



# **Marine Ecological Modelling Global Climate Change**

## **Principles of Ecological Niche Modelling**

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# Ecological Niche Modelling\*\*

Process of using **computer algorithms to estimate and predict the relationship between the distribution of biodiversity and the environmental conditions.**

Provides insights about **species environmental tolerances and habitat preferences**, and allows **making spatial predictions** of geographical distributions.

\*\* also known as environmental niche modelling, species distribution modelling, habitat distribution modelling, ...



# Main approaches in ENM

## Mechanistic modelling

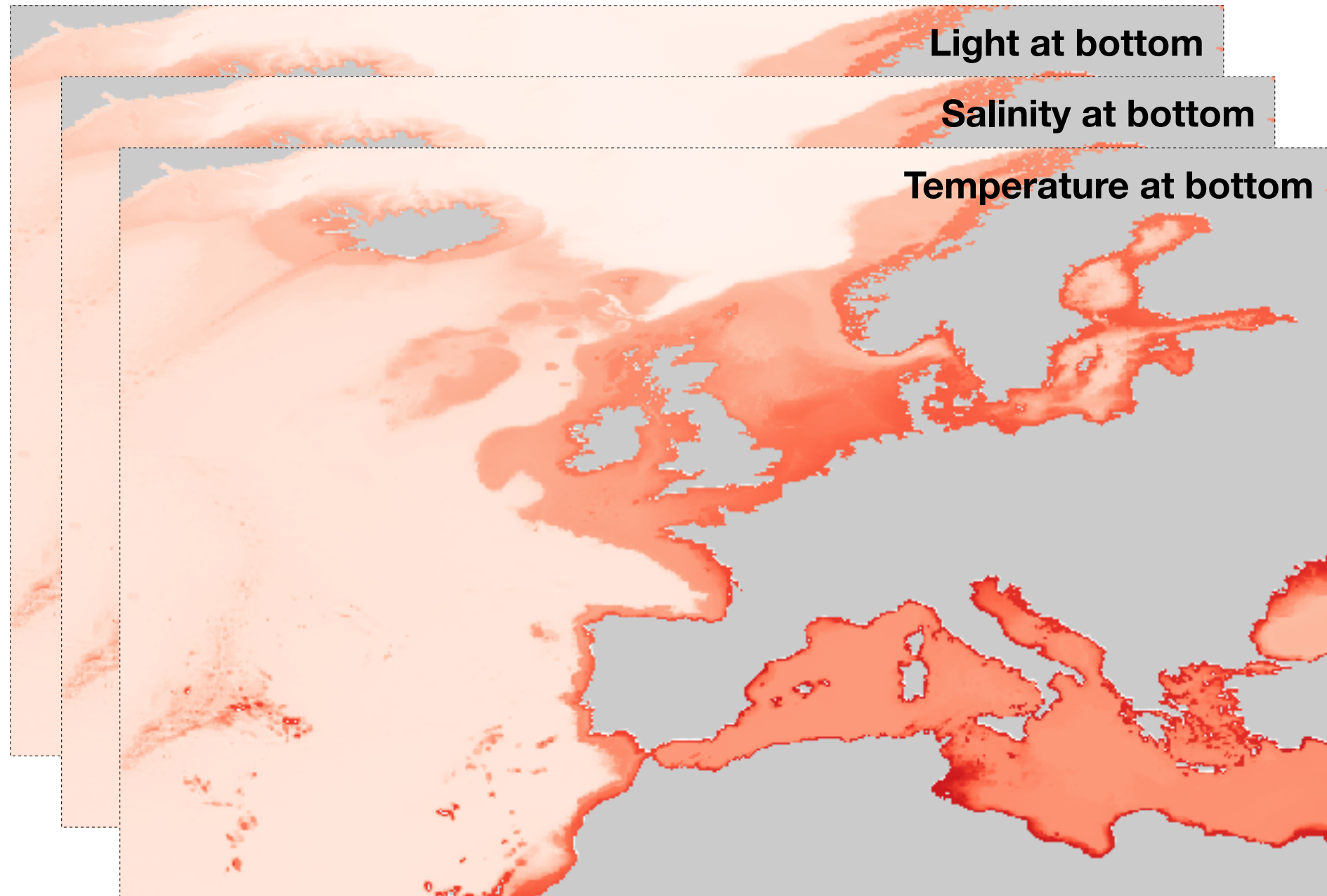
**Uses information about the physiological response of species to environmental conditions.**

(e.g., needs data on the effect of temperature on species survival; not always available).

## Correlative modelling

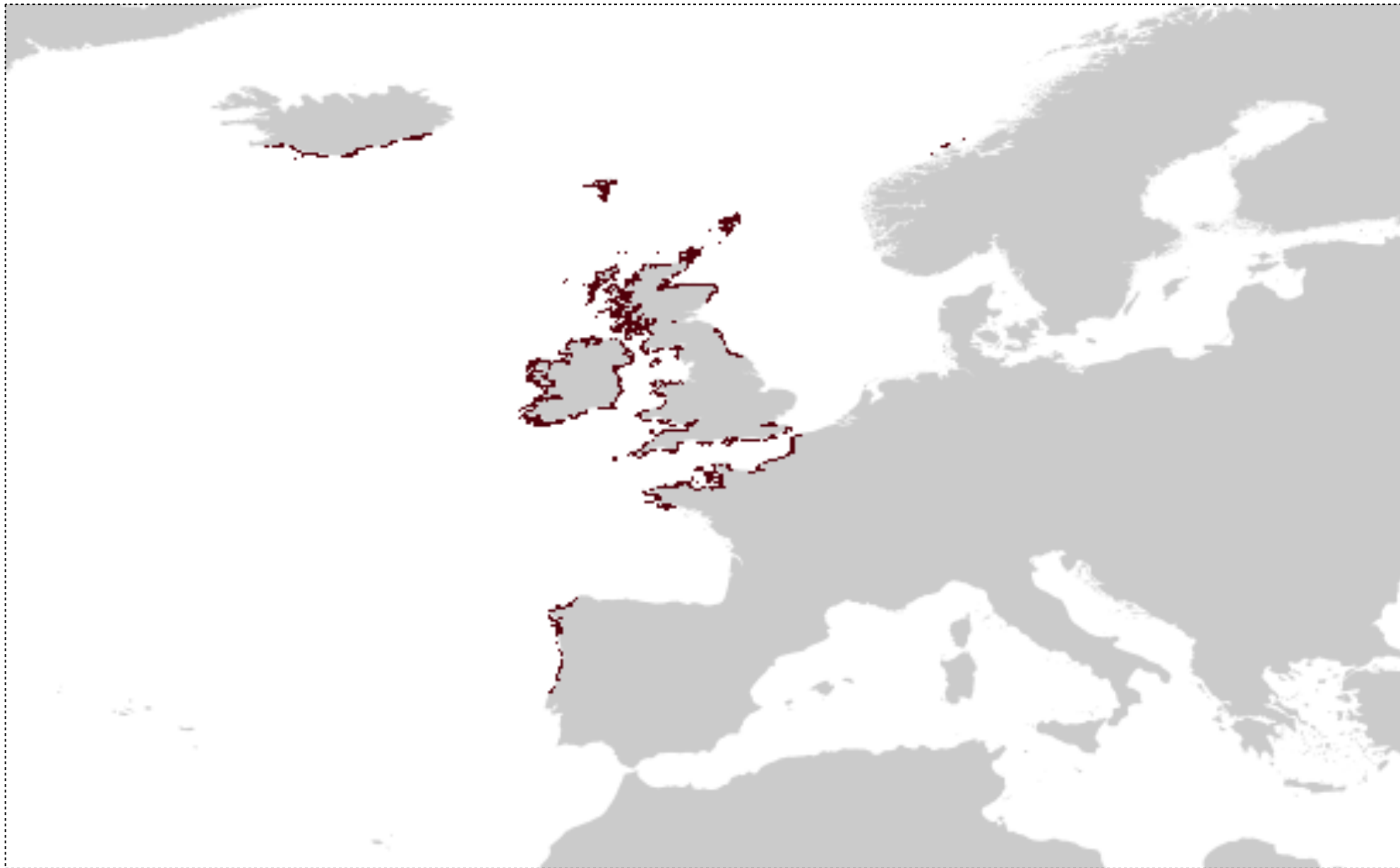
**Based on the statistical correlation between presence records and the environment, under the assumption that the distribution of a species is an indicator of its environmental requirements.**

(i.e., niche theory; the fact that a species occurs in a particular place is linked to its tolerance to the conditions of such place).



