



IMPALHA

IMPACTO DEL PALANGRE EN LOS
HÁBITATS MARINOS BENTÓNICOS



IDENTIFICATION OF ONGLINE FISHING GROUNDS USING MACHINE LEARNING FOR BENTHIC HABITAT IMPACT ASSESSMENT.

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SUMMA RY

1 INTRODUCTION

- IMPALHA project.
- Study area.
- Longlines.

3 MACHINE LEARNING

- Learning processes.
- Inputs / outputs.
- Types of models.
- Decision trees and random forest.

2 CHARACTERIZATION

- Importance of impact studies and fisheries.
- Fishing grounds identification and longlines soaking time.
- Data types.

4 FIRST RESULTS

- Speed distribution.
- Relevance of selected variables.
- Predictions.
- Identified fishing grounds.

1

INTRODUCTION



IMPALHA PROJECT

IMPACT OF THE BOTTOM LONGLINE ON BENTHIC HABITATS IN THE SCIs OF THE NATURA 2000 NETWORK

Theme:

Protected áreas.

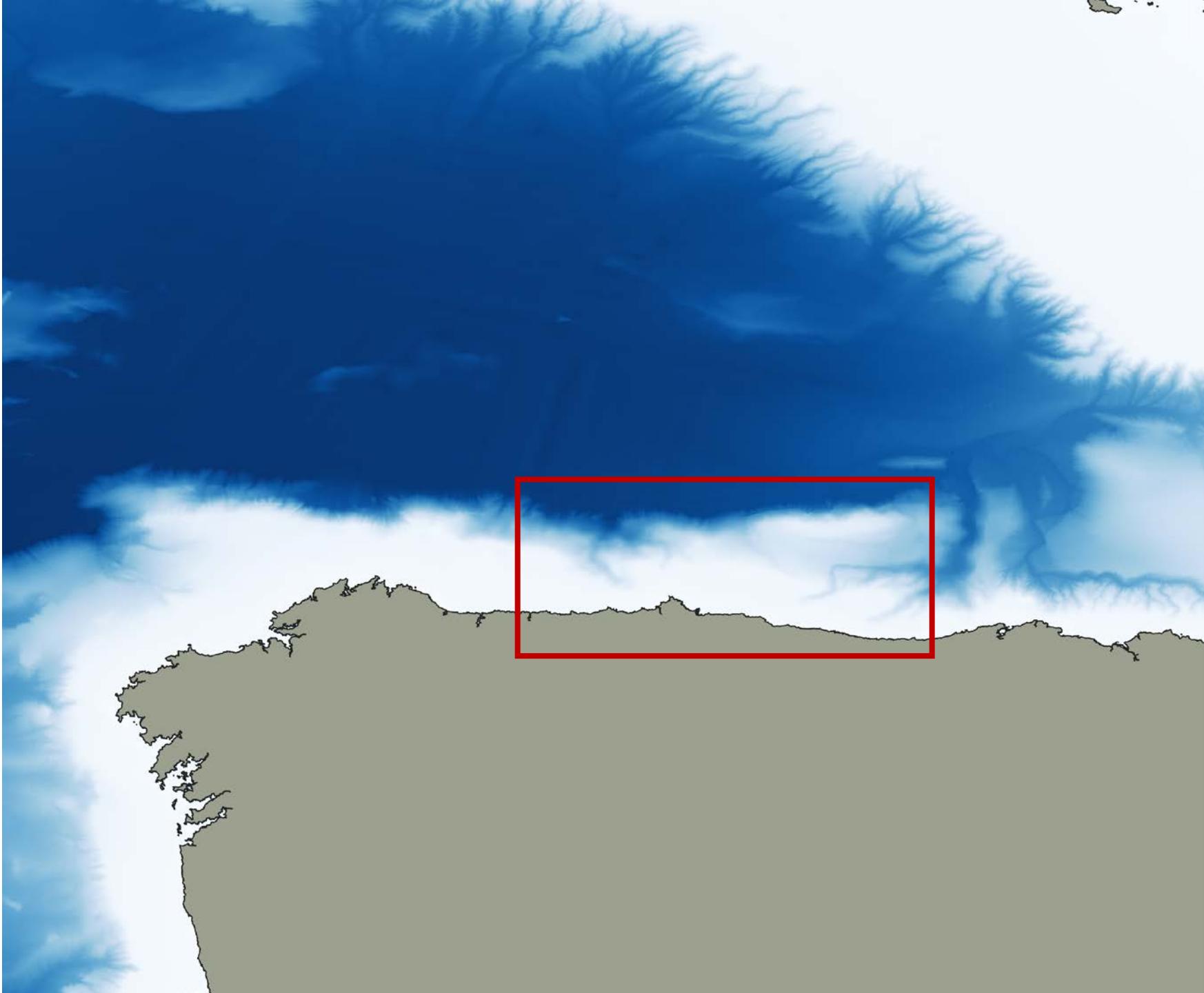
Goals:

1. Spatio-temporal characterization of longline fishing in the Avilés Canyon SCI and interaction with benthic habitats.
2. Design of the BACI (Before-After-Control-Impact) Campaign, to be developed in the second phase of the project.
3. Communicate the results of the project and the main methodologies and concepts for the study of the impact of fishing activity on habitats and advice on its application.

