



# Unlocking the past to predict the future

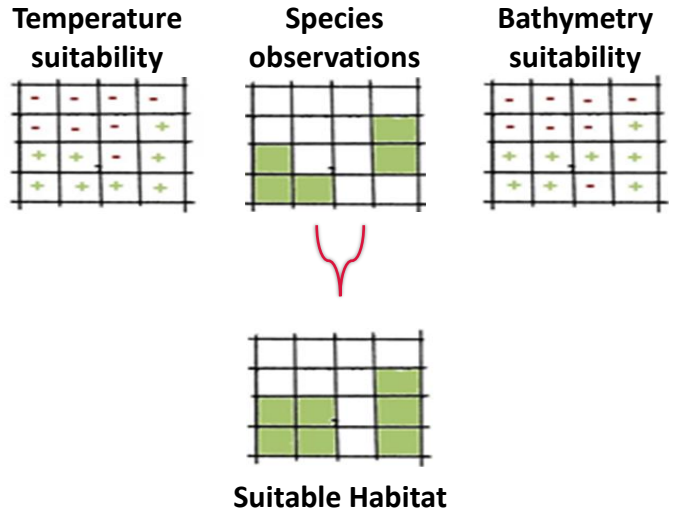
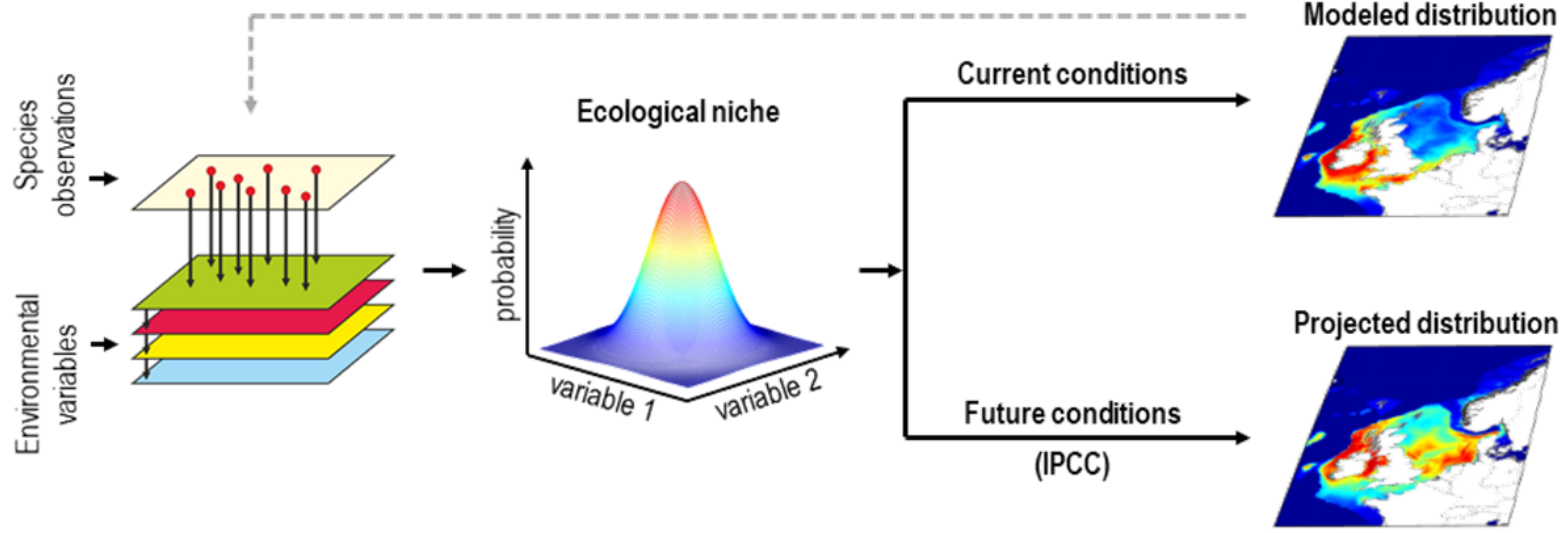
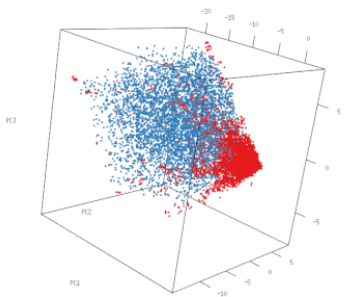
## Enhancing species distribution modelling with long-term data

**Eric Goberville**

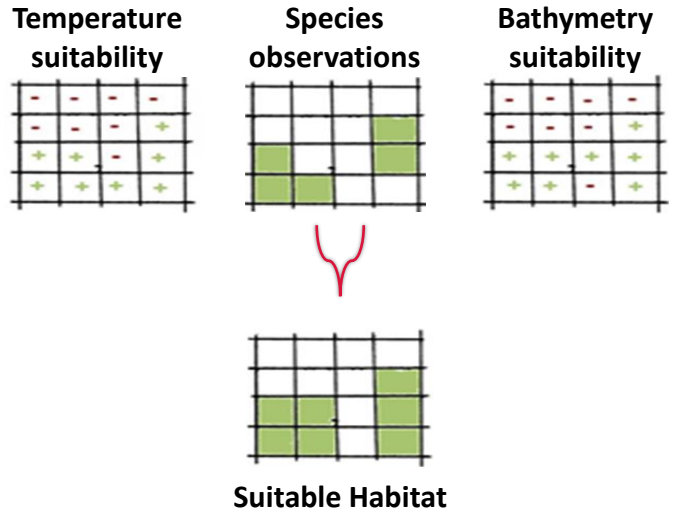
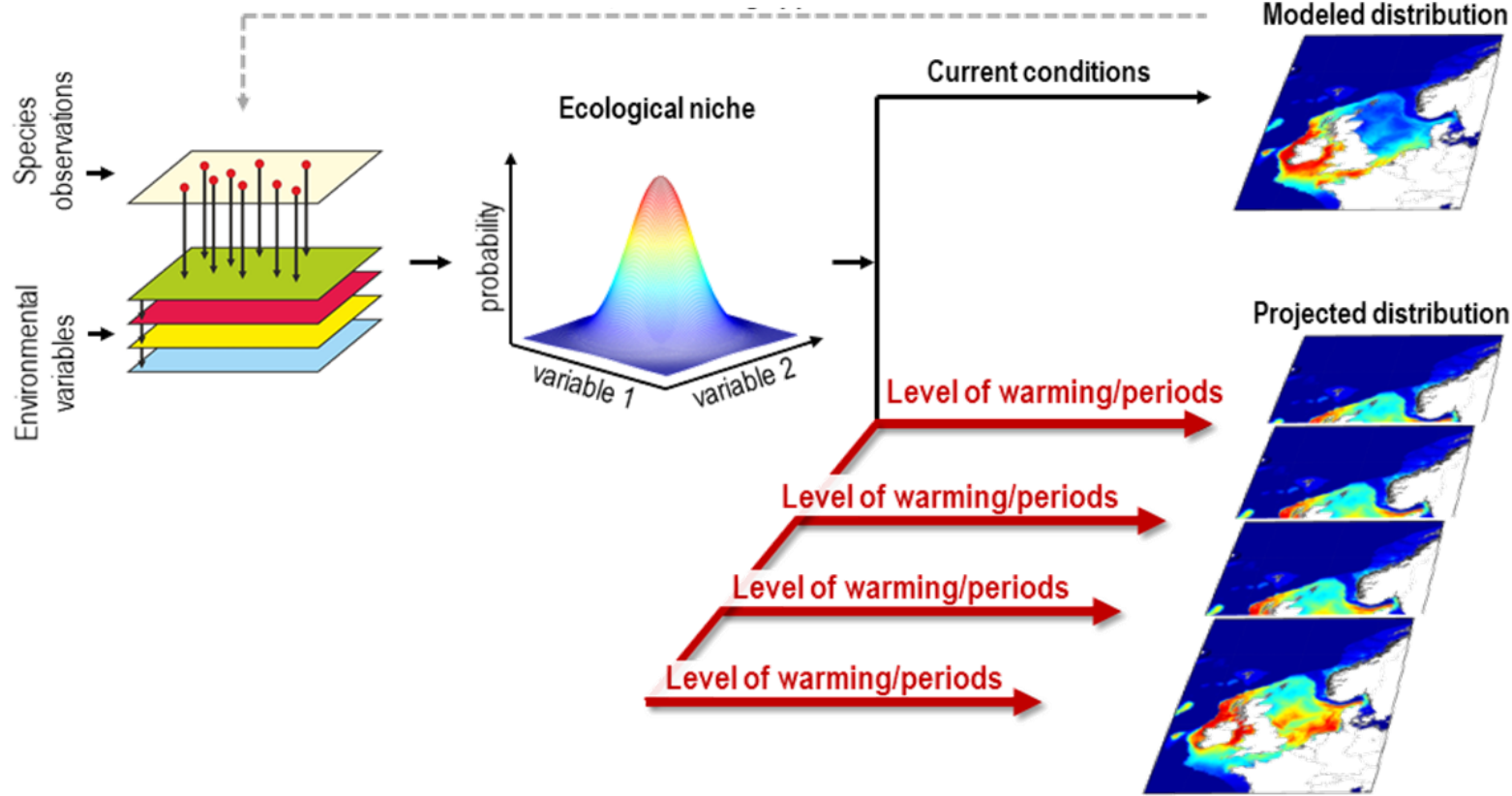
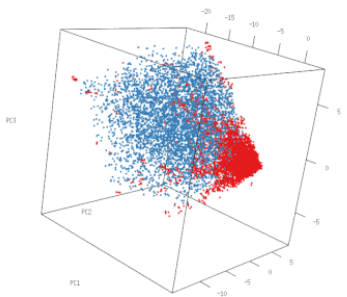
[eric.goberville@sorbonne-universite.fr](mailto:eric.goberville@sorbonne-universite.fr)