## **Mechanistic distribution models**

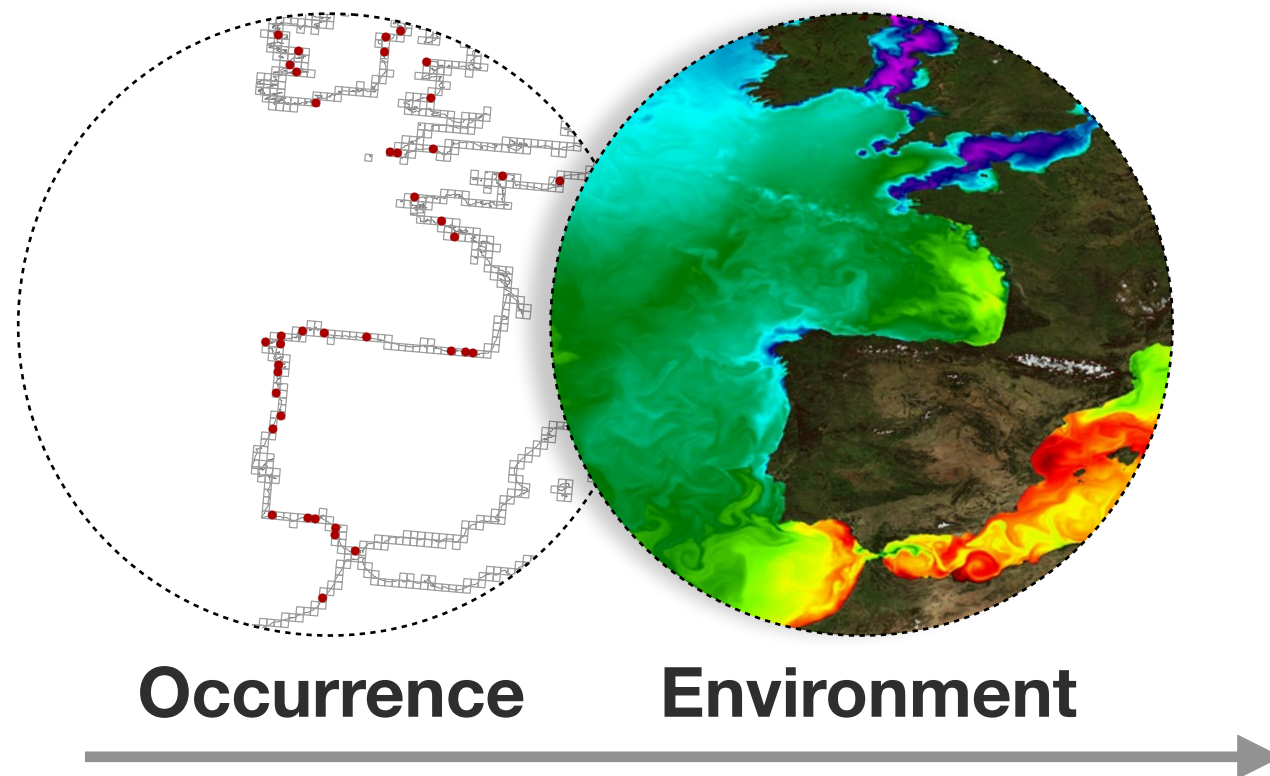
Built by **reclassifying environmental gradients with tolerance limits inferred from physiological experiments.**



## Mechanistic distribution models

A straightforward approach to predict the distribution of species.

$$\text{Presence} = [\text{Light} > 5 \text{ E.m}^2.\text{year}^{-1}] \cap [5^\circ\text{C} \leq \text{Temperature} \leq 20.5^\circ\text{C}]$$

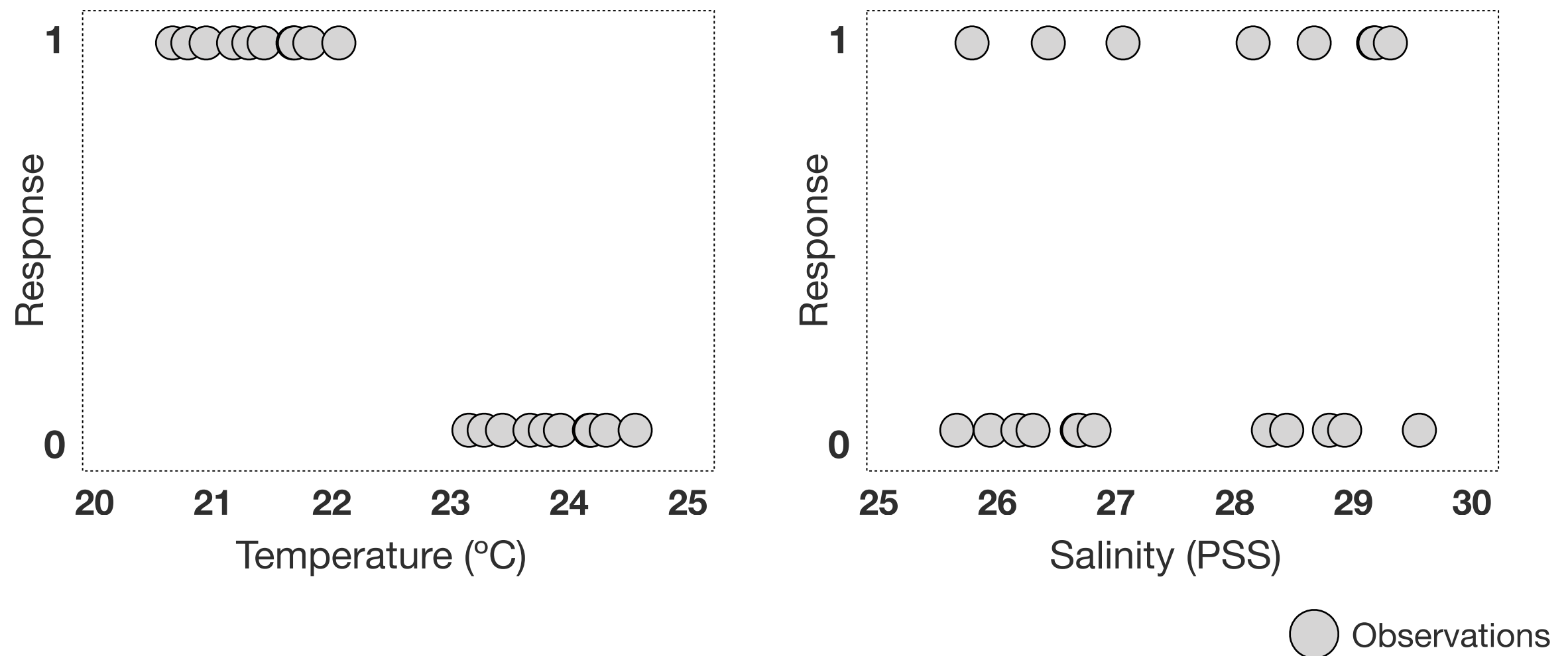


Resp	TempMax	Nitrate	Salinity
1	21	3	27
1	22	2	28
1	21	3	30
1	20	3	26
1	21	2	26
1	22	2	26
0	23	1	27
0	24	0	30
0	23	0	28
0	25	1	27
0	23	0	26
0	23	0	26

