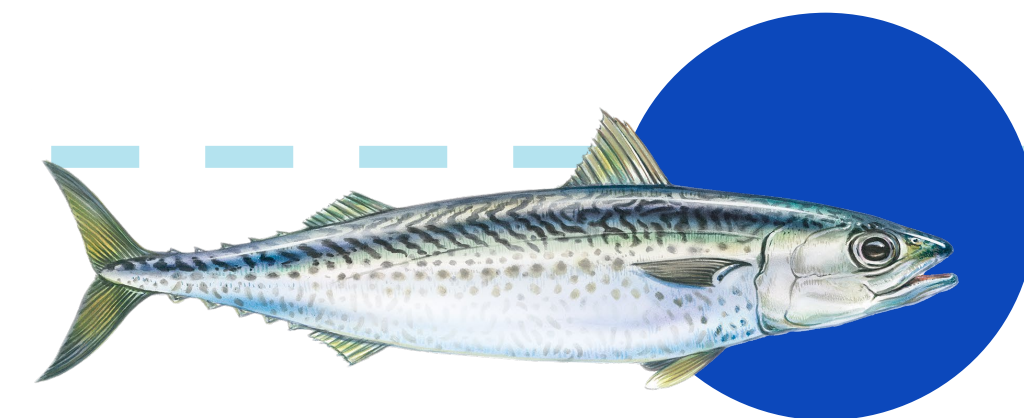


# Assessing external environmental drivers for the Moroccan Chub Mackerel ( *Scomber colias* ) population dynamics

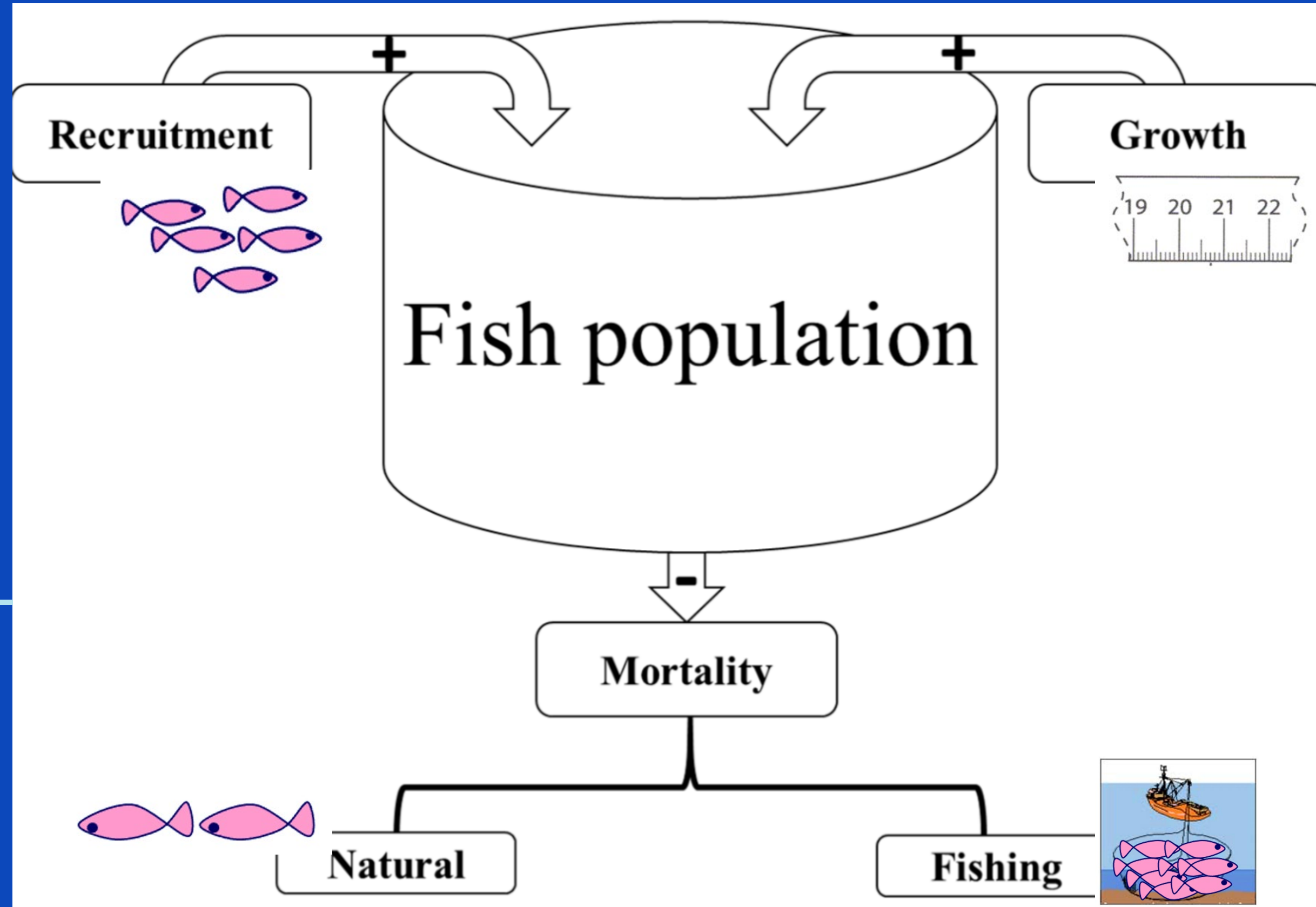
Ghoufrane Derhy, Diego Macías , Karima Khalil , Khalid Elkalay, Margarita María Rincón



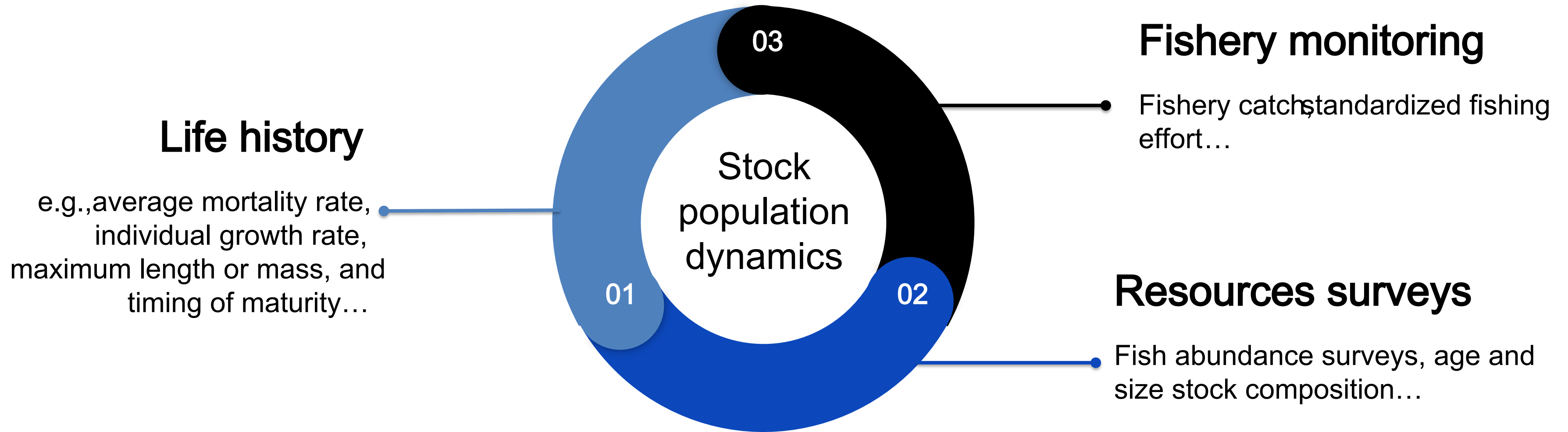
# Stock population dynamics

Changes in stock population dynamics may be caused by a combination of natural and fishing effects or other factors.

$$B_{t+1} = B_t + R + G - M - C$$



# Estimating population dynamics trends ?



One keyword is missing, which is the environment.

