

The use of artificial intelligence algorithms to  
assess ecological patterns from marine  
landscapes.  
The RIOSAMP experience.

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# RIOSAMP Project

## Coastal Resource Management Council (CRMC)- Fugate

### Ocean Engineering

- Hydrodynamics and sediment transport [Grilli S. and Harris]
- Acoustic [Miller & Potty]
- Geotechnics [Baxter]
- Wind Resource Assessment [Grilli A. and Spaulding]
- Siting and Environmental Impact [Grilli A. and Spaulding]

### Graduate School of Oceanography

- Oceanography [Codiga and Ullman]
- Mammals [Kinney]
- Geology and Habitat (King)
- Fisheries (Coolie)
- Primary productivity (Nixon&Oviate)
- Meteorology(Merril)

### College of Environment and Science

- Birds [Paton]
- Fisheries [Smythe]
- GIS [Damon]

### College of Art and Science

- Archeology[Mather]

### Coastal Research Institute

- Outreach (McCann)

## SAMP Management Team

Fugate – McCann- Spaulding - Nixon – De Bow

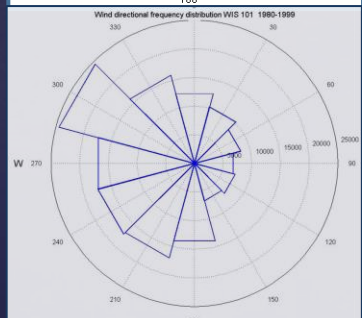
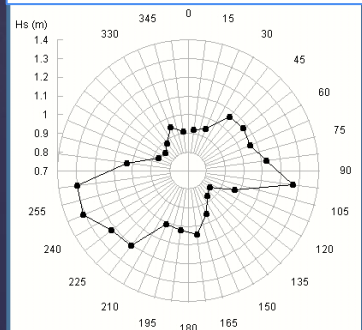
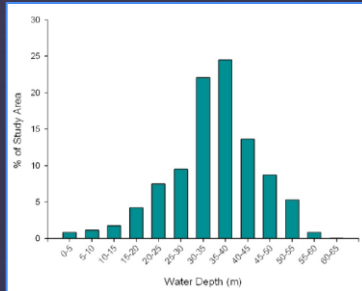
Full list contributors and complete flow chart: [http://seagrant.gso.uri.edu/oceansamp/pdf/documents/about\\_org\\_chart.pdf](http://seagrant.gso.uri.edu/oceansamp/pdf/documents/about_org_chart.pdf)

## Objectives:

- Provide up to date scientific information to guide CRMC in its renewable energy permit policy
- Provide a zoning map of the SAMP area identifying optimal area for wind farm siting