STUDY AREA

AVILES CANYON

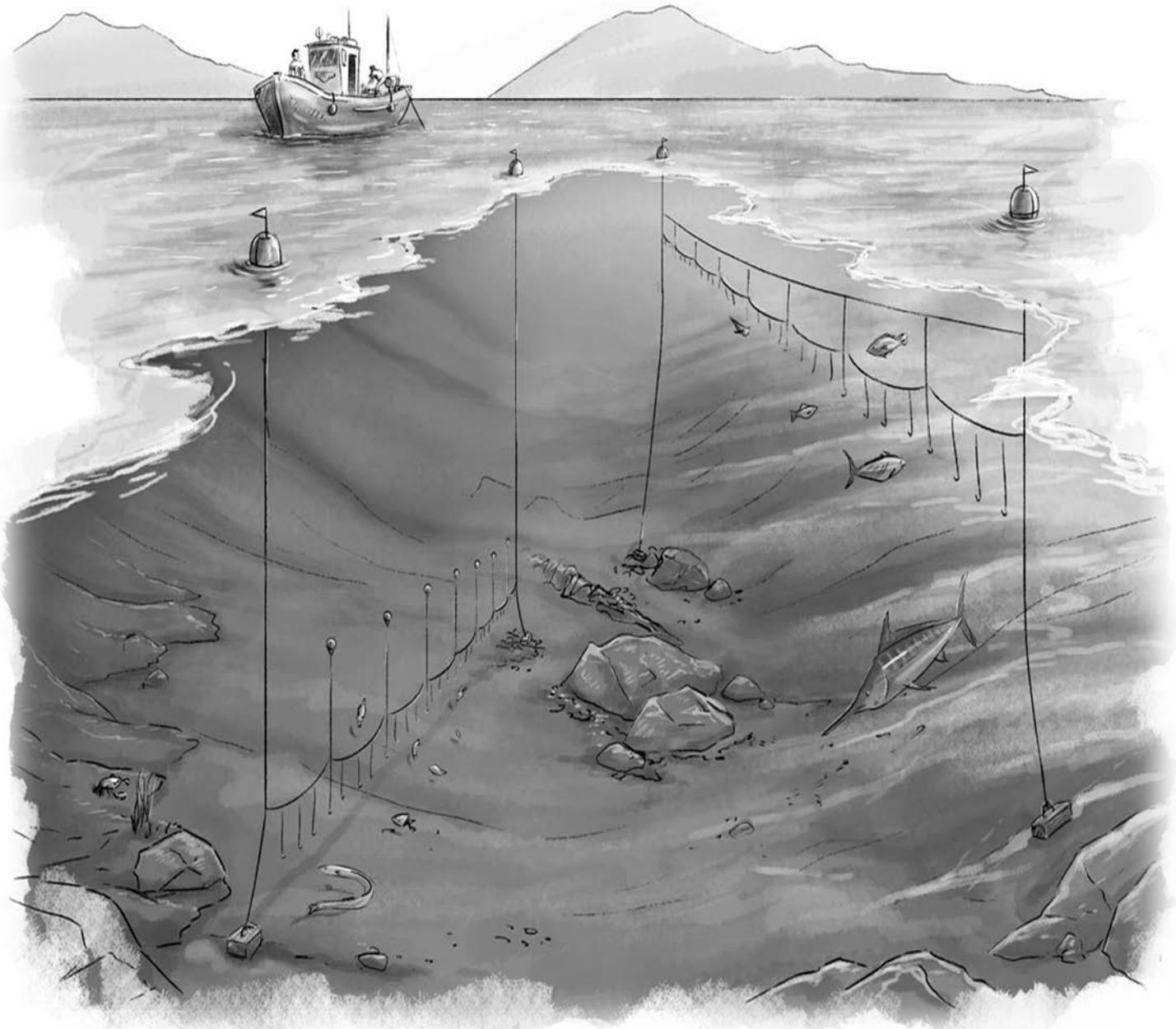
The study area was expanded in order to increase the available data.



LONGLINE FISHING

Description

- Line of hooks
- Variable length
- Two basic types
 - Surface
 - Bottom
- Selective gear



2

CHARACTERIZATIO N

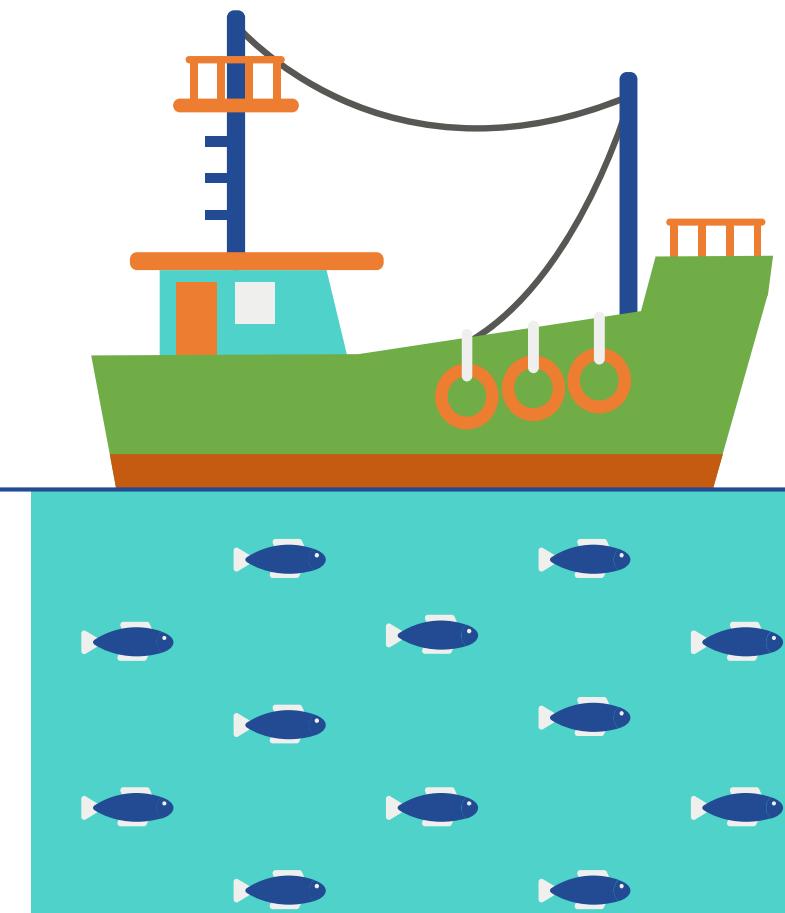
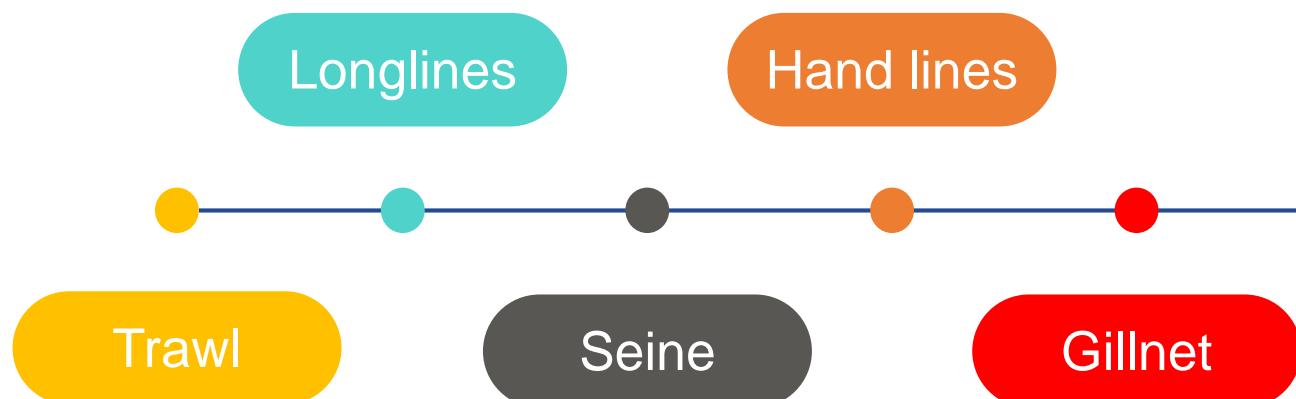