# Changing in environmental conditions



↳ May violates the key assumptions of traditional stock assessment models

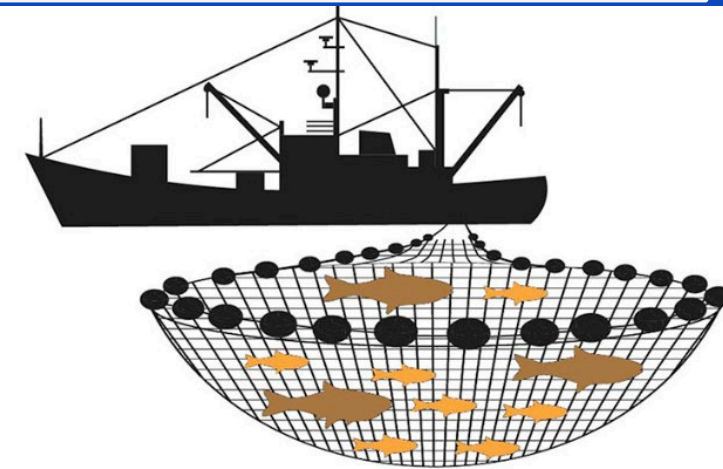
Change in primary productivity

Change in fish growth

Change in fish distributions

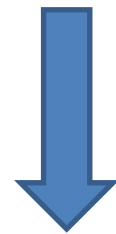
Newly opened fisheries

- Changing fishery yields
- More fishers

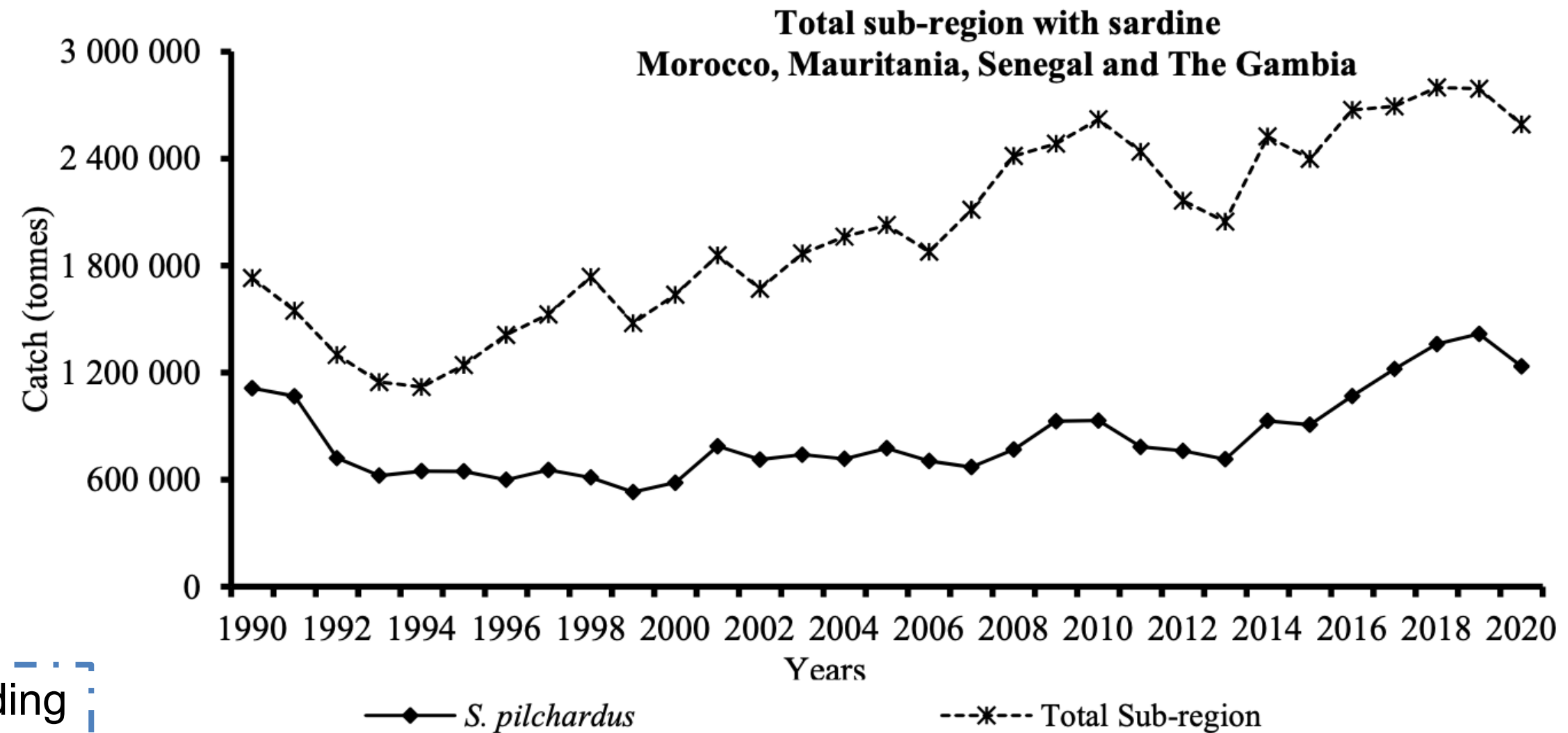


# Small pelagic stocks in CECAF areas

- Small pelagic fish are highly dynamic fish populations, influenced by complex processes;
- Their dynamic and complex life-cycles often cover wide ocean areas, which is why many small pelagic fish stocks are shared between different coastal states;
- Understanding the factors triggering the observed changes in small pelagic stocks entails research and deeper analyses.



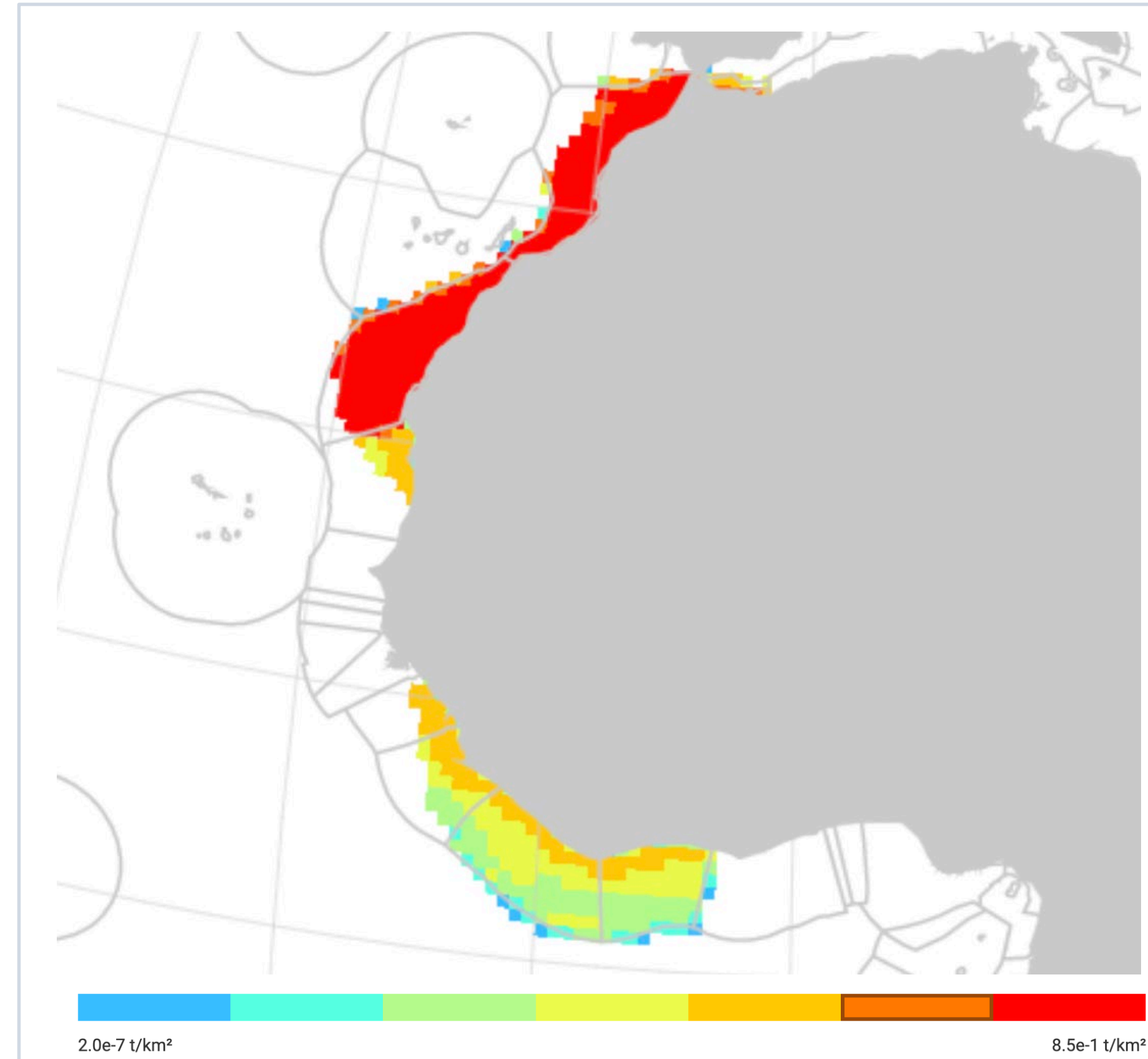
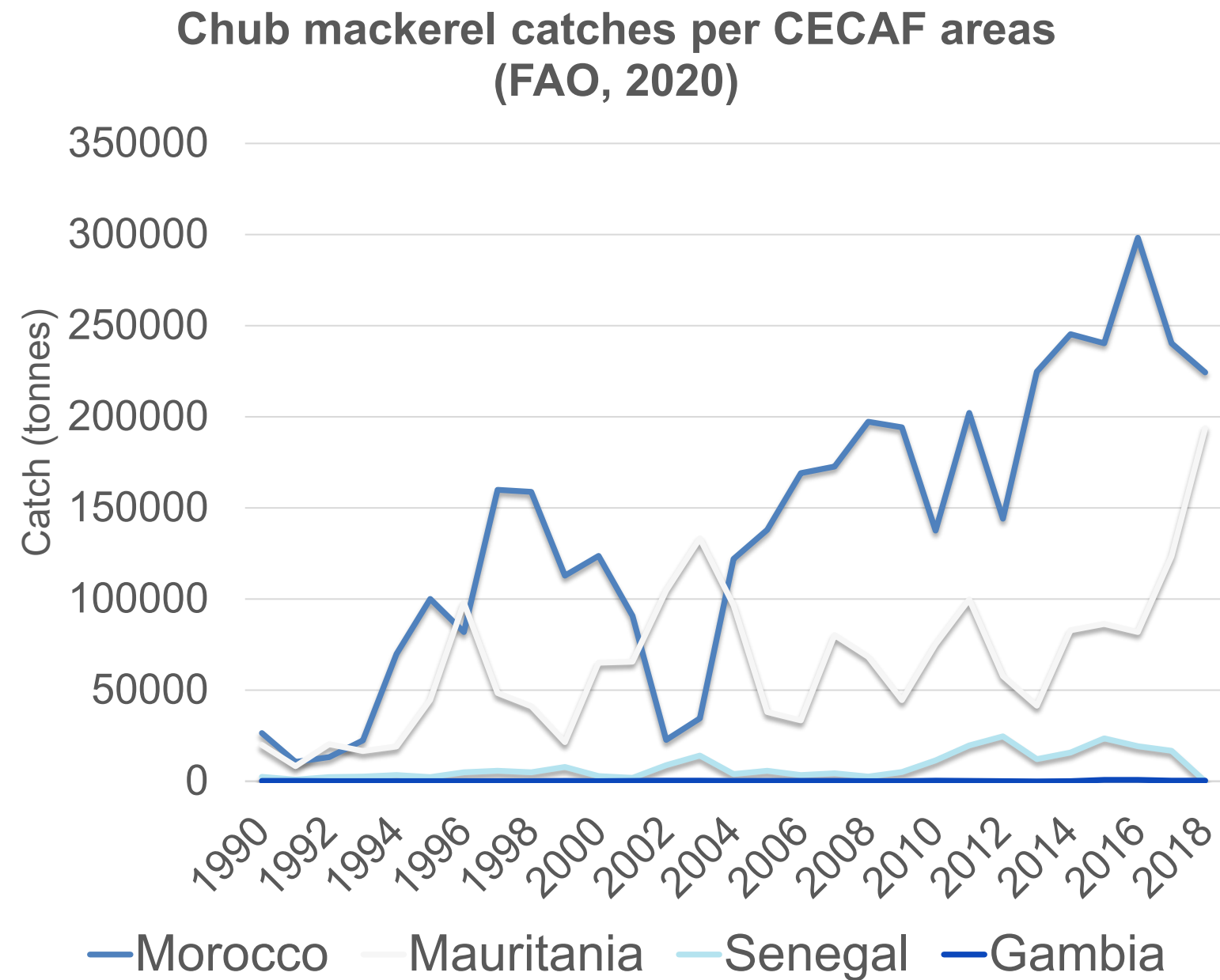
Improve oceanographic modeling for understanding fisheries biology and population dynamics.