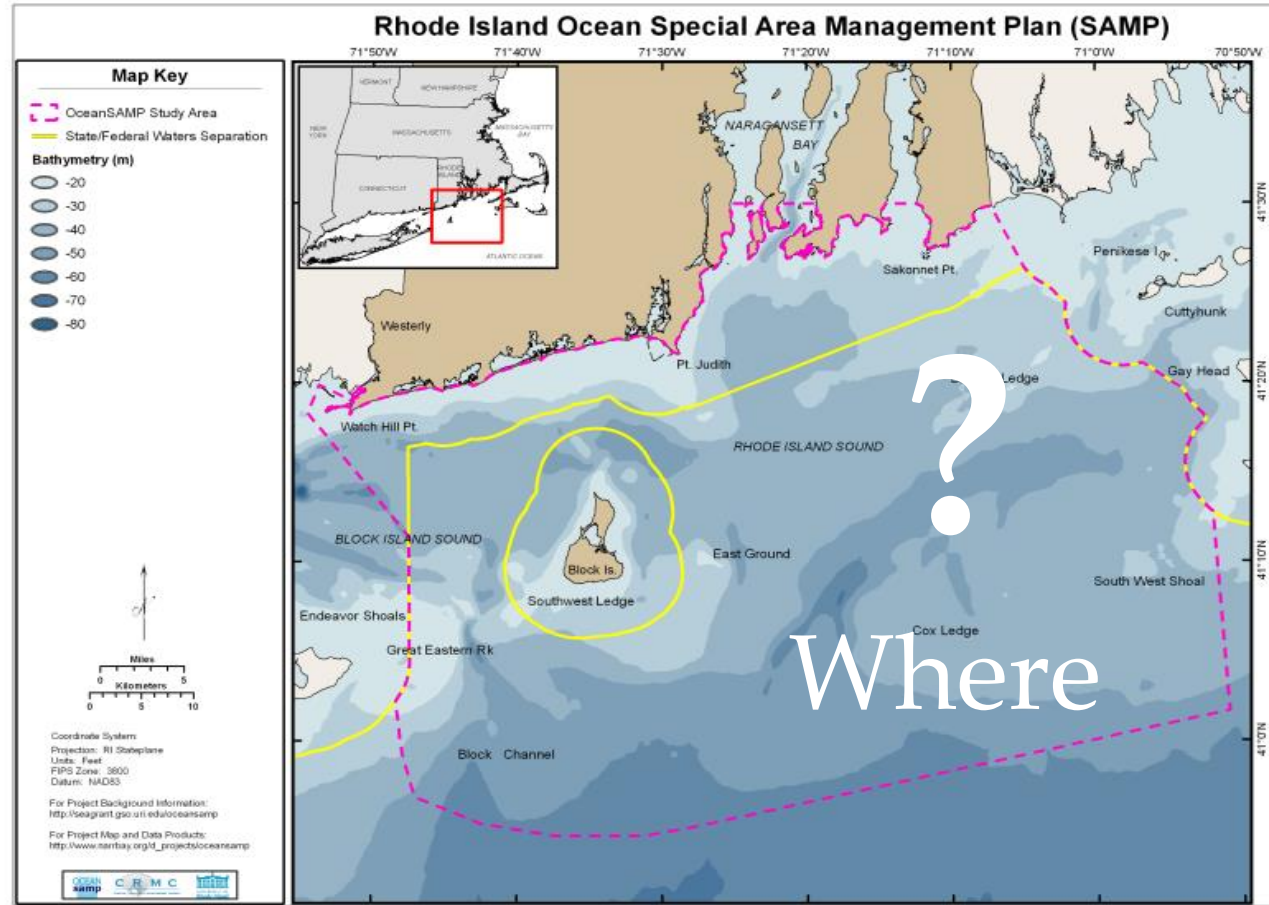
# Wind Farm Siting Optimization

38 000  
Km<sup>2</sup>  
Depth  
15 - 60 m

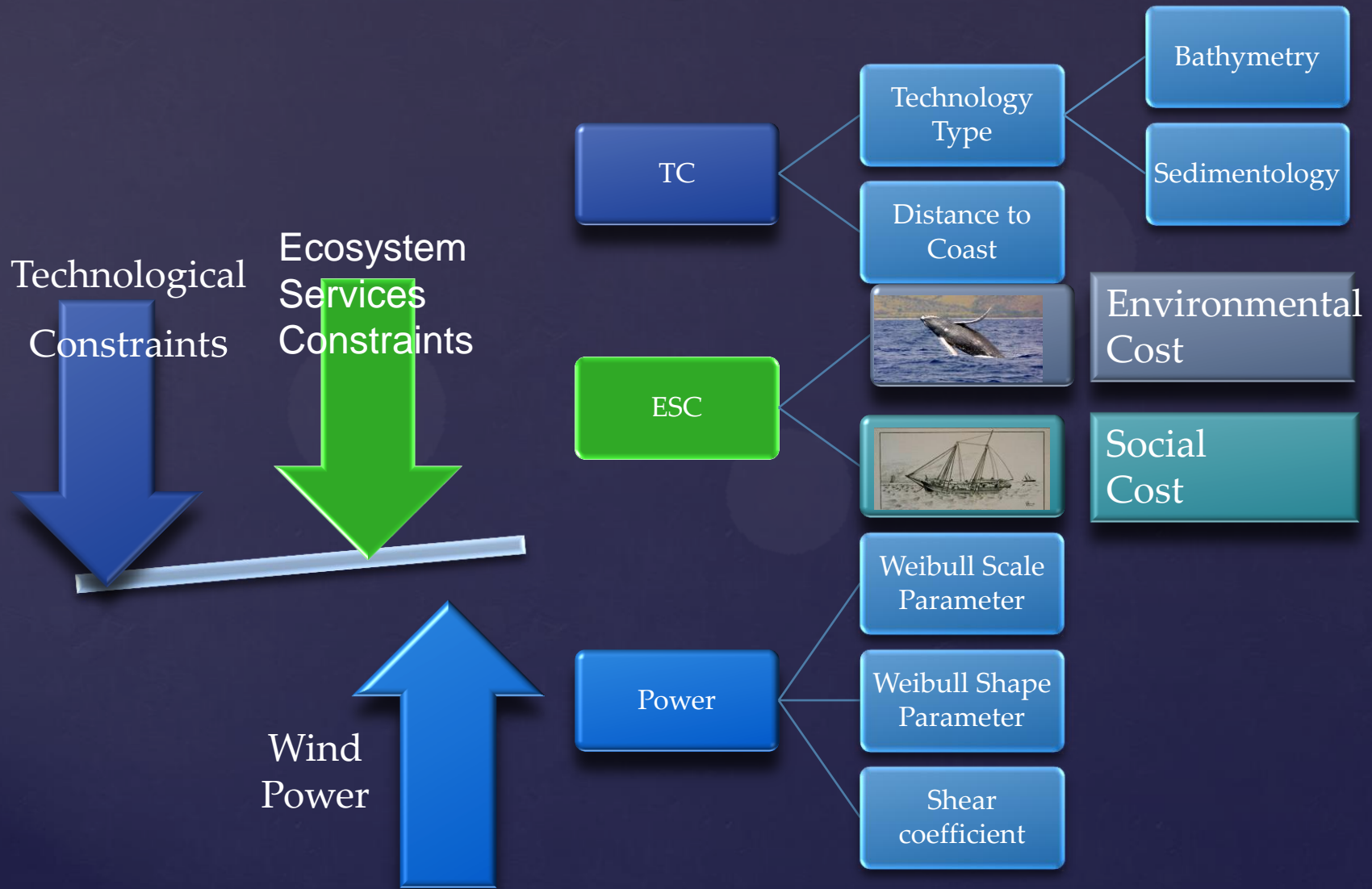


Wave  
Climate

Wind  
Climate



# Wind Farm Siting Index (WiFSI)



# SAMP Ecosystem based Management (EBM) conceptual Framework

**GOAL:** *keep a healthy, productive and resilient ecosystem which provides to human the services they need (ecosystem services)*

