

# The distribution of diadromous fish at sea and the role of Marine Protected Areas (MPAs)

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Pôle OFB-INRA-Agrocampus Ouest-UPPA  
pour la gestion des migrateurs amphihalins  
dans leur environnement



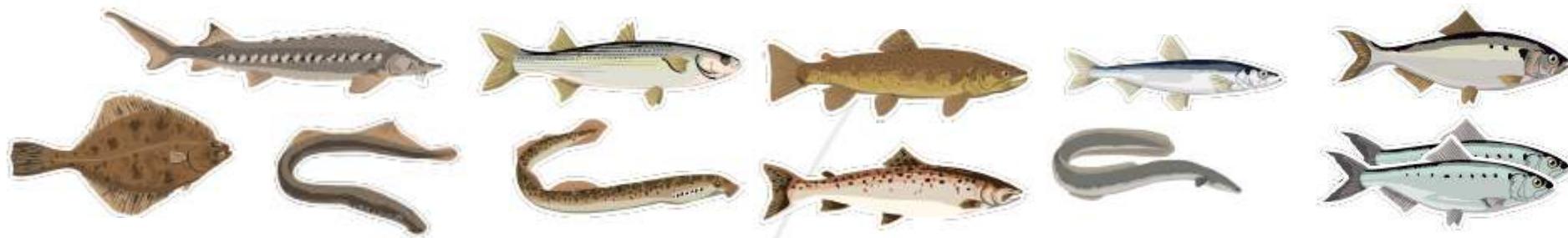
**INRAE**

**l'institut Agro**  
agriculture • alimentation • environnement



# Objectives

1. Understand the distribution of diadromous fish at sea.
  2. Evaluate the value of MPAs for diadromous fish.
  3. Evaluate their sensitivity to bycatch.
  4. Connect their freshwater and marine habitats.
- To respond to the Habitat Directive & Marine Strategy Framework Directive requirements.

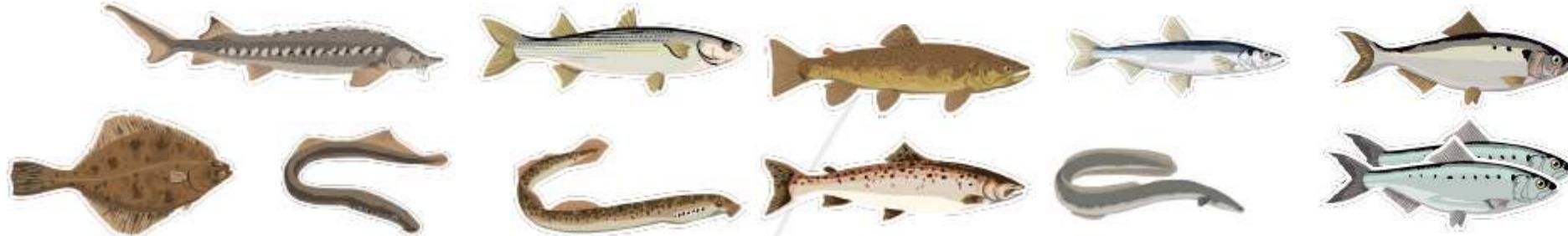


# Diadromous fish studied



Fr – 2019  
EU - 2011

Latin name	Common name	EU IUCN	Fr IUCN
<i>Acipenser sturio</i>	Atlantic sturgeon*	CR	CR
<i>Alosa alosa</i>	Allis shad*	LC	CR
<i>Alosa fallax</i>	Twait shad*	LC	NT
<i>Alosa agone</i>	Mediterranean twaite shad*	LC	NT
<i>Anguilla anguilla</i>	European eel*	CR	CR
<i>Lampetra fluviatilis</i>	River lamprey*	LC	VU
<i>Petromyzon marinus</i>	Sea lamprey*	LC	EN
<i>Chelon ramada</i> *	Thinlip mullet	LC	LC
<i>Osmerus eperlanus</i>	Smelt	LC	NT
<i>Platichthys flesus</i>	European flounder	LC	DD
<i>Salmo salar</i>	Atlantic salmon*	VU	NT
<i>Salmo trutta</i>	Sea trout	LC	LC



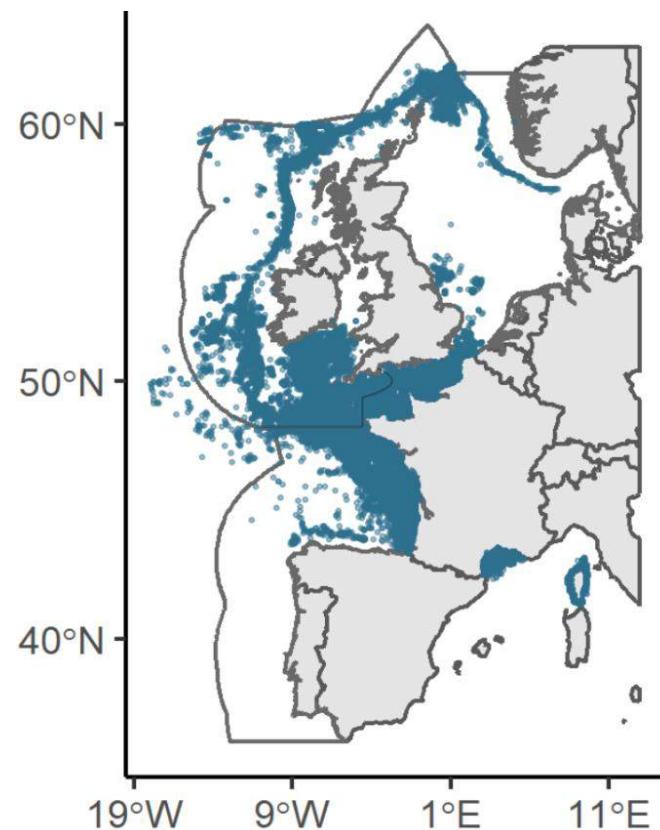
# Surveys

42 surveys, 1965-2019, 168 904 hauls

## Fisheries dependent data

### OBSMER

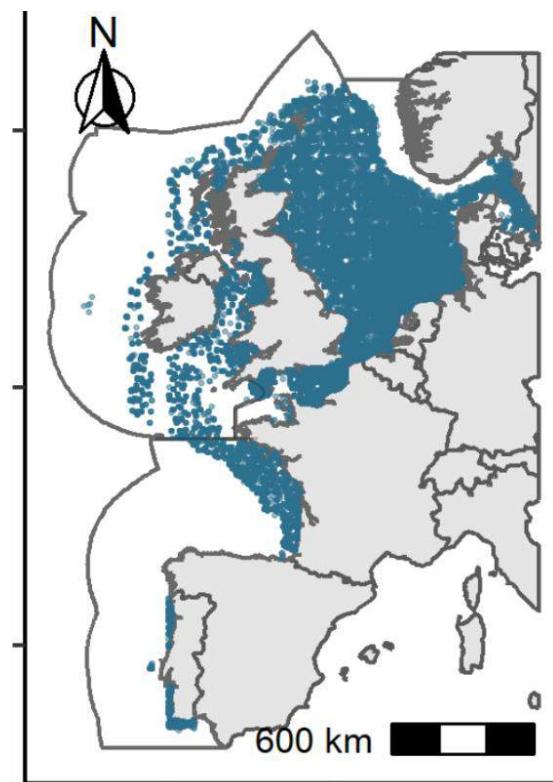
2003 – 2019, 100 617 hauls



## Scientific surveys

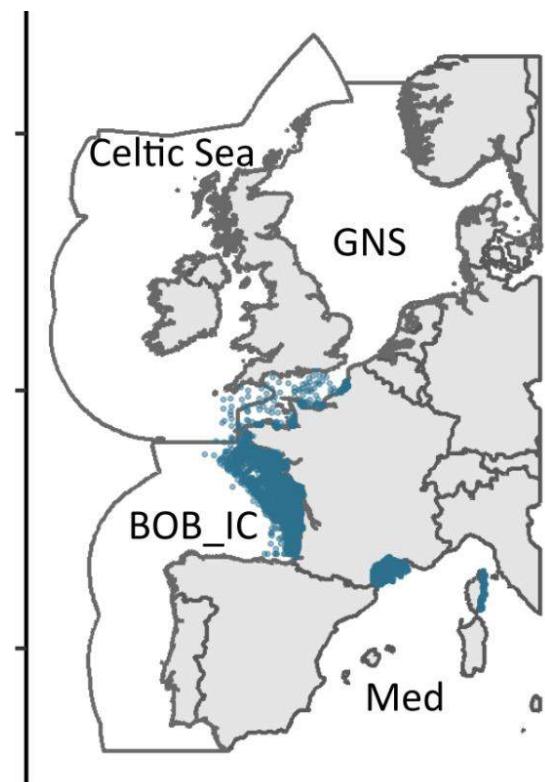
### ICES DATRAS

1965 - 2018, 54 865 hauls



### IFREMER

1980 - 2018, 13 422 hauls

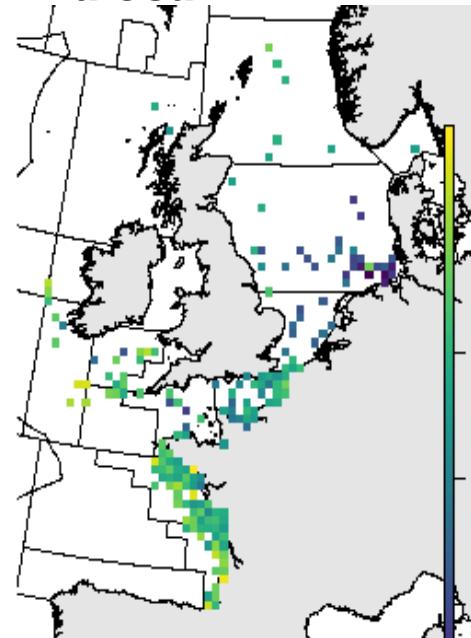


# Data cleaning

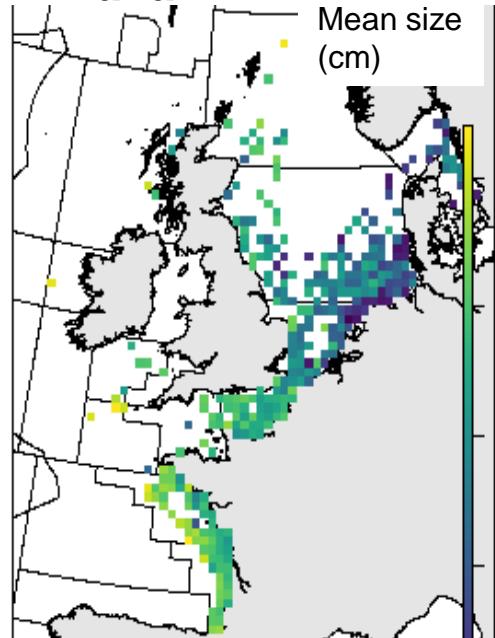
- Merging datasets.
- Exploring outliers and NAs (cm/mm, points on land, etc.).
- Data analysis (length, month, distribution, gear effect) & comparison with literature.

e.g.