Total small pelagic species and sardine catches in the subregion by year with and without Sardine catches (FAO, 20021)

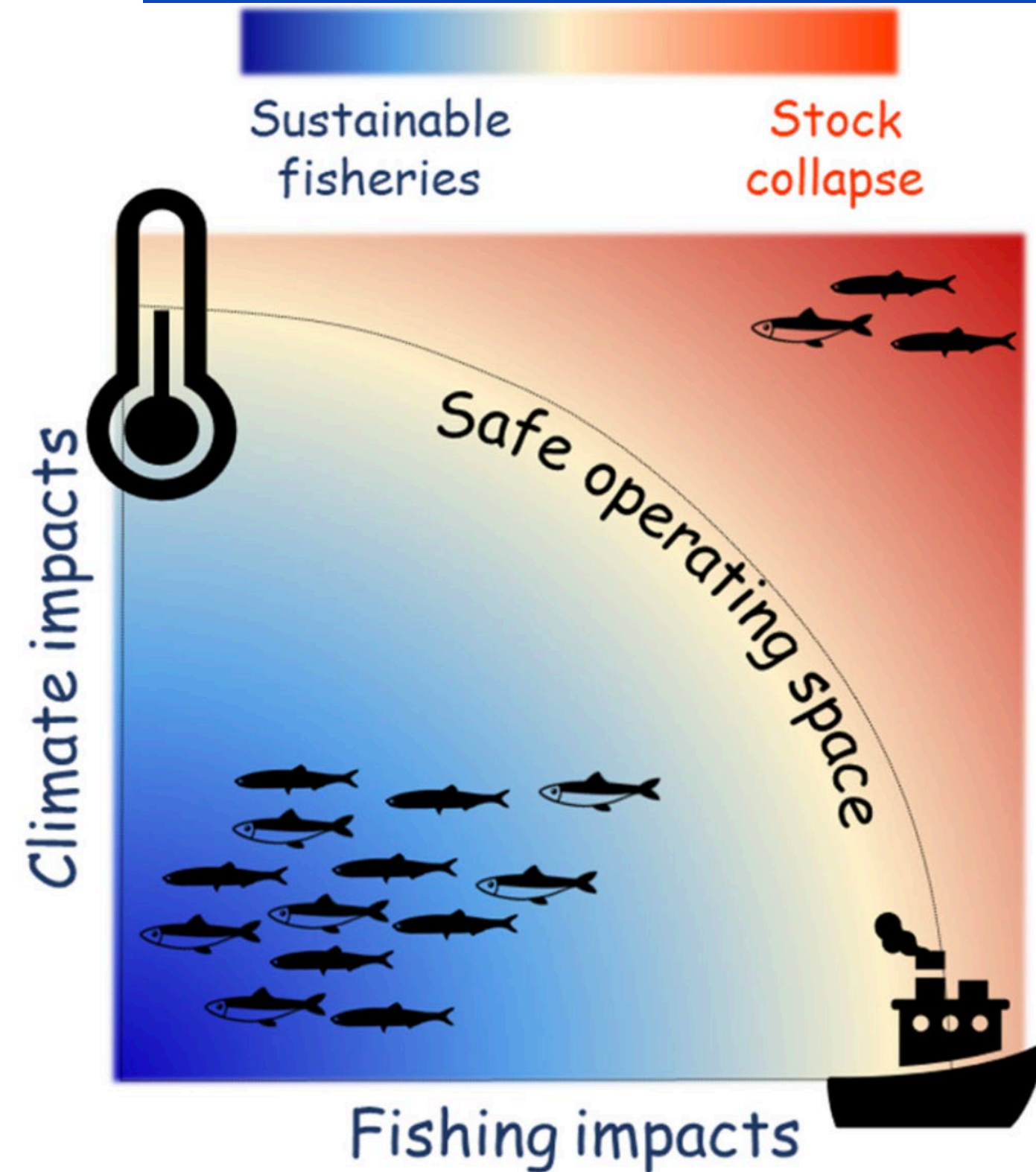
# Chub mackerel population dynamics

Environmental impacts on recruitment tend to be significant drivers of population variability for small pelagic species such as **Chub mackerel (*Scomber colias*)**.



# However, fishing can affect natural dynamics

A population's response to its environment may in fact be changed by the impacts of fishing so the processes are interrelated



# Methodology

## Adopted approach

---

- Collecting all available data  
Fisheries and surveys data
- Estimating chub mackerel population trend  
Using a stock assessment model (SRiCTmodel)
- Correlation analysis between stock abundance trend and environmental covariates  
Define the environmental factors affecting the stock abundance
- Develop an integrated stock assessment model  
Including the impact of environmental variables



# Chub Mackerel stock

Center and south of Moroccan Atlantic coast

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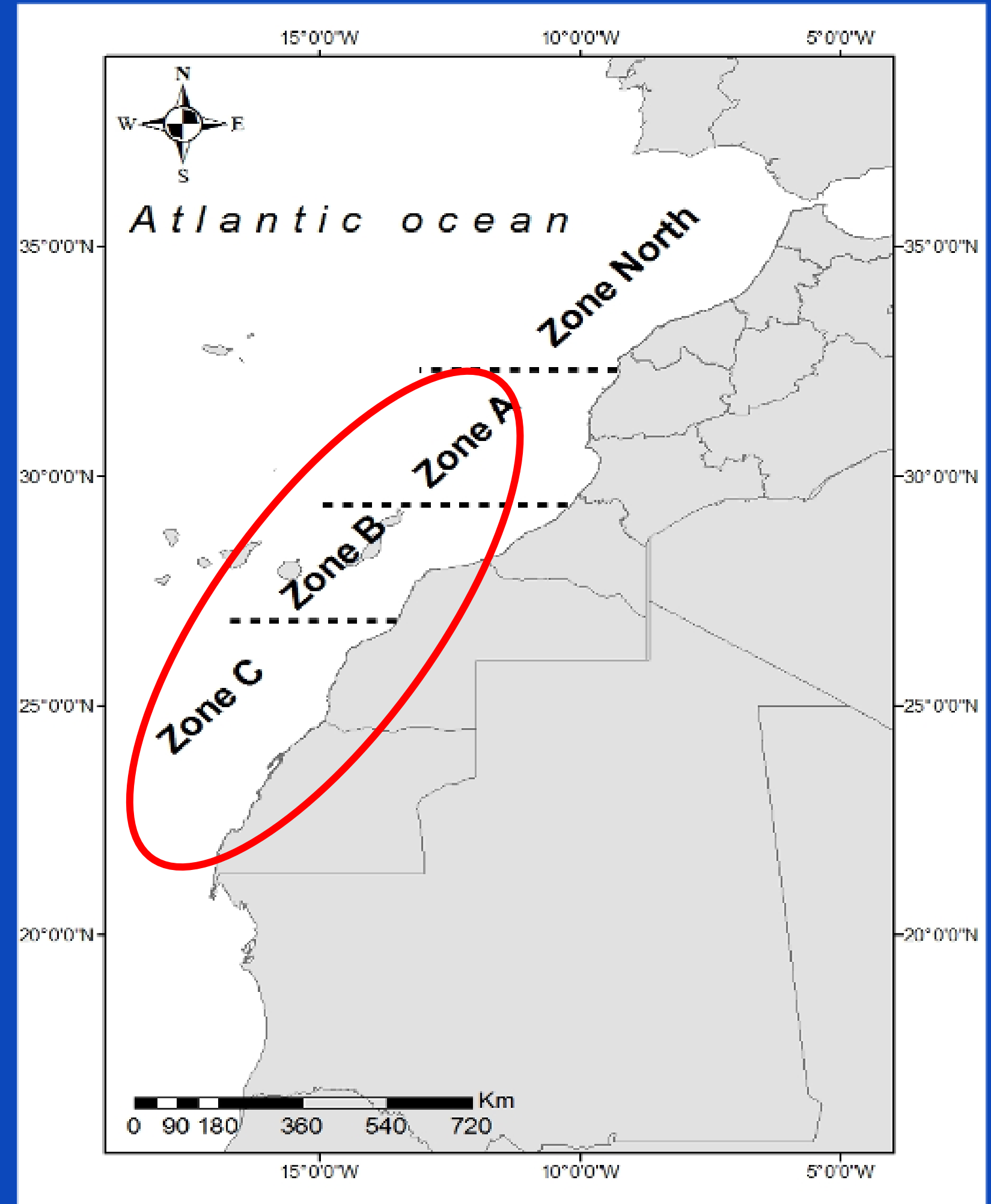
17%

Of Morocco's total small pelagic stock

---

High level of organic production  
(1997.419 mgCm<sup>-2</sup>day<sup>-1</sup>)

---



# Methodology

## Adopted approach

- Collecting all available data

Fisheries and surveys data

- Estimating chub mackerel population trend

Using a stock assessment model (RiCTmodel)

- Correlation analysis between stock abundance trend and environmental covariates