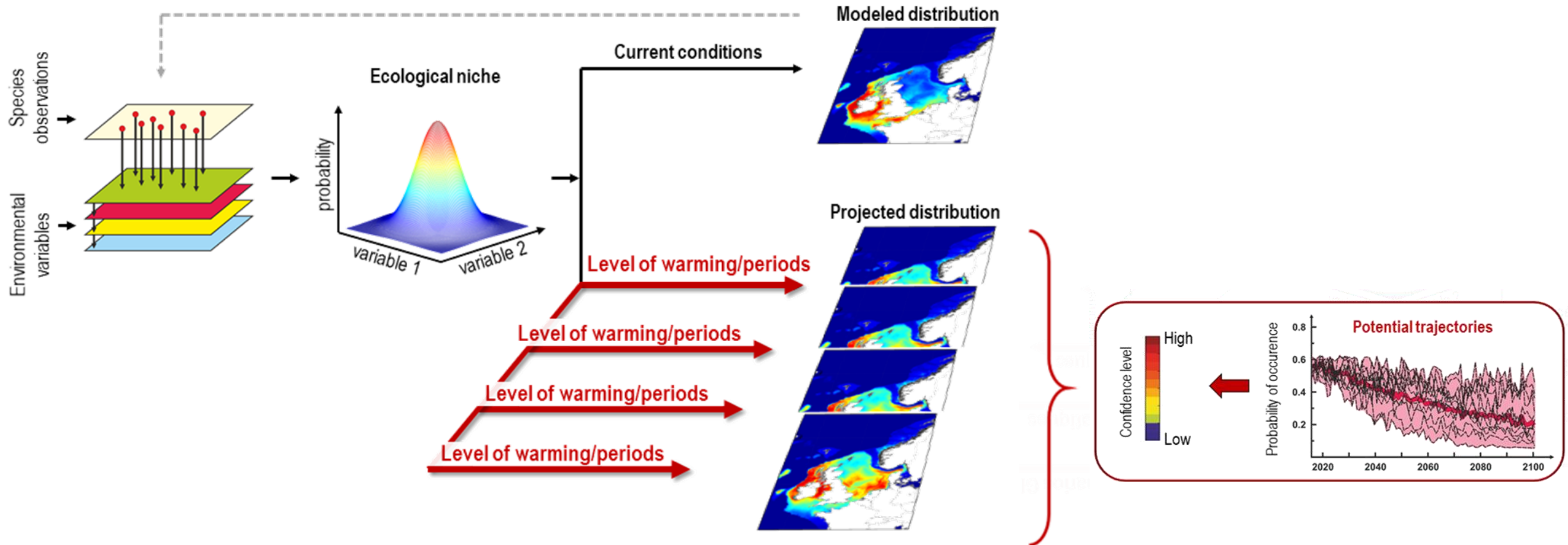
# From species occurrences to niche modelling and biogeographical distributions