Data for modelling

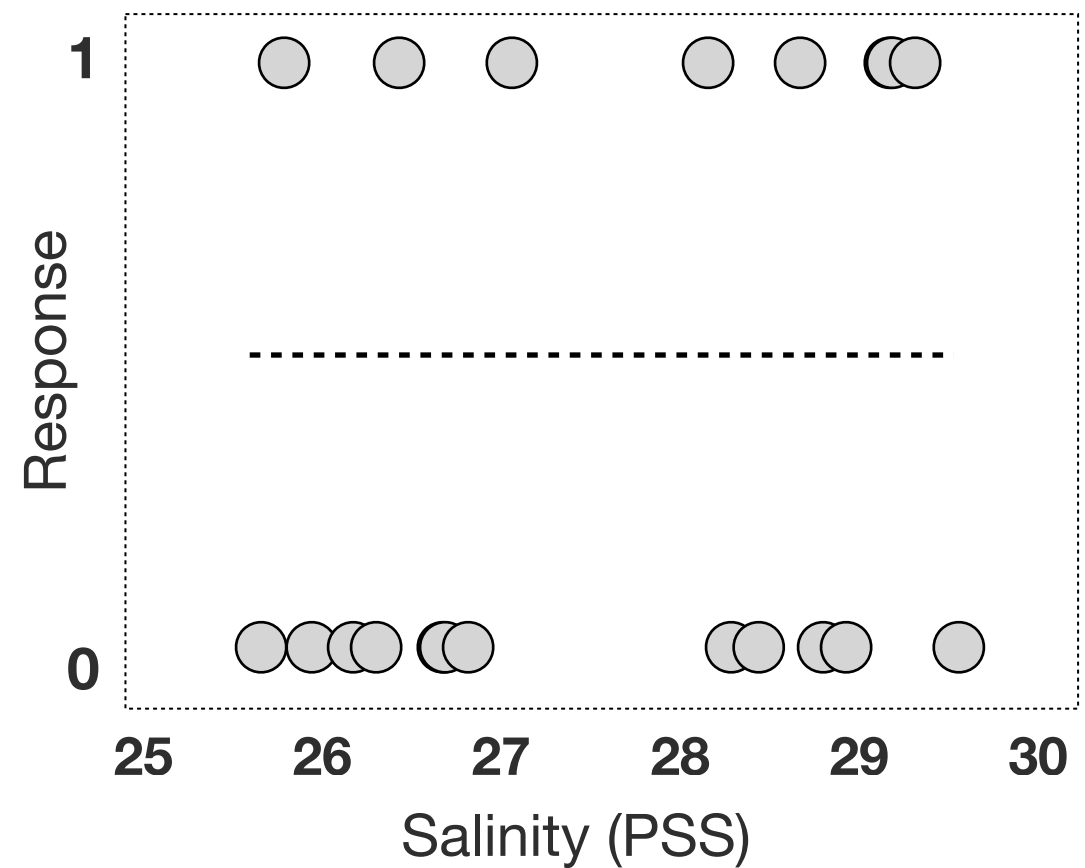
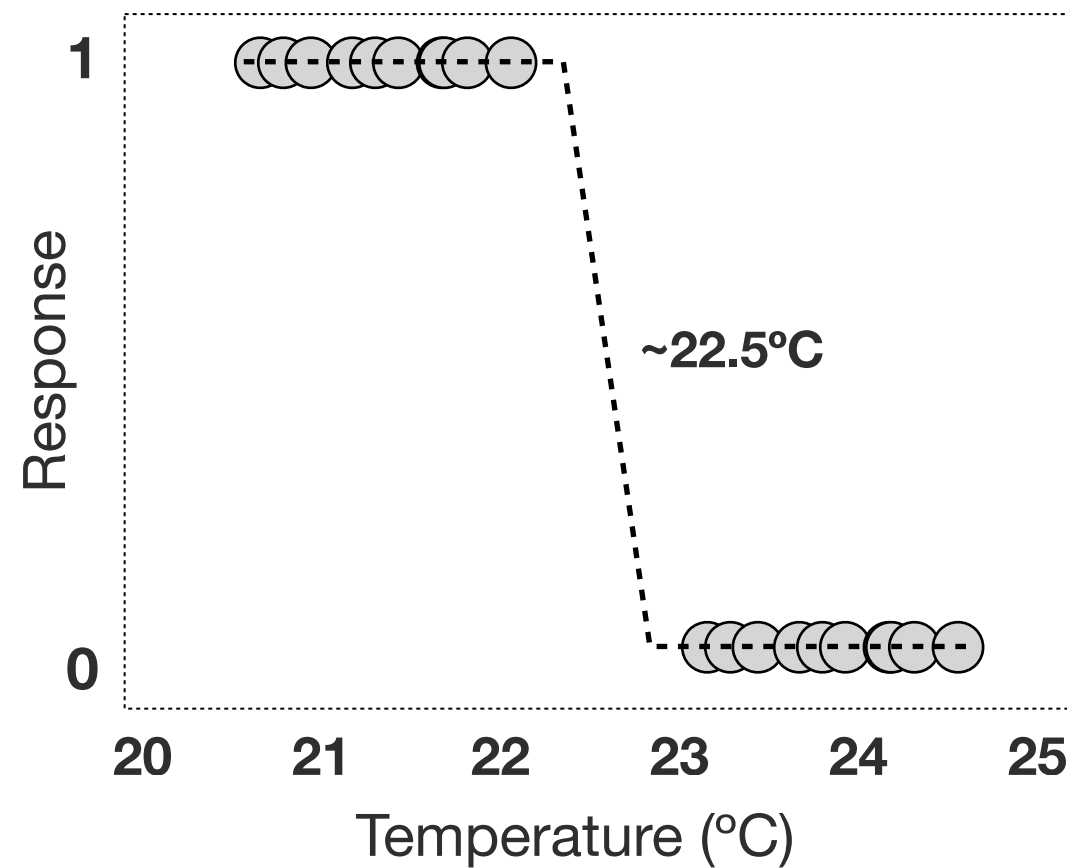
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Describe the **statistical relationship between distribution records and environmental conditions at those sites**. The models should be evaluated for “ecological realism” - consistency with ecological knowledge of limiting factors and species response curves.