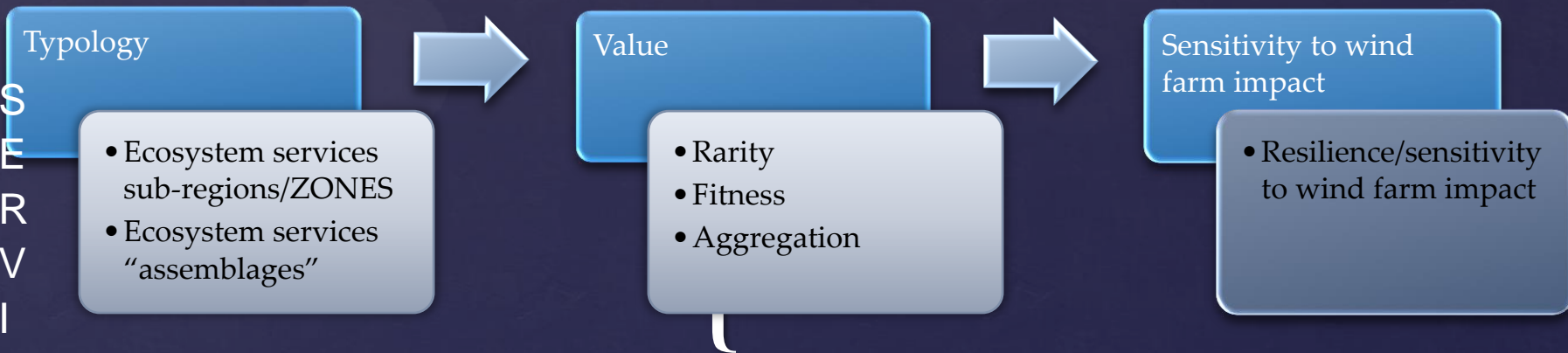


*Nomenclature of Karen McLeod and Heather Leslie (2009).*

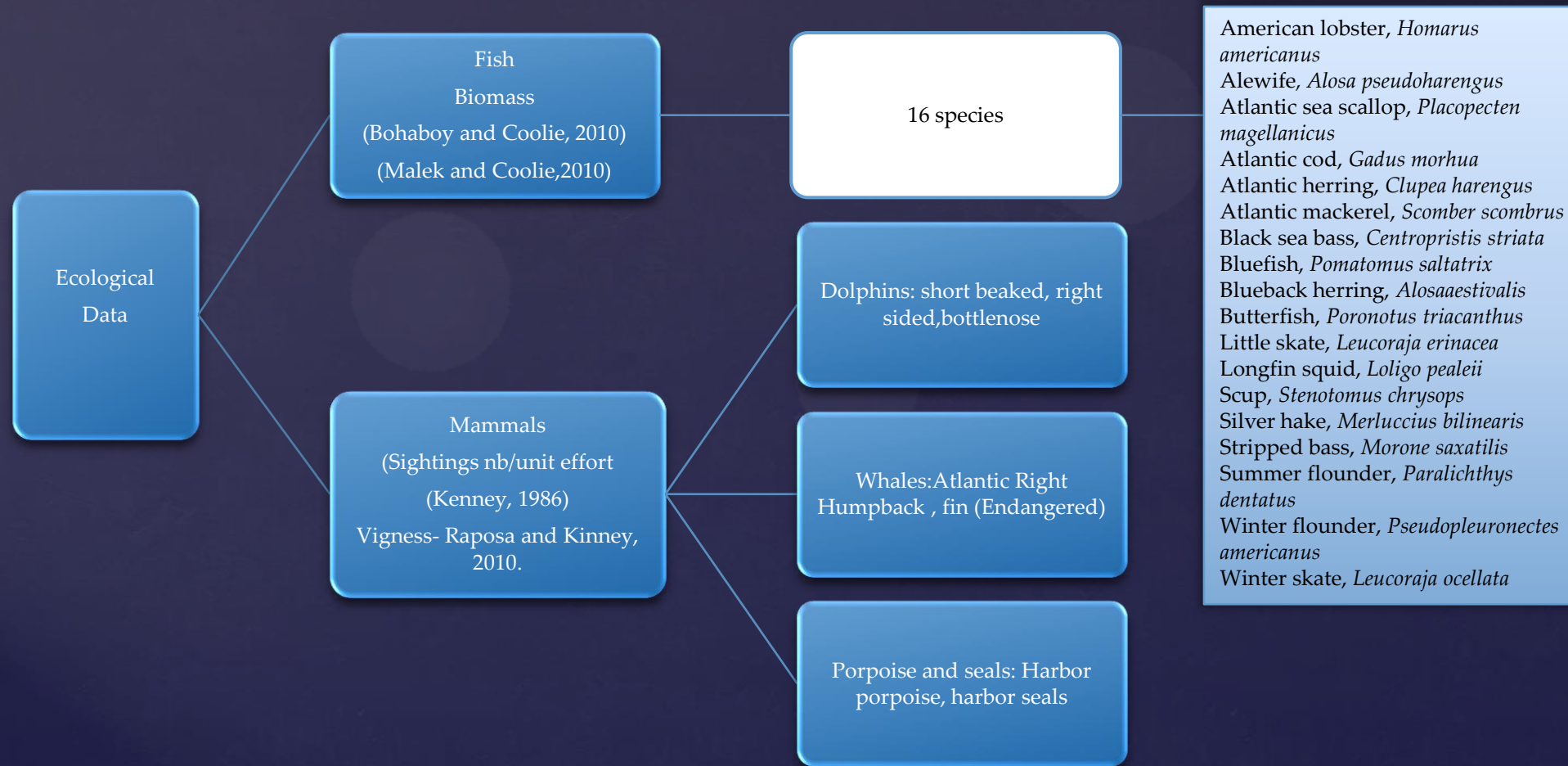
*Modified according to Oumeraci, 2011*

# Ecosystem Services Valuation

E  
S  
C  
O  
L  
O  
G  
I  
C  
A  
L



# Ecological Service



# Ecological Data Analysis

## STATISTICAL DISTRIBUTION

- LOG-NORMAL DISTRIBUTION
- NORMALISATION

## SPATIAL INTERPOLATION

- MINIMUM 30 POINTS
- KRIGGING ON 250X 250 M GRID
- STATISTICAL DISTRIBUTION CONSERVED

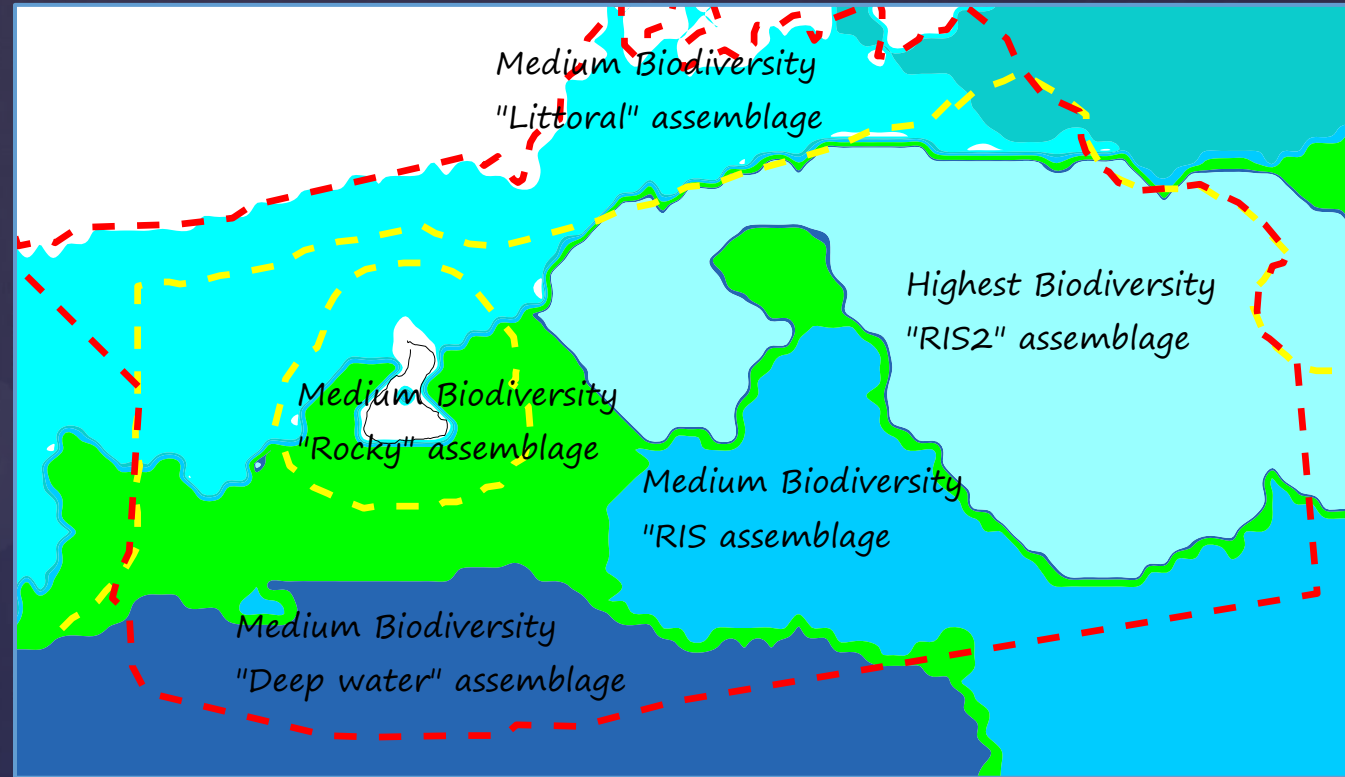
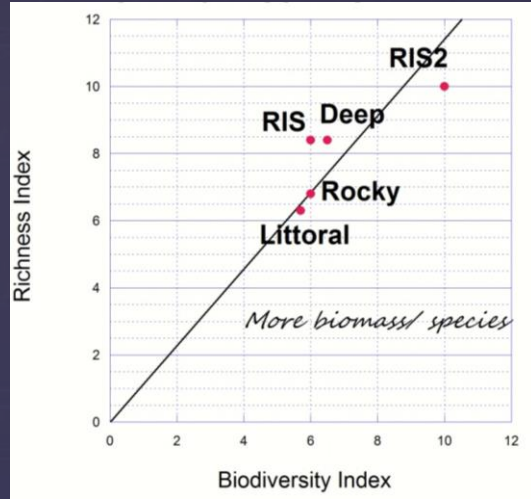
## MULTIVARIATE SPATIAL ANALYSIS

- PRINCIPAL COMPONENTS
- CLUSTER ANALYSIS (K-MEANS)



# Ecological Typology

# Spring



- Deep
  - Mammals
- Rocky
  - Demersal
- RIS
  - Herring
  - Mammals
- RIS2
  - Demersal
  - Herring
- Littoral
  - Demersal
  - Lobster

# Ecological Service Sensitivity To Wind Farm Impact

Species Sensitivity coefficient

Based on French McKay

10

• Death

8

• Health Damage

6

• Habitat Modified

4

• Behavior Disturbance

2

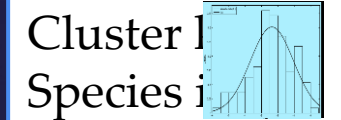
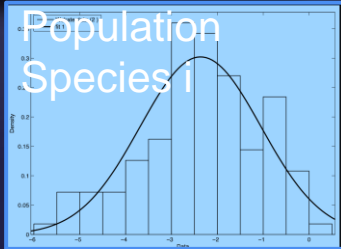
• Non significant impact

0

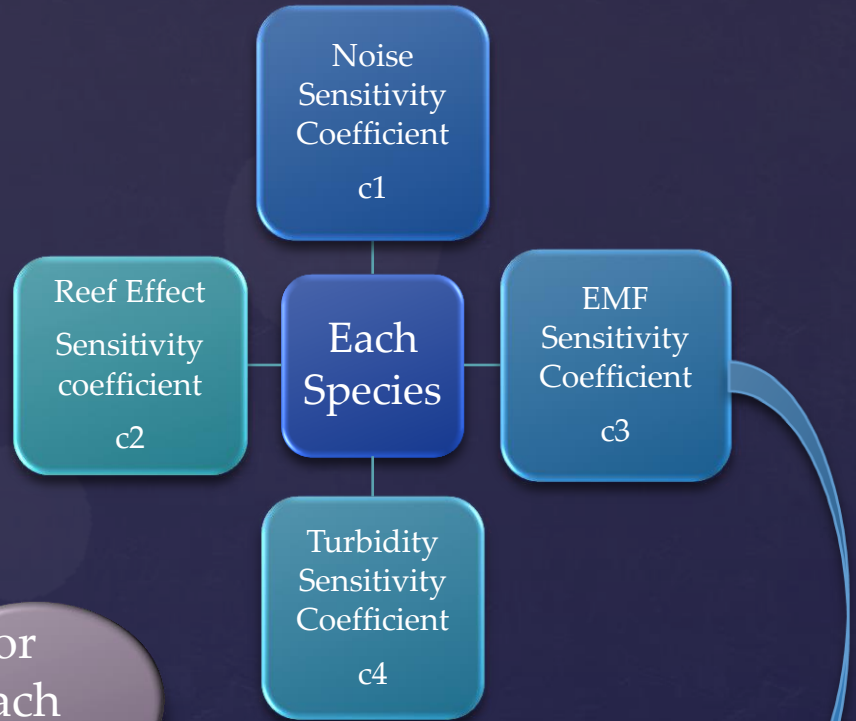
• No impact

-2

• Attraction



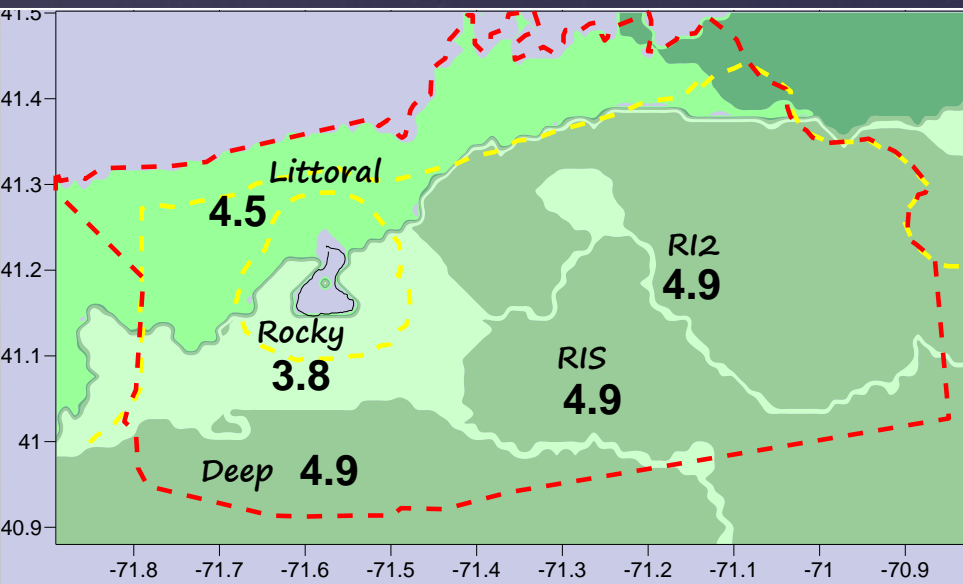
“S”  
 Each species in a cluster/zone obtains a score S based on the combination of its relative abundance probability distribution (PD) to the population PD. and to its local relative richness