Define the environmental factors affecting the stock abundance

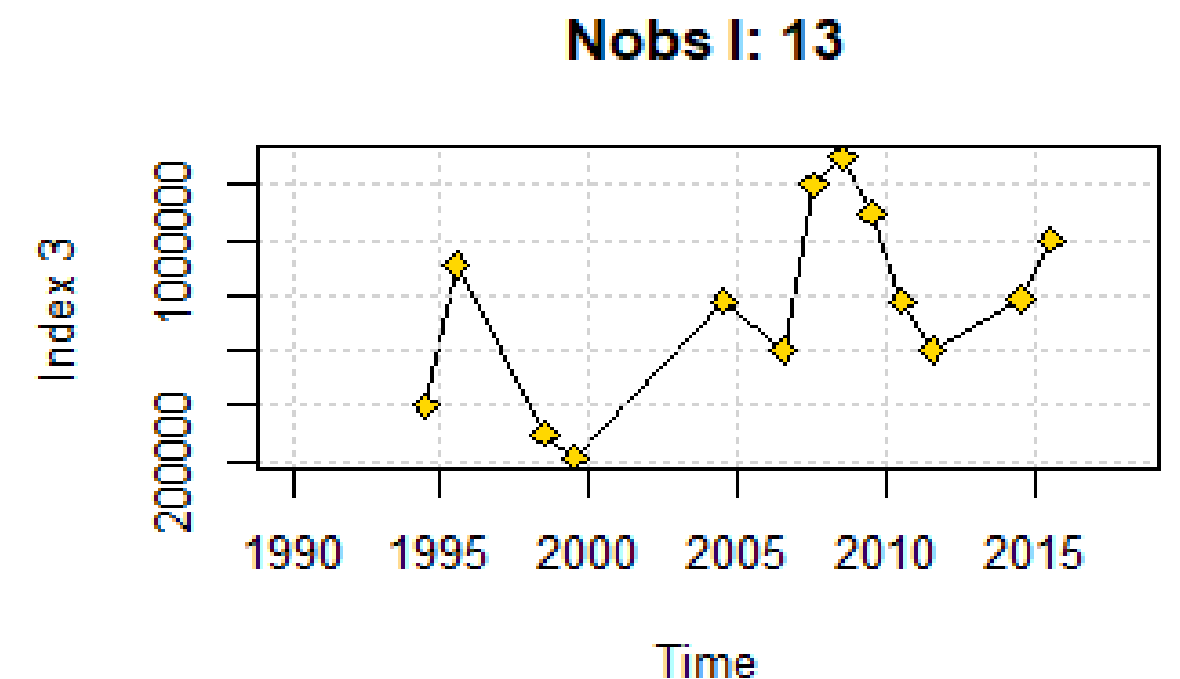
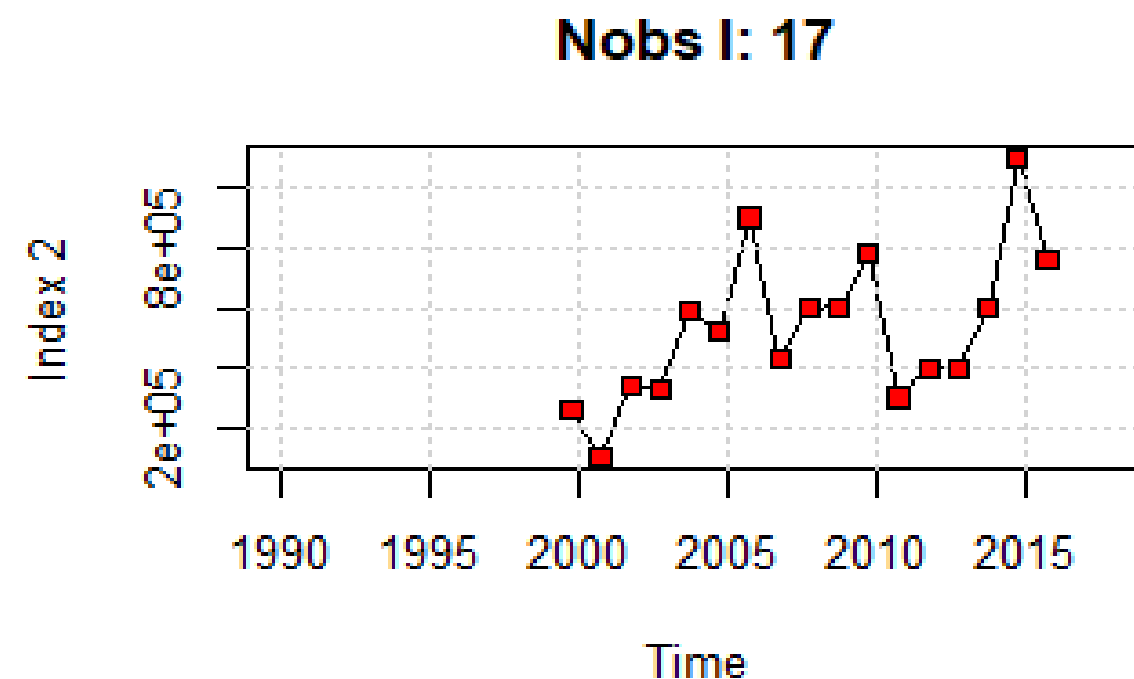
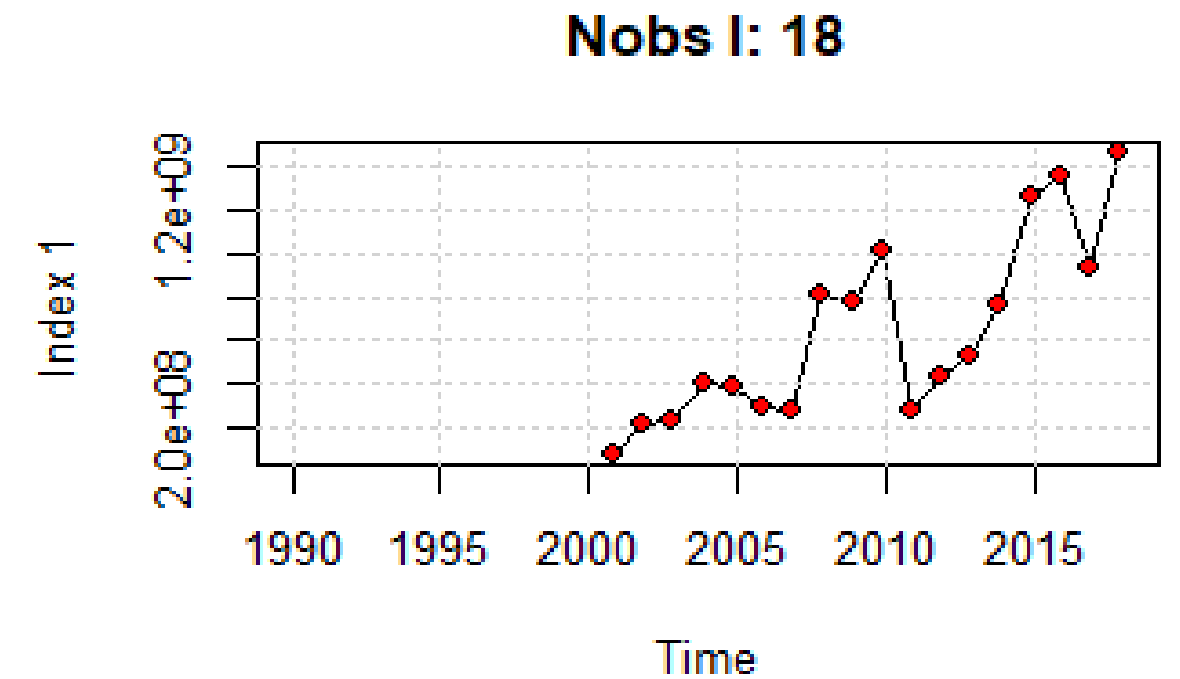
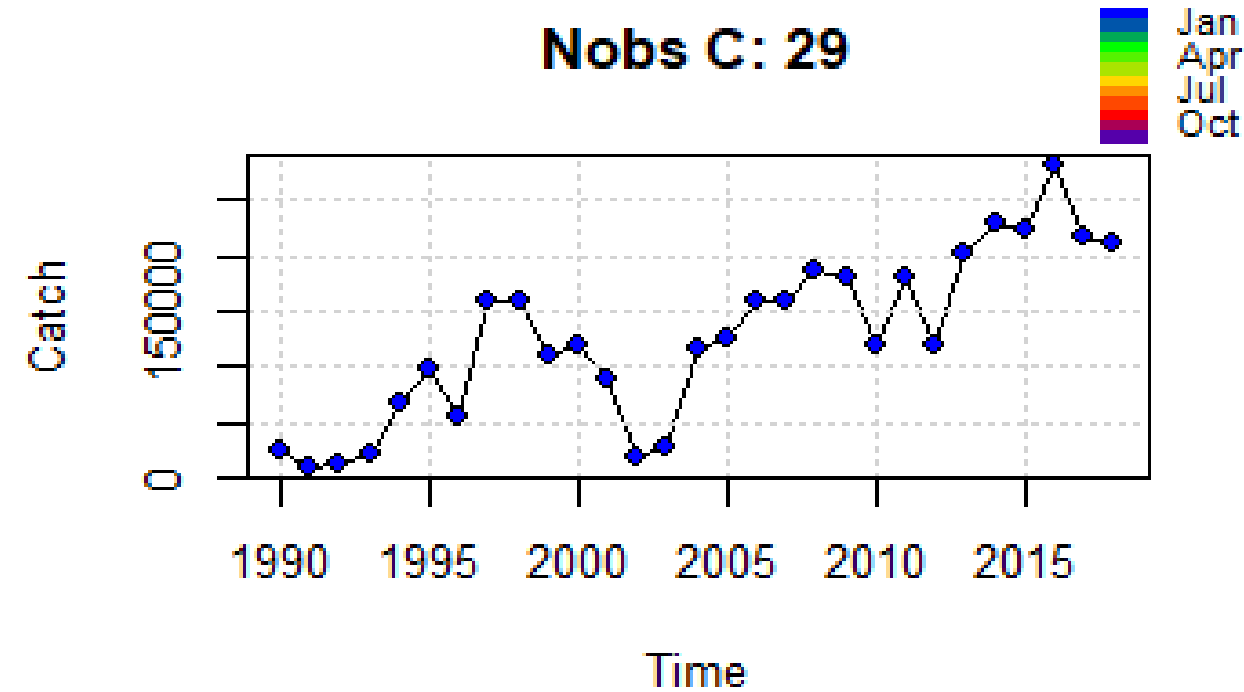
- Develop an integrated stock assessment model