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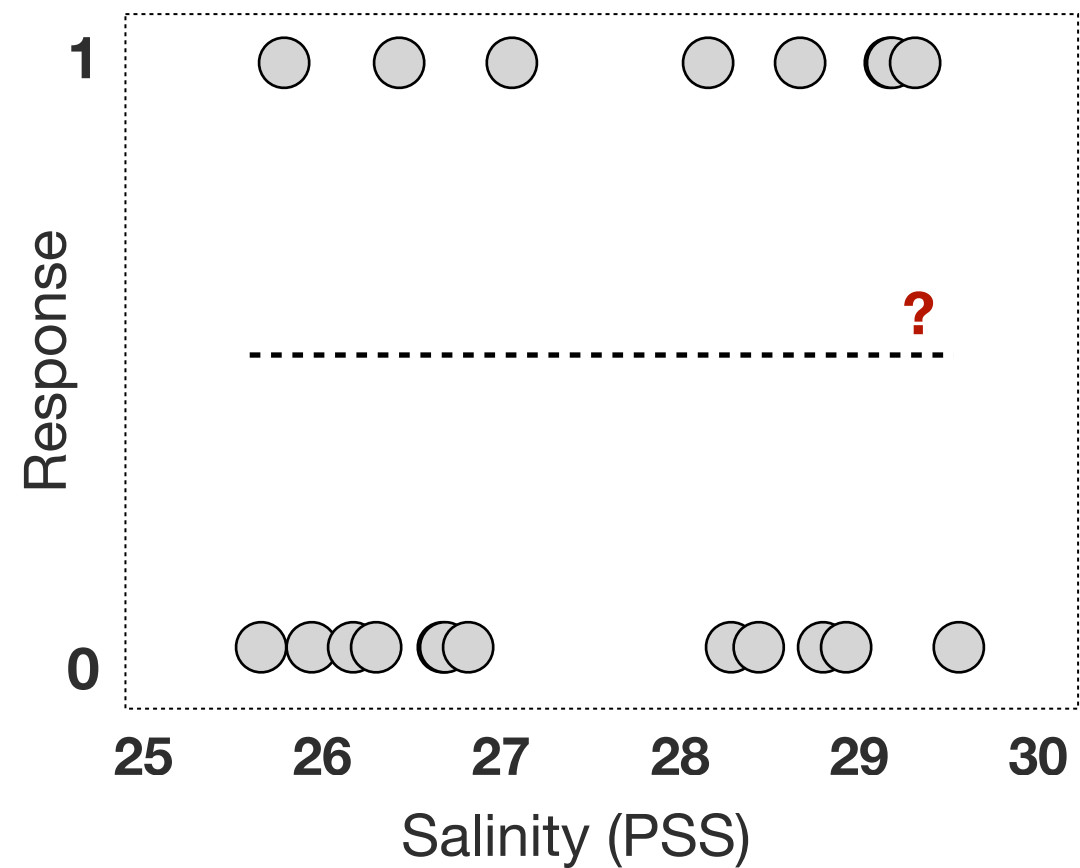
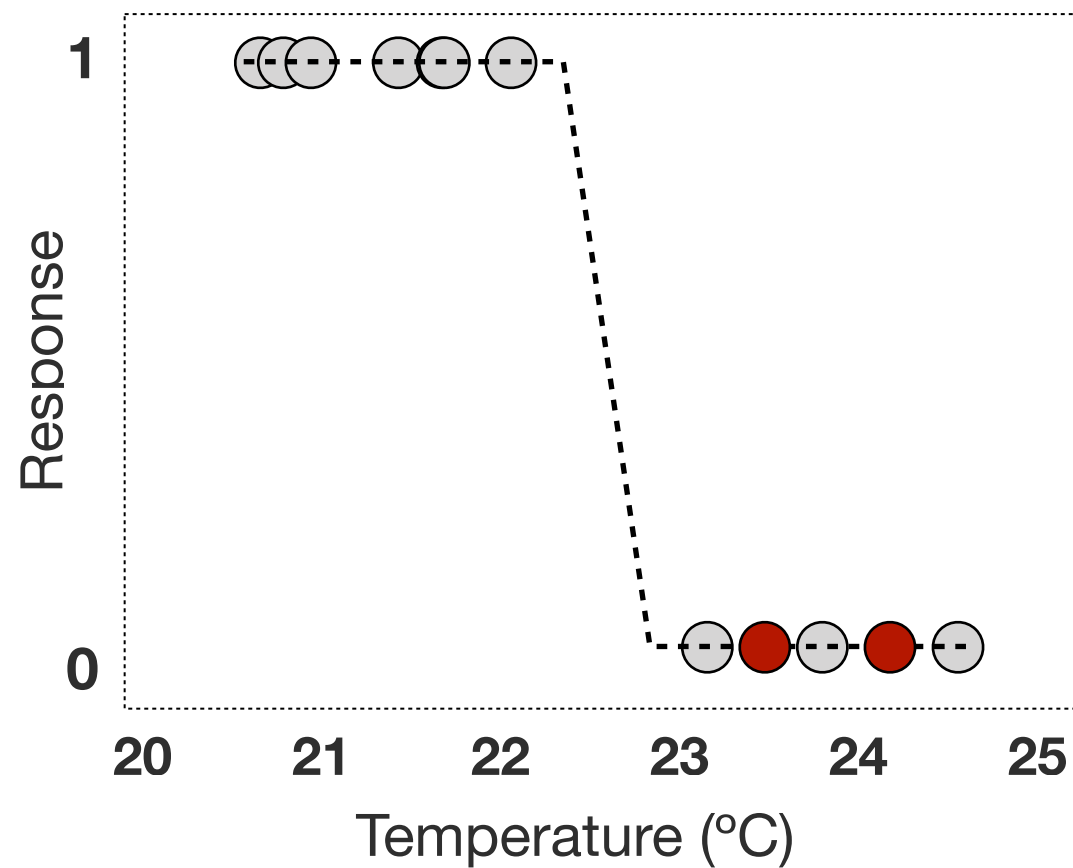


● Observations

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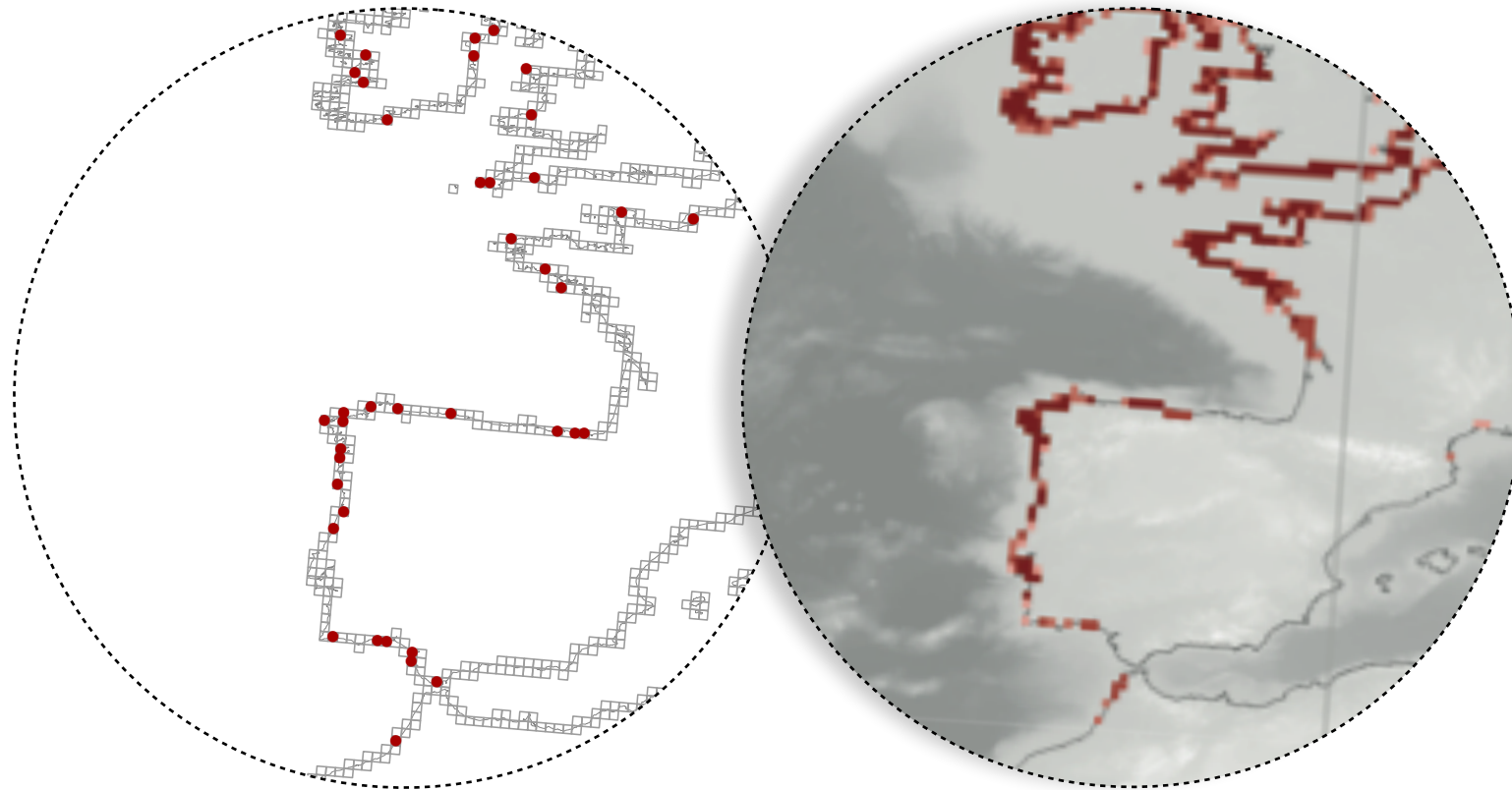




● Unknown data    ● Observations

**If a model can explain** the relationship between distribution records and environmental variables, we can **make predictions to unknown samples (unsurveilled regions)**.

e.g., Temp. = 23.5°C or 24.5°C, response is 0 (i.e., absence).