*A. alosa*

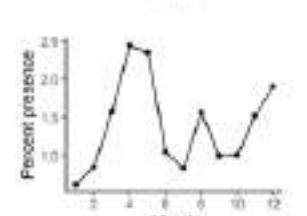
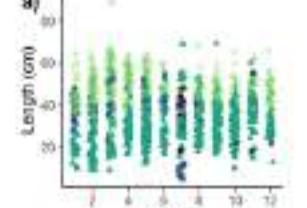


*A. fallax*

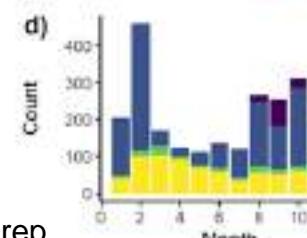


Mean size  
(cm)

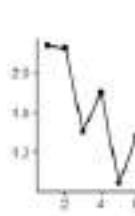
*A. alosa*



Benthic  
Demersal



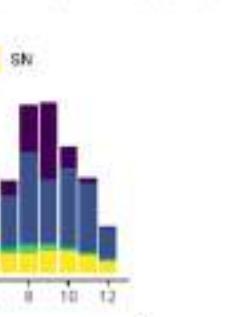
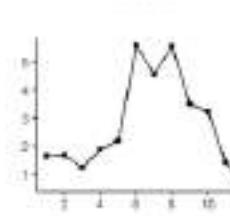
*A. fallax*



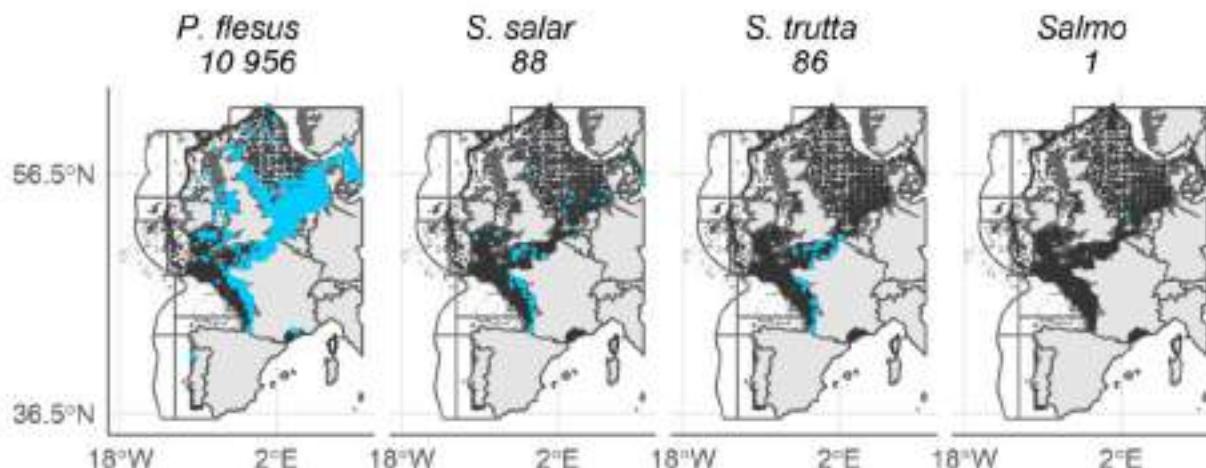
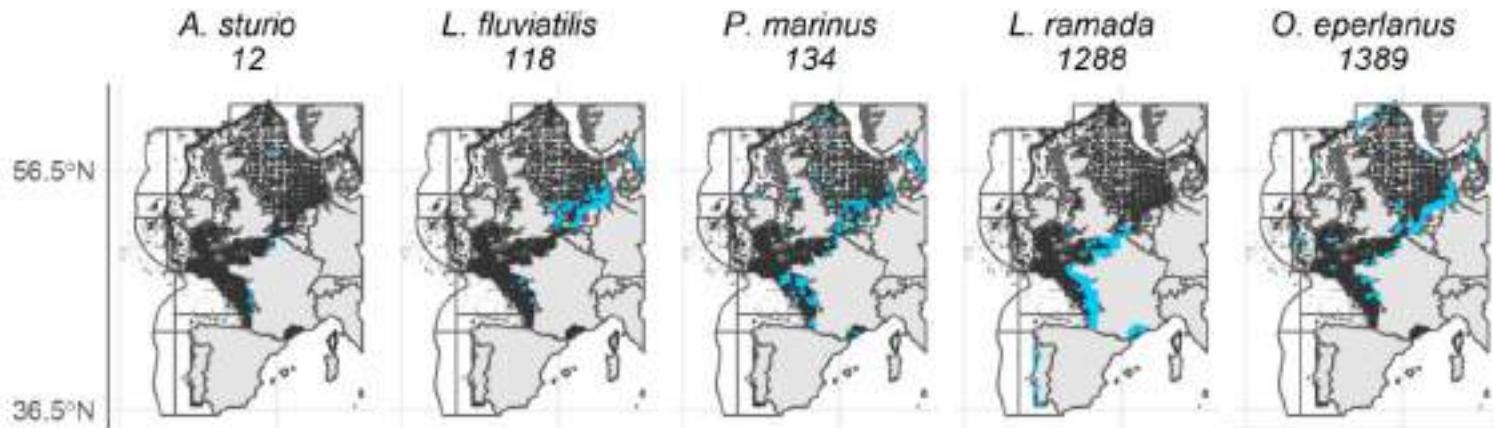
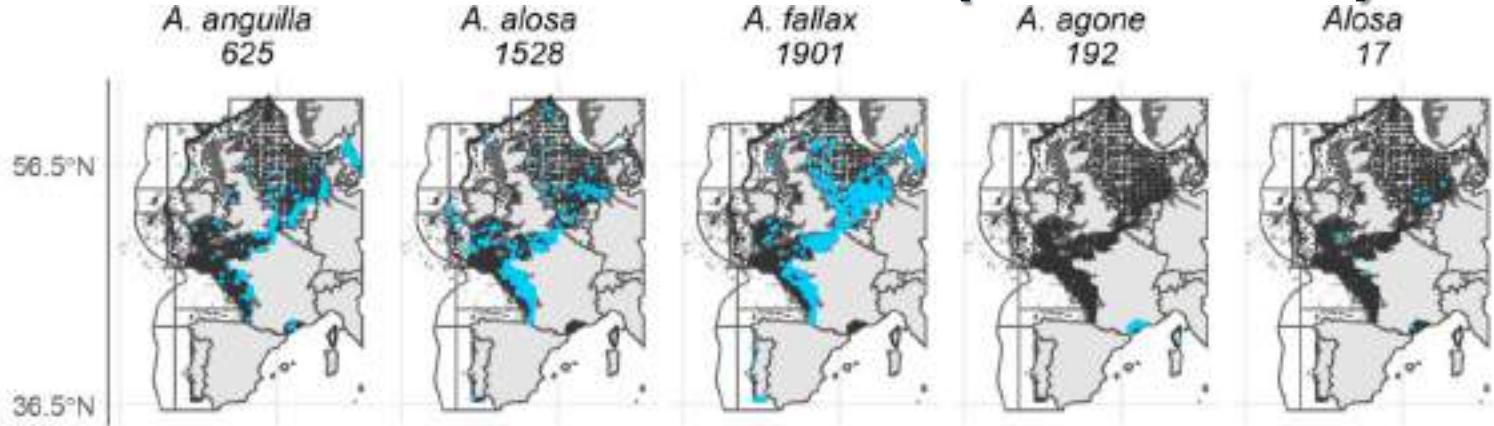
Pelagic Seine



*A. agone*



# Presence of diadromous fish (1965-2019)



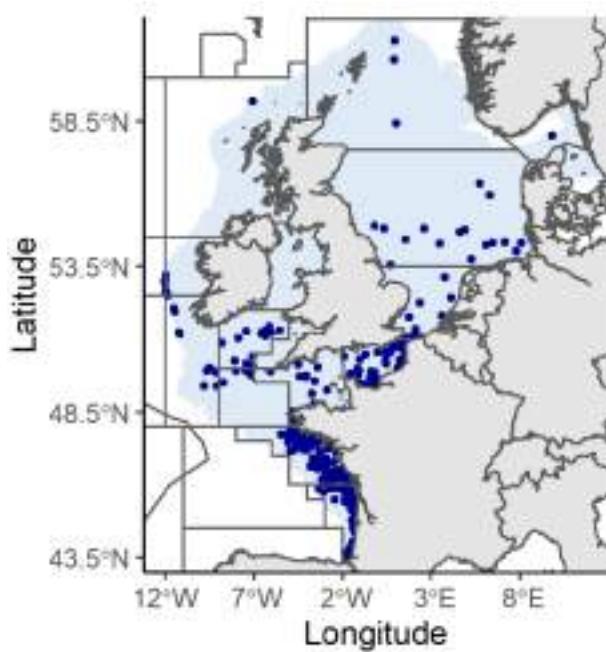
**Potentiel  
Miss-identification**



# 1) Aim

- i. Model the present distribution of diadromous fish.
- ii. Taking into account imperfect detection, gear bias & spatial autocorrelation.

