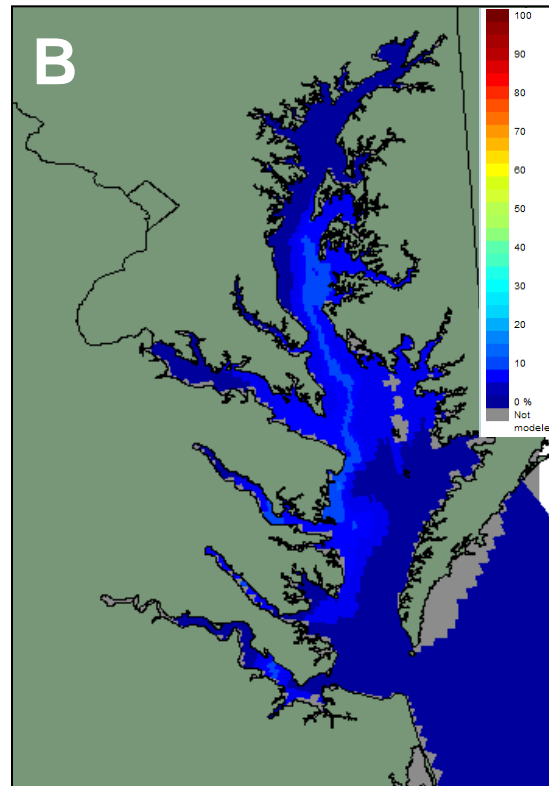
Examples of Chesapeake Bay Species Forecasts

Sea Nettle



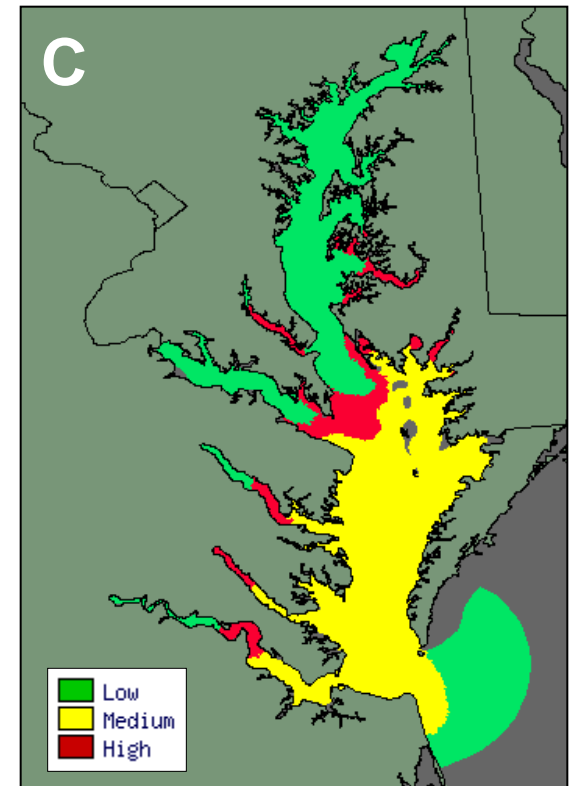
Predicted likelihood of encountering sea nettles on 17 August 2007.

Water-borne Pathogen



Predicted likelihood of *Vibrio vulnificus* on 20 April 2011.