IMPORTANCE OF IMPACT STUDIES AND FISHERIES

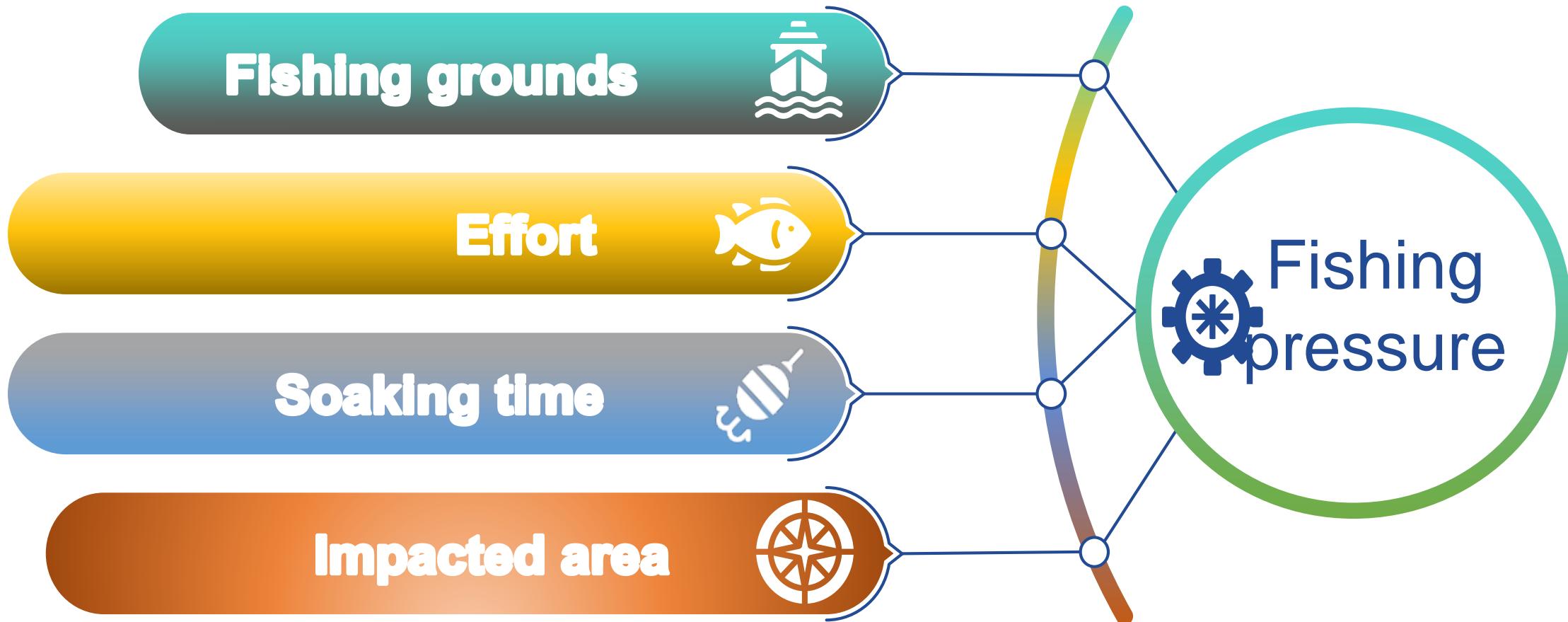
Quantify actual impacts

Identify its spatial extent

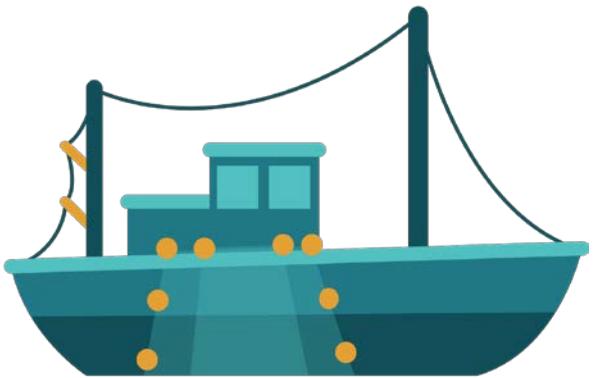
Mitigation



FISHING GROUNDS IDENTIFICATION AND LONGLINES SOAKING TIME

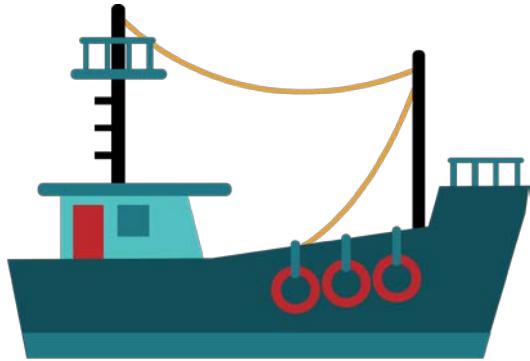


DATA TYPES



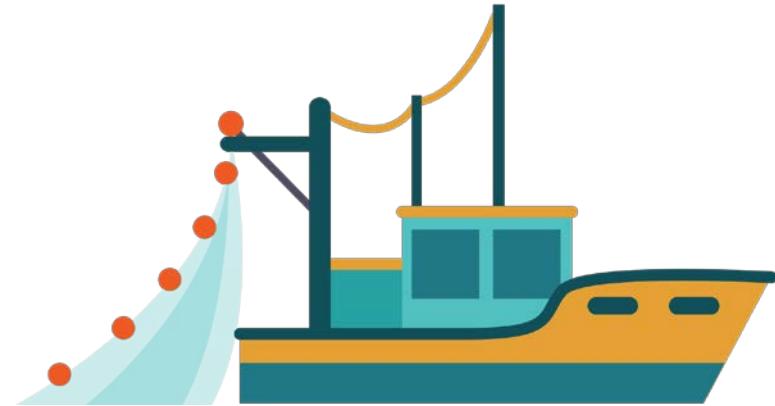
VMS

- Fishing vessels $\geq 15m$
- Resolution ~ 2 hours



AIS

- All vessels
- Variable resolution
 - Typically 3-5 minutes
- Yearly AIS $\sim 400M$



GREEN
BOXES

- Some fishing vessels
- GPS information

3

MACHINE LEARNING



LEARNING PROCESS

01 Data collection

Different sources

02 Data curation

Filter, select, homogenization ...