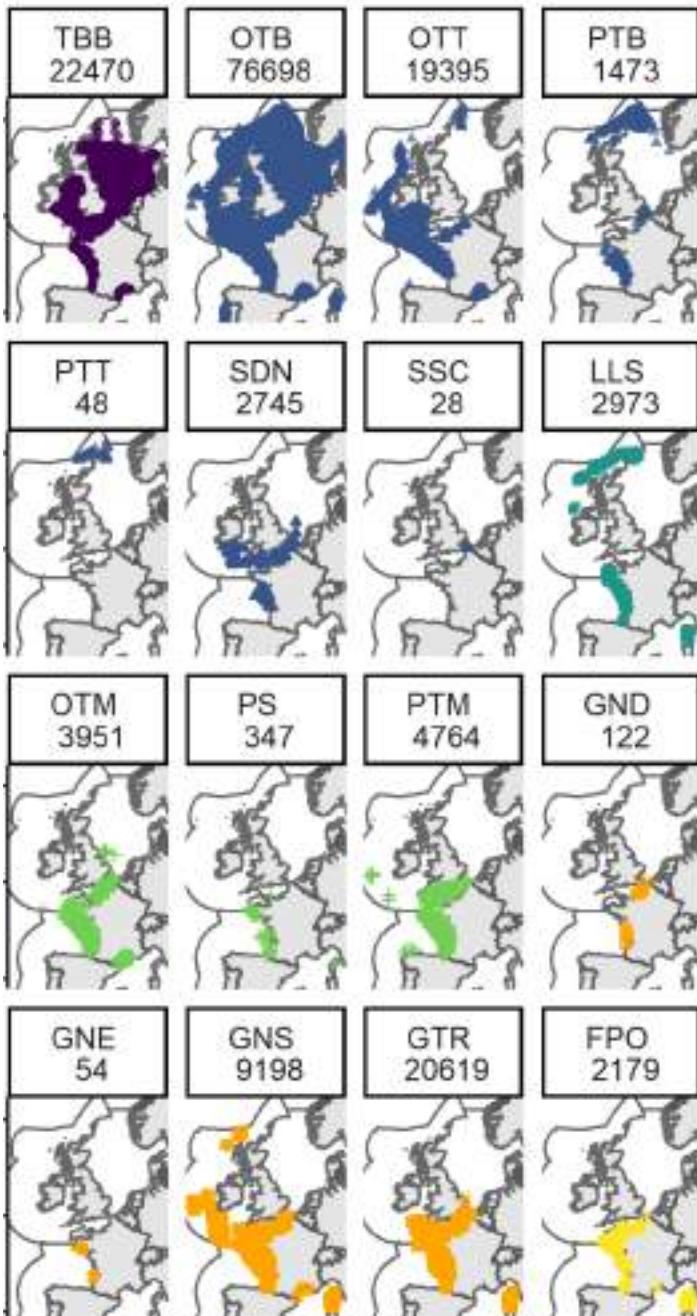
*A. alosa*



*A. fallax*

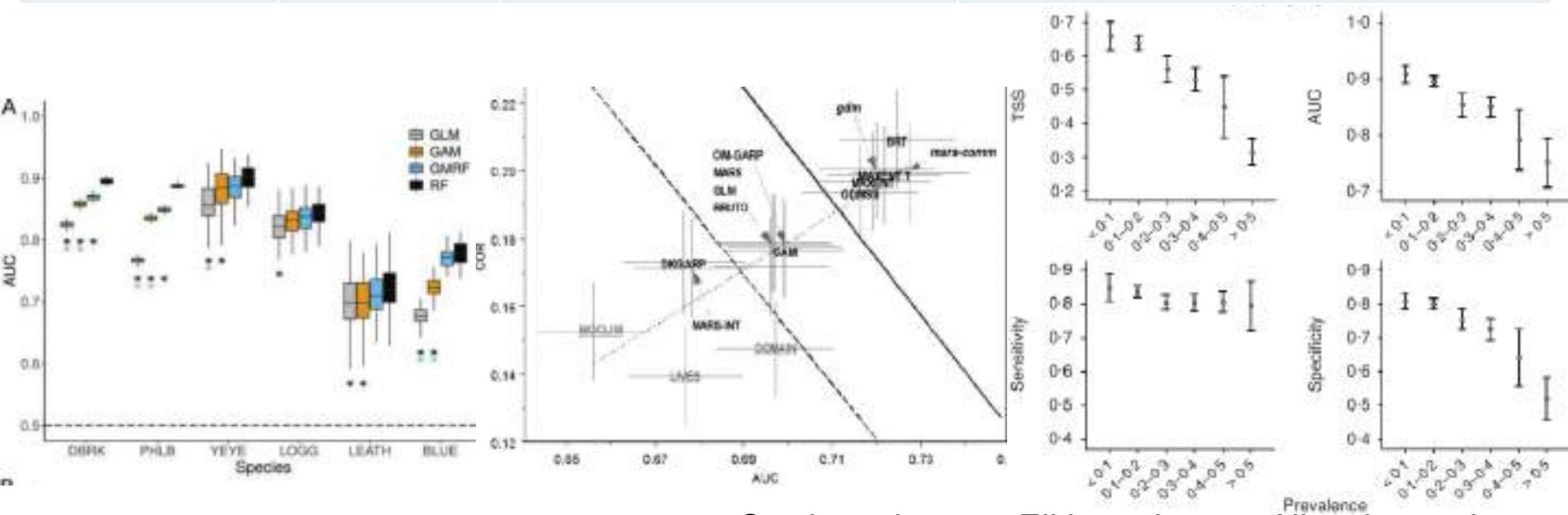


█ Benthic mobile  
█ Demersal mobile  
█ Line  
█ Pelagic mobile  
█ Seine net



# Classic distribution models

Model type	Examples	+	-
Multiple regressions	GLM, GAM, MARS	<ul style="list-style-type: none"> <li>Existing packages to compare</li> </ul>	<ul style="list-style-type: none"> <li>Imperfect detection not considered</li> </ul>
Regression trees	BRT, RF, CTA	<ul style="list-style-type: none"> <li>Commonly used</li> <li>Computational intensity = lower</li> </ul>	<ul style="list-style-type: none"> <li>Zero imbalance</li> <li>Doesn't consider sampling bias of different fear types</li> </ul>
Discriminate analysis	ANN, FDA		
Presence only	MaxEnt, BIOCLIM, DOMAIN	<ul style="list-style-type: none"> <li>Pseudo absences</li> <li>Reduce zero inflation</li> </ul>	<ul style="list-style-type: none"> <li>Already have true zeros</li> <li>Pseudo absences can cause problems</li> </ul>



Stock et al, 2020; Elith et al 2006; Allouche et al 2006

# Hierarchical SDM (Bayesian)

## Site occupancy intrinsic conditional autoregressive model (SO iCAR)

# Hierarchical SDM (Bayesian)

## Site occupancy intrinsic conditional autoregressive model (SO iCAR)

### Habitat suitability - favourable

$z_i$  = variable describing presence/absence (PA) at site  $i$

$\theta_i$  = probability of presence – habitat suitability

$X_i$  = environmental covariates

$\beta$  = how much the environmental variable contribute to the suitability process

$P_j$  = spatial random effect in cell  $j$  at observation  $i$  (iCAR)



$$z_i \sim \text{Bernoulli}(\theta_i)$$
$$\text{logit}(\theta_i) = X_i \beta + P_{j(i)}$$

### iCAR – probability of presence depends on that of the nearest site

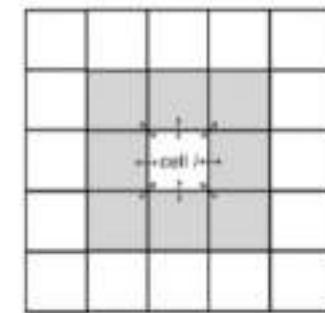
$p_j$  = spatial random effect in cell  $j$

$\mu_j$  = mean  $p$  in the neighbourhood

$V_p$  = variance of the spatial random effect

$P_j$  = spatial random variable

$n_j$  = number of neighbours for cell  $j$



$$P_j \sim \text{Normal}(\mu_j, \frac{V_p}{n_j})$$

### Observational process

$y_i$  = PA at site  $i$

$z_i$  = habitat suitability for site  $i$  + if several visits in 1 cell

$\delta_i$  = probability of detecting species at site  $i$

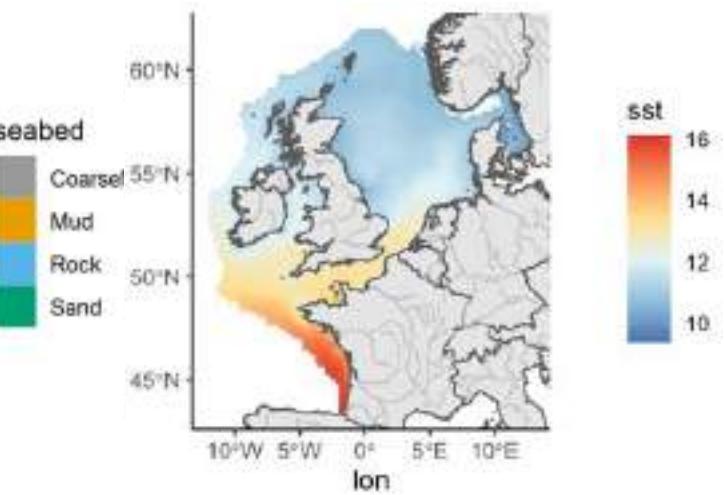
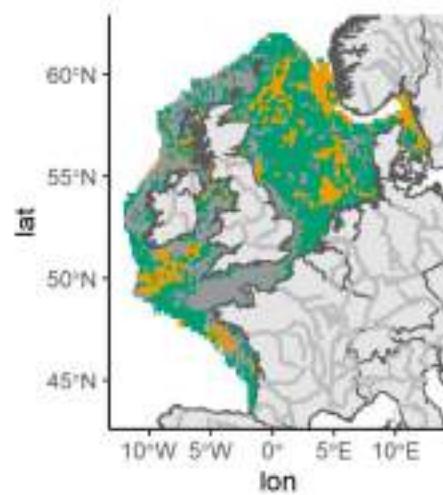
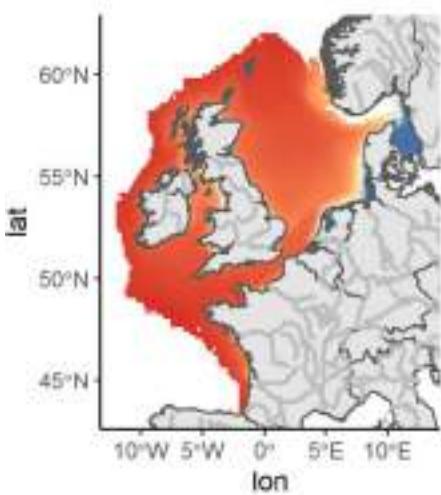
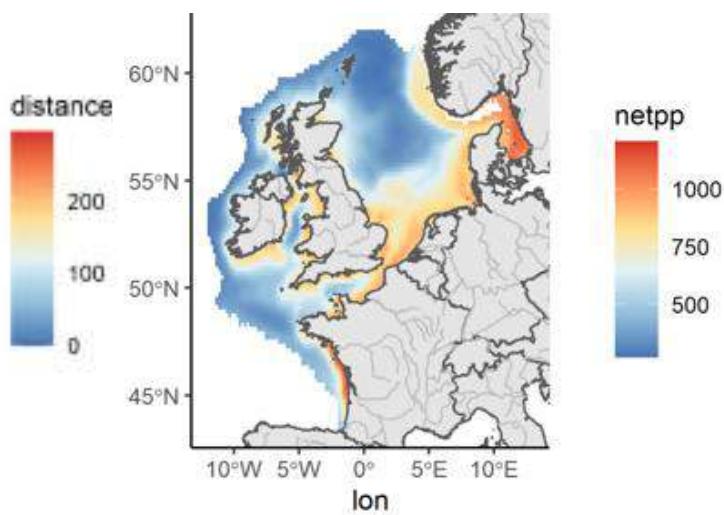
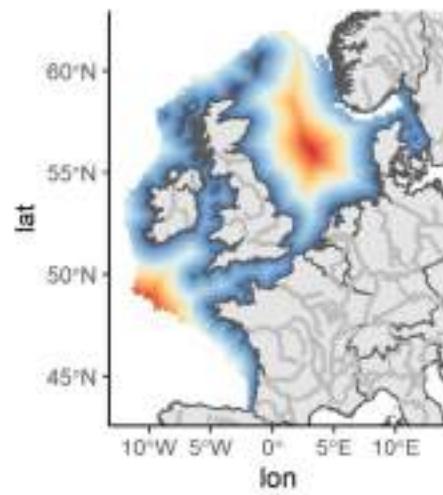
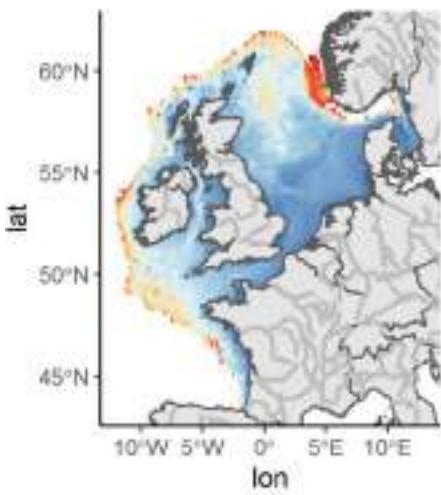
$W_i$  = observability covariate (gear)

$\gamma$  = how much the observational variable contribute to the detectability

$$y_i \sim \text{Bernoulli}(z_i \delta_i)$$
$$\text{logit}(\delta_i) = W_i \gamma$$



# Environmental variables



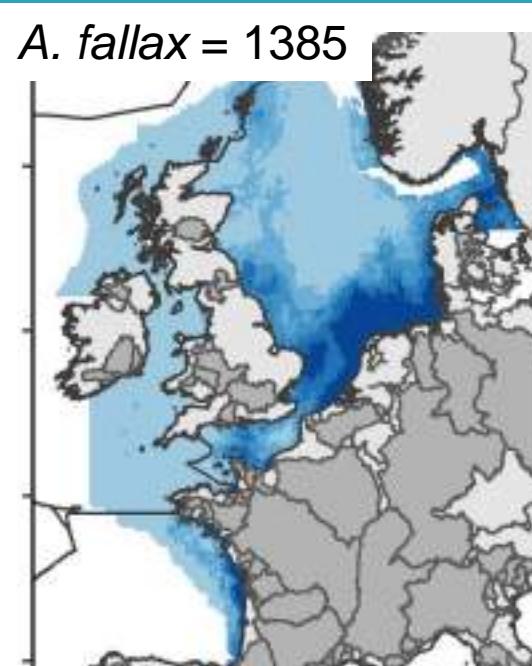
# Shad

*A. alosa* = 802



Sensitivity: 0.70  
Specificity: 0.96  
PCC: 0.94

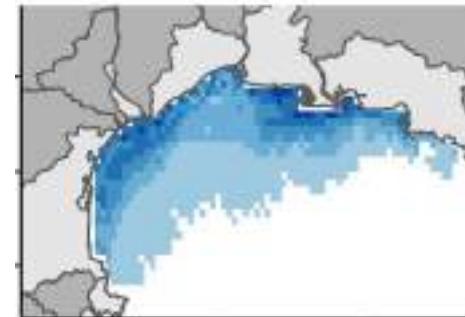
*A. fallax* = 1385



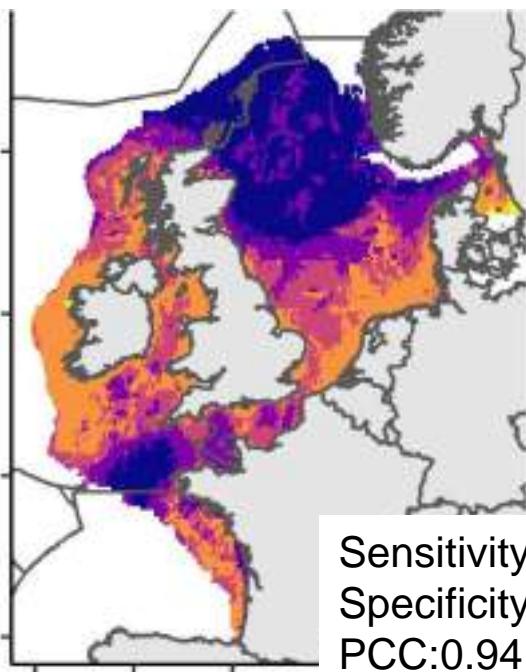
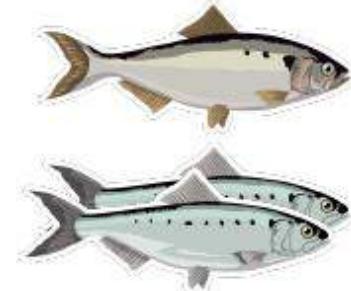
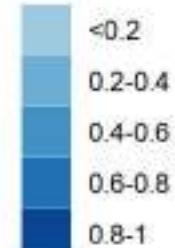
Sensitivity: 0.81  
Specificity: 0.85  
PCC: 0.85