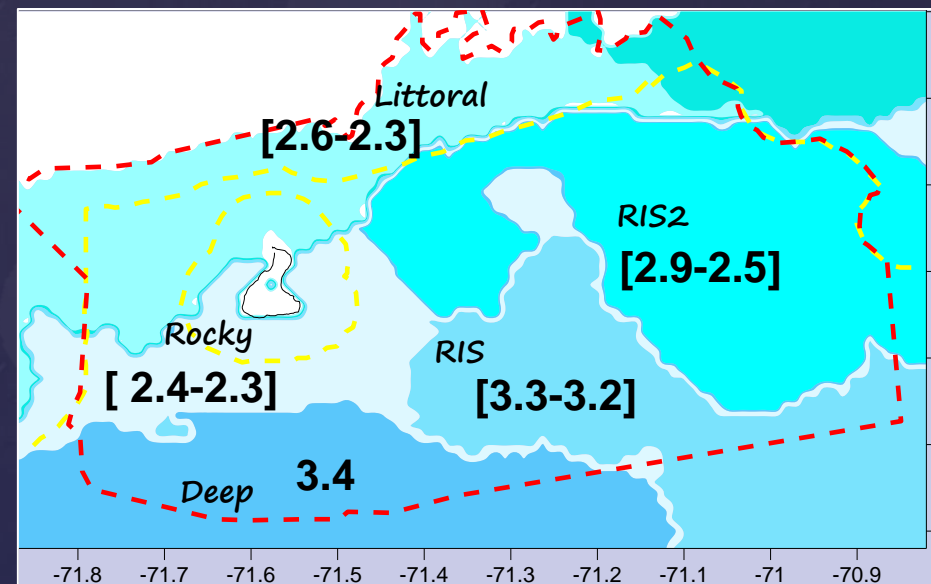
$$I_{c,o} = \dot{a}_n (s_i^2 * c_{i(c,o)})$$

$$ESI_{c,o} = sign(I_{c,o}) * \frac{\sqrt{|I_{c,o}|}}{N \sqrt{\max |(s_i^2 * c_{i(c,o)})|}}$$

# Ecological Service Sensitivity Index To Wind Farm Impact



Construction phase (Spring)

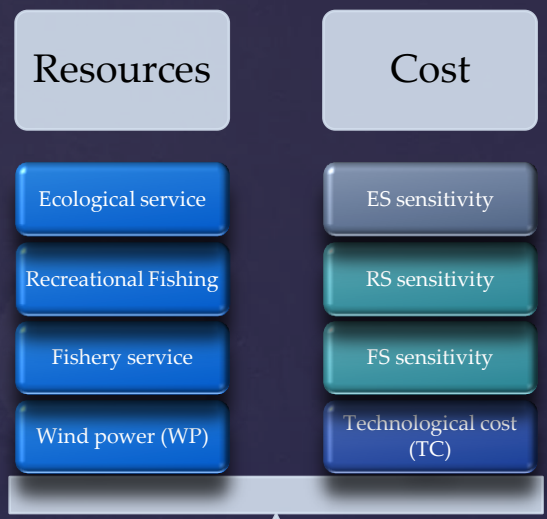


Operation Phase (Spring)

# Wind Farm Siting Index

[WiFSI]

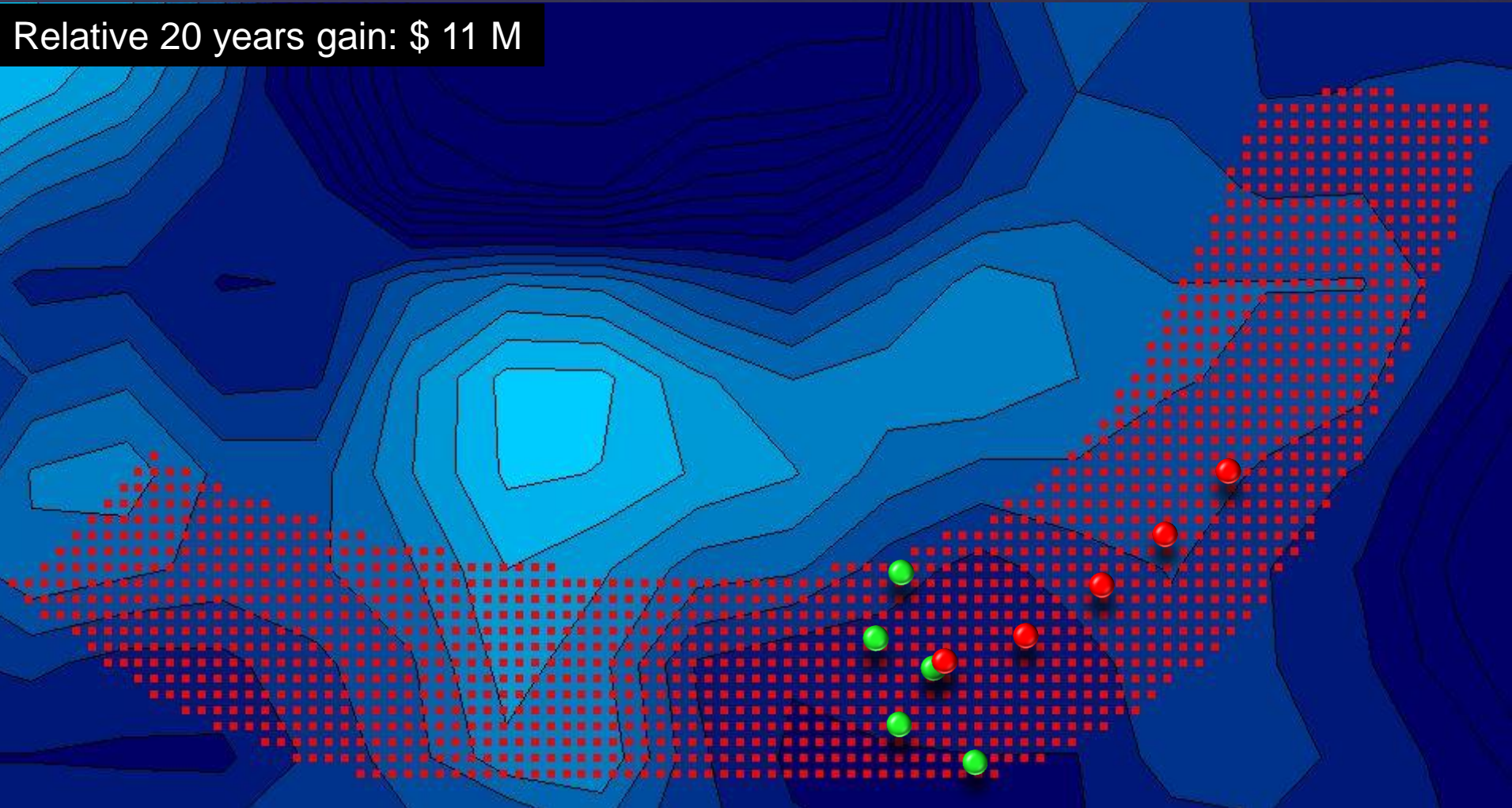
$$WiFSI = \frac{w_1 * TC + w_2 * ESI + w_3 * FSI}{WP}$$