03 AI training

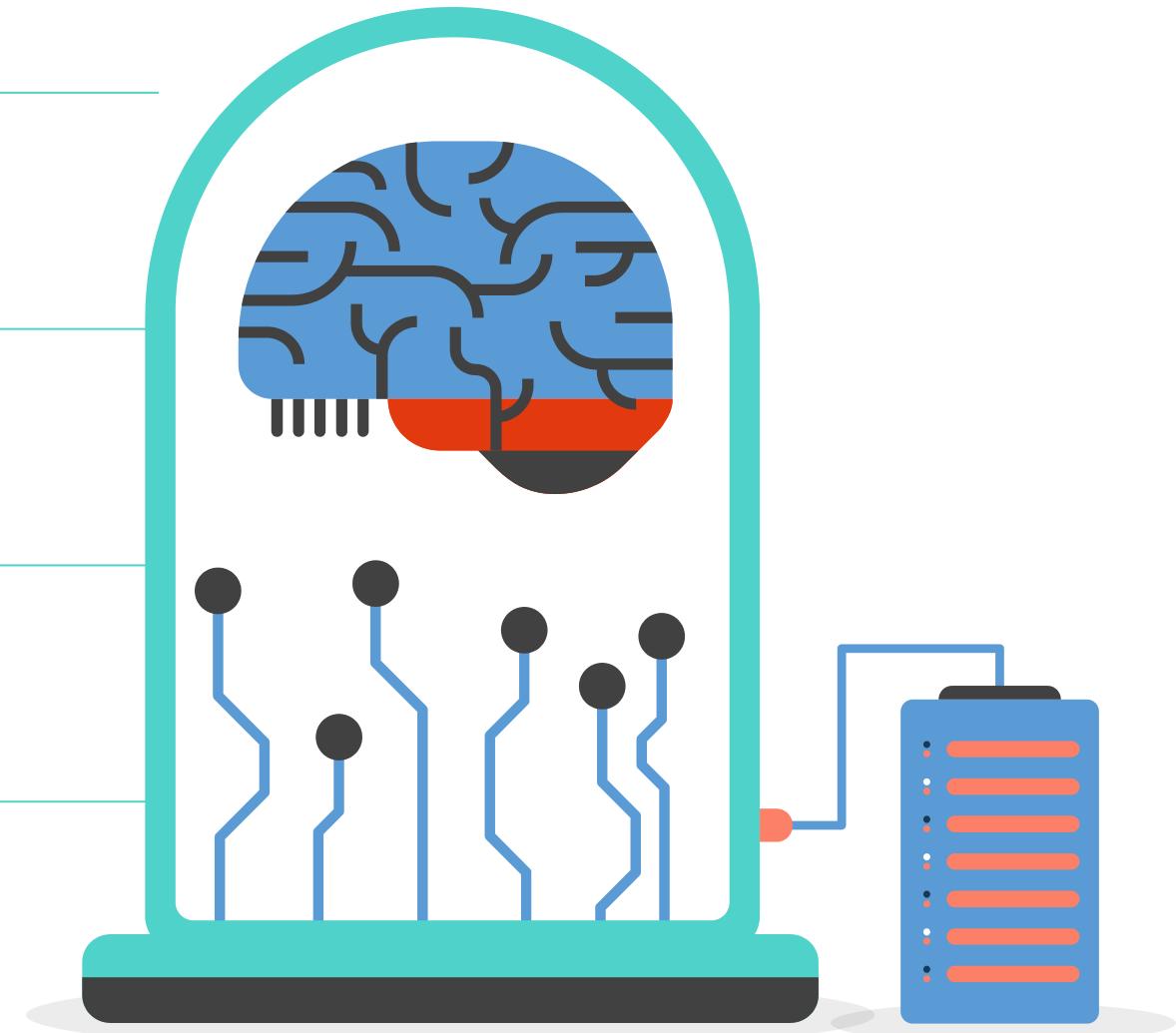
Relevant variable selection

04 AI evaluation

Sensitivity, specificity, errors ...