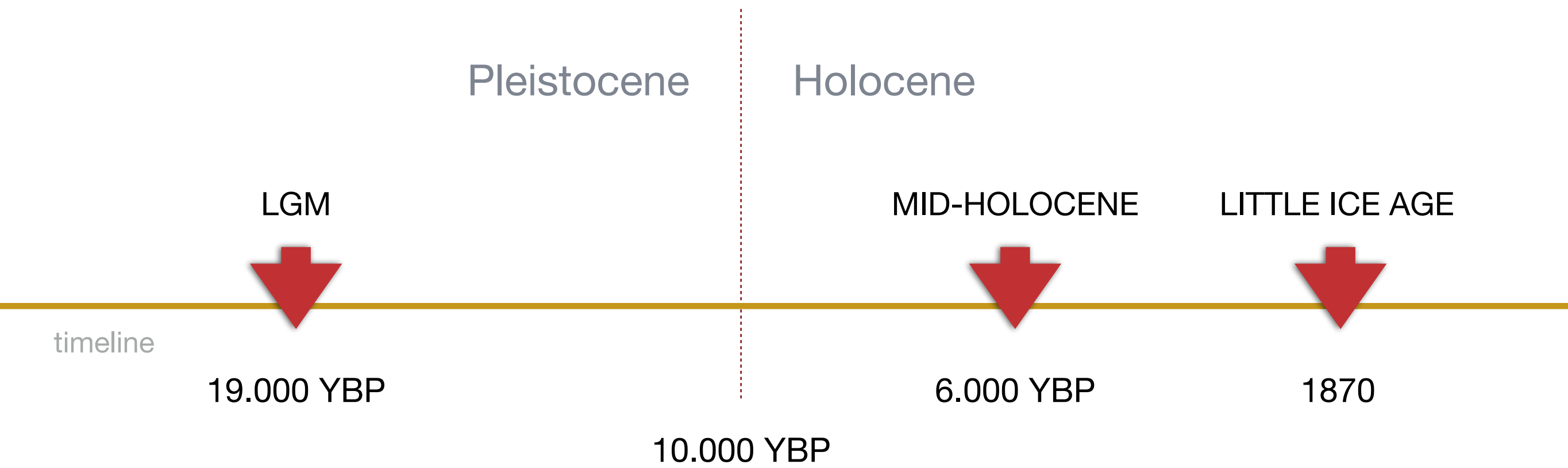


## **Predictive model-based interpolation**

Made to **new sites within the range of values of environmental conditions sampled in the training data** and within the **same time window** of sampling.

From **scattered records in space** to **continuous distribution surfaces**.

Typical **applications** include **mapping species' present distributions**, **important tools for ecology and conservation planning**.

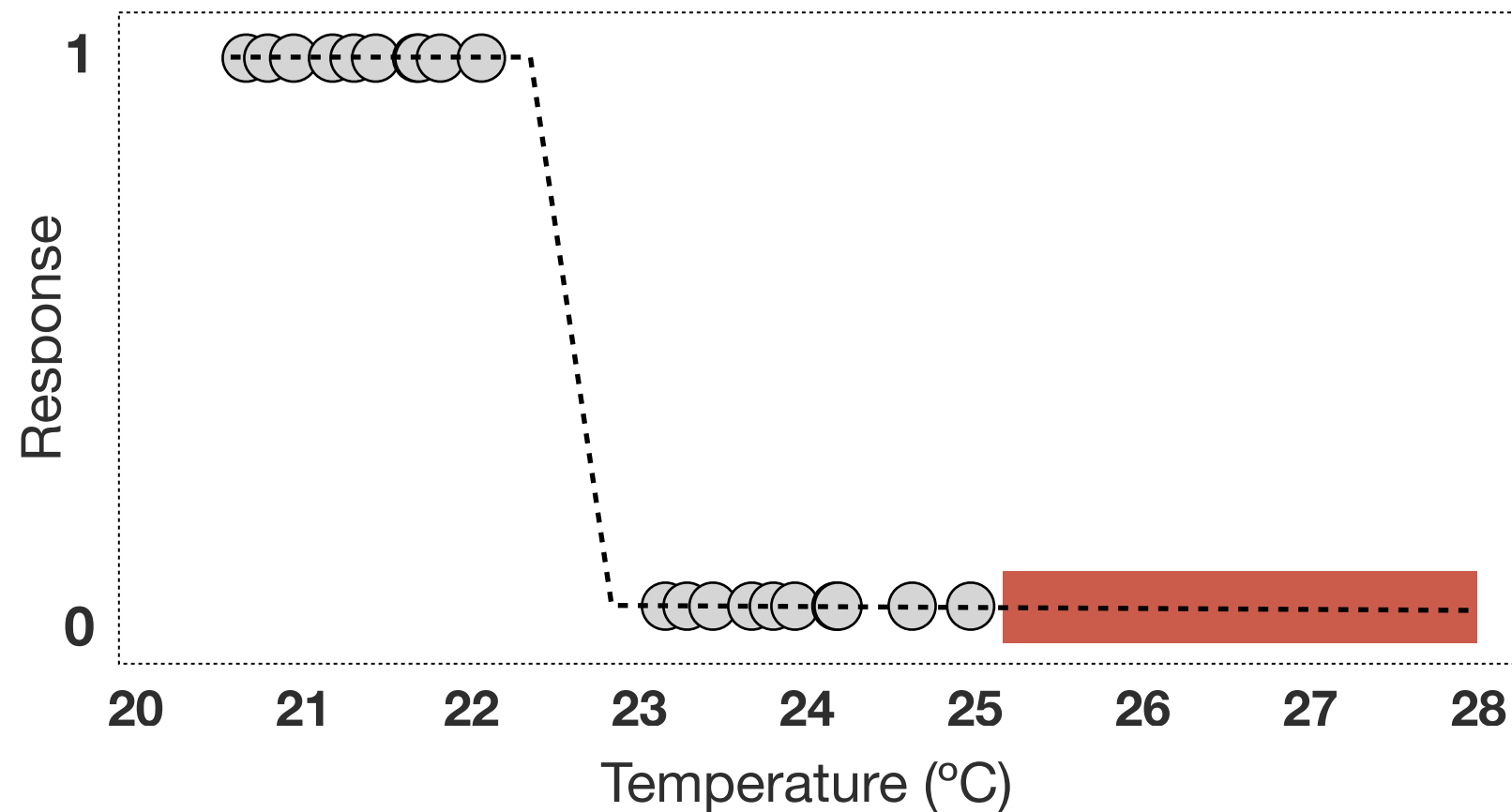


## **Predictive model-based transferability**

Made to **unsurveilled geographic or temporal domains.**

**No information on the similarity between the environment of training data and the predictions** (present-day vs future conditions).