# OPTIMIZATION RESULTS

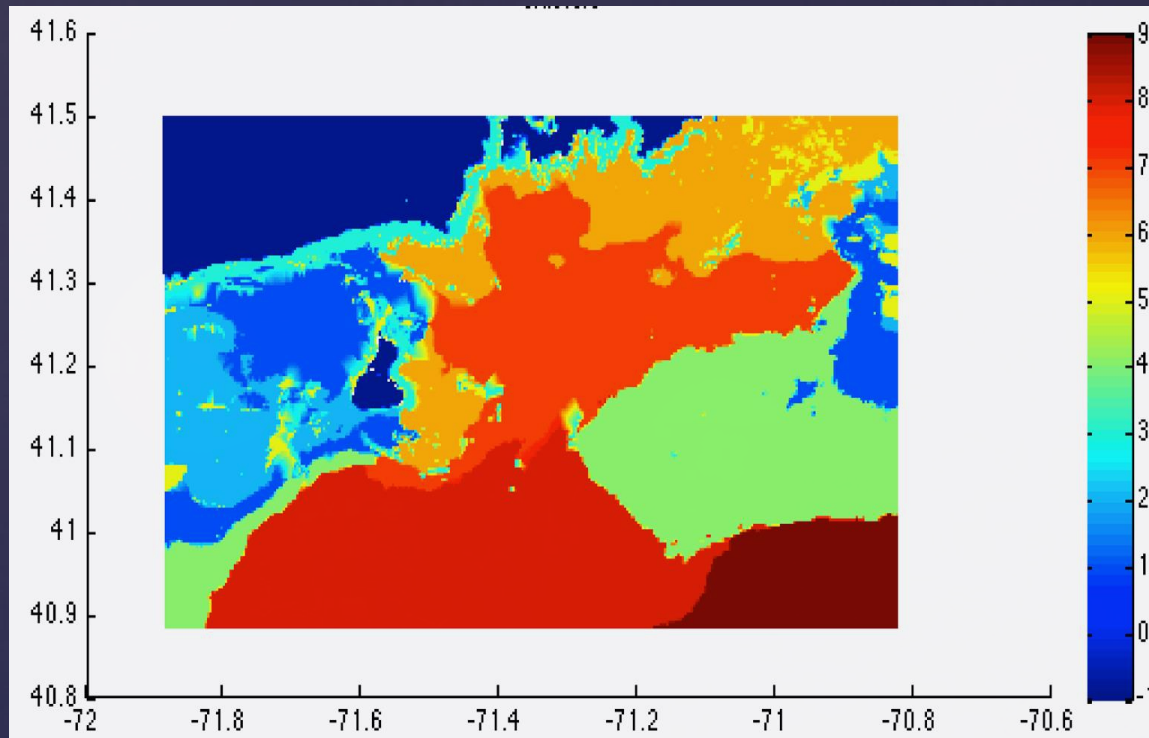
Genetic Algorithm – Include WaSP Wake model in optimization function

Relative 20 years gain: \$ 11 M



O'Reilly C., Grilli A. and Potty G. 2013. Micrositing Optimization of the Block Island Wind Farm, RI, USA.  
Proceeding of the International Conference on Ocean, Offshore and Arctic Engineering (OMAE 2013), Nantes June 9-14, 2013

# Marine Landscapes



# Data

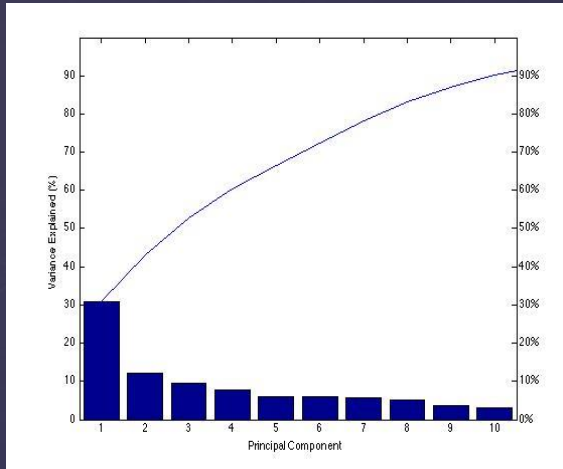
Variable name	Description	Unit	Source
<b>Tidal velocity*</b>	Maximum tidal velocity	m/s	ROMS modeling Grilli S. et al. 2010; Harris et al., 2012.
<b>Significant wave height*</b>	95 % Significant wave Height in a 50 year storm event	m	STWAVE modeling Grilli A. et al 2008
<b>Depth*</b>	Water Depth	m	NGDC Coastal Relief Model
<b>Distance to shore</b>	Distance from each grid cell to closest point to shore	km	Grilli A. et al, 2010
<b>Slope</b>	Maximum slope between 2 grid cells (200 m apart)	Deg.	NGDC Coastal Relief Model ; SURFER toolbox
<b>Roughness</b>	Slope Standard deviation in 1000 m radius		LaFrance et al. 2010
<b>Phi median</b>	Sediment median diameter (on a phi scale ; $\Phi = -\log_2 D_{mm}$ )	$\Phi$	SEABED: Atlantic coast offshore surficial sediment data. US Geological Survey Reid et al. 2005
<b>Clay</b>	Fraction of clay in sediment	%	SEABED: Atlantic coast offshore surficial sediment data. US Geological Survey Reid et al. 2005
<b>SST Spring</b>	Mean Seasonal Sea surface Temperature (Spring)	Degree Celsius	Satellite data NASA Terra and Aqua (MODIS sensors) Codiga et Ullman, 2010
<b>Stratification Spring</b>	Buoyancy frequency squared 0.25 to 2.5 km resolution	$s^{-2}$	FVCOM modeling. Codiga et Ullman, 2010 Chen et al (2006)
<b>SST Fall</b>	Mean Seasonal Sea Surface Temperature (Fall)	Degree Celsius	Satellite data NASA Terra and Aqua (MODIS sensors) Codiga et Ullman, 2010

17 variables

Similar variables to Degraer's Belgium environment typology (2012)

<b>Stratification Fall</b>	<b>Buoyancy frequency squared 0.25 to 2.5 km resolution</b>	$s^{-2}$	<b>FVCOM modeling Codiga et Ullman, 2010 Chen et al (2006)</b>
<b>Aspect Ratio</b>	Slope directionality	Degree [0-360]	NGDC Coastal Relief Model ; Satellite data NASA Terra and Aqua (MODIS sensors) SURFER toolbox
<b>BPI fine scale</b>	Bathymetric position index fine scale [negative values indicate a canyon; positive values indicate a ridge; around 0, flat or constant slope]		NGDC Coastal Relief Model and GIS
<b>BPI large Scale</b>	Bathymetric position index large scale [negative values indicate a through; positive values indicate a ridge; around 0, flat or constant slope]		NGDC Coastal Relief Model and GIS
<b>North-ness</b>	North -South component in slope $\sin(\text{Aspect Ratio})$ positive value indicates North-ness		NGDC Coastal Relief Model and GIS
<b>East-ness</b>	West-East component in slope $\cos(\text{Aspect Ratio})$ positive value indicates East-ness		NGDC Coastal Relief Model and GIS