05 Improve

Continuous process



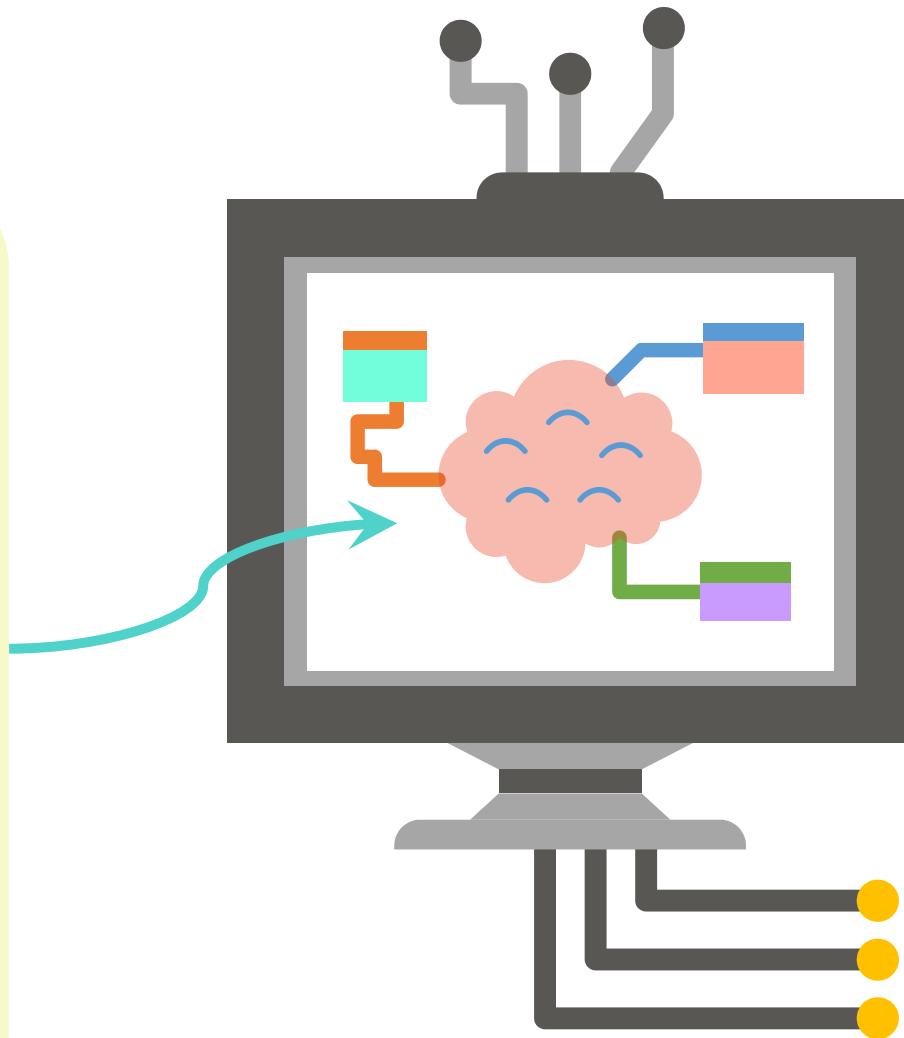
Inputs

Relevant variables

- Speed
- Day or night Arg.
- acceleration
- Date & time
- Turning angle Mod.
- acceleration Change rates

:

INPUTS / OUTPUTS



Outputs

Binary prediction

- Fishing / no Fishing

MODEL TYPES

USED MODELS

Artificial neural networks

Improveable results

Decision trees

Good results

Support-vector machines

Bad results

Random forest

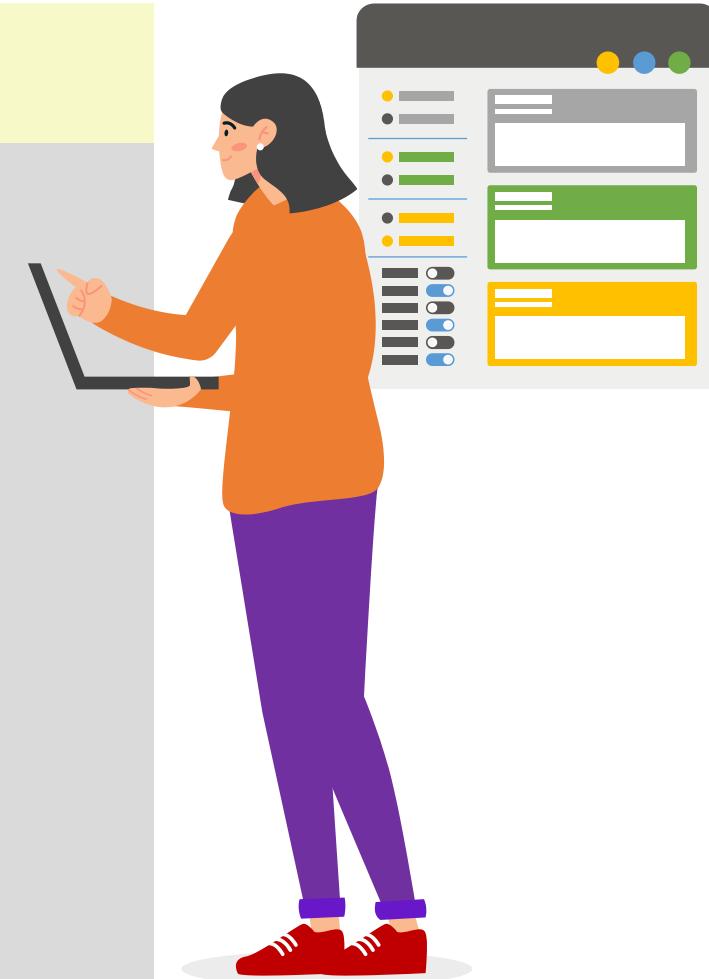
Better results

Bayesian networks

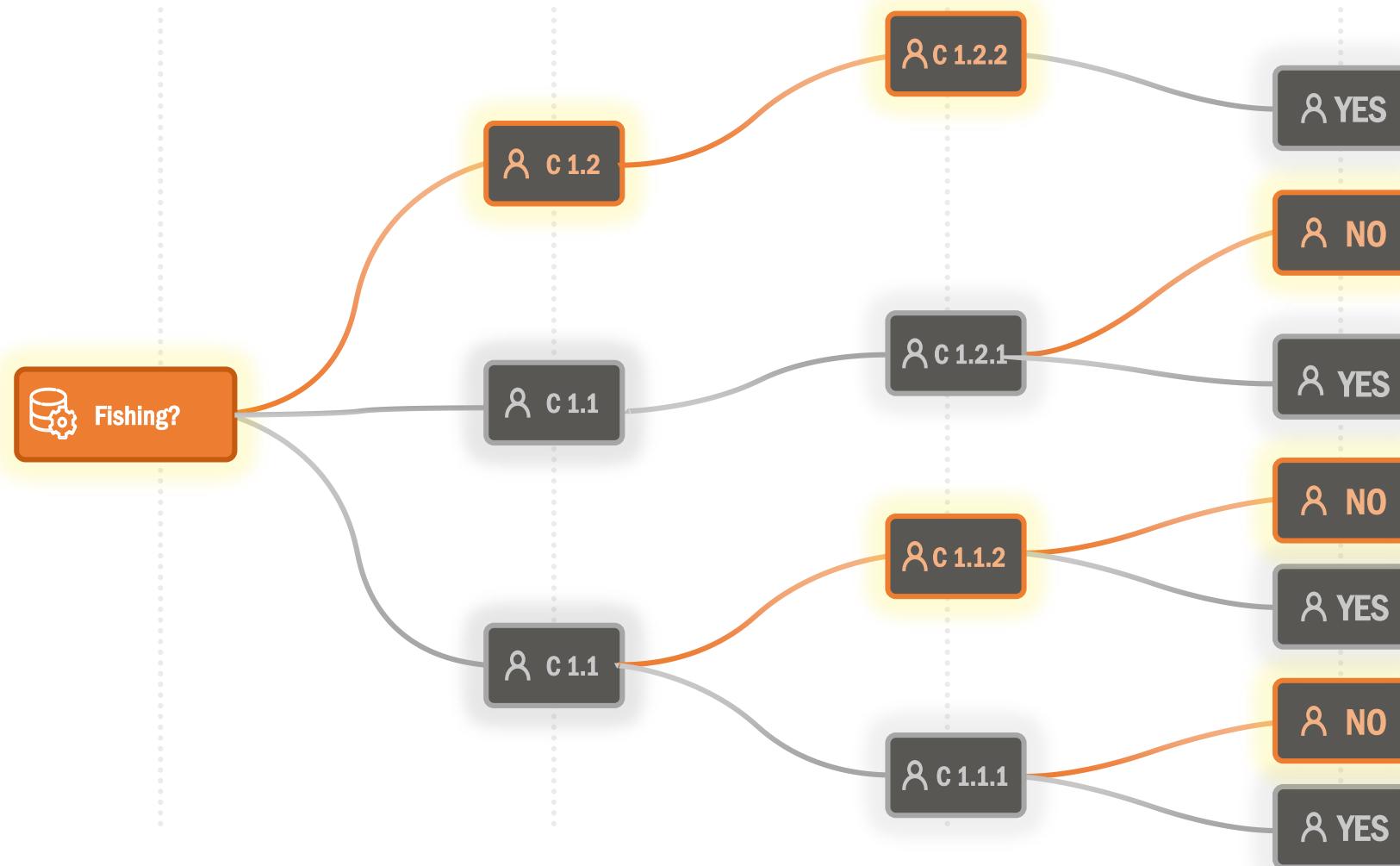
Improveable results

Genetic algorithms

Under development



DECISION TREES AND RANDOM FOREST



Random forest: The decision tree “democratic” version

4