Model **transferability** may lead to **extrapolation.**

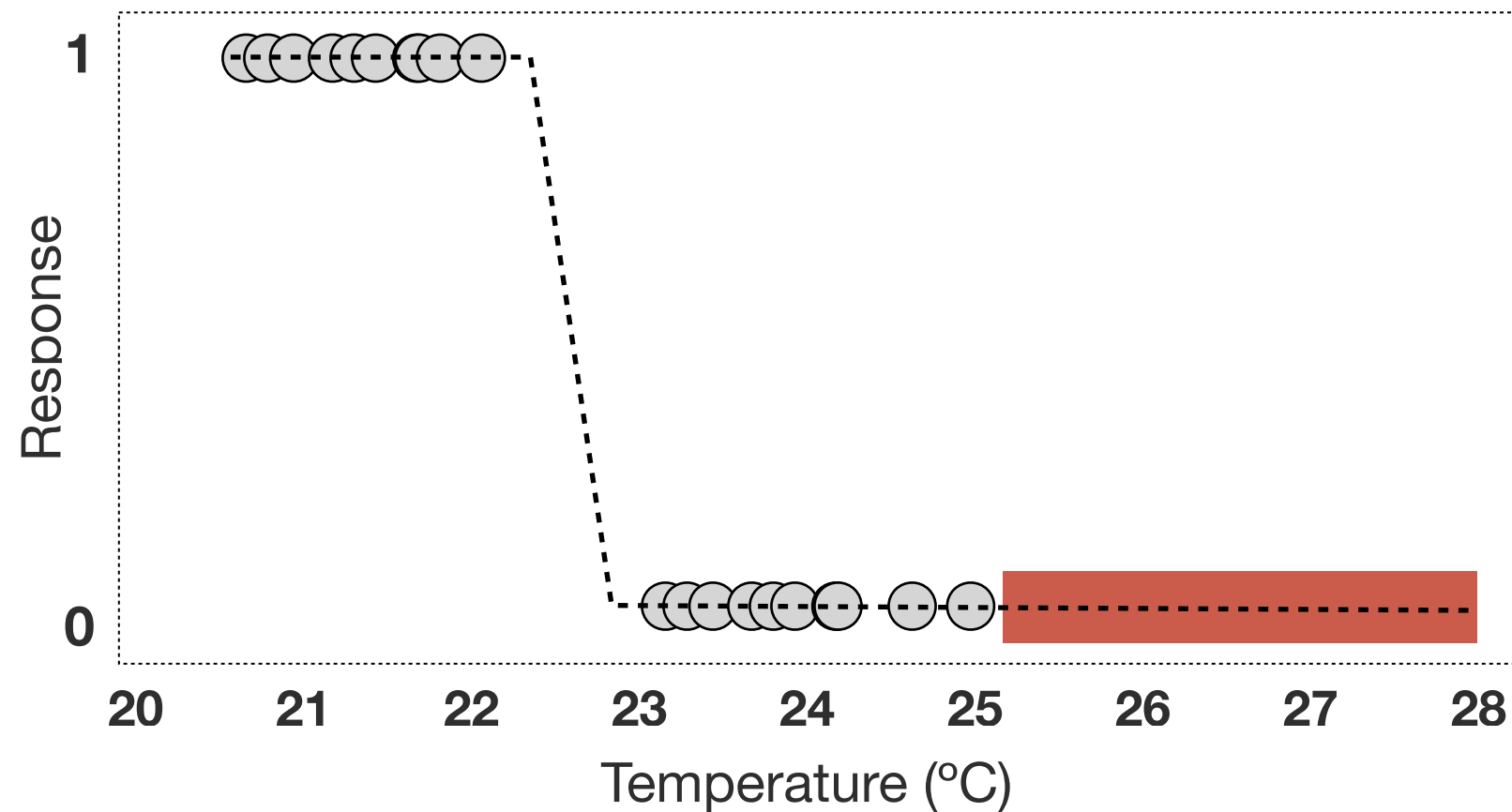


**Extrapolation** refers to **predictions for environmental values** that are **outside the range of environmental values used to fit the model**.

e.g.,

**A model used records with temperatures of 20-25°C.**

If **predictions are made for temperatures  $> 25^{\circ}\text{C}$** , then the model will extrapolate. No information exists on the probability of occurrence at  $> 25^{\circ}\text{C}$ , so the **predictions will be uncertain**.



## Avoid extrapolation in favour of interpolation

But when extrapolation exists (e.g., future climate changes), model should be treated with a great deal of caution.

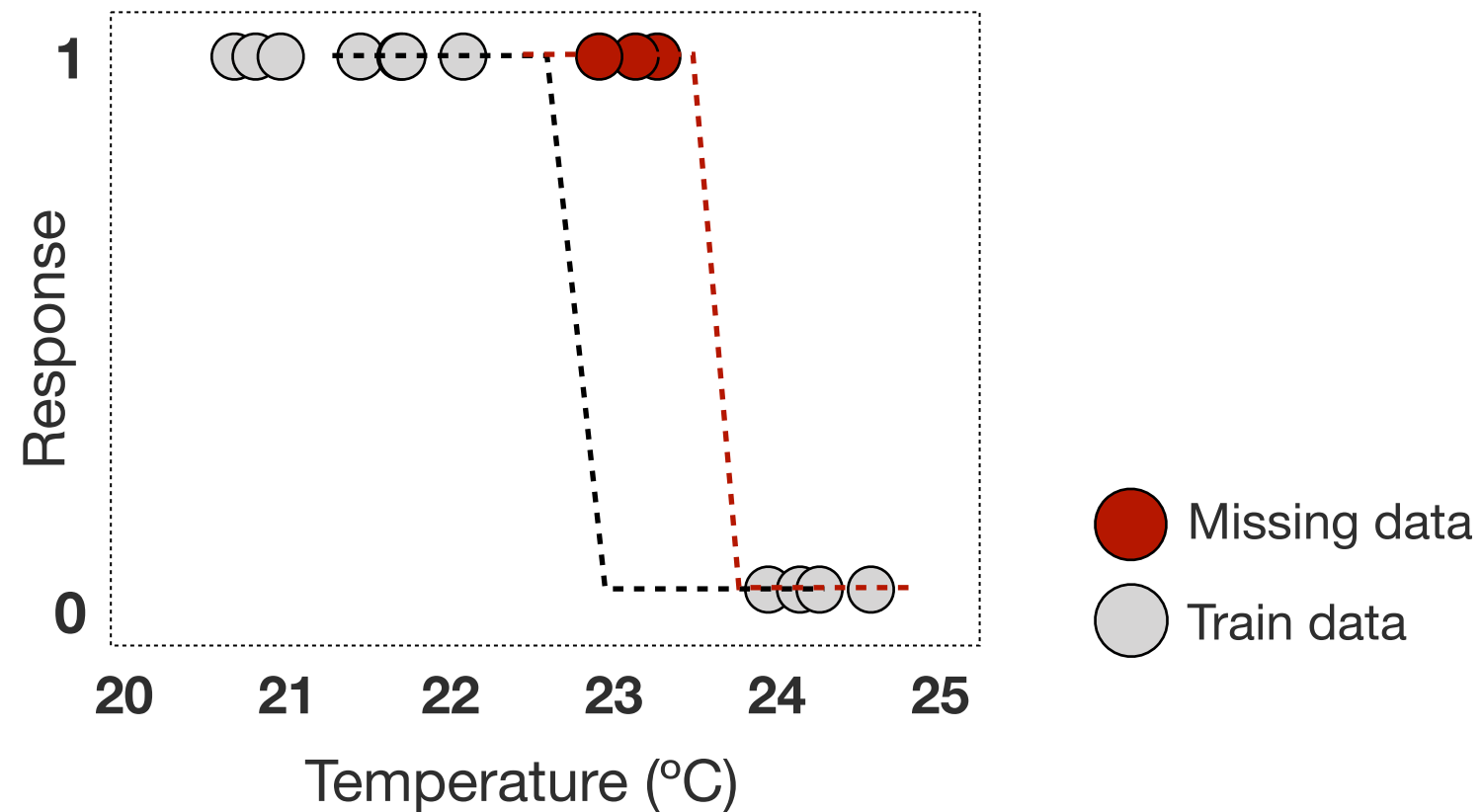
1. Avoid predicting with complex functions;
2. Use a parsimonious models (reduced number of predictors).
3. Interpret models with sound ecological knowledge;



Corrective models can identify the **niche of a species only if the records used to fit the models cover the distribution of the species.**

When mapped, it represents the **potential distribution or the habitat suitability.**