EuroDiad v. 4.0

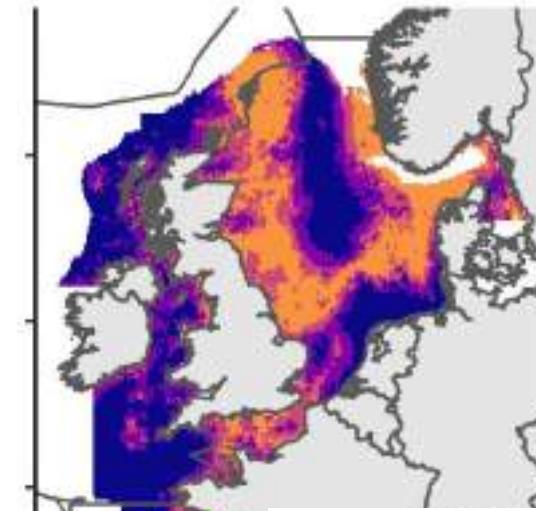
*A. agone* = 176



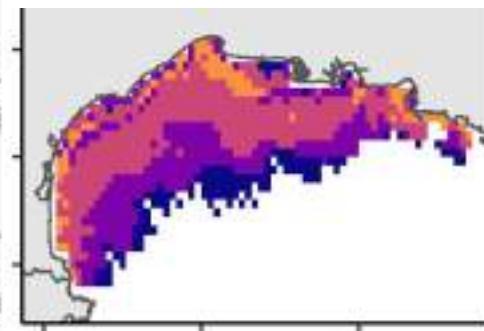
Probability of presence



Sensitivity: 0.70  
Specificity: 0.96  
PCC: 0.94



Sensitivity: 0.81  
Specificity: 0.85  
PCC: 0.85



Sensitivity: 0.72  
Specificity: 0.82  
PCC: 0.81

# European eel & Lampreys

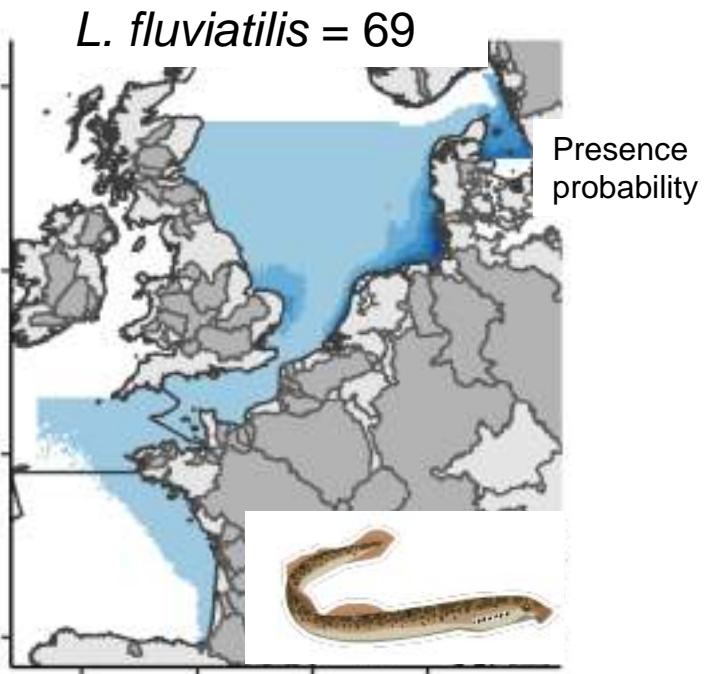
A. anguilla = 176



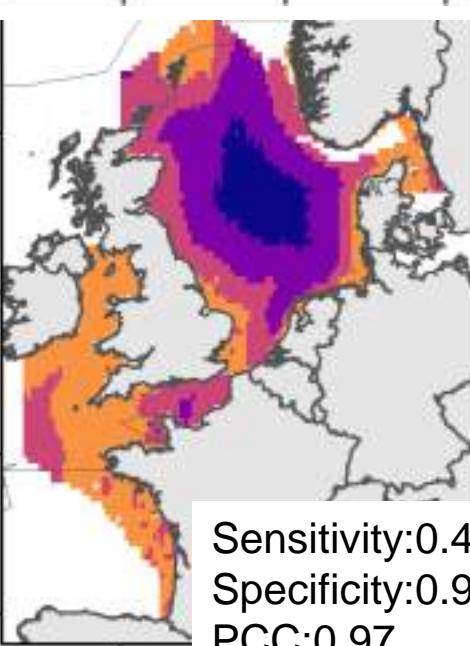
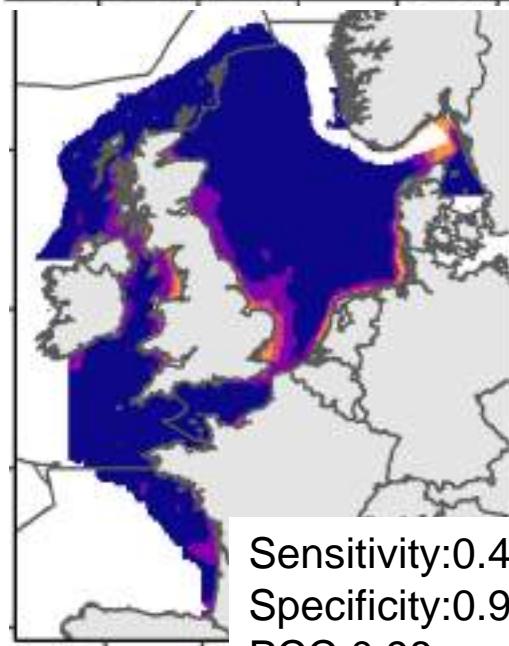
P. marinus = 74



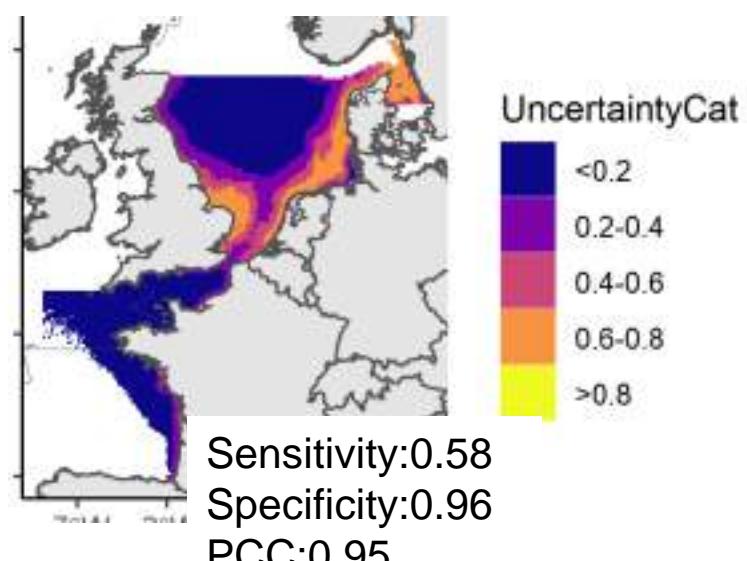
L. fluviatilis = 69



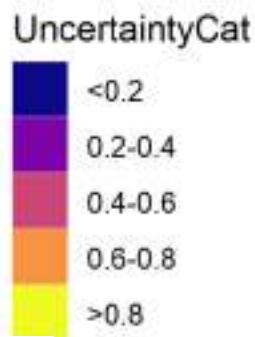
Sensitivity:0.48  
Specificity:0.99  
PCC:0.98



Sensitivity:0.40  
Specificity:0.99  
PCC:0.97

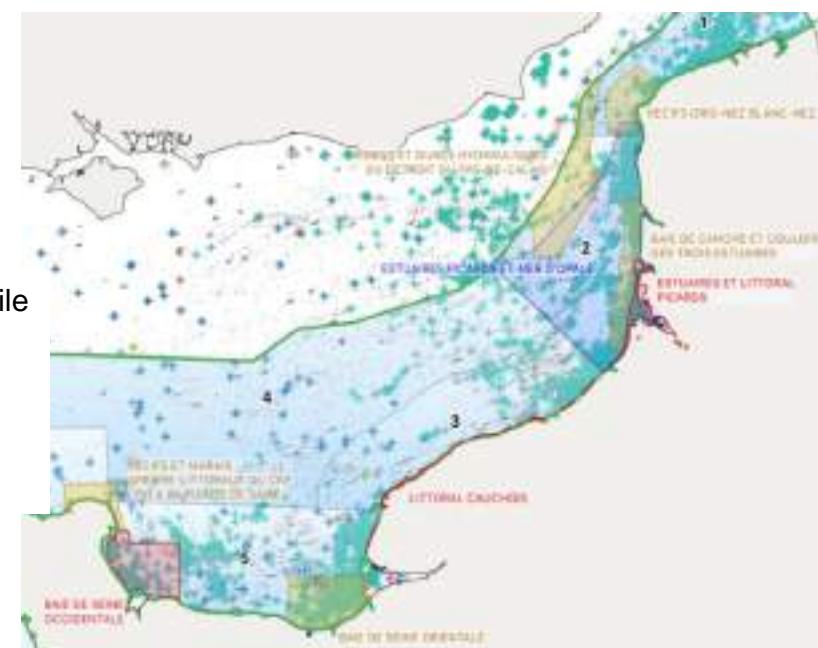
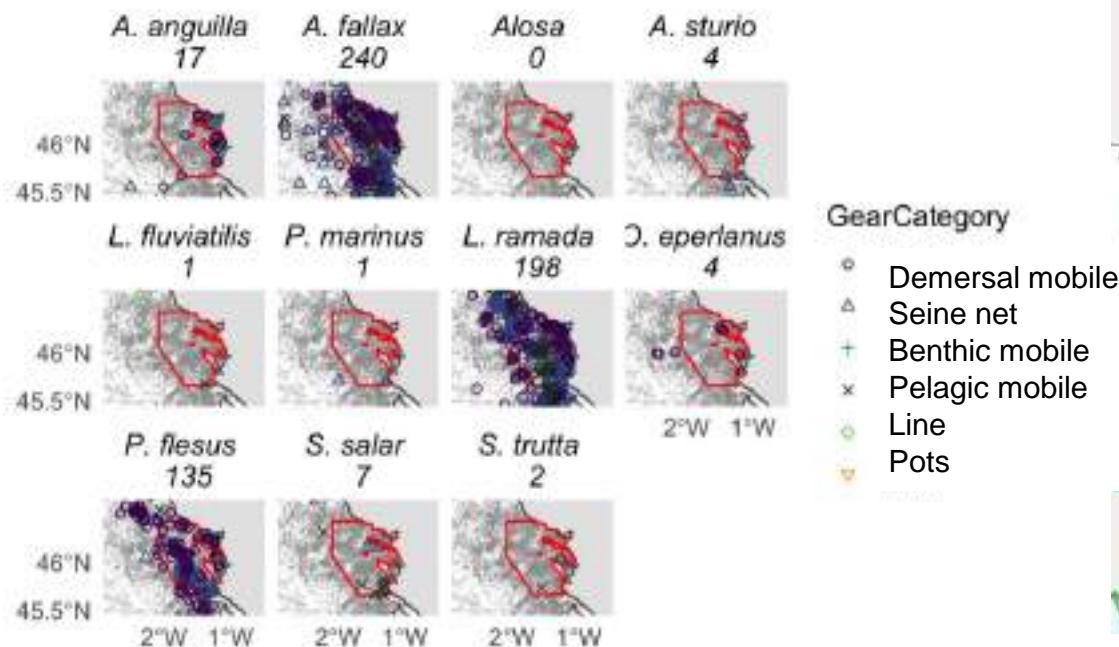