Including the impact of environmental variables

# Collecting all available data

Fisheries and surveys data

- Catches data (1990-2018) [FAO, 2020]
- Abundance index data (3 indices from Amir Moulay Abdellah survey, Nansen survey and Atlantida acoustic survey)



# Methodology

## Adopted approach

---

- Collecting all available data  
Fisheries and surveys data

- Estimating chub mackerel population trend  
Using a stock assessment model (RiCTmodel)

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Including the impact of environmental variables

# Estimating chub mackerel population trend

Using a stock assessment model -Surplus Production model in iContinuous Time (SPiCT)-

- State space model, incorporating dynamics of fisheries and biomass;

- Two statistical parts:

## The process equations

**Biomass:** 
$$dB_t = rB_t \left(1 - \left[\frac{B_t}{K}\right]^{n-1}\right) dt - F_t B_t dt + \sigma_B B_t dW_t$$

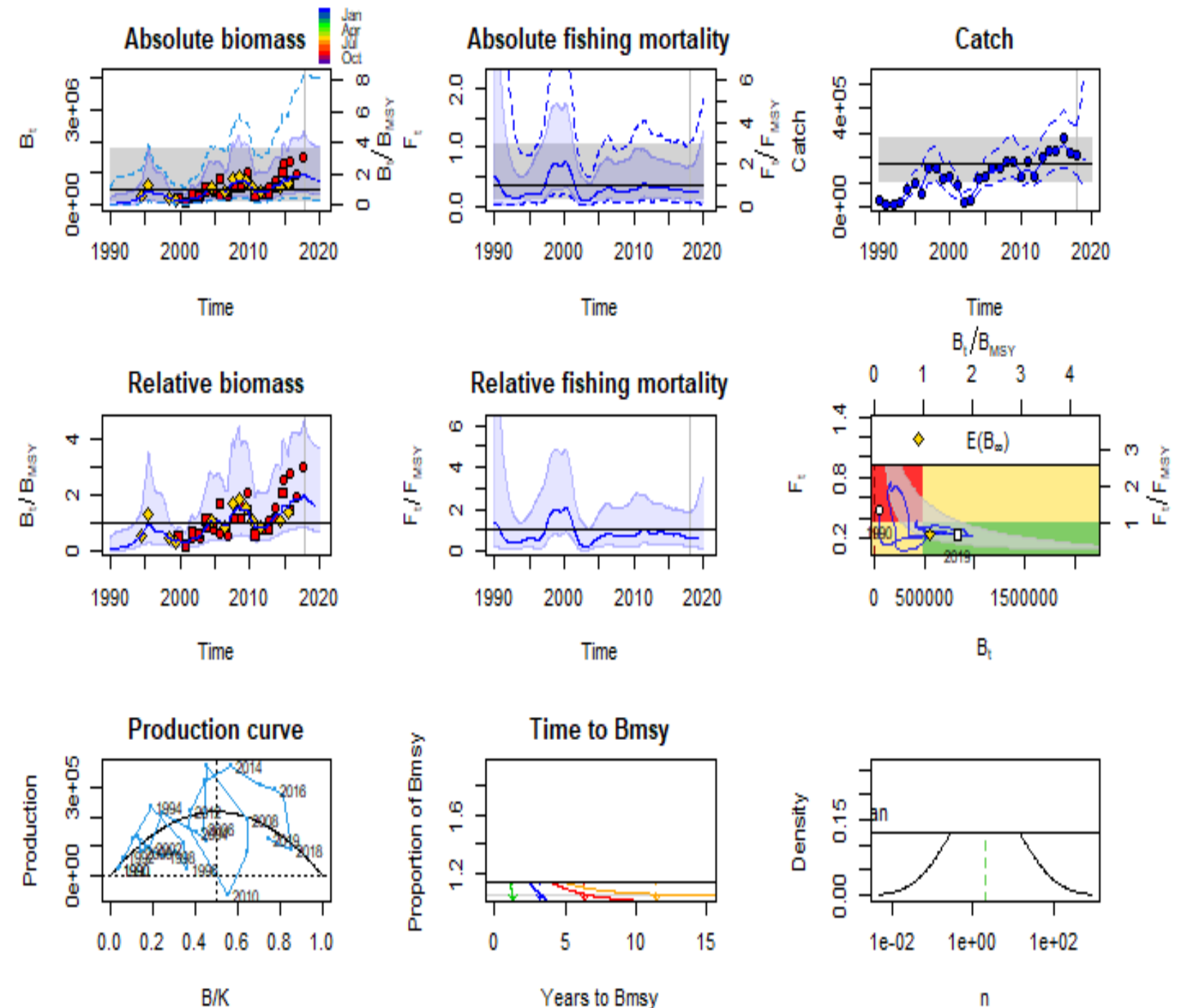
**Fishing:** 
$$d\log(F_t) = f(t, \sigma_F c)$$

## The observation equations

**Catch:** 
$$\log(C_t) = \log\left(\int_t^{t+\Delta} F_s B_s ds\right) + \epsilon_t, \quad \epsilon_t \sim N(0, [\beta \sigma_F]^2)$$