# Principal Components Analysis



**PC1:** *Offshore-ness/coastal-ness*

**PC2:** *Sedimentology*

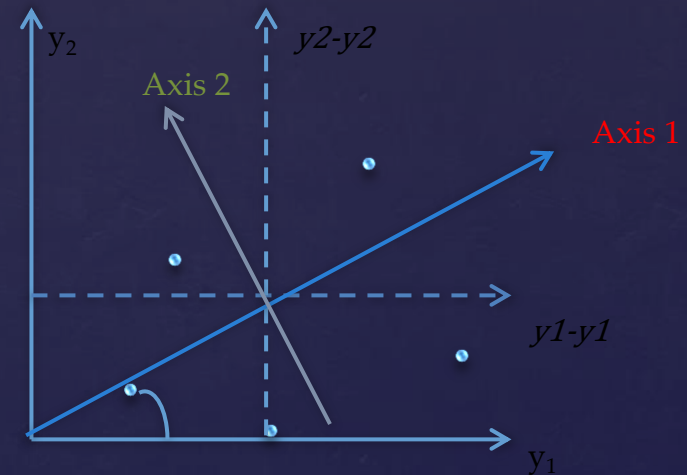
**PC3:** *Fresh Water input*

**PC4:** *Large scale geomorphology*

**PC5:** *Fine scale geomorphology*

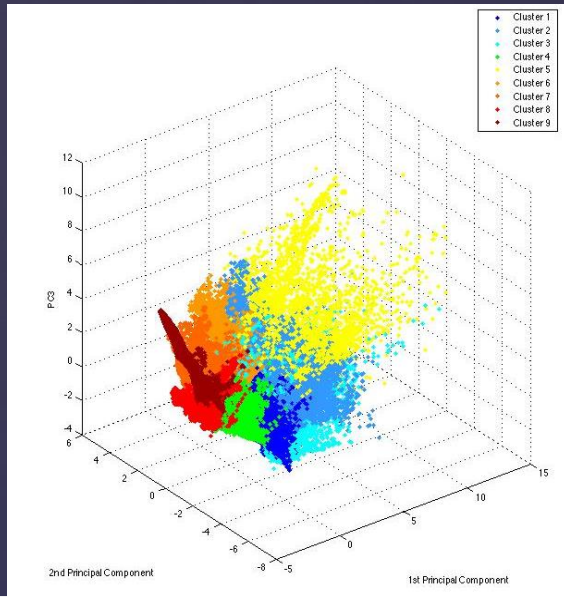
**PC6:** *Upwelling*

- PC analysis is a rotation of the original axes in the directions which explains the maximum variance.
- The new rotated axis or PC are independent. Few explains most of the variance.

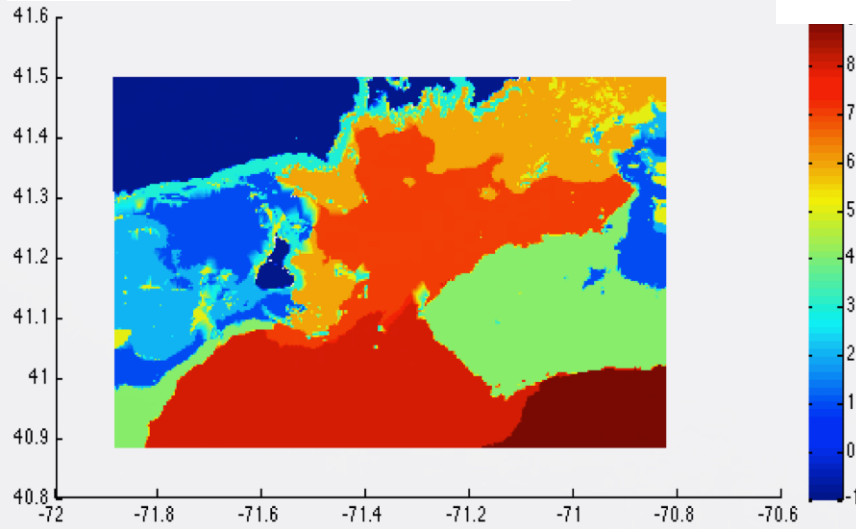




# Typology: Cluster Analysis



Cluster	PC 1 Offshoreness	PC 2 Sedimentology	PC 3 Fresh Water [cold currents]	PC 4 Large Scale Geomorphology	PC 5 Small Scale Geomorphology	PC 6 Upwelling stratification
1	Intermediate	Medium sand	X	Relatively smooth		
2	Intermediate	Fine Sand/clay	XX	Relatively rough		
3	Coastal shallow water					
4	Offshore shallow water	Medium sand		Smooth	Smooth	
5		Fine sand/clay	XX	High roughness	High roughness	
6	Coastal	Fine sand/clay	-- warm			
7	Intermediate		--warm			X
8	Offshore deep water	Medium sand				X
9	Offshore deep water	Fine sand/clay				



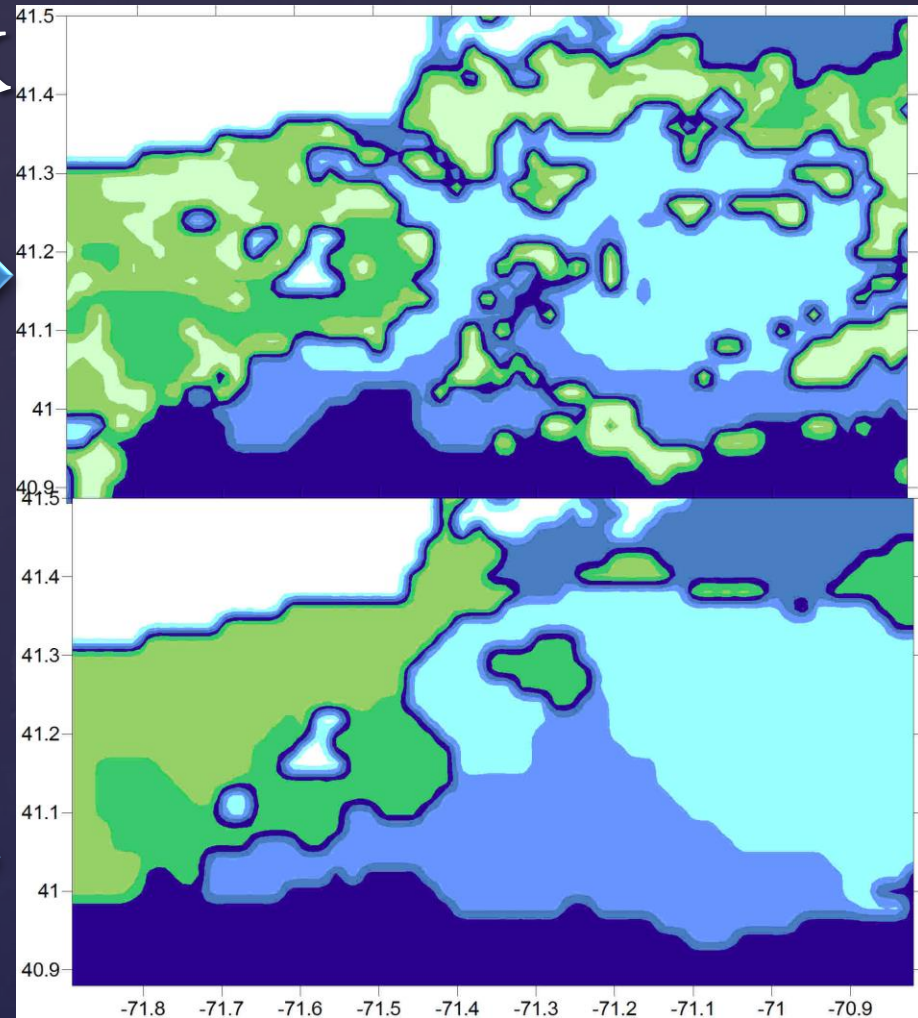
# Pattern Recognition(AI)

## Neural Network

Automatic recognition from 5  
Abiotic variables:

Depth  
Temperature  
Stratification  
Roughness  
Phi median

Target ecological regions



# Support Vector Machine(SVM)

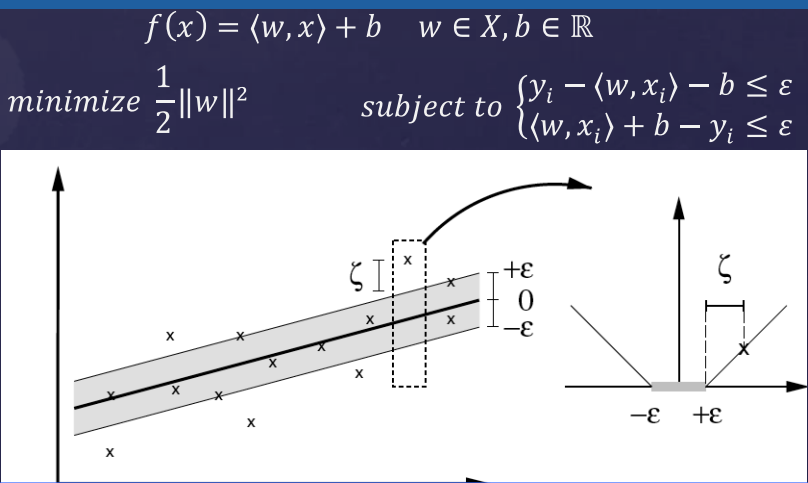
Vapnick, 1995

*Non Linear generalisation of the Generalized Portrait [ Vapnick and Chervonenkis, 1964]*

Support vector machines = dual to maximum margin classifiers (MMC)  
 Dual obtained by applying Lagrangian optimization theory to MMC optimization problem