Sensitivity:0.58  
Specificity:0.96  
PCC:0.95



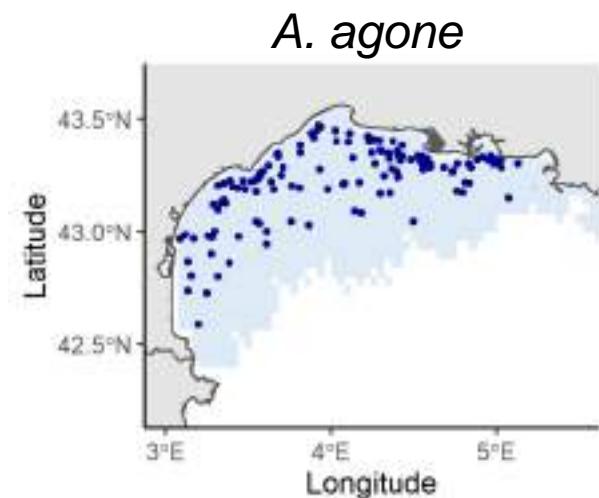
## 2) Aim

- Model the present distribution of diadromous fish at a finer resolution to evaluate the pertinence of MPAs.
- Provide information to help with the management of MPAs.



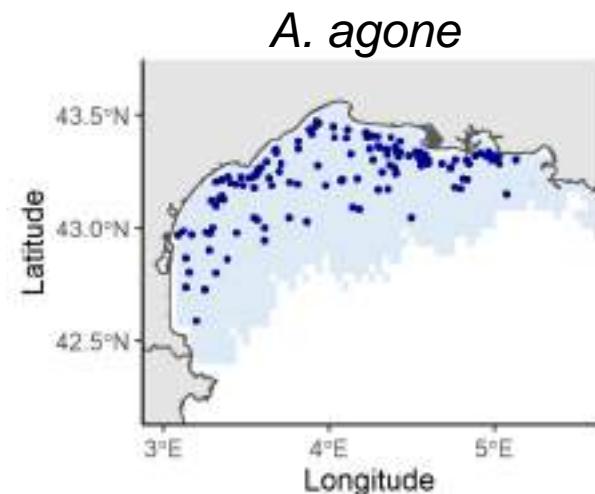
# Combined Model for Accurate Predictions (CMAP)

1. Gridded Binomial (BN) iCAR
2. Gridded Zero Inflated Binomial (ZIB) iCAR
3. Site Occupancy (SO) iCAR

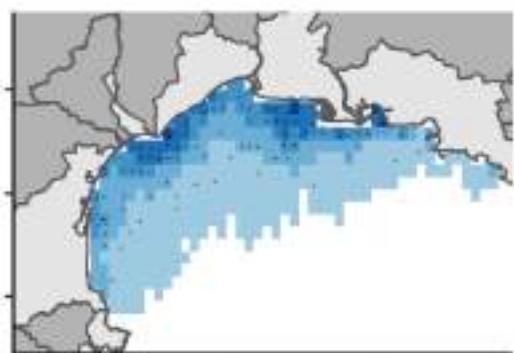


# Combined Model for Accurate Predictions (CMAP)

1. Gridded binomial (BN) iCAR
2. Gridded zero inflated binomial (ZIB) iCAR
3. Site occupancy (SO) iCAR



Gridded BN iCAR



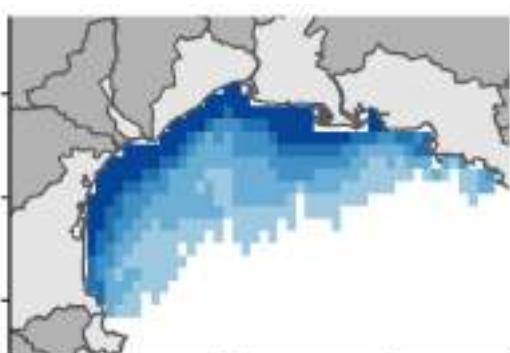
PPV:0.72  
NPV:0.89

Gridded ZIB iCAR

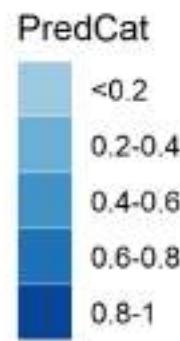


PPV:0.27  
NPV:0.99

SO iCAR

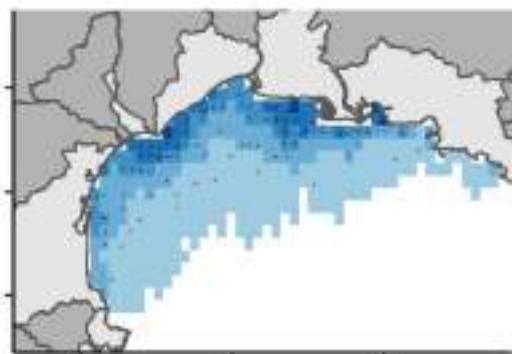


PPV:0.45  
NPV:0.94

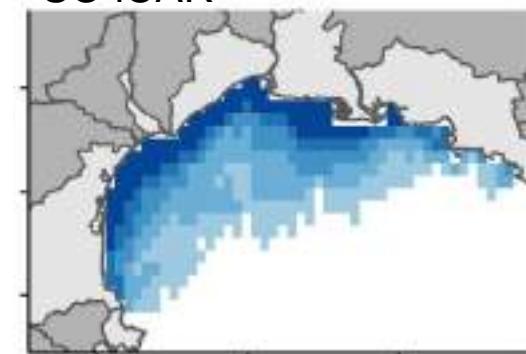


# Combined Model for Accurate Predictions

BN iCAR



SO iCAR



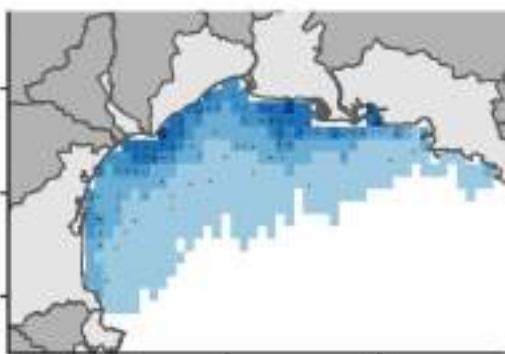
Separate  
objectives

Core  
habitats

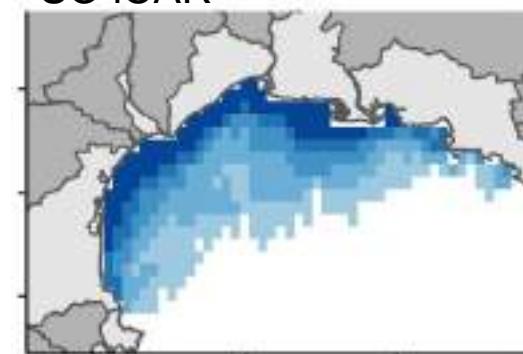
'Unsuitable'  
habitats

# Combined Model for Accurate Predictions

BN iCAR



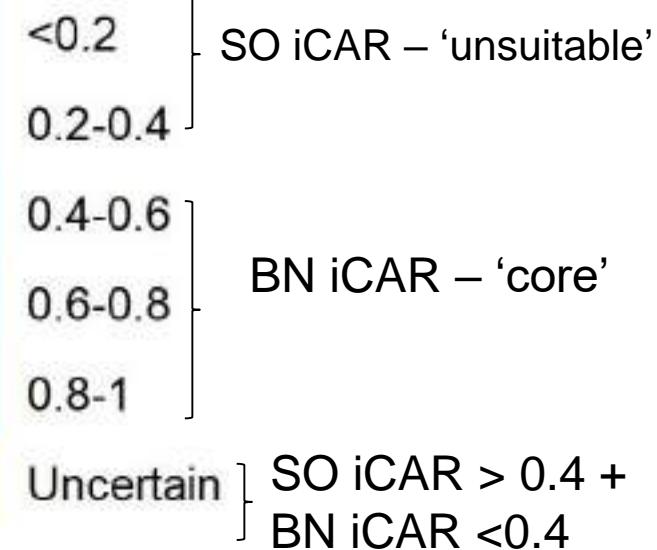
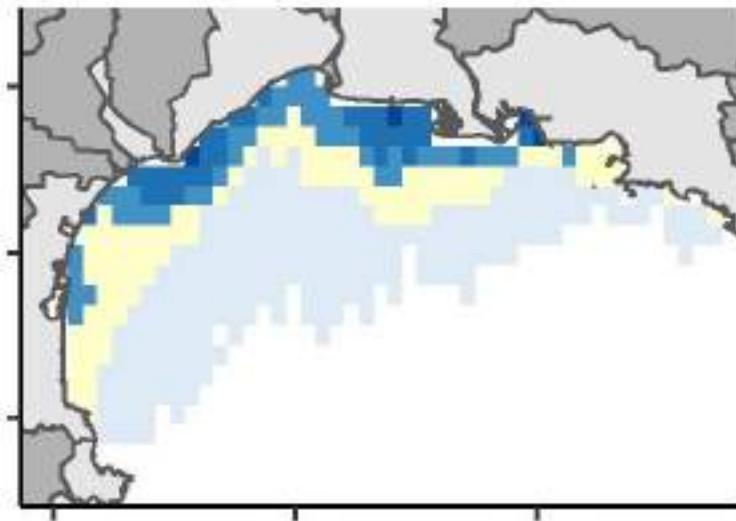
SO iCAR



Separate objectives

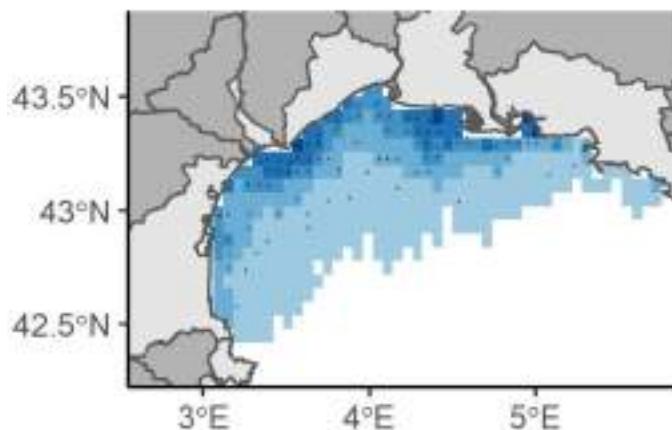
Core habitats

'Unsuitable' habitats

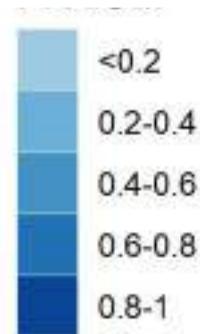
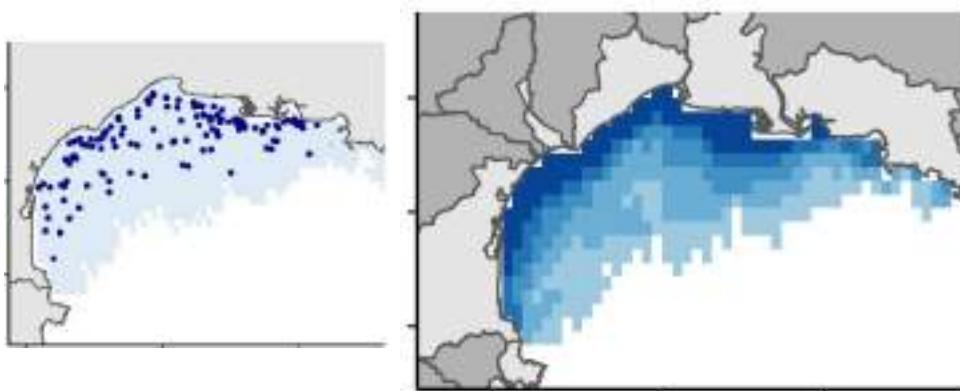