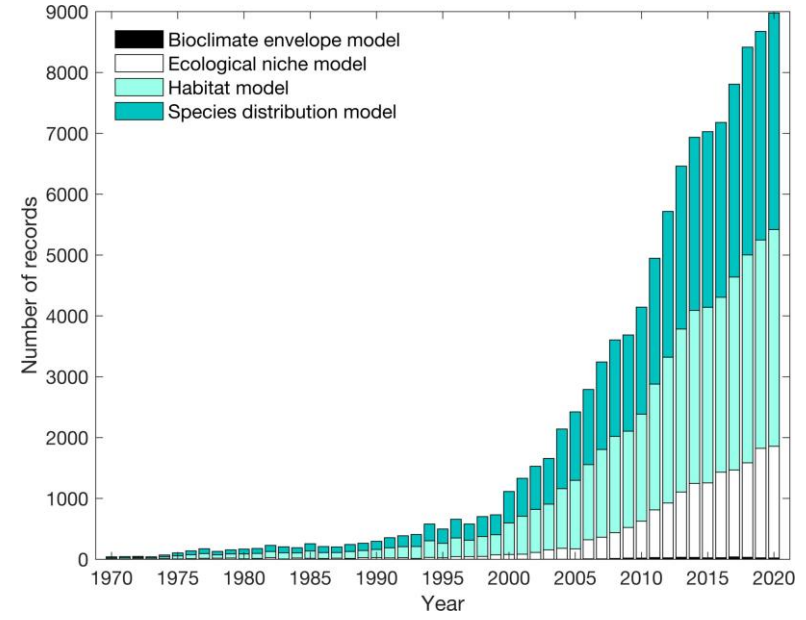
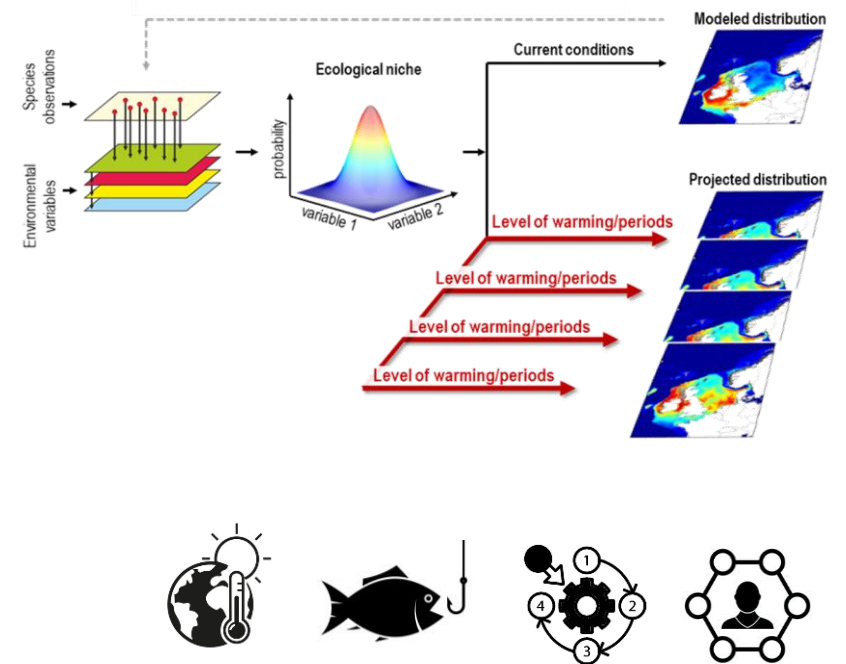
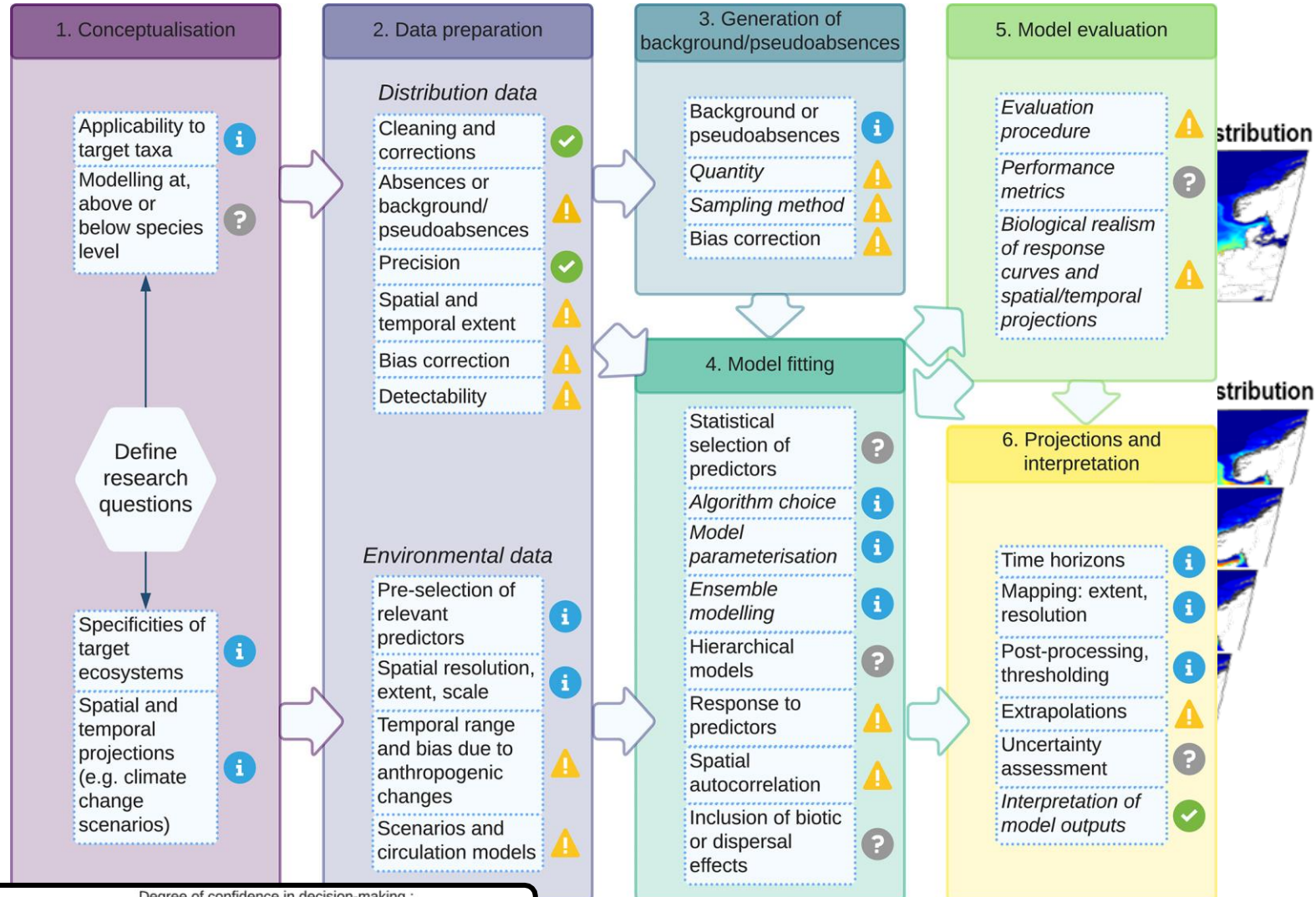
# From species occurrences to niche modelling and biogeographical distributions



# From species occurrences to niche modelling and biogeographical distributions



# The major uncertain steps in the construction of Species Distribution Model

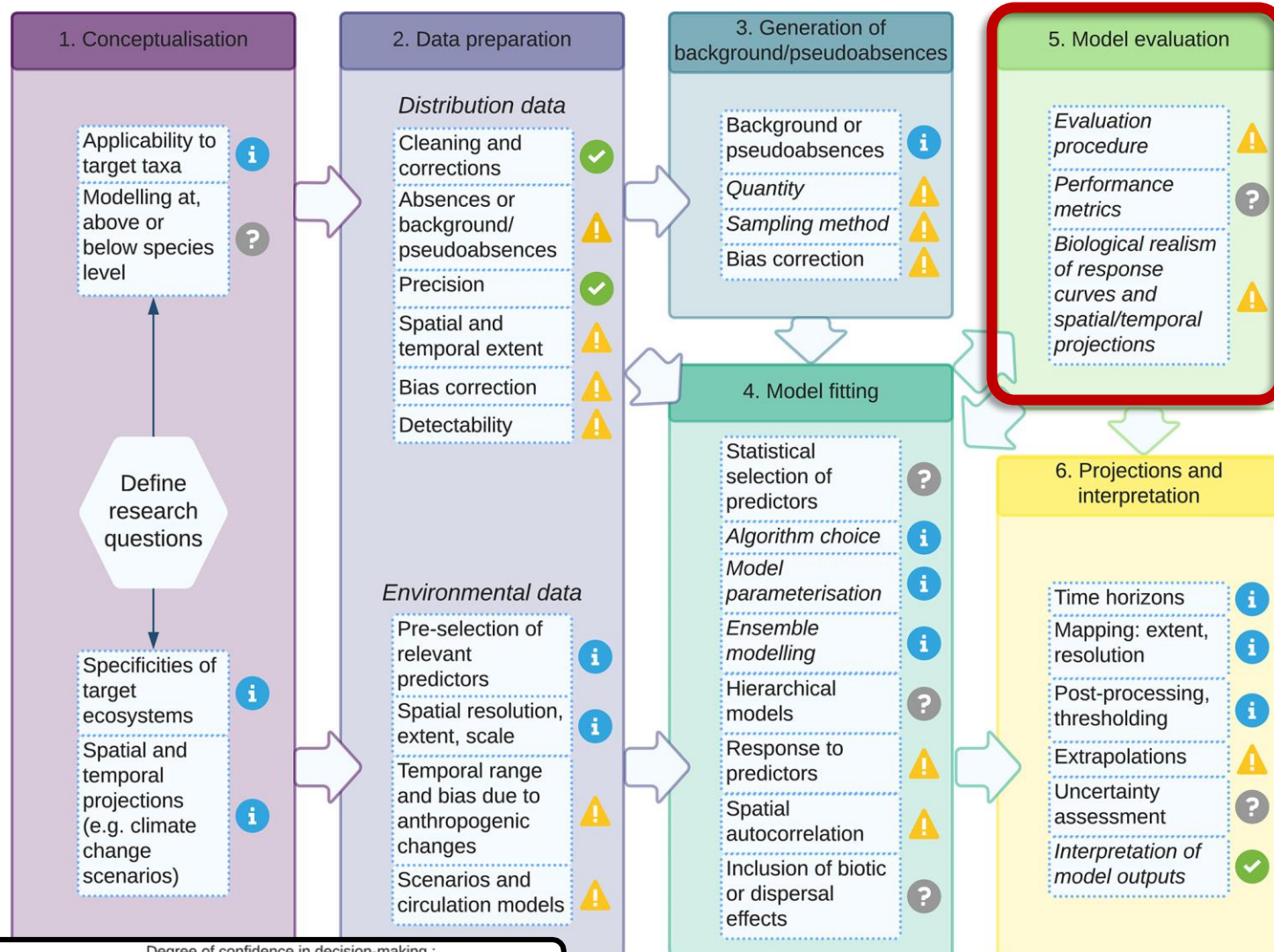


Aspect which will require one or several decisions to make

Degree of confidence in decision-making:

- ✓ Well established
- ⚠ Limited due to the incompleteness of guidelines
- i Established but incomplete guidelines exist, confidence will vary depending on study features
- ⊘ Unresolved

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- Unresolved (?)