**Index:** 
$$\log(I_t) = \log(qB_t) + e_t, \quad e_t \sim N(0, [\alpha \sigma_B]^2)$$

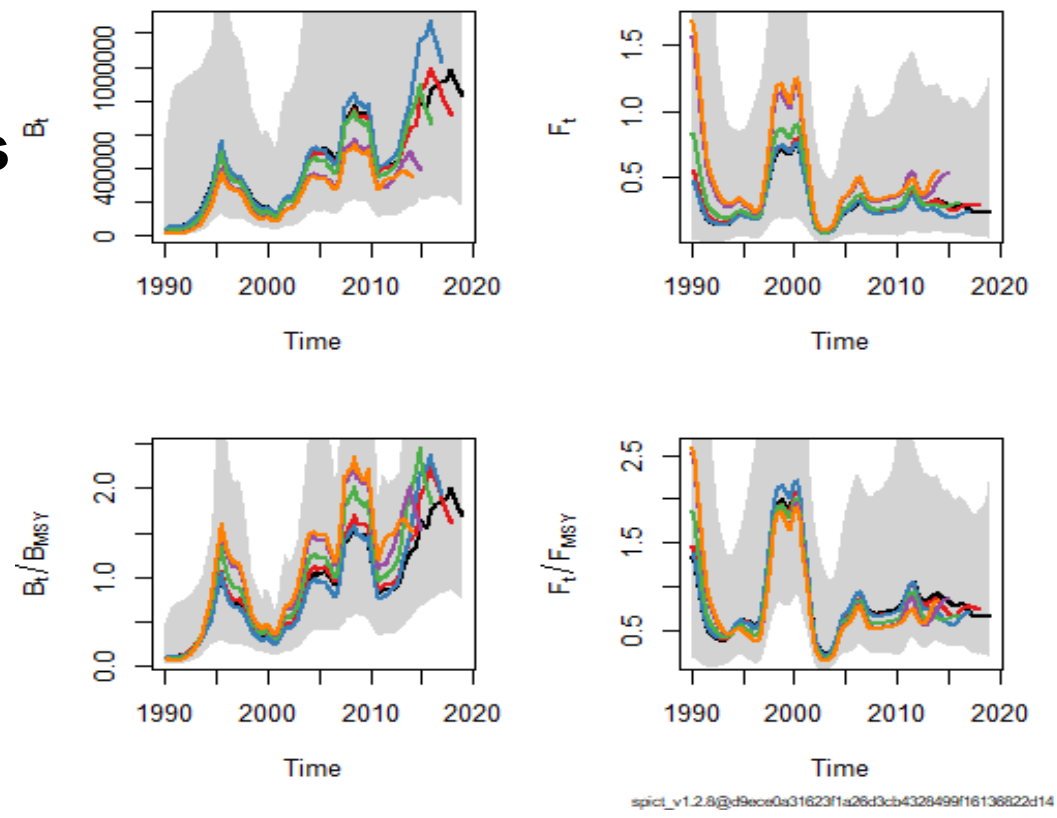
## Model outputs



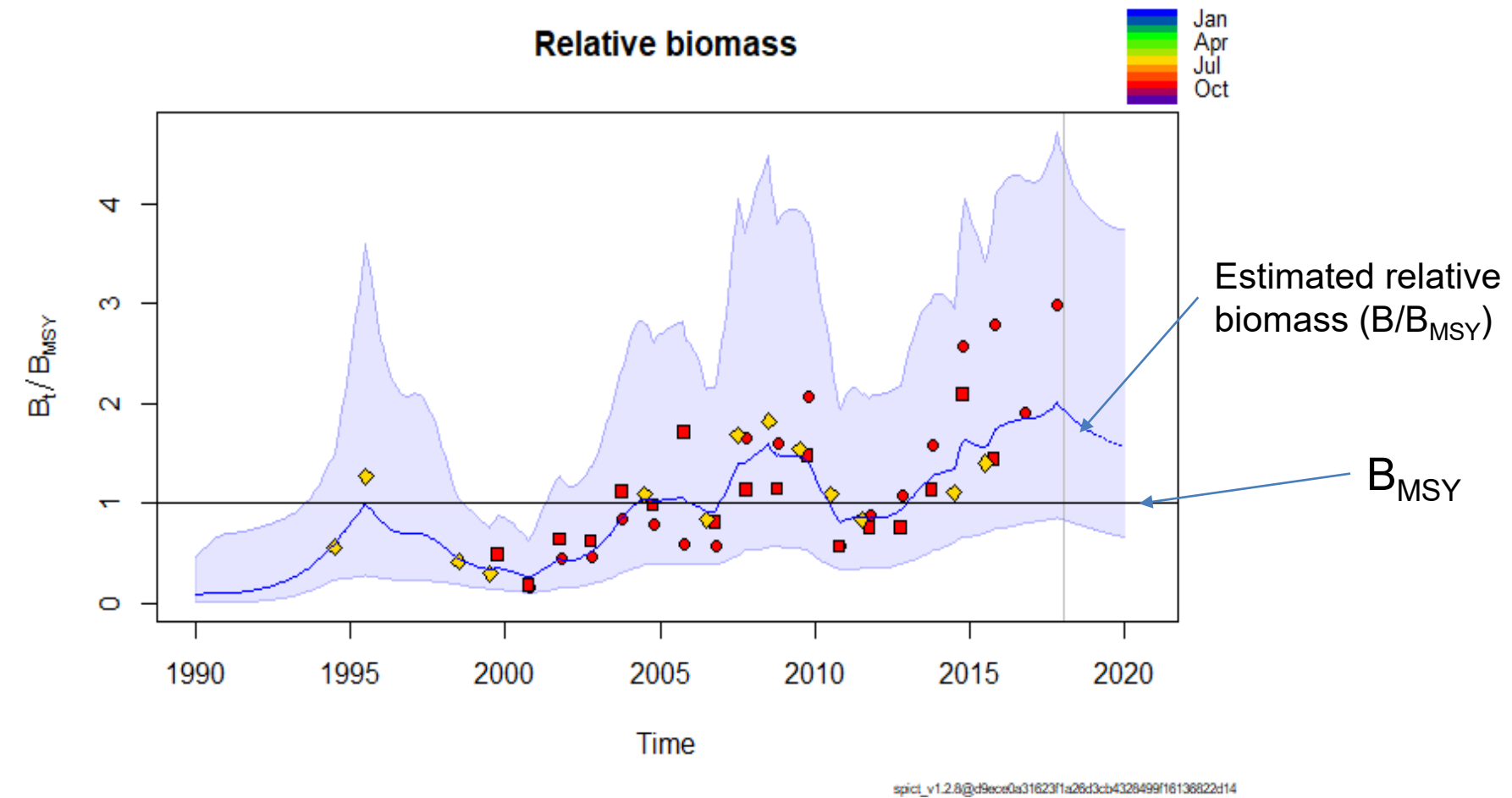
# Estimating chub mackerel population trend

Using a stock assessment model -Surplus Production model in iContinuous Time (SPiCT)-

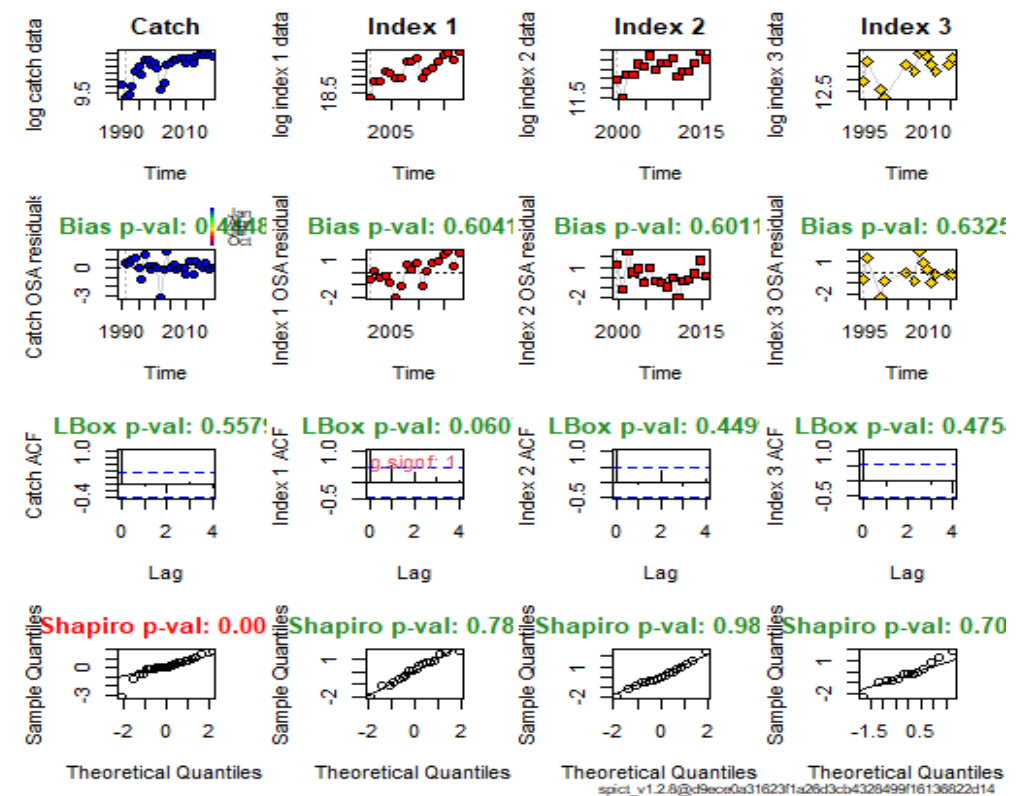
- Retrospective analysis



Estimated relative biomass ( $B_t/B_{MSY}$ )



- Model diagnostic



➤ The relative biomass ( $B/B_{MSY}$ ) trend used to explore the relationship between the estimated chub mackerel stock abundance and different external environmental drivers

# Methodology

## Adopted approach

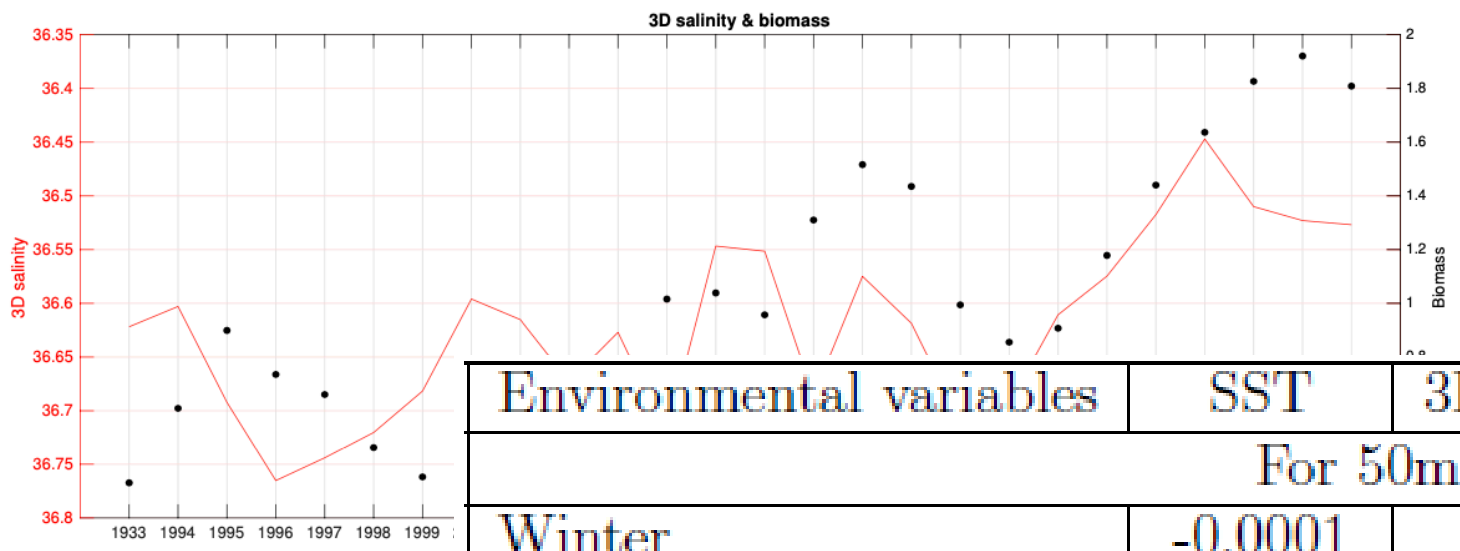
- Collecting all available data  
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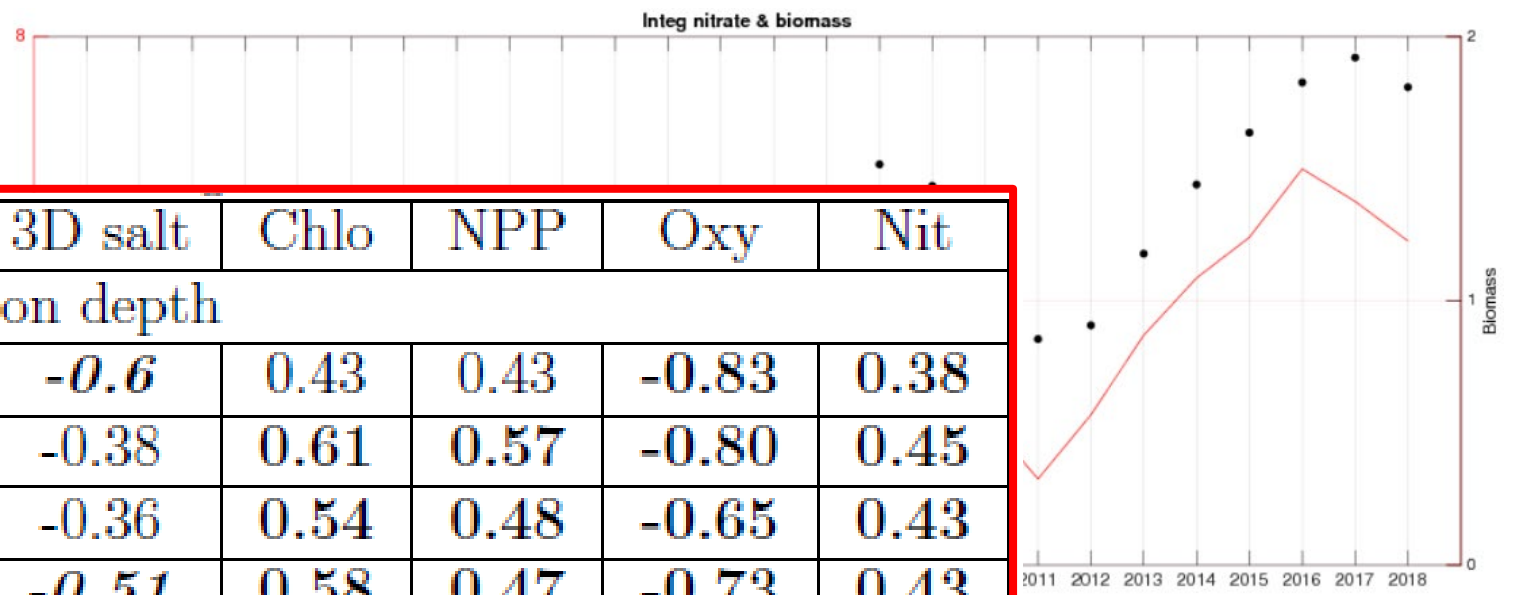
# Correlation analysis between stock abundance trend and environmental covariate