# CMap for Marine Protected Areas (MPA) management

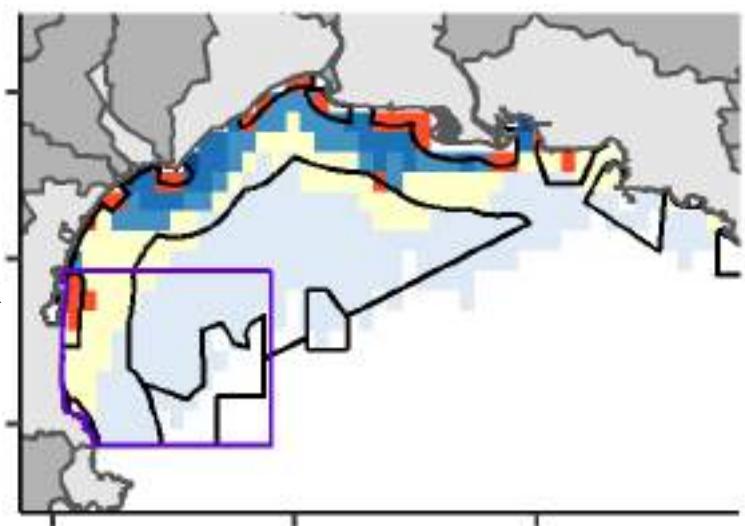
BN iCAR



SO iCAR



— Natura 2000 SCI  
— National Marine Park



<0.2  
0.2-0.4  
0.4-0.6  
0.6-0.8  
0.8-1 } SO iCAR

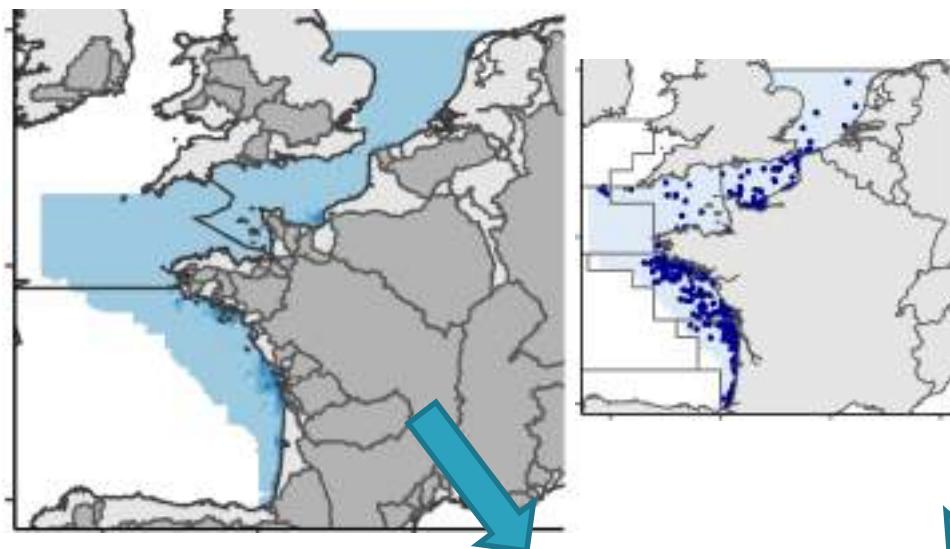
<0.2  
0.2-0.4  
0.4-0.6  
0.6-0.8  
0.8-1 } BN iCAR

<0.2  
0.2-0.4  
0.4-0.6  
0.6-0.8  
0.8-1 } SO iCAR > 0.4 + BN iCAR < 0.4  
Uncertain }

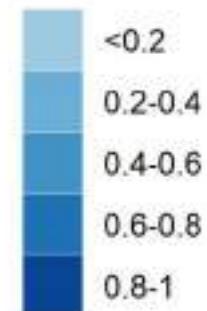
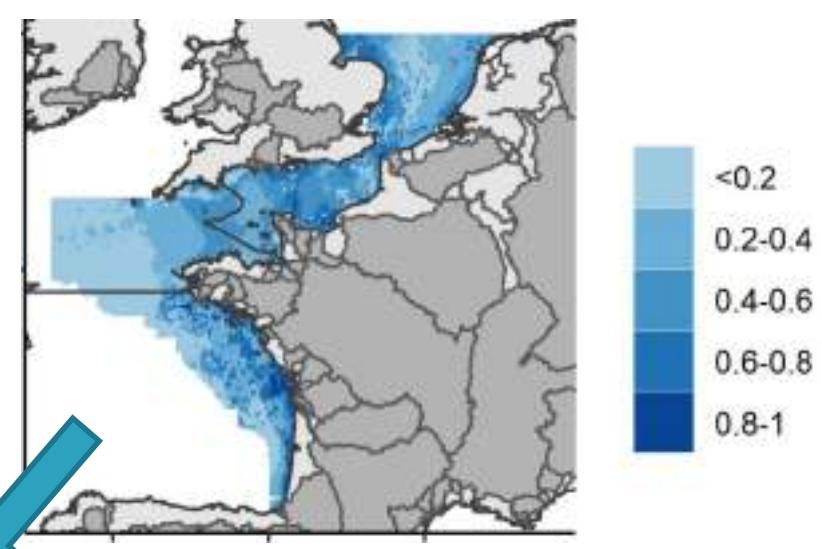
● Core areas within an MPA

# *A. alosa*

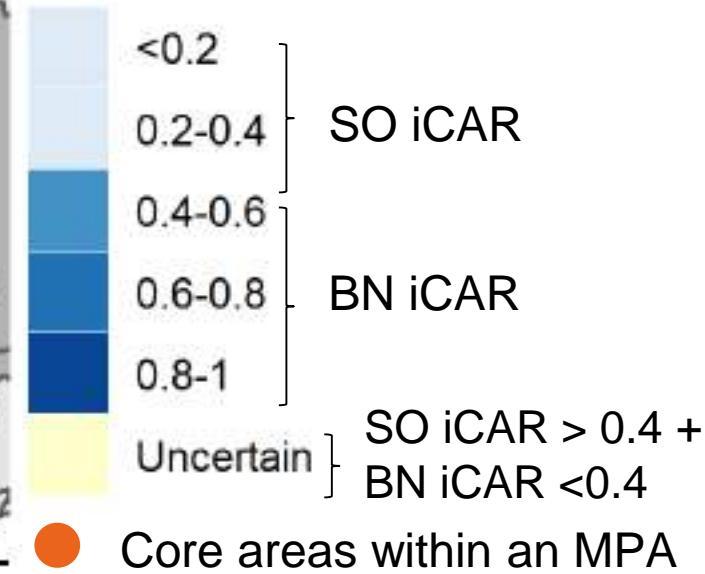
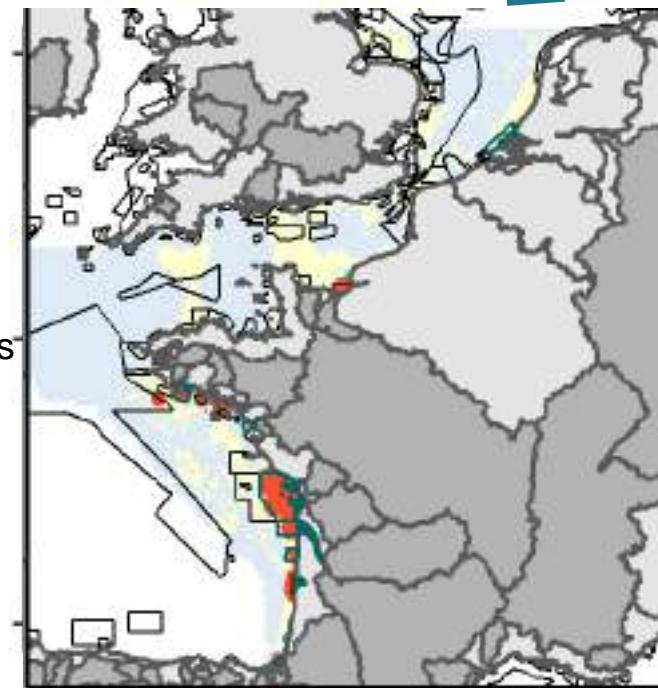
**BN iCAR**



**SO iCAR**

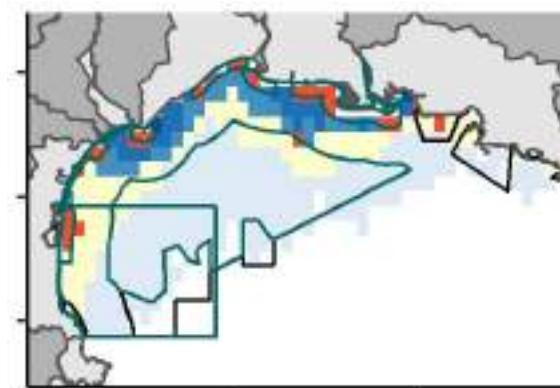


— With management measures  
— Without management measures

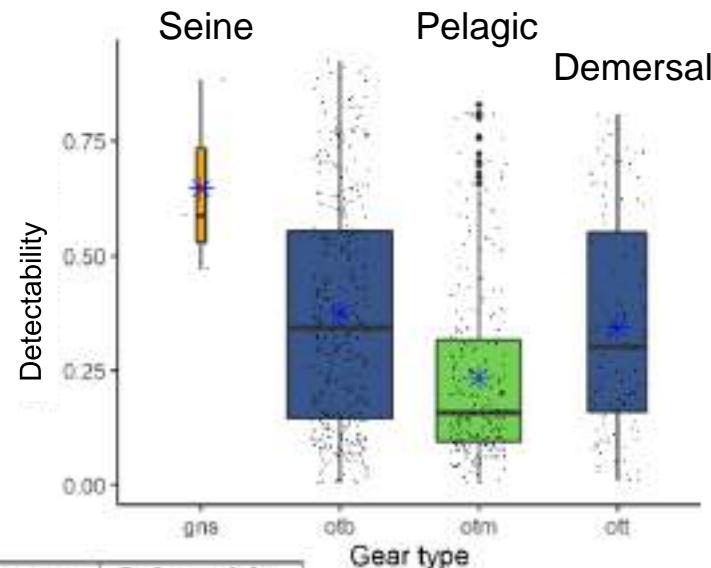


# Value of MPAs for the protection of diadromous fish

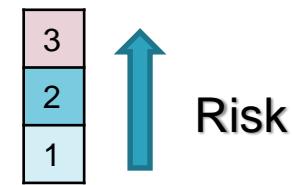
- 1) Are their MPAs which don't protect diadromous fish and which may be of benefit?
- 2) Do the present management measures protect them sufficiently?
- 3) What factors affect the present of diadromous fish within MPAs?