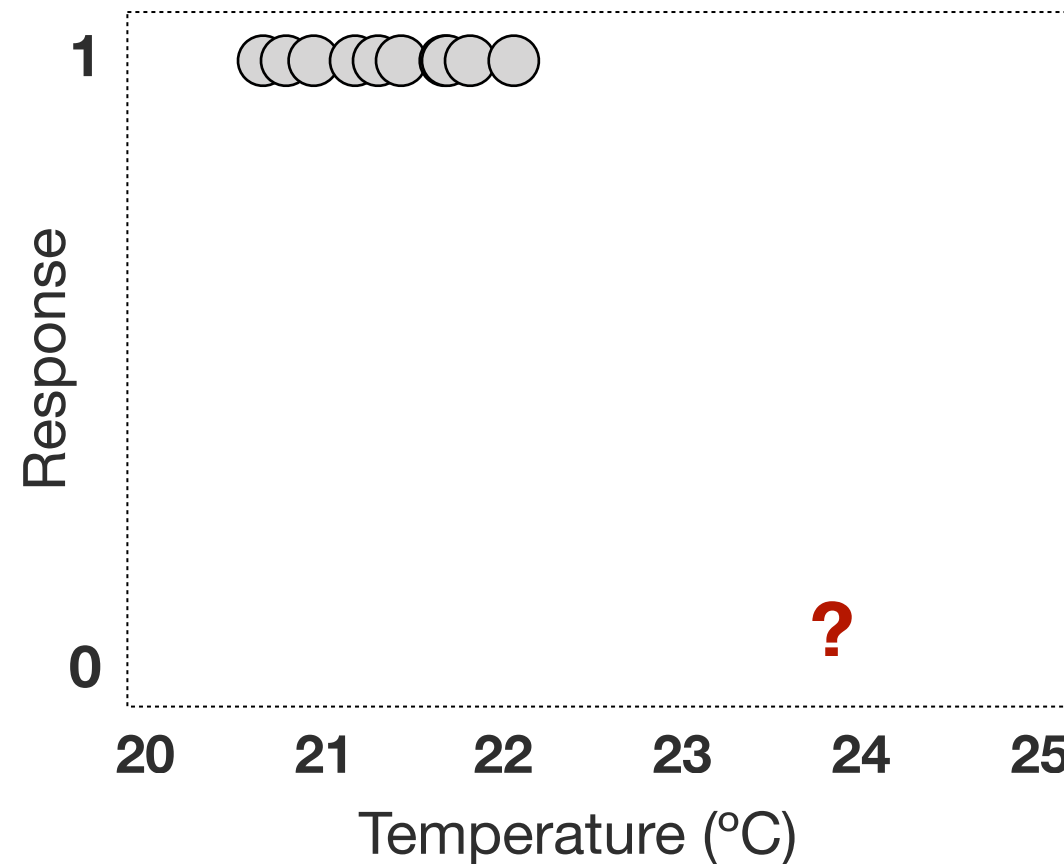
If records are insufficient, the models do not identify the fundamental niche; **the model fits only to the portion of the niche that is represented by the observed records** (truncated niche).



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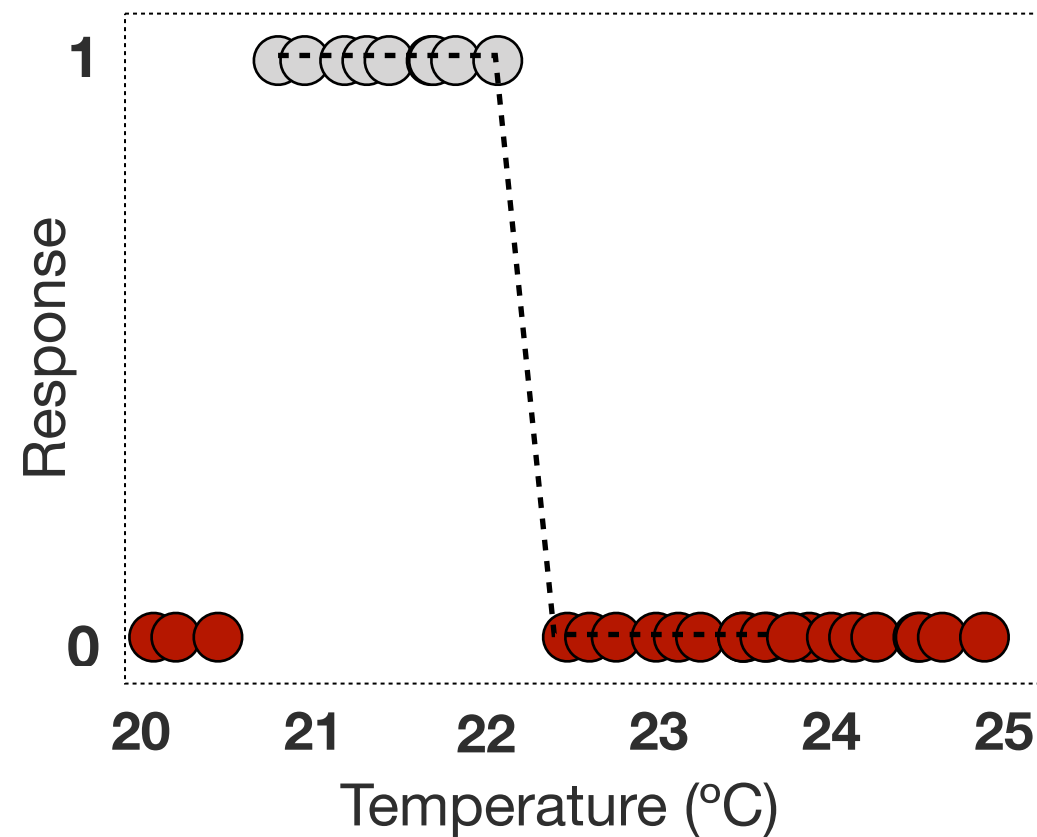


**Correlative models** infer the **relationship** between the occurrence of biodiversity and the environment, but **absence records are often unavailable** or unknown, leading to **presence-only datasets** and to the need of developing **presence-only models**.