Define the environmental factors affecting the stock abundance

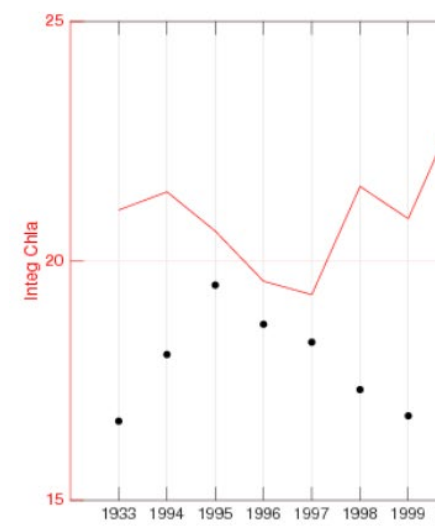
Salinity



I/m3

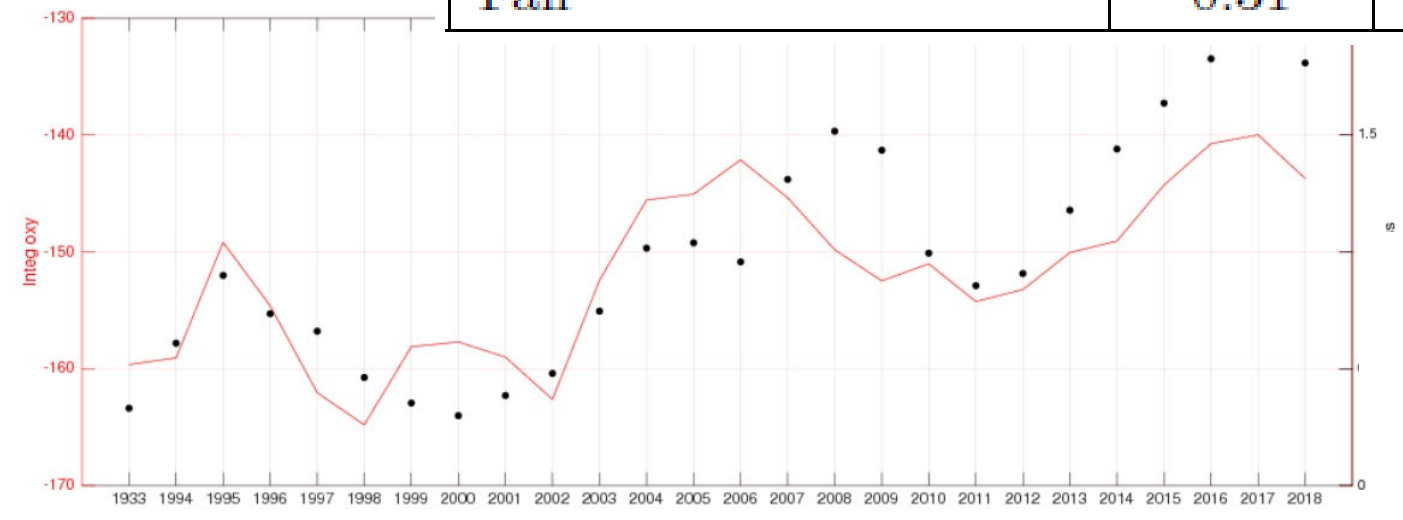


Chlorophyll mg/m3

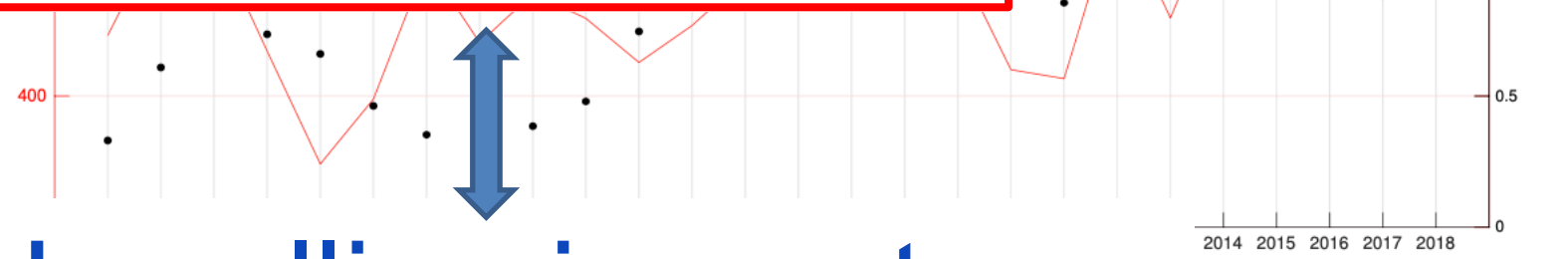


Environmental variables	SST	3D temp	3D salt	Chlo	NPP	Oxy	Nit
For 50m integration depth							
Winter	-0.0001	0.009	<b>-0.6</b>	0.43	0.43	<b>-0.83</b>	<b>0.38</b>
Spring	-0.05	-0.10	-0.38	<b>0.61</b>	<b>0.57</b>	-0.80	<b>0.45</b>
Summer	0.26	0.17	-0.36	<b>0.54</b>	<b>0.48</b>	-0.65	<b>0.43</b>
Fall	0.31	0.22	<b>-0.51</b>	<b>0.58</b>	<b>0.47</b>	-0.73	<b>0.43</b>
For 150m integration depth (except for oxygen with 100-200m)							
Winter	-0.00014	-0.05	<b>-0.49</b>	0.36	0.34	<b>-0.83</b>	<b>0.69</b>
Spring	-0.06	-0.08	-0.32	<b>0.64</b>	<b>0.56</b>	-0.80	<b>0.68</b>
Summer	0.27	0.06	-0.16	<b>0.58</b>	<b>0.48</b>	-0.65	<b>0.56</b>
Fall	0.31	-0.07	<b>-0.42</b>	<b>0.64</b>	<b>0.44</b>	-0.73	<b>0.69</b>

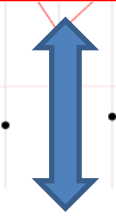
Oxygen (nmol/m3)