## SVM Non Linear Regression

- A data universe  $X$
  - A sample set  $S \subset X$
  - Some target function  $f: X \rightarrow \mathbb{R}^n$
  - A training data set  $D$   
 $D = \{(x, y) \mid x \in S \text{ and } y = f(x)\}$
  - Compute a model
- $\hat{f}: X \rightarrow \mathbb{R}$  using  $D$  such that  $\hat{f}(x) \cong f(x)$   
 for all  $x \in X$



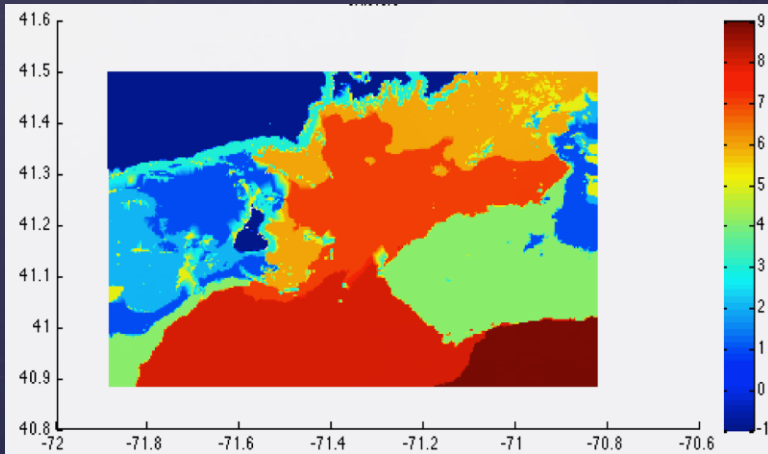
- Objectives**
1. Automatic Pattern recognition
  2. "Novelty" detection (outliers)

# Pattern Recognition

## Support Vector Machine (AI)

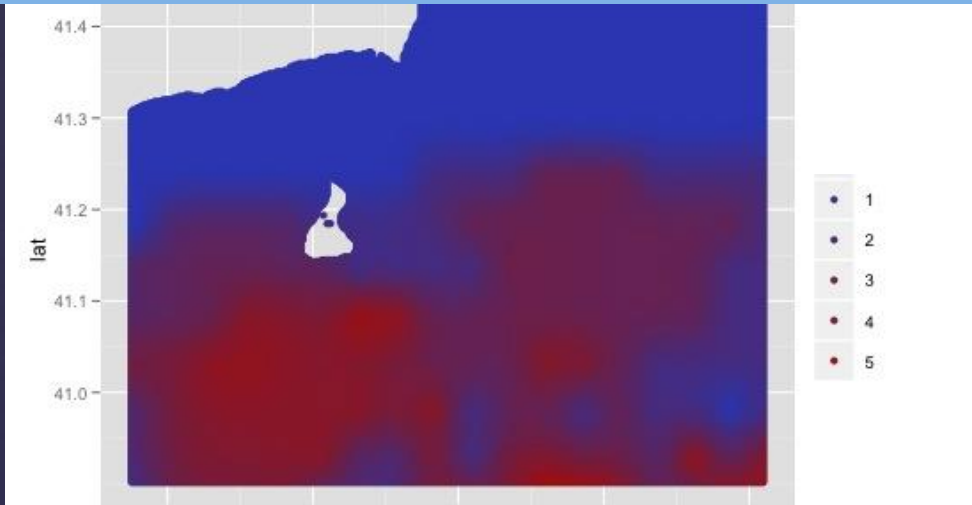
Use of non-linear regression to predict mammals distribution from abiotic variables

$R^2 = 88\%$  5-fold

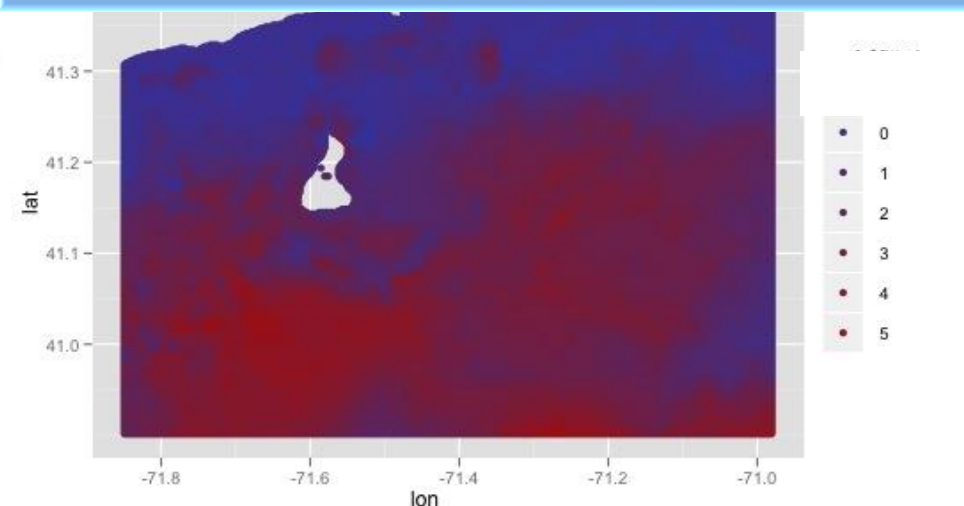


Abiotic Variables: depth, surface temperature, stratification, bottom Roughness, % clay, phimedian

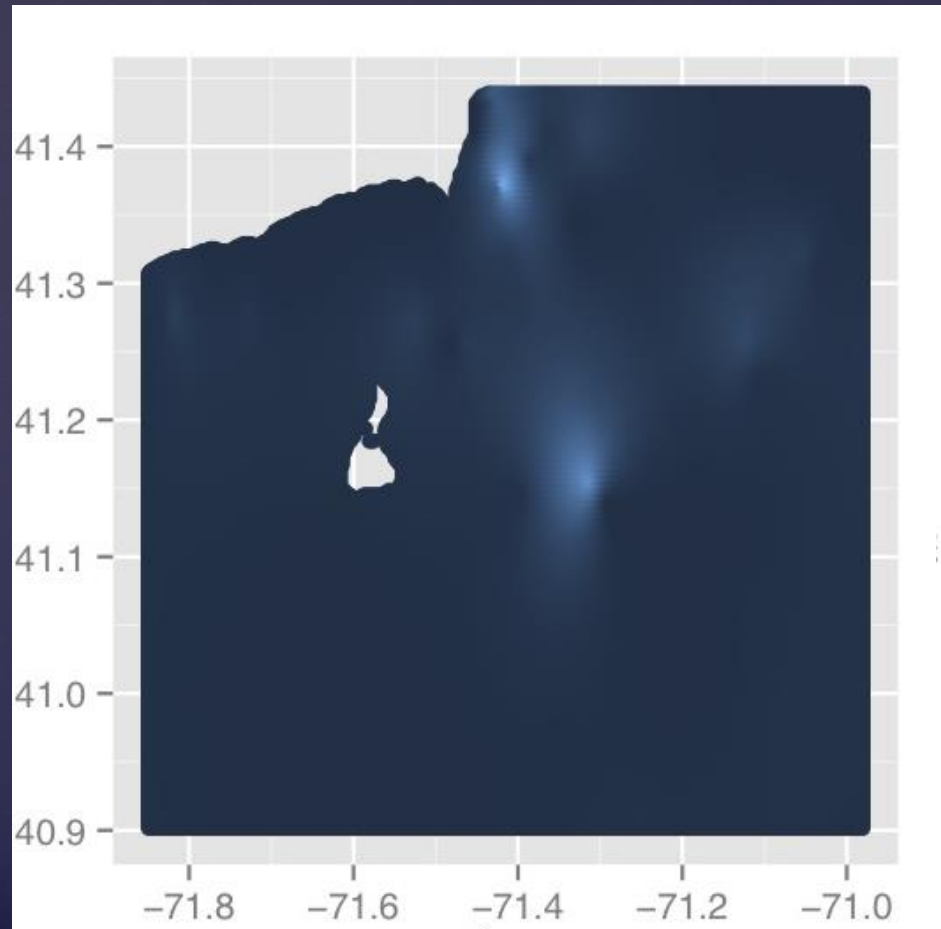
## Mammals Abundance Spring Observations