With management measures  
Without management measures



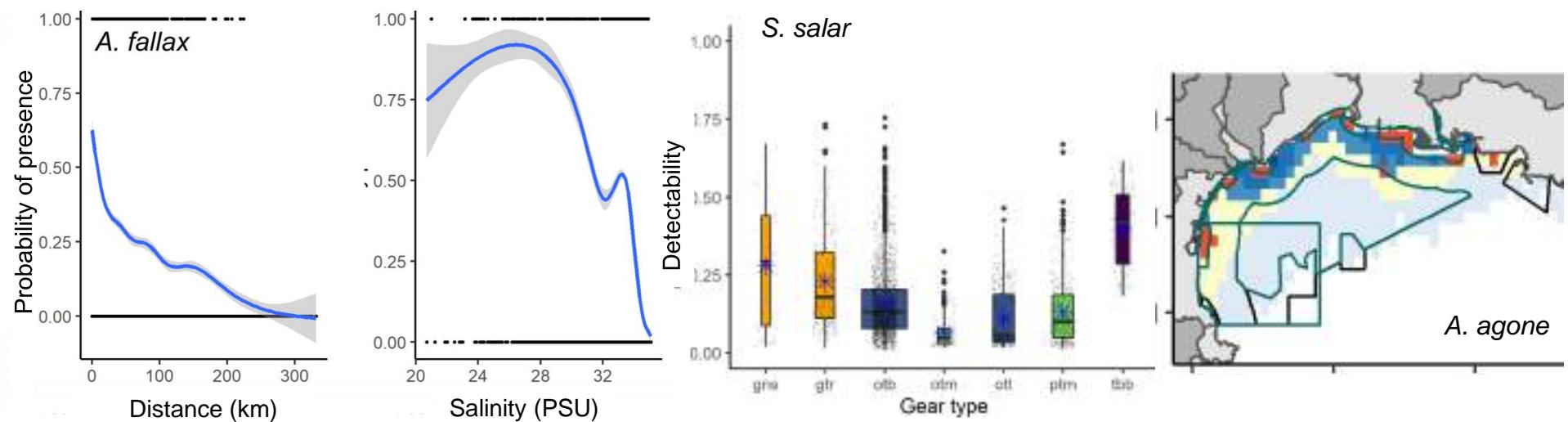
Gear category	Standard abbreviations	Shad	Eel	Lamprey	Sturgeon	Salmonids
Purse Seine	PS	1	1	1	1	1
Seines	SPR, SSC	1	3	1	2	1
Trawls	OTB, OTT, PTB, PTM, TBB	2	3	2	3	2
Dredge	DRB	1	1	1	1	1



# Conclusion

- All species have a coastal distribution.
- All species caught by a variety of gear types.
- Relatively little bycatch BUT even a small amount may have a large impact on the population...
- Limiting access to certain gear types with higher probability of capture may help protect diadromous fish.
- Need for targeted data collection on diadromous fish to better understand their marine life history stages.

e.g.



# Outputs

1. Elliott et al, 2021. Shedding light on the river and sea lamprey within western European waters. Endangered species research. DOI:10.3354/esr01113.
2. Elliott et al, In review. Modelling the distribution of low occurring diadromous fish at sea from fisheries dependent and independent data. Biological Conservation.
3. Elliott et al, In review. Data paper: Fisheries dependent and independent data used to model the distribution of diadromous fish. Biological Conservation.
4. Elliott et al, In prep. Could MPAs do more for diadromous fish at sea?
5. Dambrine et al, In prep. Connecting diadromous fish freshwater and marine habitats to assess climate change vulnerability.



# Thanks! Question?

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Etienne Rivot et Sophie Elliott - Agrocampus Ouest  
Elodie Réveillac - La Rochelle Université

Laurent Beaulaton - OFB

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Pôle AFB-INRA-Agrocampus Ouest-UPPA  
pour la gestion des migrants amphihalins  
dans leur environnement



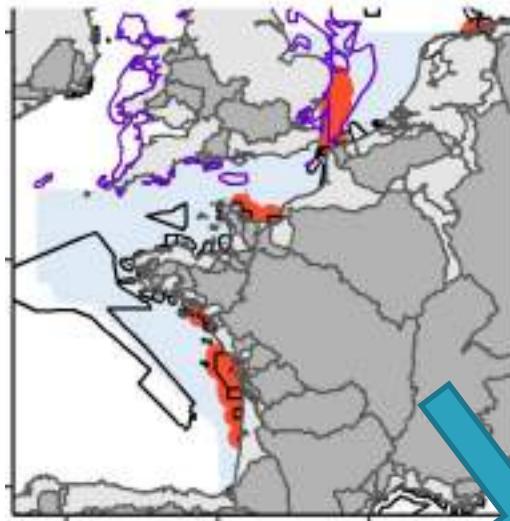
INRAe

l'institut Agro  
agriculture • alimentation • environnement

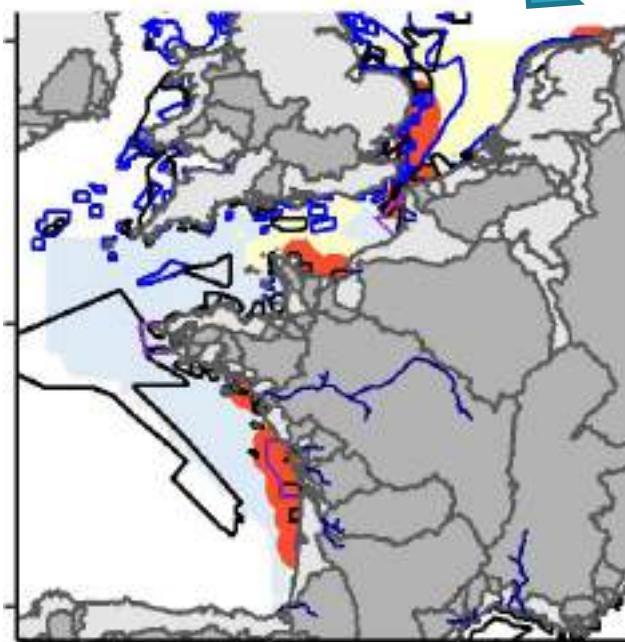
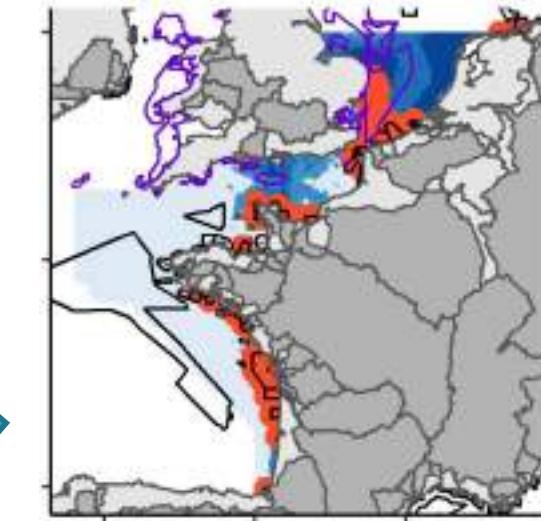
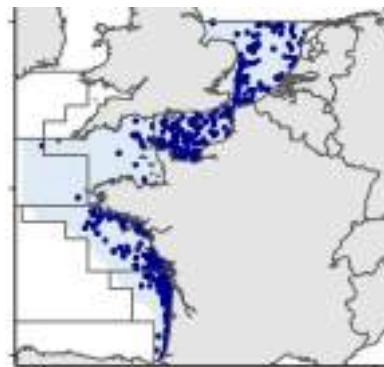
AGRO CAMPUS  
OUEST  
UNIVERSITÉ  
DE PAU ET DES  
PAYS DE L'ADOUR

# CMAP for MPA management - *A. fallax*

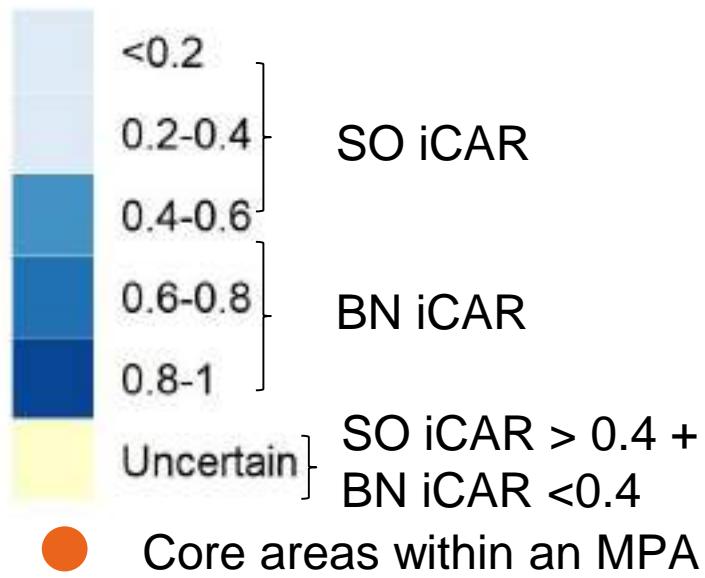
BN iCAR



SO iCAR



Probability of  
presence



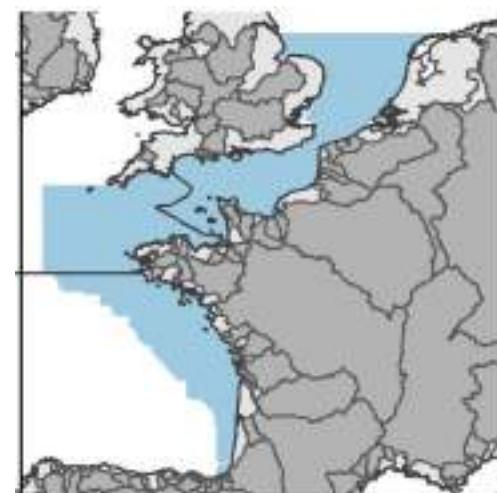
● Core areas within an MPA

# CMAP for MPA management - *A. anguilla*

BN iCAR

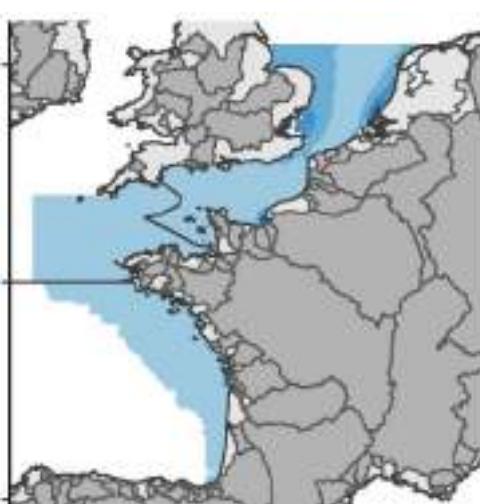
ZIB iCAR

SO iCAR



	1	0
1	4	1
0	56	4451

Sens:0.03  
Spec:0.99  
PPV:0.80  
NPV:0.99



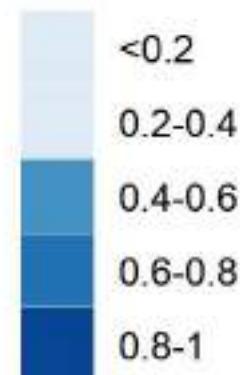
36	160
24	4292



26	32
34	4420

Sens:0.23  
Spec:0.99  
PPV:0.45  
NPV:0.94

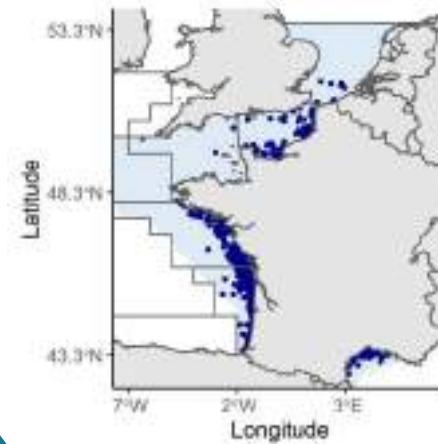
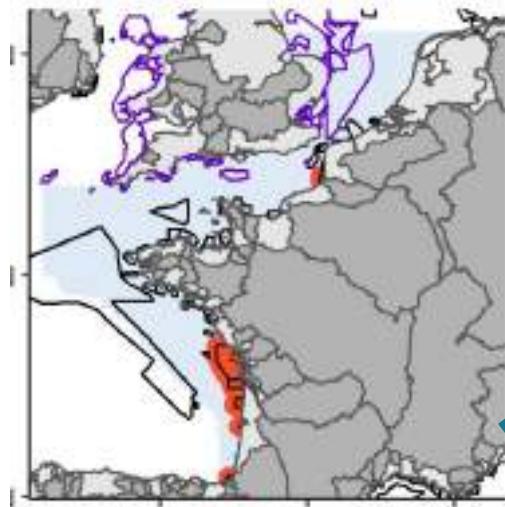
Probability of presence