Leroy et al. 2023



## AUC: a misleading measure of the performance of predictive distribution models

Jorge M. Lobo<sup>1\*</sup>, Alberto Jiménez-Valverde<sup>1</sup> and Raimundo Real<sup>2</sup>

PERSPECTIVE

WILEY Journal of Biogeography

Without quality presence–absence data, discrimination metrics such as TSS can be misleading measures of model performance

Selecting thresholds of occurrence in the prediction of species distributions

Canran Liu, Pam M. Berry, Terence P. Dawson and Richard G. Pearson

Presence-only species distribution models are sensitive to sample prevalence: Evaluating models using spatial prediction stability and accuracy metrics

Liam Grimm<sup>a,b,\*</sup>, Rachel Whitt<sup>a,b</sup>, Ana Horta<sup>a,b</sup>

<sup>a</sup>School of Environmental Sciences, Charles Sturt University, Albury, Australia

<sup>b</sup>Institute for Land, Water and Society, Charles Sturt University, Albury, Australia

BIODIVERSITY REVIEW

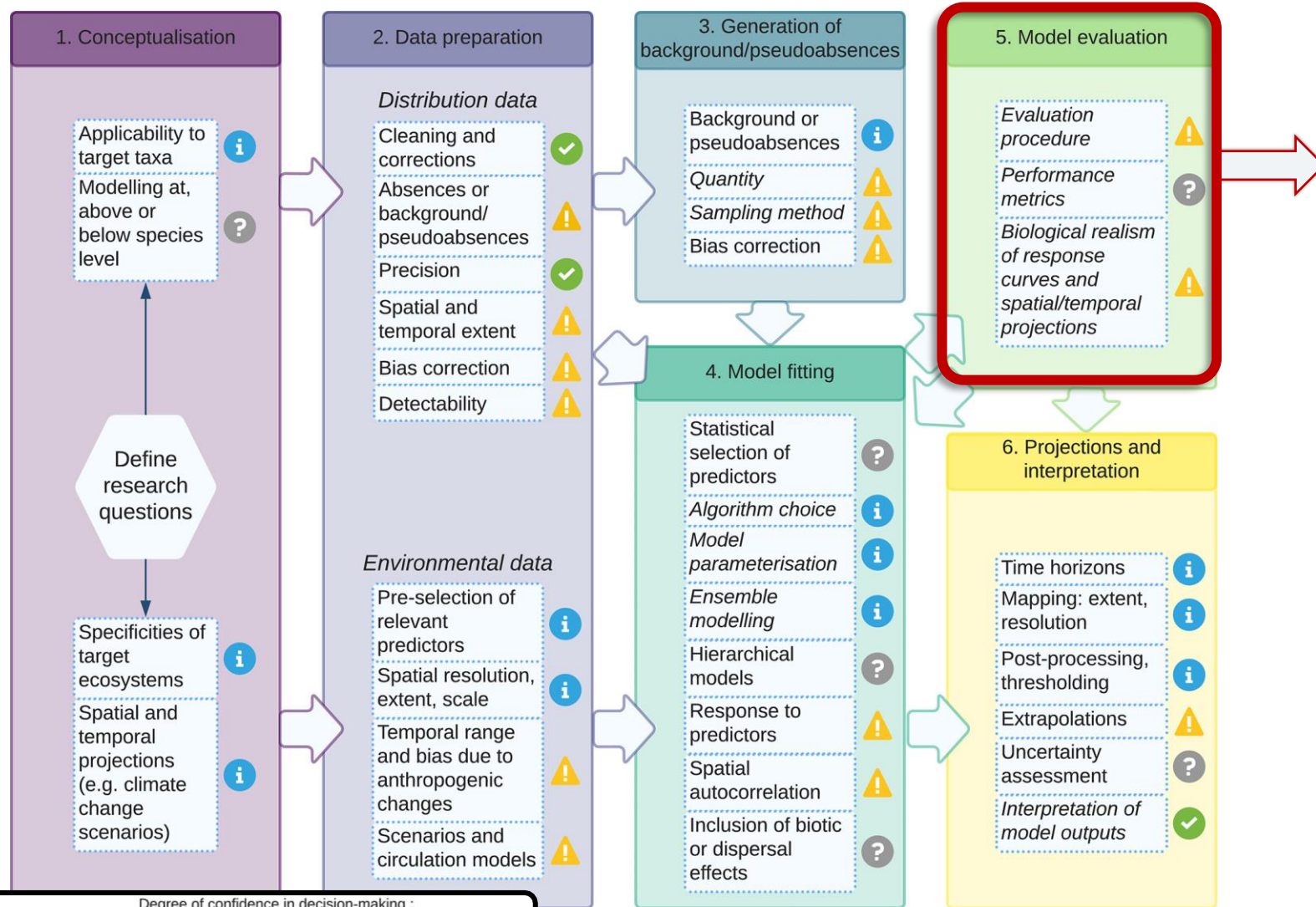
WILEY Diversity and Distributions

A review of evidence about use and performance of species distribution modelling ensembles like BIOMOD

Tianxiao Hao<sup>id</sup> | Jane Elith<sup>id</sup> | Gurutzeta Guillera-Arroita<sup>id</sup> | José J. Lahoz-Monfort<sup>id</sup>

And much more...

# The major uncertain steps in the construction of Species Distribution Model



How can long-term time series be used for the evaluation of niche modelling?



Aspect which will require one or several decisions to make

Degree of confidence in decision-making :

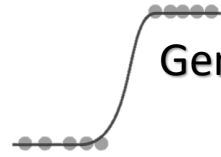
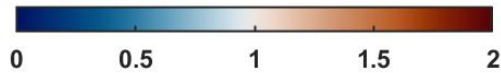
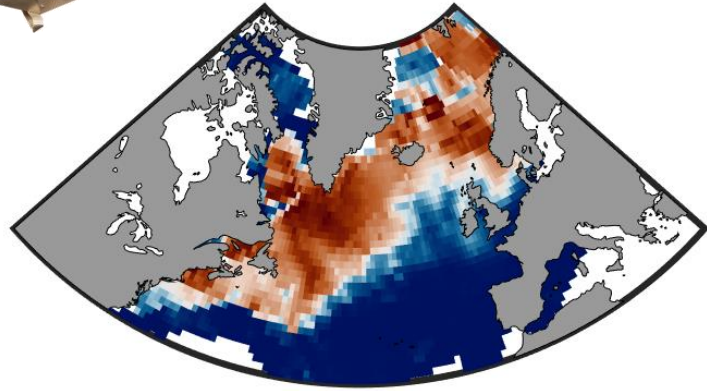
- Well established (✓)
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- Unresolved (?)