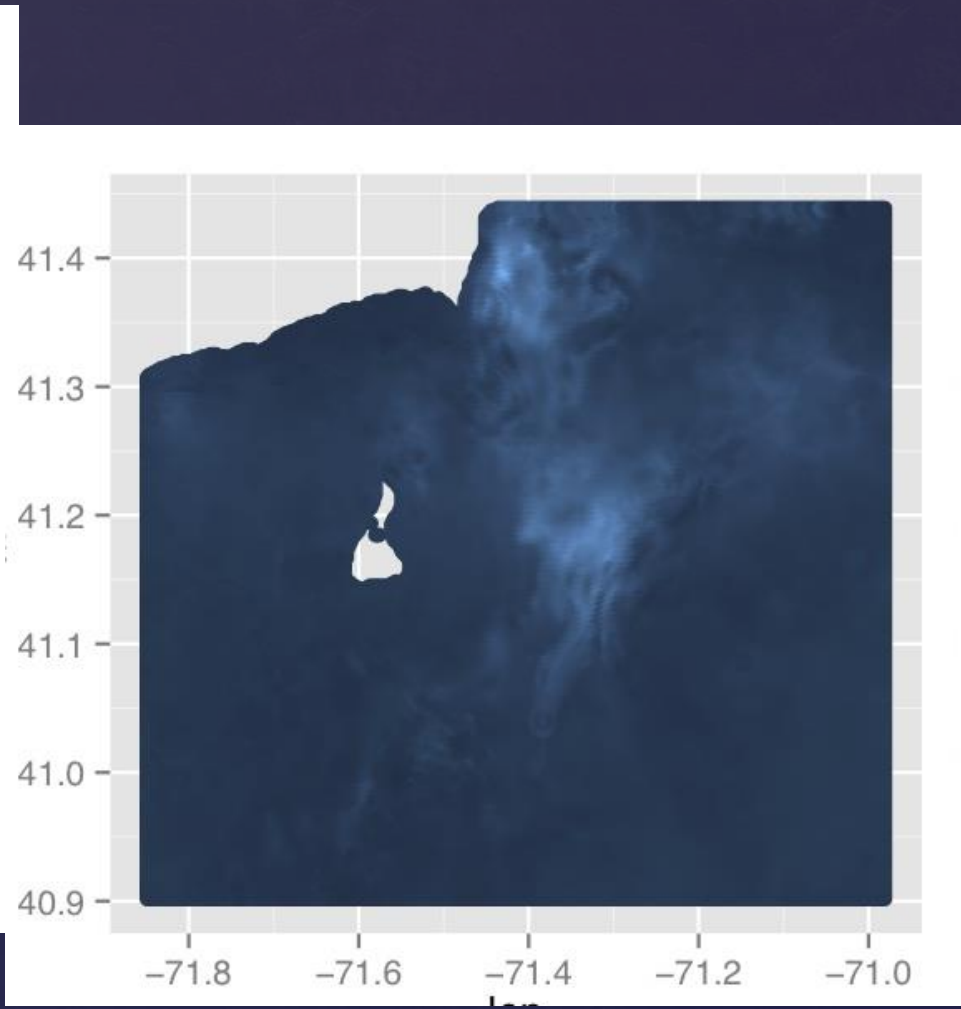
## Mammals Abundance Spring Predictions



# Homarus Americanus Spring

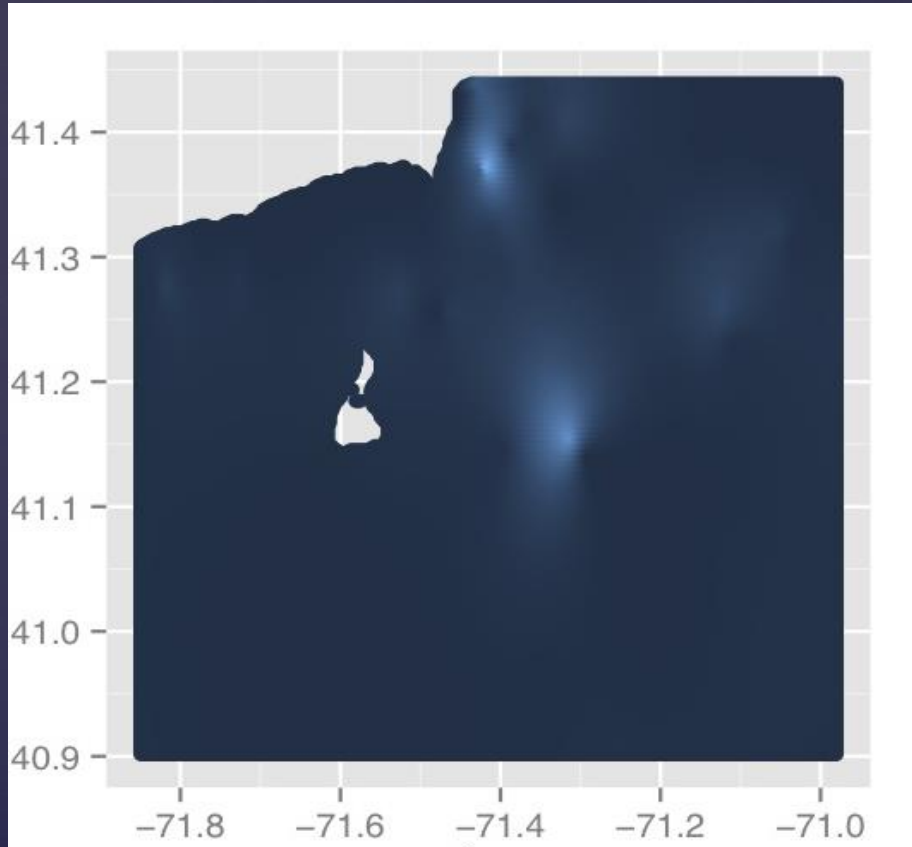


Observations

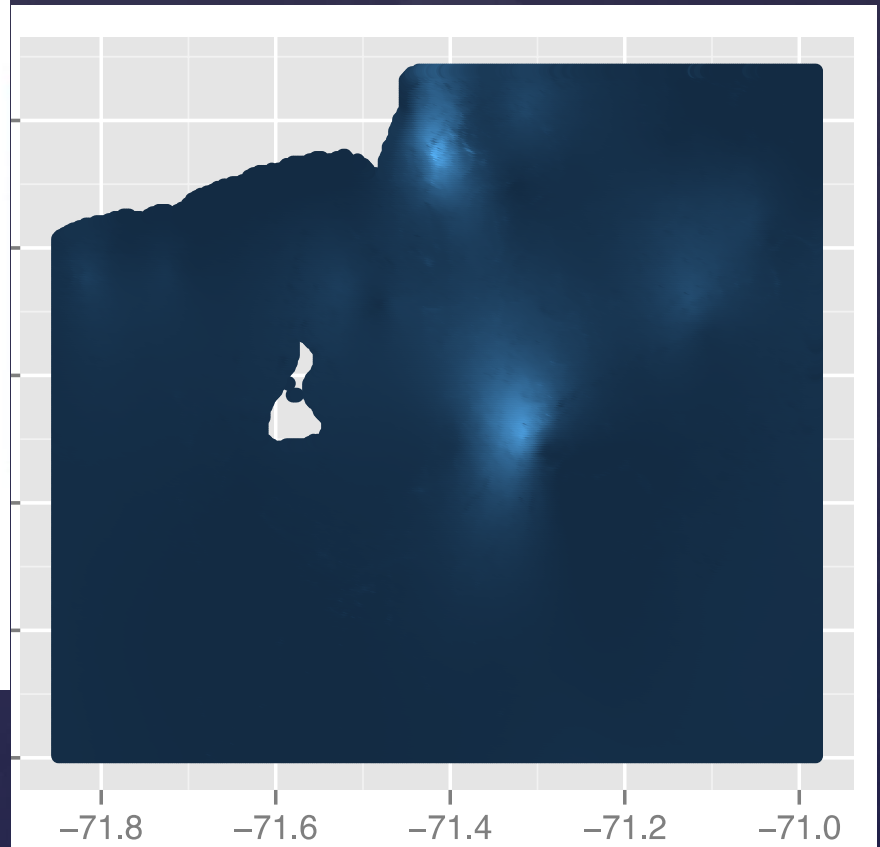


Predictions SVM 3-Fold

# Homarus Americanus Spring



Observations



Predictions Random Forest



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Photograph by Andrew Wong, 2008 International Photography Contest

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