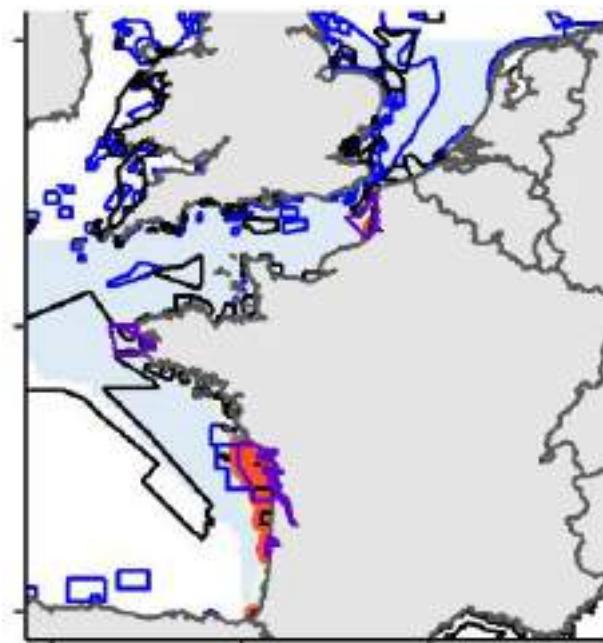
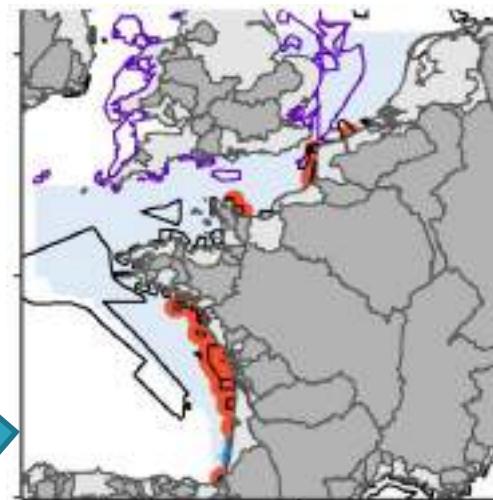
● Core areas within an MPA

# CMAP for MPA management - *C. ramada*

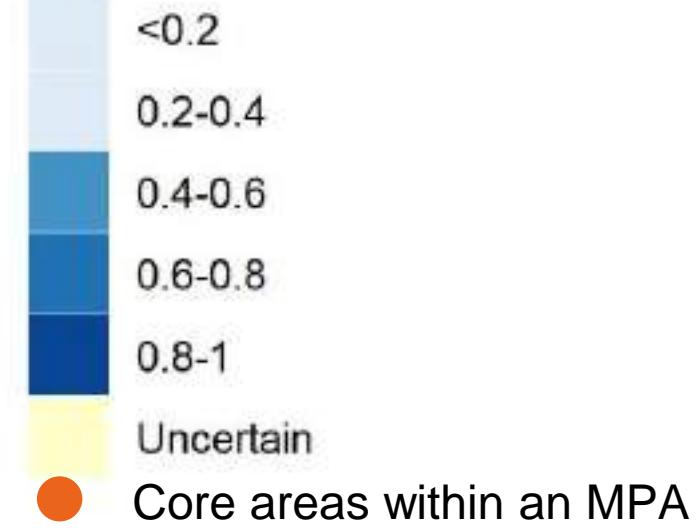
BN iCAR



SO iCAR

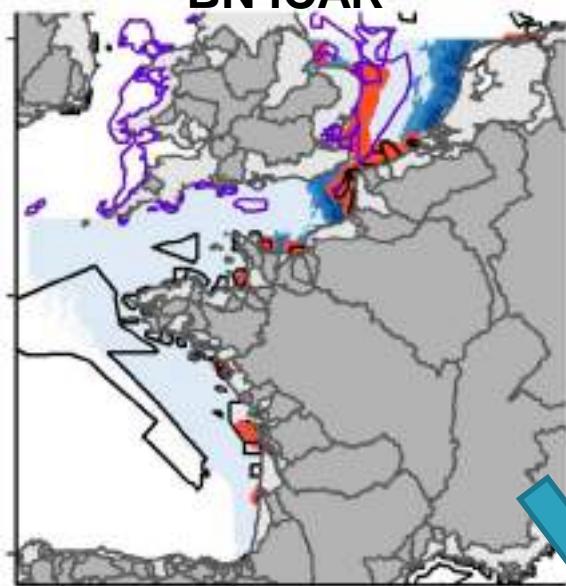


Probability of  
presence

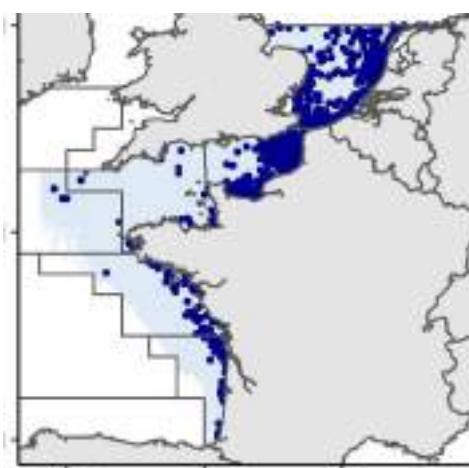


# 'CMAP for MPA management – *P. flesus*

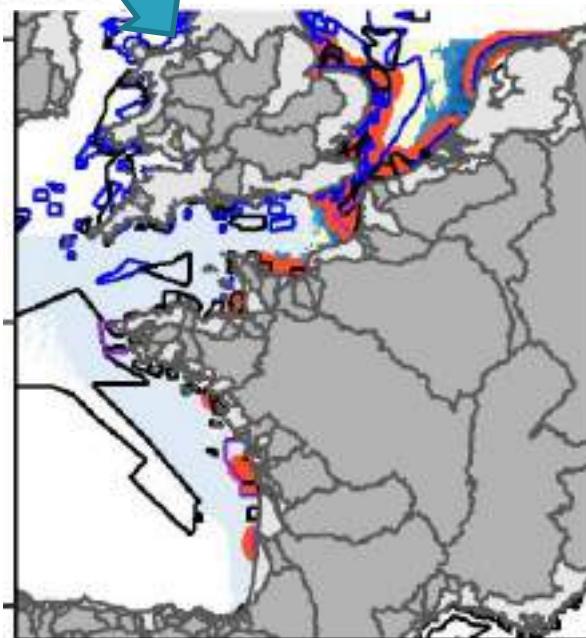
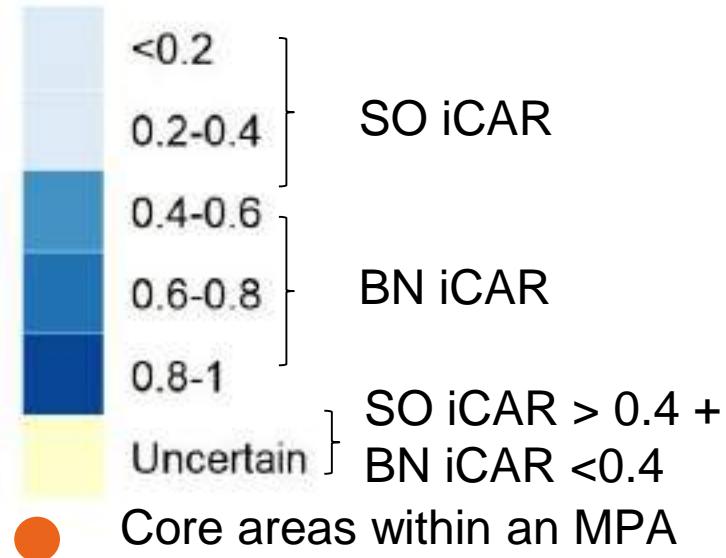
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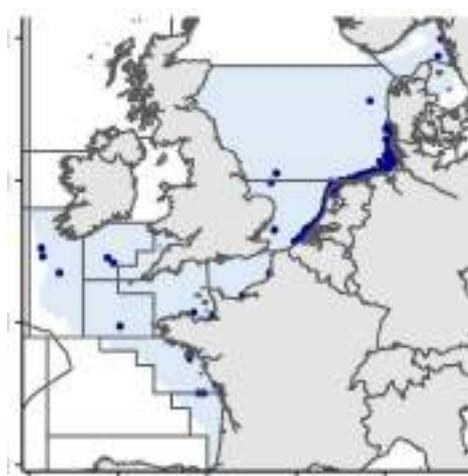
SO iCAR



Probability of presence



# CMAP for MPA management – *O. eperlanus*

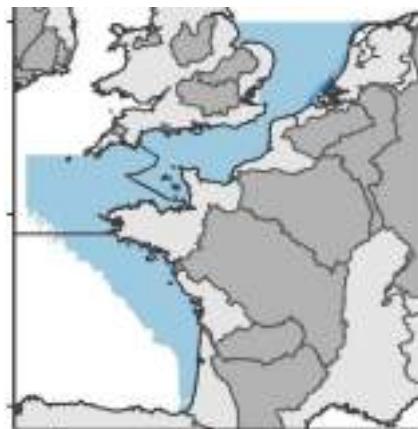


ZIB iCAR

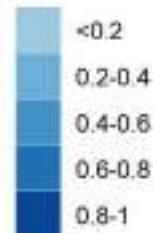
Sens:0.51  
Spec:0.98  
PPV:0.32  
NPV:1

	1	0
1	60	126
0	19	5199

SO iCAR

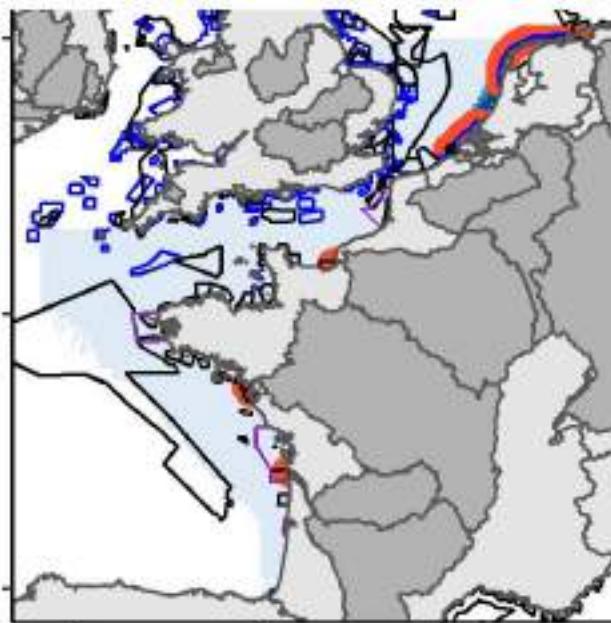


Probability of presence

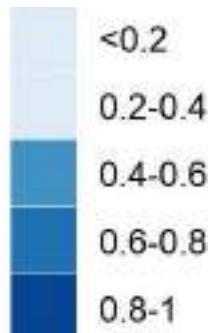


Sens:0.66  
Spec:0.99  
PPV:0.64  
NPV:1

58	32
21	5293



Probability of presence



● Core areas within an MPA