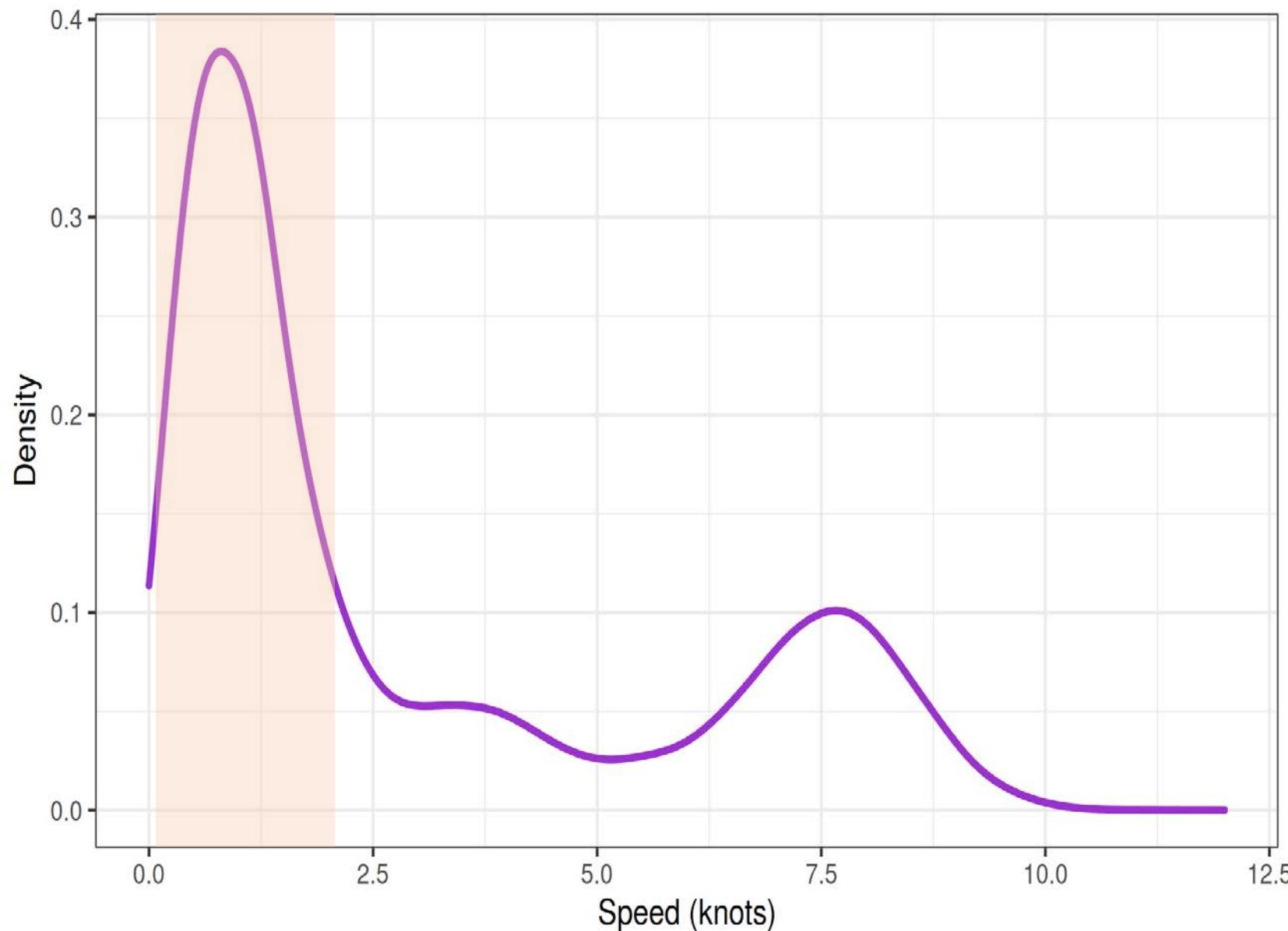
FIRST
RESULTS



SPEED DISTRIBUTION

Bimodal distribution

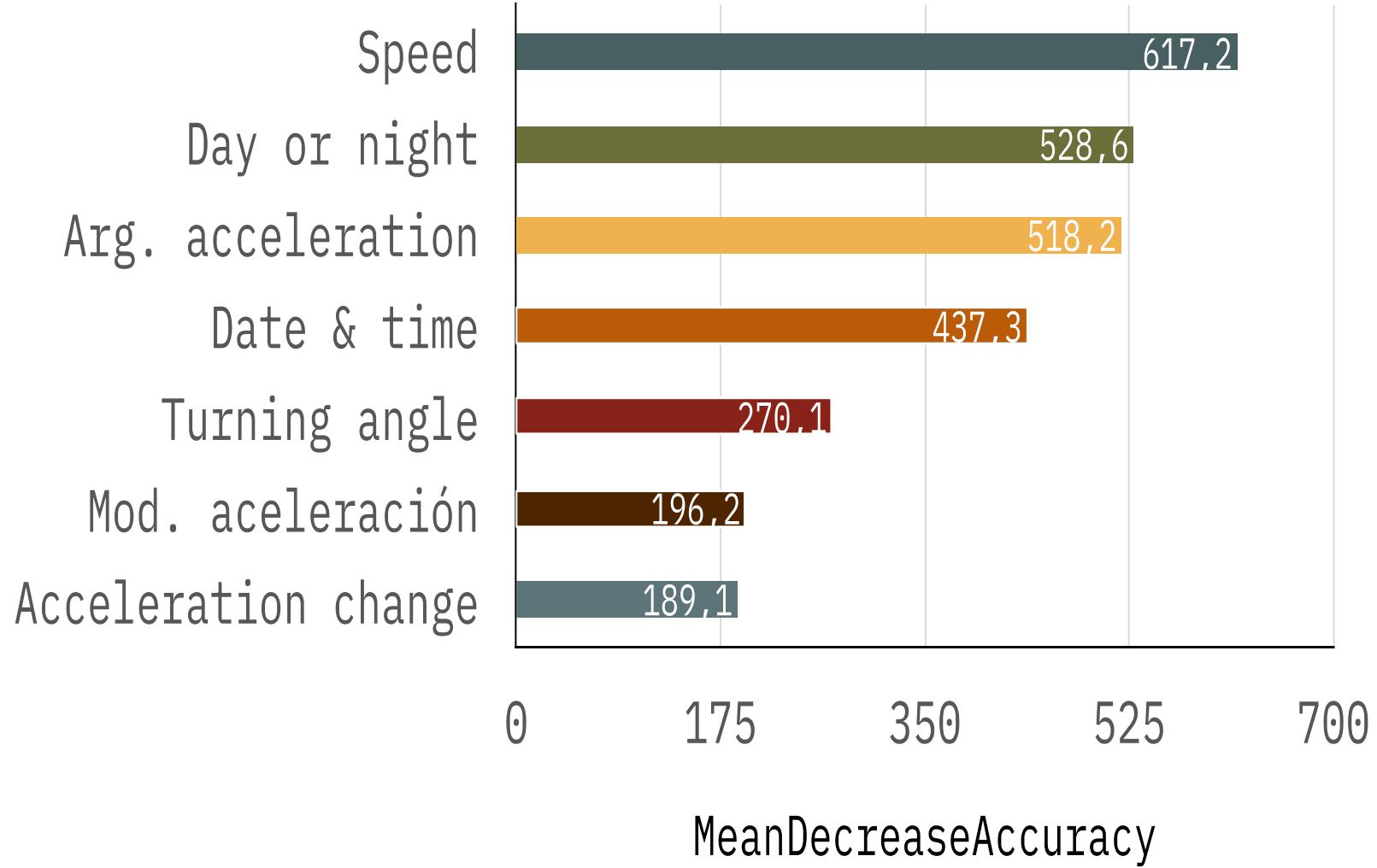
- First mode: High peak for fishing speeds
- Second mode: Sailing speeds.



RELEVANCE OF SELECTED VARIABLES

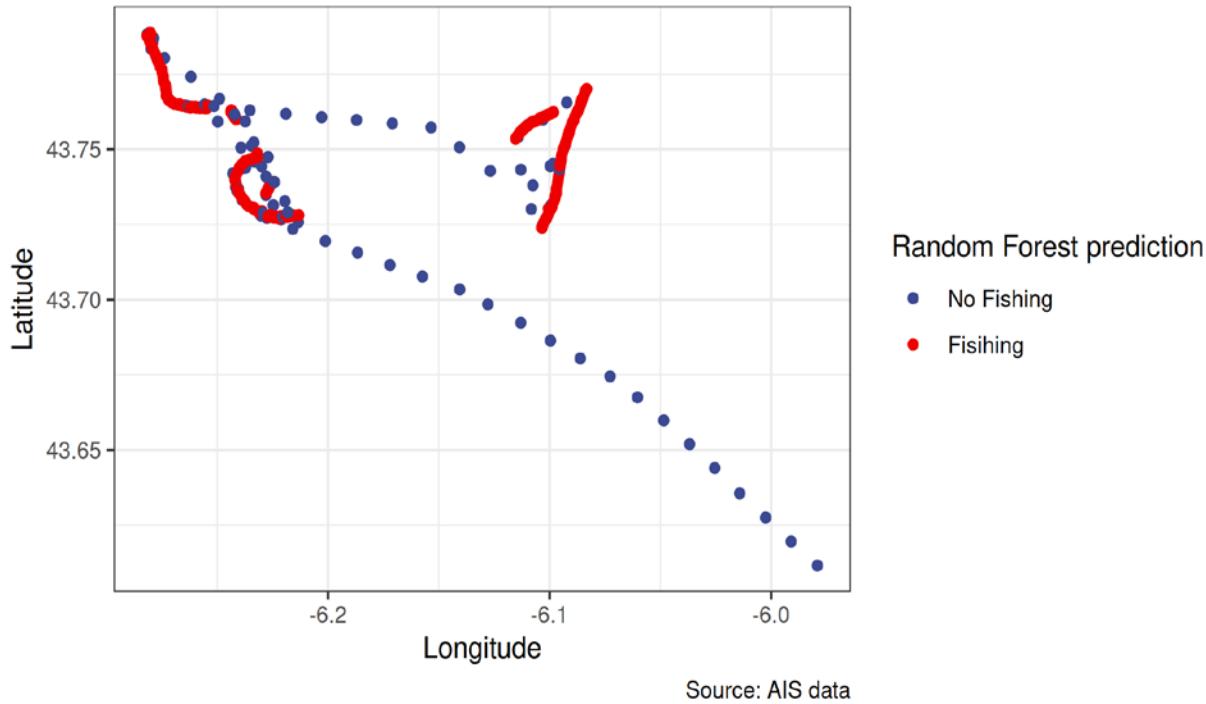
Direct impact over prediction accuracy

- Each variable has a different impact on the accuracy of the model.
- Variable selection is delicate and very iterative.
- Most relevant variables related to speed and date.