# Using long-term time series to model species distribution

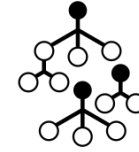
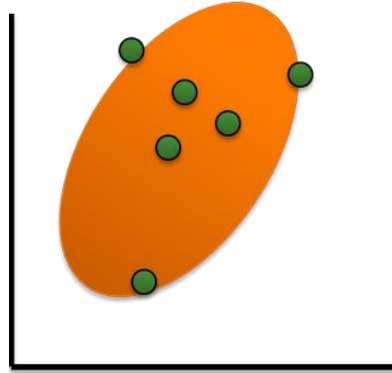
*Calanus finmarchicus*, “The Rolls Royce of zooplankton”<sup>©</sup>



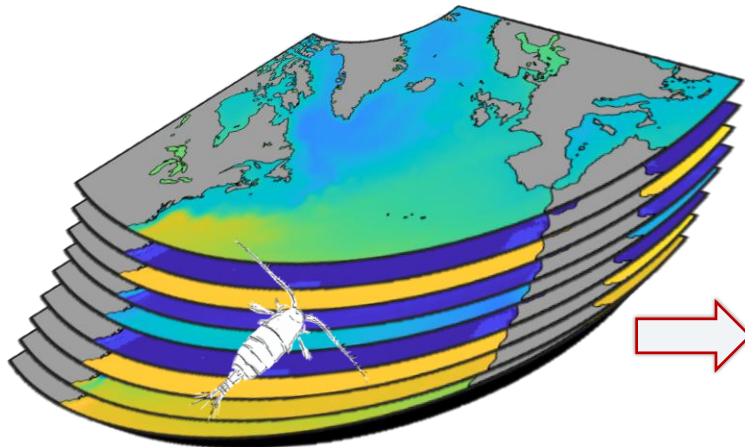
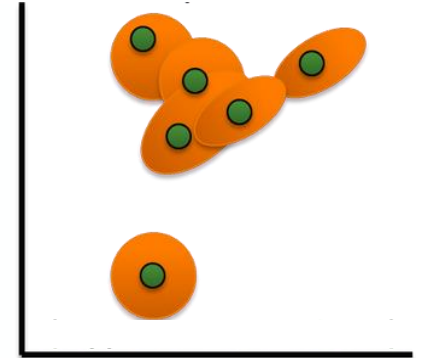
Observations



Generalized linear model



Random Forest



13 environmental parameters reflecting monthly thermohaline conditions, primary production, stratification, oceanic/atmospheric circulation, from 1958 onwards

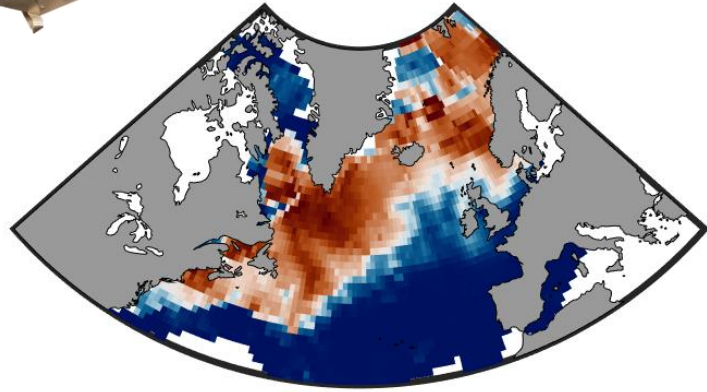


# Using long-term time series to model species distribution

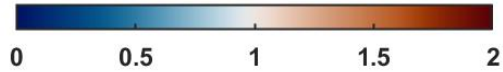
*Calanus finmarchicus*, “The Rolls Royce of zooplankton”<sup>©</sup>



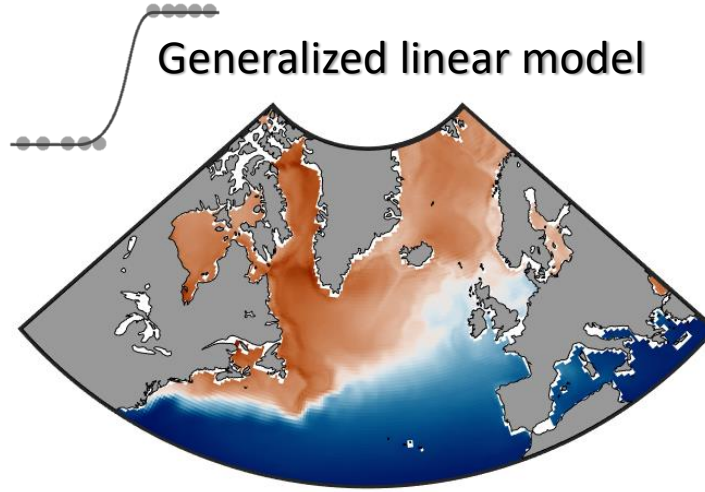
Observations



Abundances (log X+1)



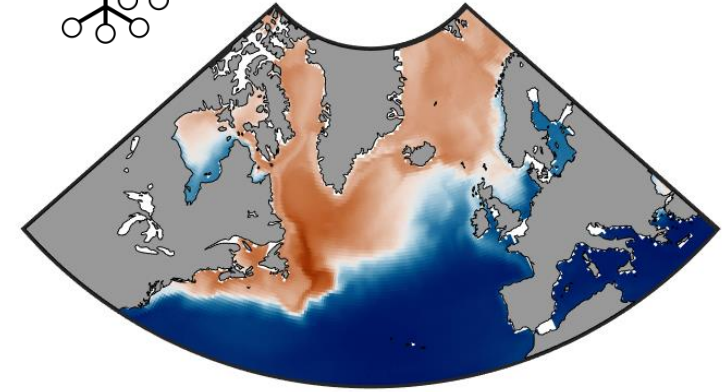
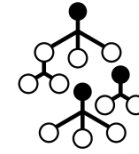
Generalized linear model



Environmental suitability



Random Forest



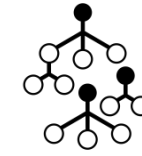
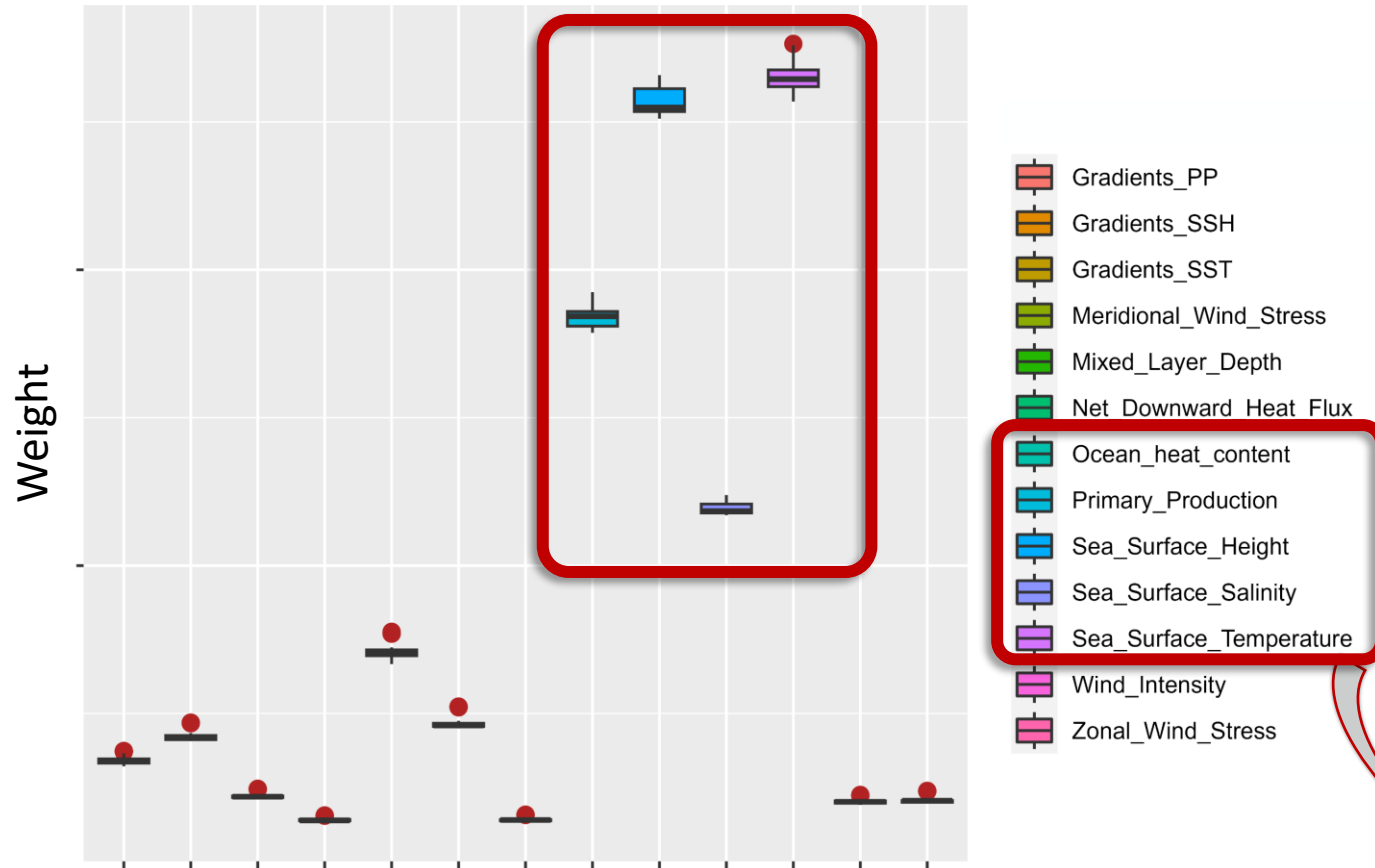
Environmental suitability