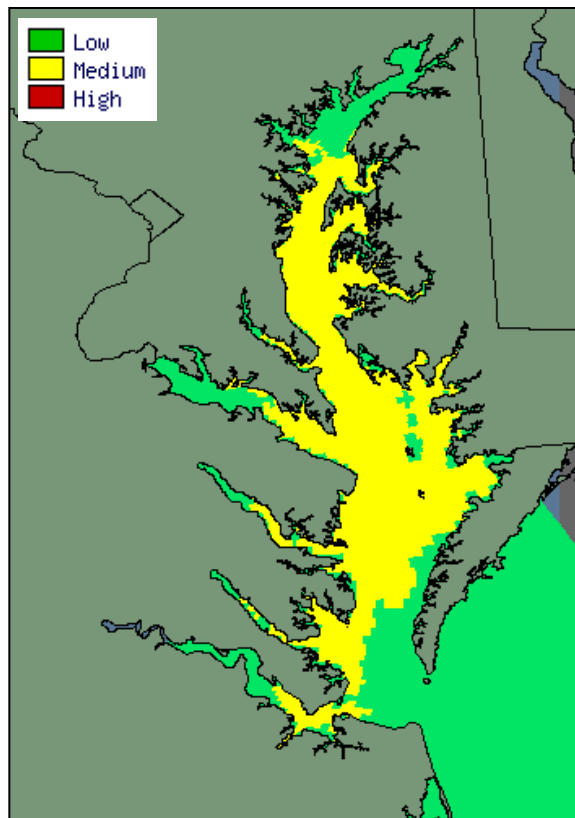
Harmful Algal



Predicted relative abundance of *Karlodinium veneficum* on 20 April 2005

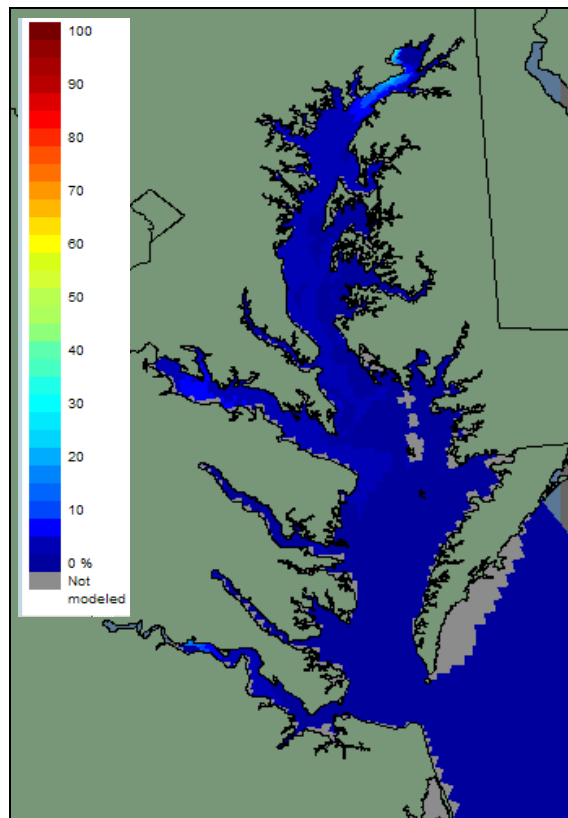
Current Chesapeake Bay HAB Forecasts

Karlodinium veneficum



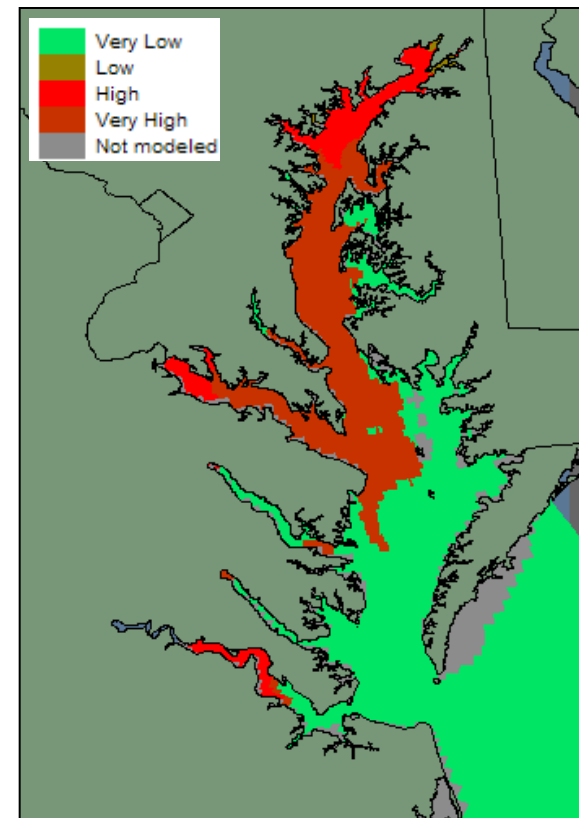
Predicted relative abundance of *K. veneficum* on June 20, 2011. Low: 0-10, med: 11-2000 cells/ml, high: > 2000 cells/ml.

Prorocentrum minimum



Predicted probability of a *P. minimum* bloom on June 20, 2011. Bloom: $\geq 3,000$ cells/ml.