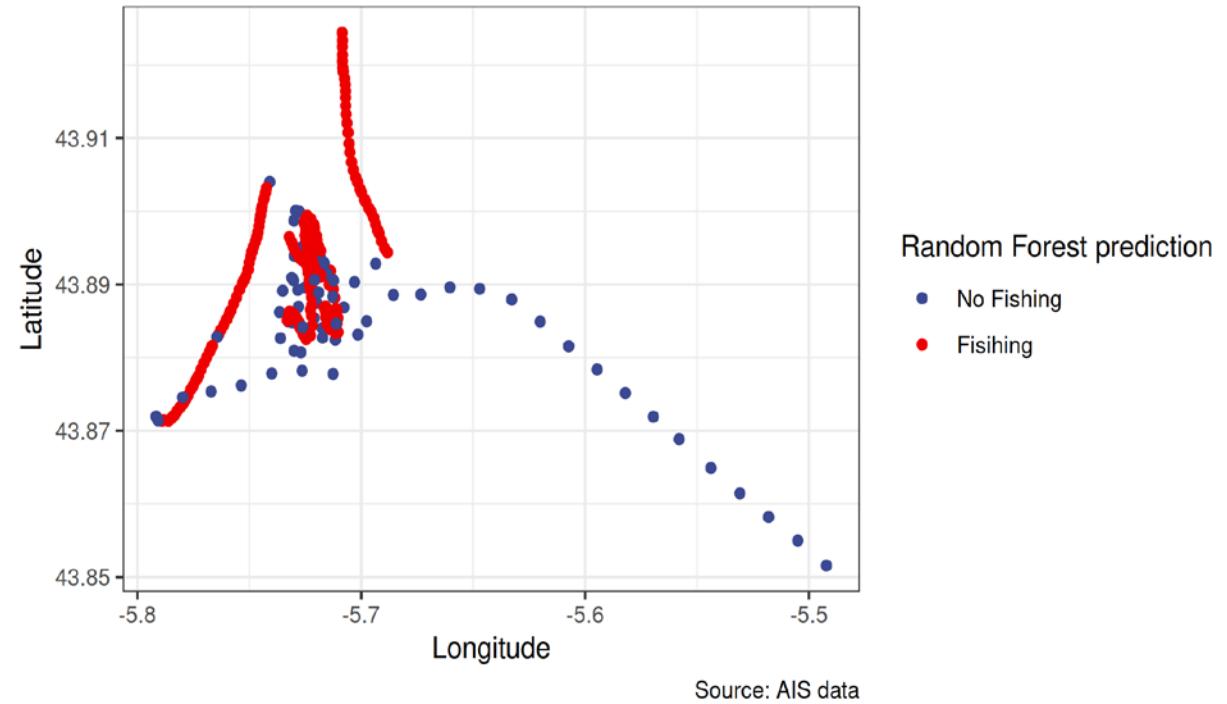


PREDICTIONS

Fishing trip with identified fishing events
Random Forest prediction

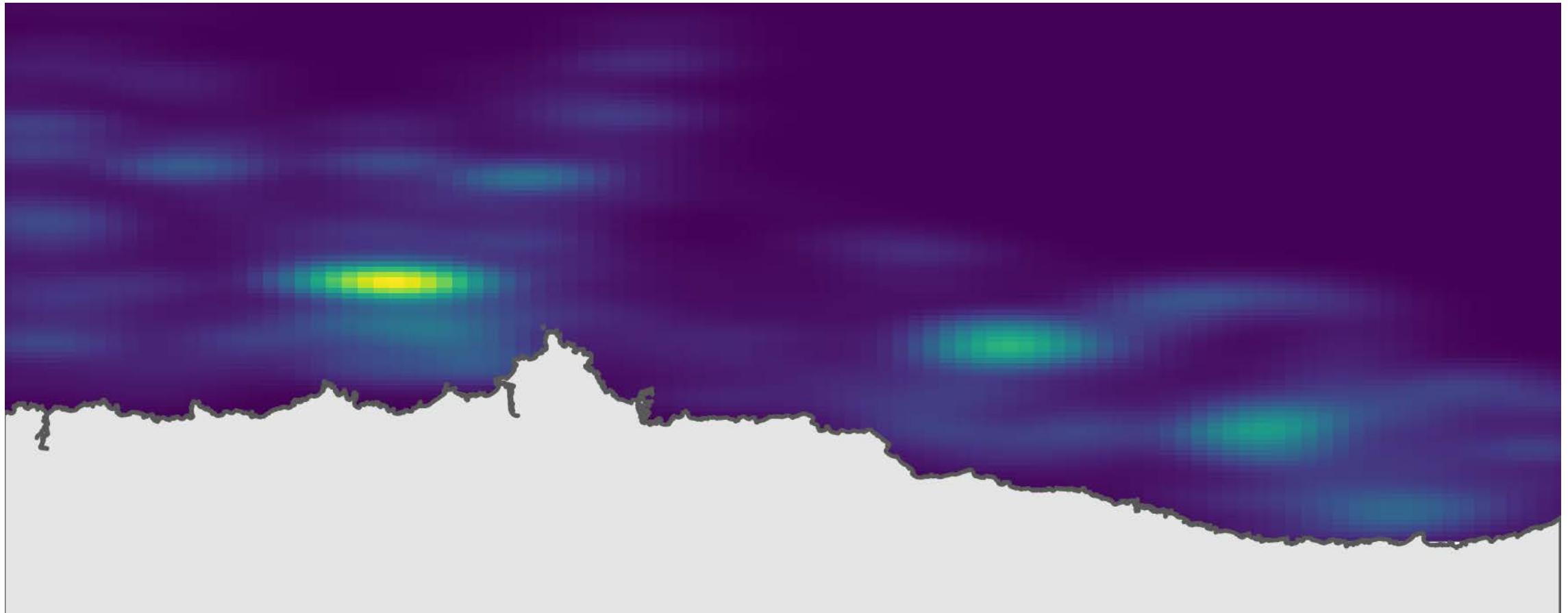


Fishing trip with identified fishing events
Random Forest prediction



Fishing event predictions for two different fishing trips from two different ships and in two different places.

IDENTIFIED FISHING GROUNDS





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Con la colaboración de la Fundación Biodiversidad,
del Ministerio para la Transición Ecológica y el Reto
Demográfico, a través del Programa Pleamar,
cofinanciado por el FEMP

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ROC Curve for Random Forest. AUC = 0.847553477828212