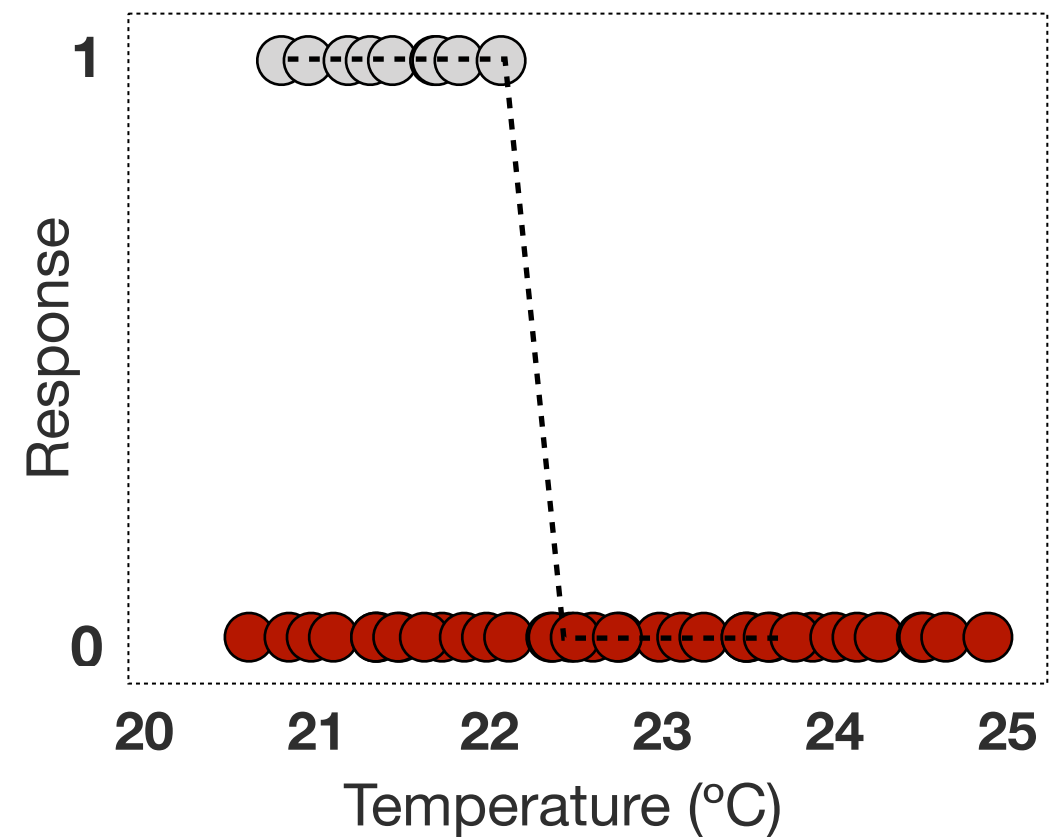




**‘pseudo-absences’ data**



**‘background’ data**



**Models based on ‘pseudo-absences’**, generated from the **study area where occurrences do not exist**. Any regression or machine learning algorithm can be implemented (e.g., GLM and BRT).

**Models based on ‘background’**, generated from the entire study area. Focus on **how the environment where the species occurs relates to the environment across the rest of the study area** (e.g., MaxEnt).



**How many absence records?**

**Linear models, additive models and maximum entropy models**

A large number of pseudo-absences / background (e.g., 10,000);

**Boosted Regression Trees models**

The same number of pseudo-absences as presences records (but never less than 1,000).



## Which environmental predictors for modelling?

Use **large datasets**, an approach with stronger criticism.

Use **pre-selected datasets**, linked to known physiological rules\*\*.

\*\* the choice of predictors should be **guided by the objectives and the hypotheses raised** about the species-environment relationship.



# Modelling with abiotic conditions

**Ecophysiological knowledge should guide the modeller** in process of choosing which factors may determine the distribution of species.

## **My main questions**

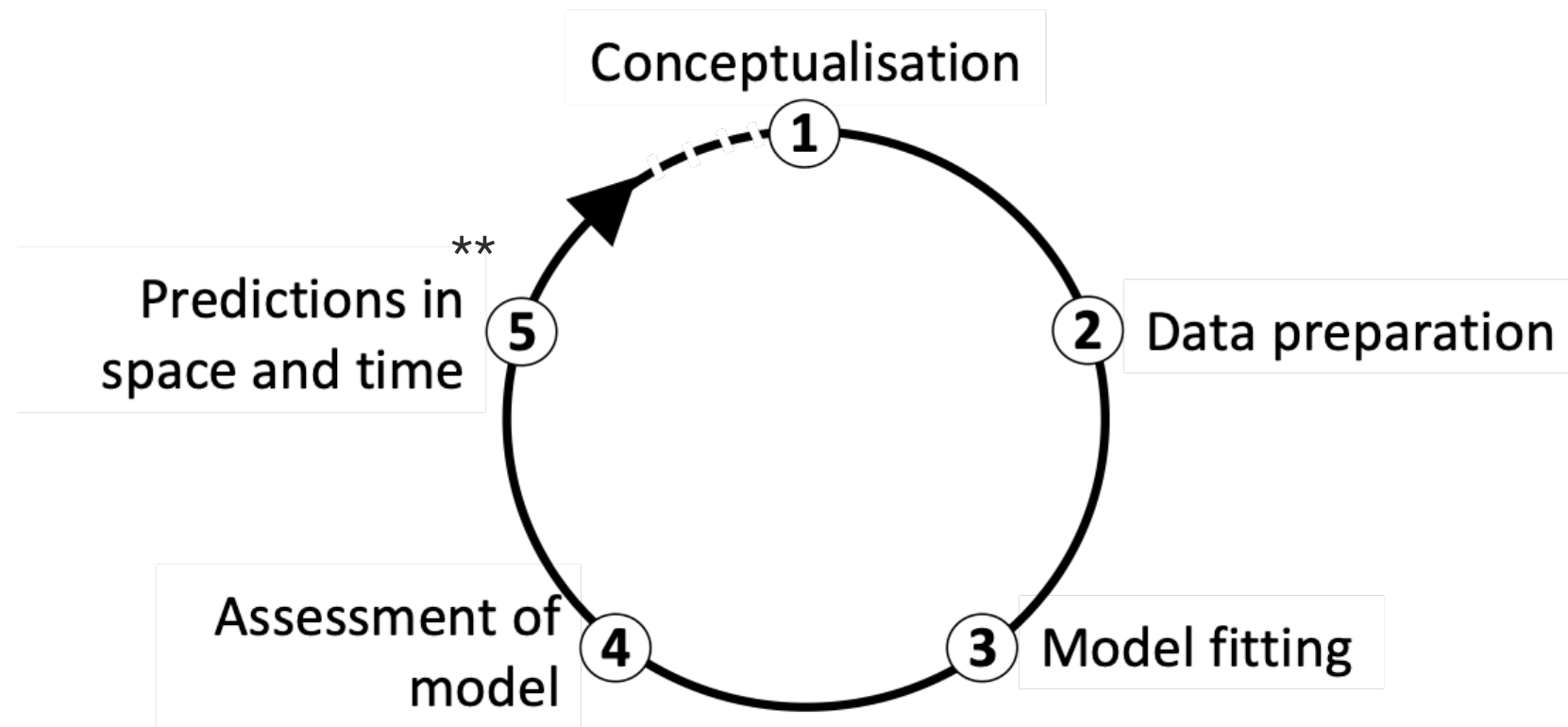
Which environmental variables drive the distribution of my model species?

Which response I expect for each environmental variable?

**My main hypotheses must be based on ecophysiological knowledge**  
(...)



## Steps for model building



**Model building is an iterative process** and there is much to learn on the way (a loop rather than a straightforward approach).

You may want to revisit and improve certain steps (e.g., improve biodiversity data collection or remove surplus environmental layers).

\*\* not always part of ENM studies; depends on the model objective.



# Model fitting

Key steps in good modelling practice include the following:

1. Gathering relevant biodiversity and environmental data, and assessing its adequacy (relevance and completeness, avoiding truncated niches);
2. Selecting an appropriate modelling algorithm;
3. Fitting the model to the training data and evaluating performance, including the realism of fitted response functions, the model's fit to data, and predictive performance on test data;
4. Mapping predictions to geographic space;
5. Apply thresholds if continuous predictions need reduction to a binary map;
6. Iterating the process to improve the model in light of knowledge gained throughout the process.



# Assumptions of ENM associated with data

Observed distributions are indicative of environmental tolerances and resource requirements since species are in equilibrium with the environmental conditions - species occur in all suitable areas and are absent from all unsuitable areas (niche theory).

**Niche space assumption :** The study contains the full range of conditions that the species can inhabit (for the examined abiotic variables).

**Dispersal / demographic assumption :** Factors related to dispersal, establishment, and persistence do not drive the species not to occupy an environmentally suitable area.

**Biotic assumption :** Biotic interactions do not drive the species not to occupy an environmentally suitable area.



# Desired properties of ENM fitting

**Deductive:** develop hypotheses about the causes of the pattern explained and predicted by the models.

**Distinct between patterns and process:** distinguish between the patterns observed and the mechanisms involved.

**Simplicity:** highlight a few mechanisms without becoming entangled in complex interactions and correlations between many variables.

**Parsimonious:** preferring the simpler of two equally adequate models; favor simple explanations over complex.

**Generality:** aimed to achieve general broad ecological conclusions.





# Main assumption of ENM

The observed distributions are an indication of the environmental tolerances and resource requirements, since species are in equilibrium with the environmental conditions.

**i.e.,**

Species occur in all suitable areas and are absent from all unsuitable areas (niche theory).