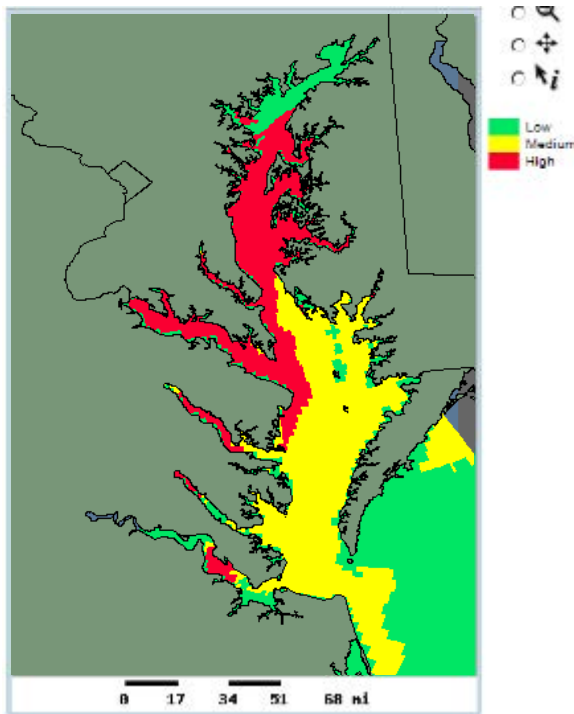
Microcystis aeruginosa



Predicted probability of a *M. aeruginosa* bloom on June 20, 2011. Bloom: $\geq 10,000$ cells/ml.

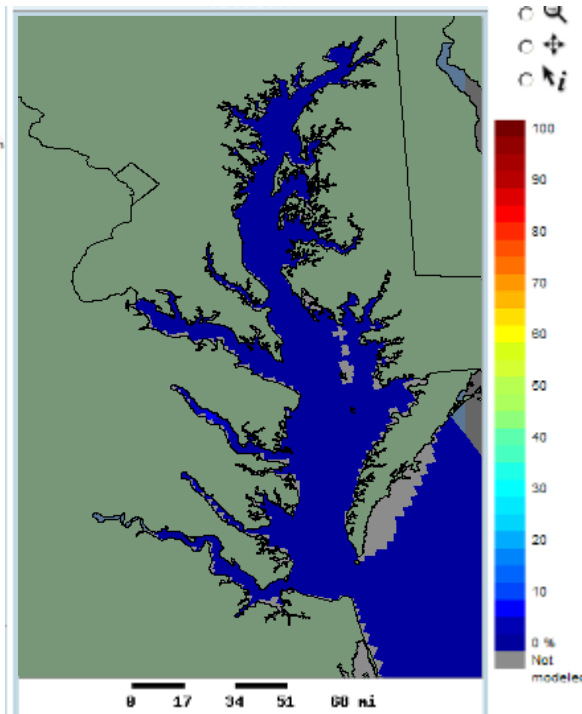
Current Chesapeake Bay HAB Forecasts

Karlodinium veneficum



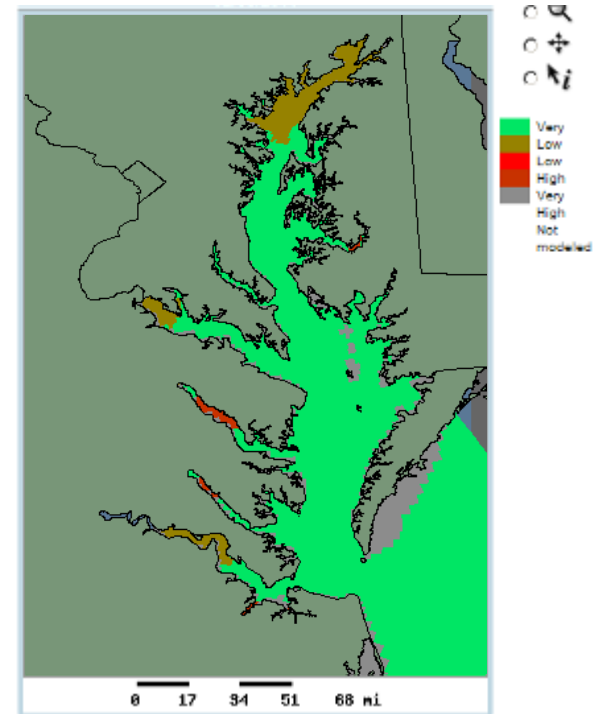
Predicted relative abundance of *K. veneficum* on March 23, 2013

Prorocentrum minimum



Predicted probability of a *P. minimum* bloom on March 23, 2013

Microcystis aeruginosa



Predicted probability of a *M. aeruginosa* bloom on March 23, 2013

Operational Status of Ecological Forecasts in Chesapeake Bay

- Sea Nettles: operational demonstration (NOAA Chesapeake Bay Office)
- Vibrios:
 - *V. cholerae*: research
 - *V. parahaemolyticus*: research
 - *V. vulnificus*: operational demonstration (NCEP)
- HABs: research