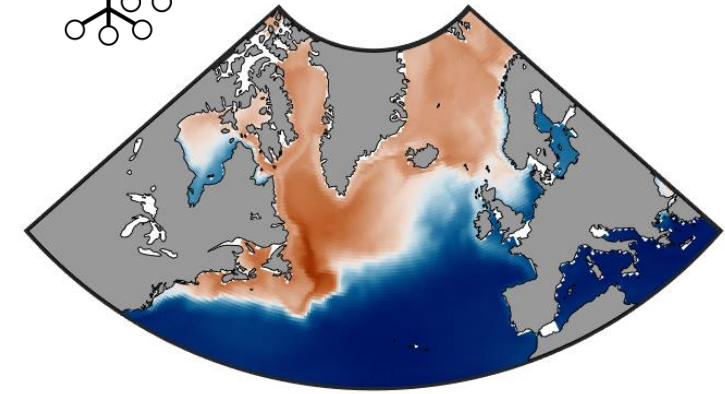
# Using long-term time series to model species distribution

*Calanus finmarchicus*, “The Rolls Royce of zooplankton”<sup>©</sup>

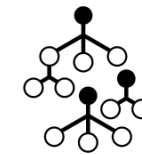
Relative importance of explanatory variables



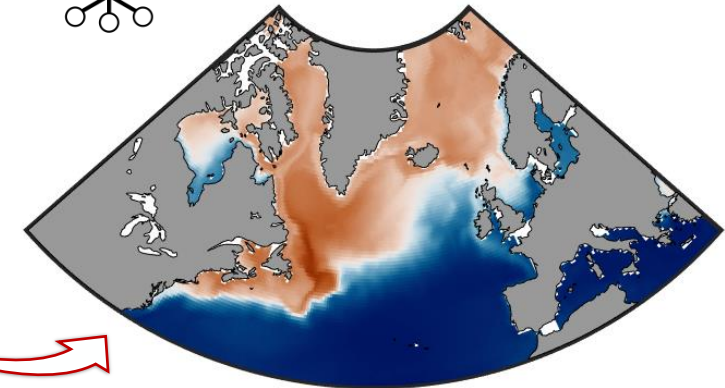
Random Forest



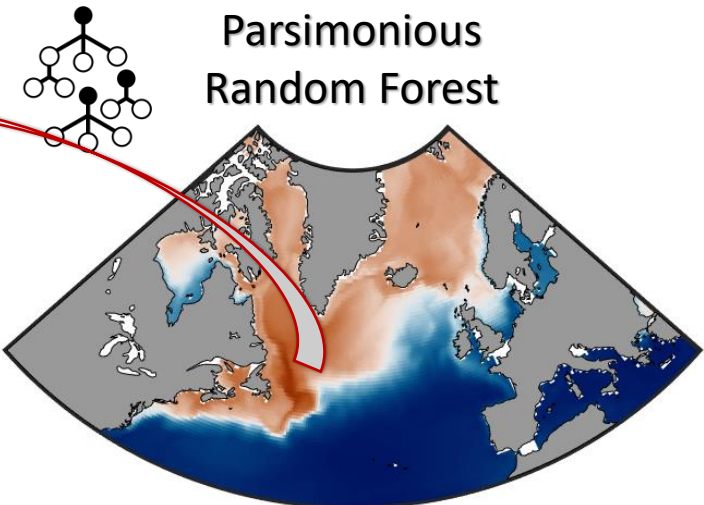
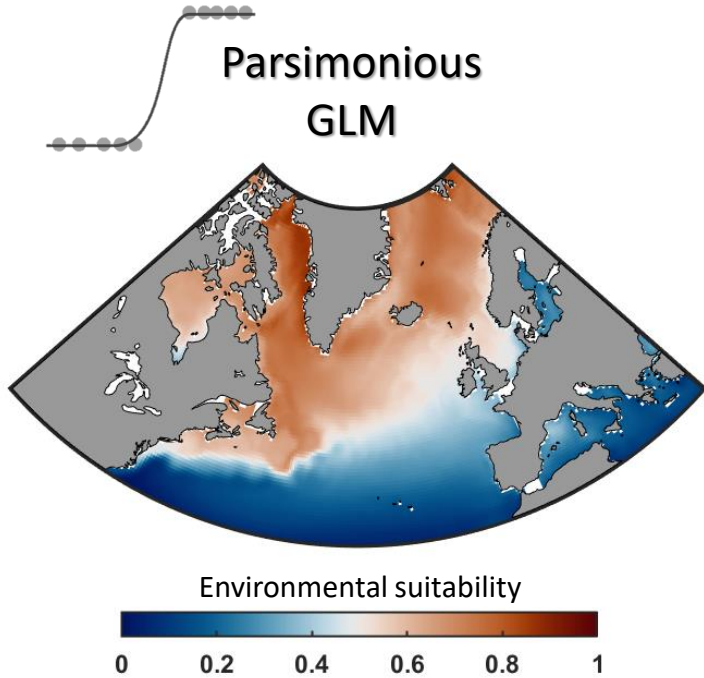
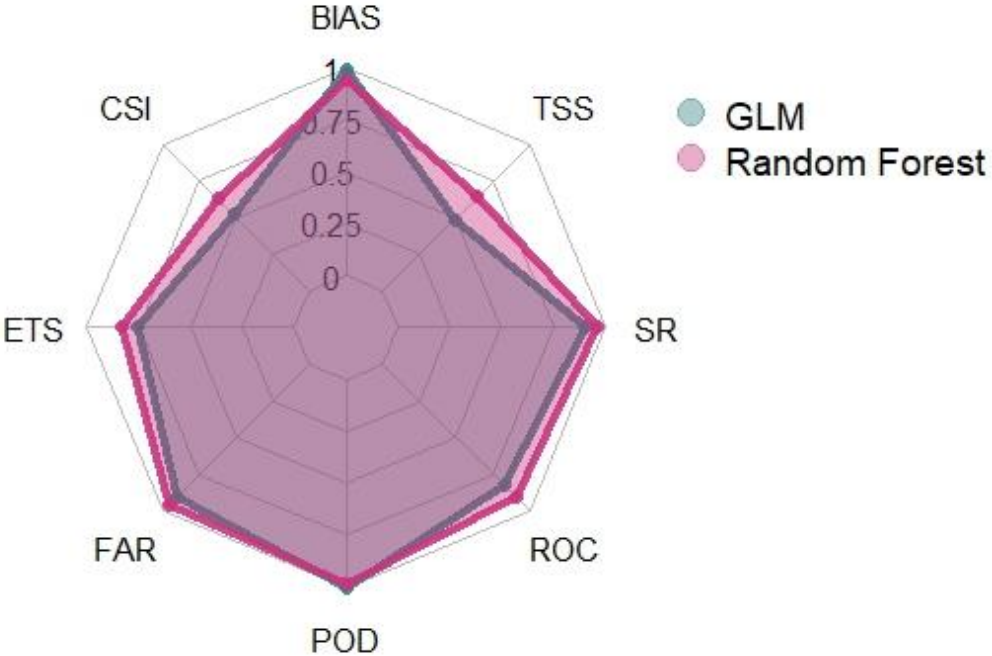
Environmental suitability