Net pro



Upwelling impact



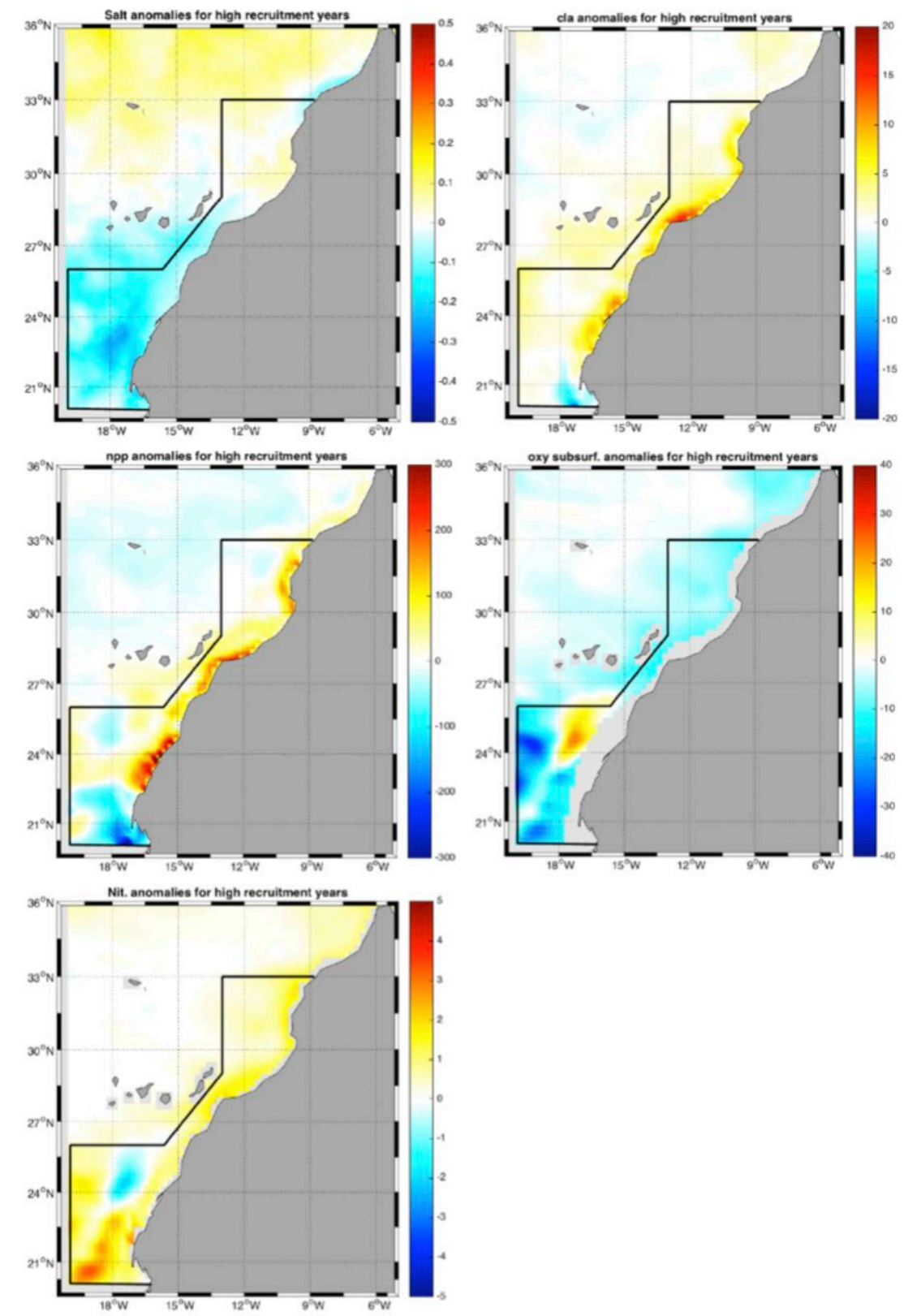
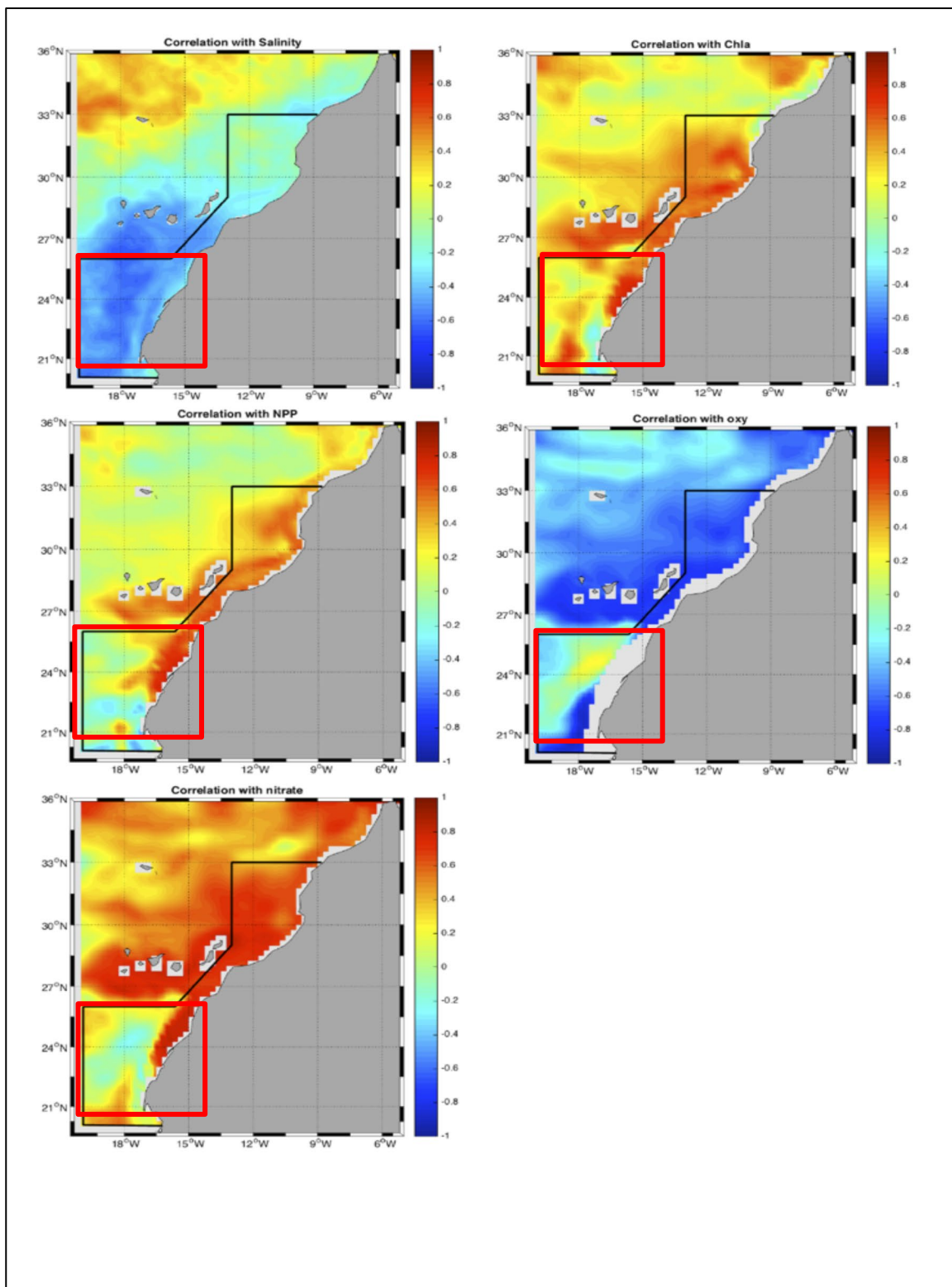




# Correlation analysis between stock abundance trend and environmental covariate

Define the environmental factors affecting the stock abundance

- Spatially explicit analysis for the correlation between  $B/BMSY$  and the selected environmental variables
- Anomalies for the different environmental variables in years of high  $B/BMSY$



# Conclusions

- Maximum R values for are located in the southern zone
- Strong upwelling intensity in the south → Accompanied with important catches (INRH, 2019; FAO, 2020)

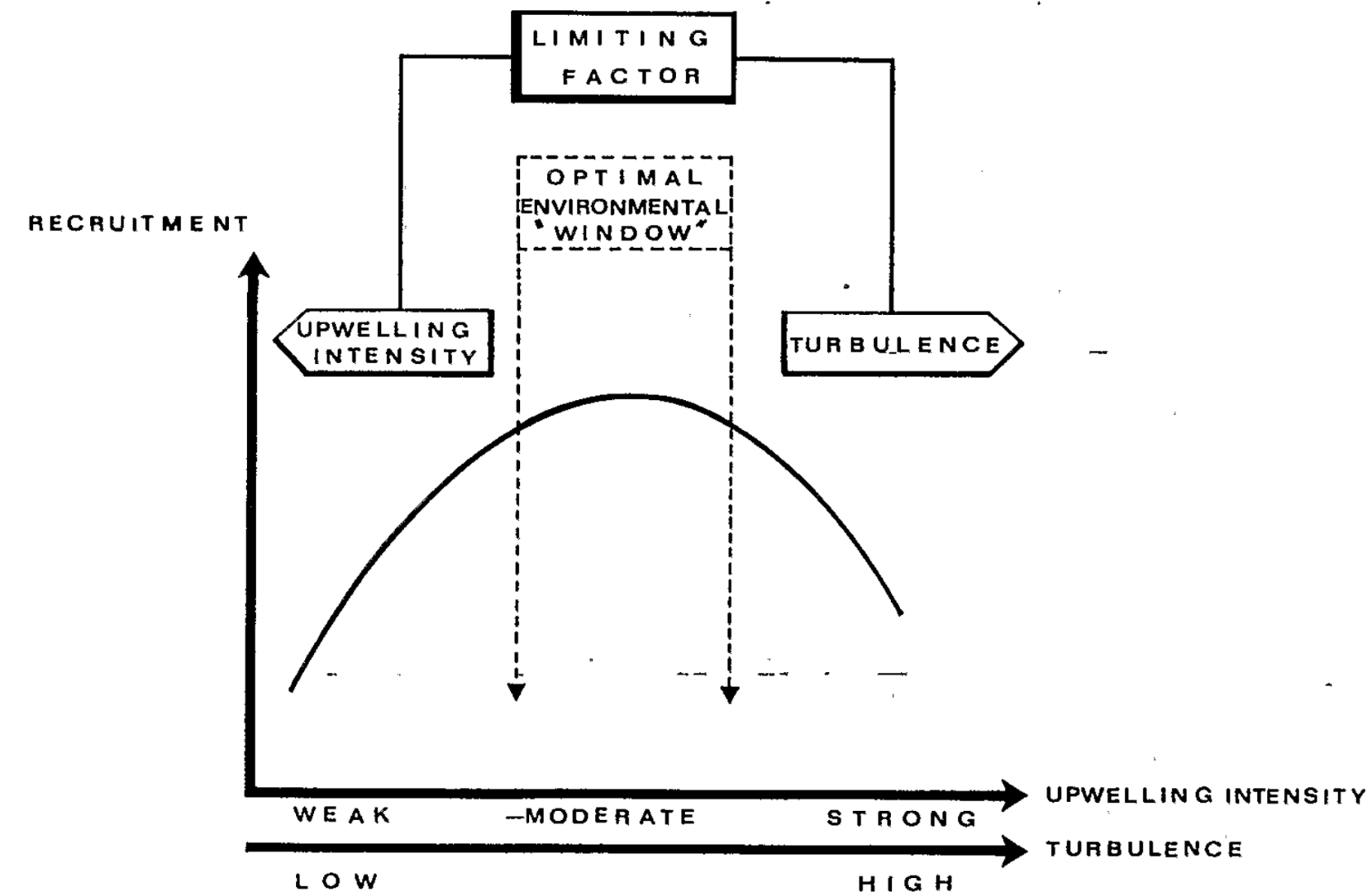
↓

**High biomass of recruited individuals**

- In our case, the turbulence is the only limiting factor for recruitment
- Adome shaped relationship between environmental factors and adult biomass

↕

**non-linear equation**



# Thanks!

Any questions?

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[Ghoufrane123.derhy@gmail.com](mailto:Ghoufrane123.derhy@gmail.com)



**Small Pelagic Fish:  
New Frontiers in Science  
and Sustainable  
Management**

November 7 - 11, 2022  
Lisbon, Portugal



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