Parsimonious Random Forest



# Assessing the robustness of models using classical evaluation metrics



Observations

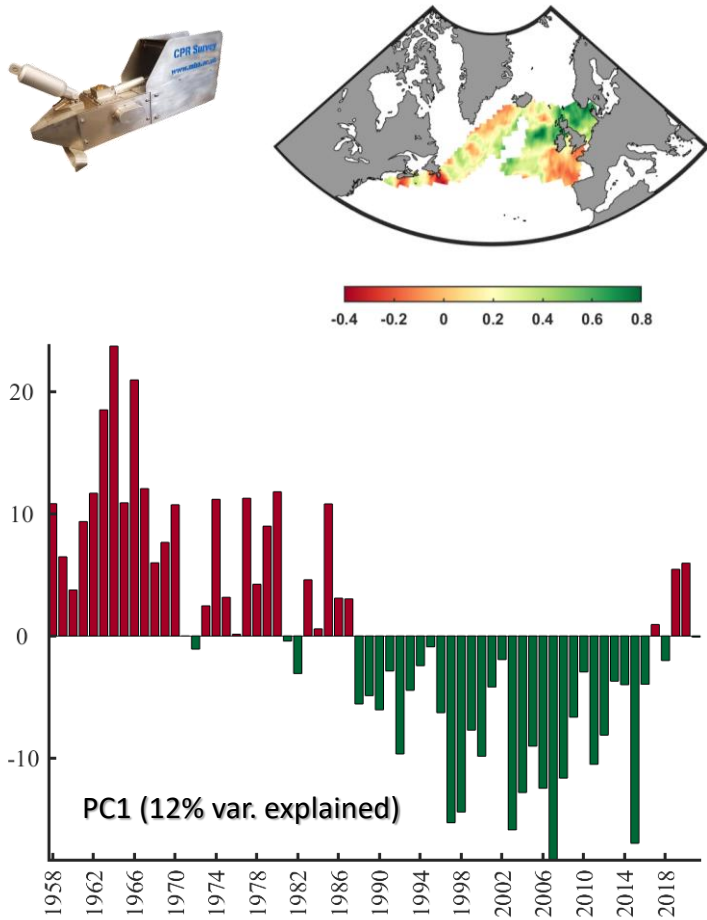
	1	0
1	tp	fp
0	fn	tn

Predictions

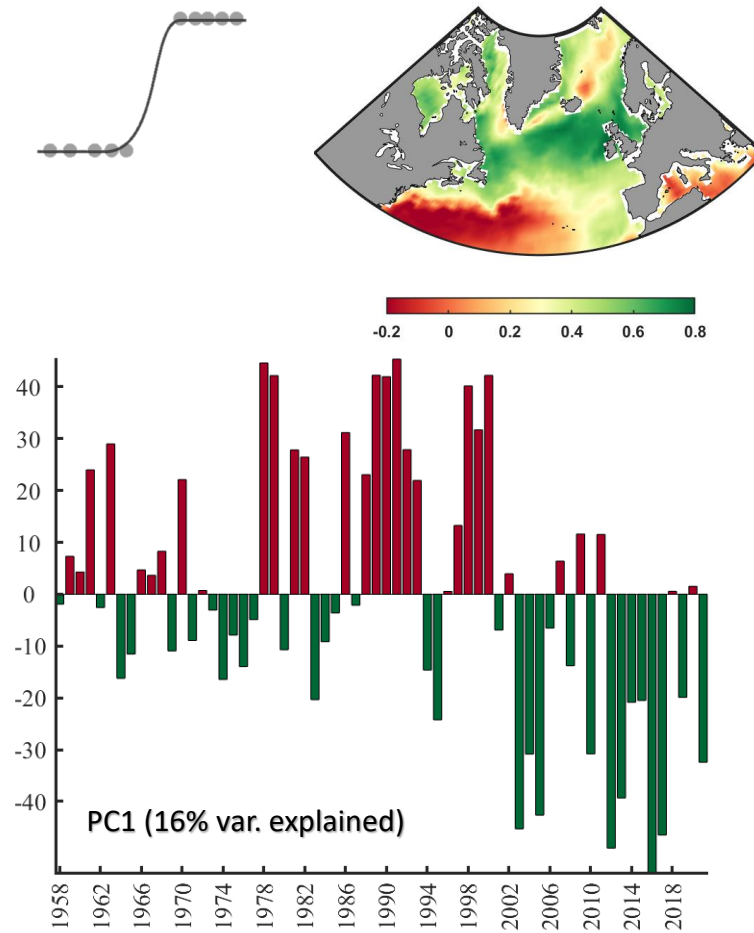


# Assessing the robustness of models using long-term time series ... Reconstructing the past for a better evaluation...

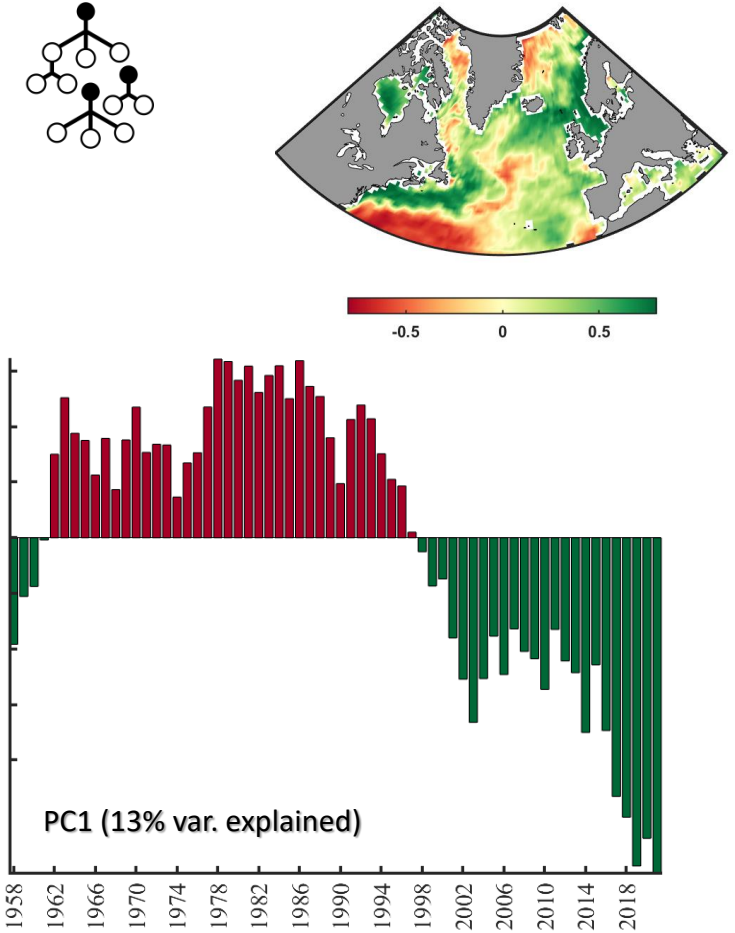
## Observations



## Generalized linear model

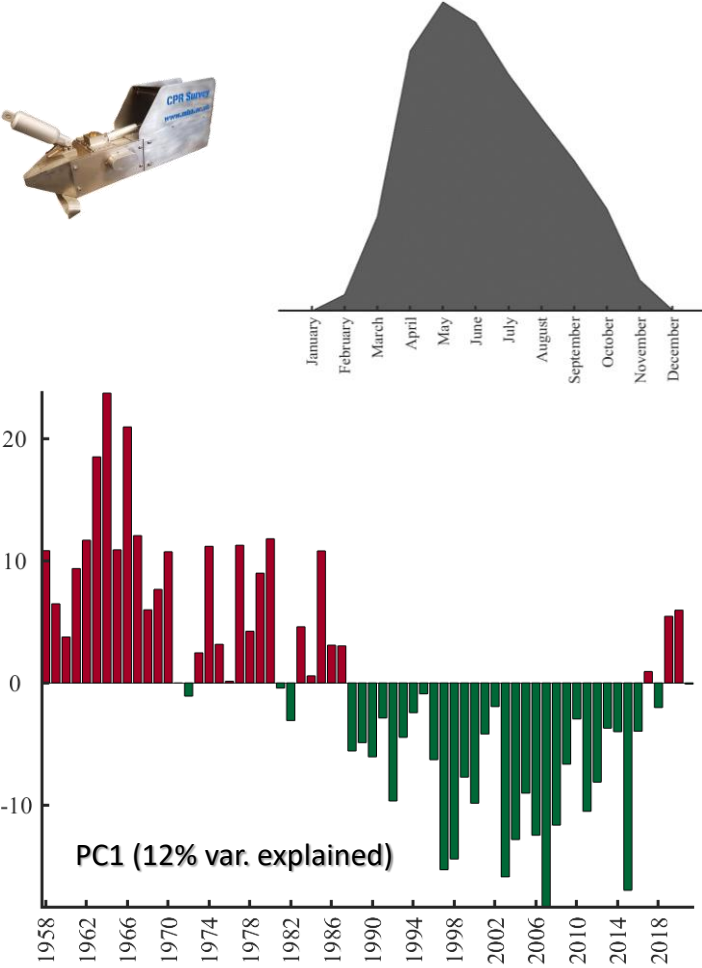


## Random Forest

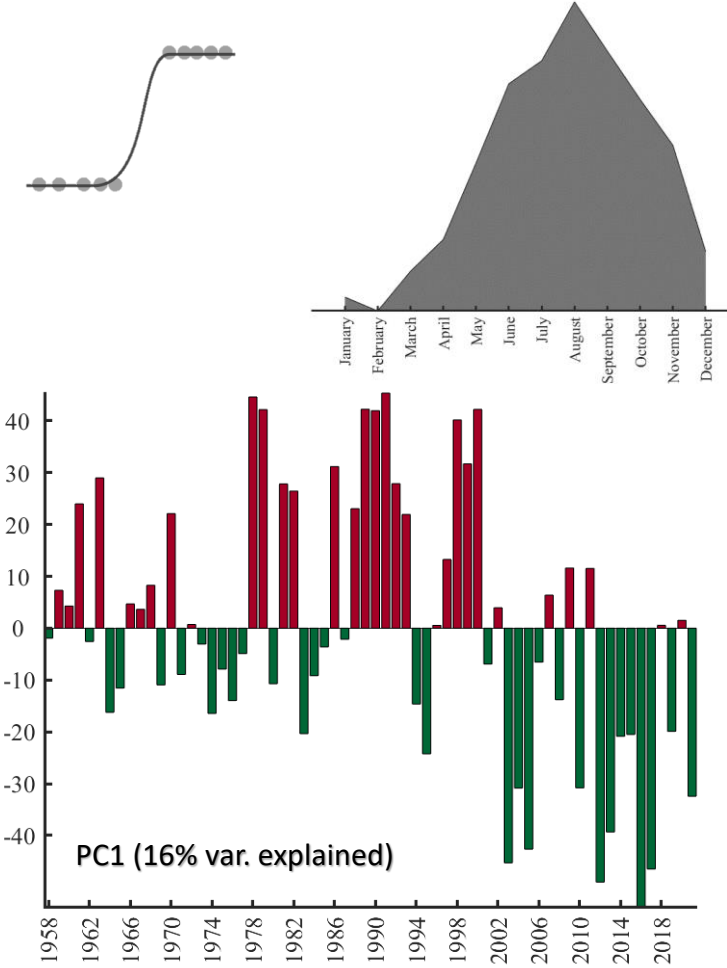


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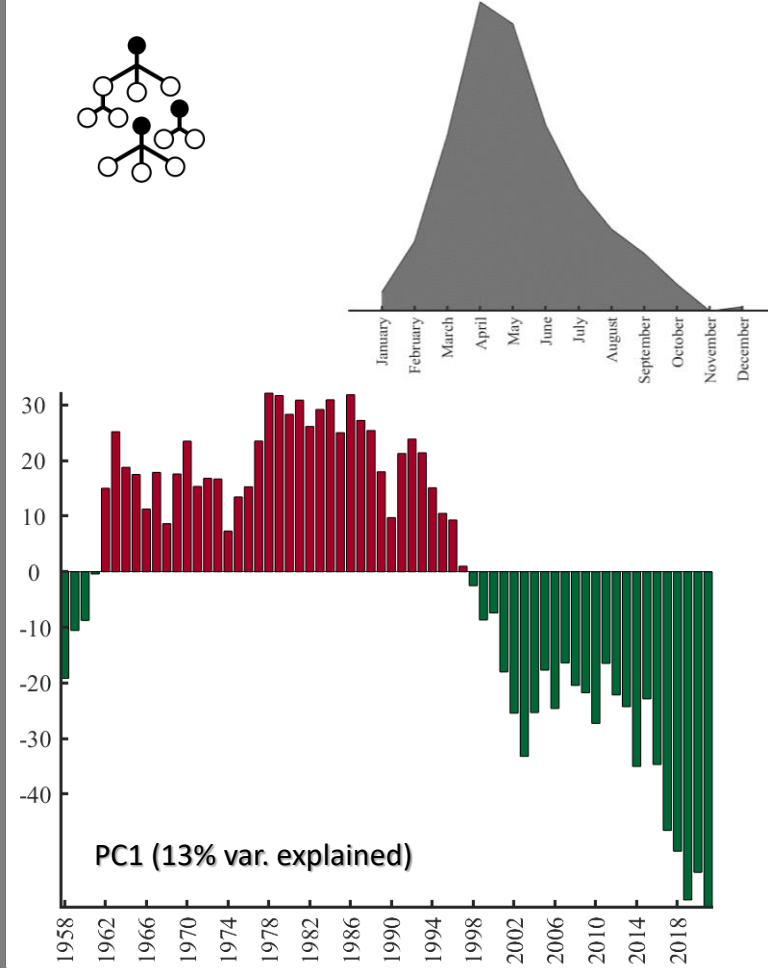
### Observations



### Generalized linear model



### Random Forest





## Take home message



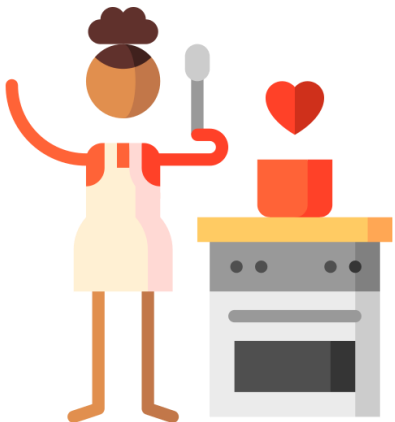
**“... All models are wrong, but some are useful ...**

George Box ... a sentence overly used in modelers' presentations during conferences...

**... The practical question is, how wrong do they have to be to not be useful ...**

**One way to test a model's forecasting capacity is to assess its accuracy through hindcasting ... relying more heavily on historical time series**

To take into account data from past ecosystem



- To implement additional algorithms to explore their sensitivity
- To consider the weight of abundances, not just presence/absence
- To broaden the spectrum of species

