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- article méthodo publié dans **Methods in Ecology and Evolution**

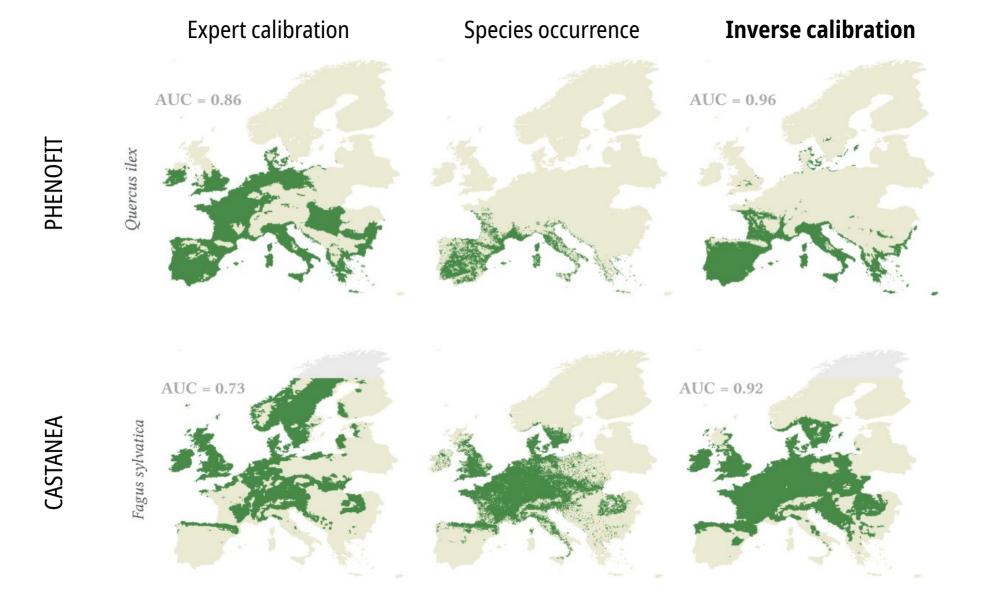
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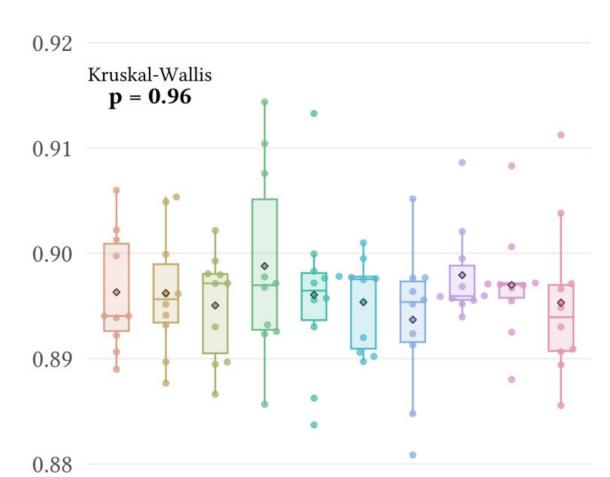
DOI: 10.1111/2041-210X.14119

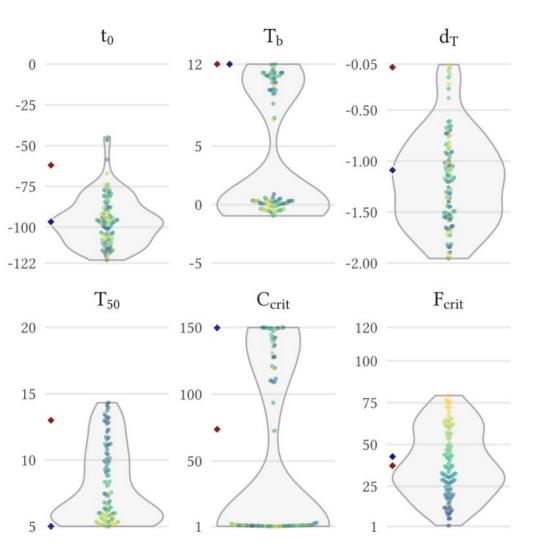
RESEARCH ARTICLE

Methods in Ecology and Evolution English South

Estimating process-based model parameters from species distribution data using the evolutionary algorithm CMA-ES







#### - approche *presence-only* inspirée de Valavi et al. (2022)

Ecological Monographs, 92(1), 2022, e01486
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Predictive performance of presence-only species distribution models: a benchmark study with reproducible code

ROOZBEH VALAVI D, 1,3 GURUTZETA GUILLERA-ARROITA D, 2 JOSÉ J. LAHOZ-MONFORT D, 2 AND JANE ELITH D2

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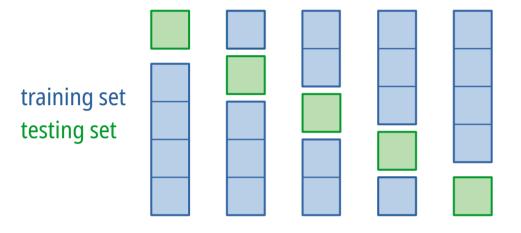
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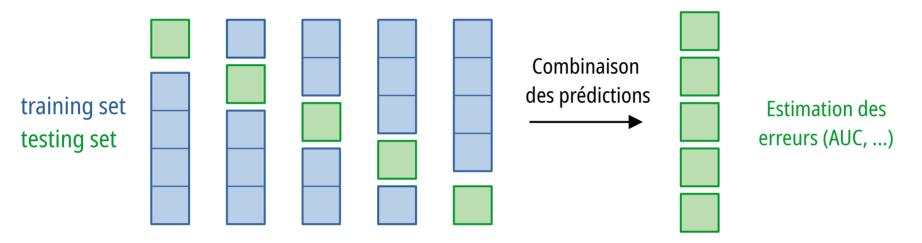
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Predictive performance of presence-only species distribution models: a benchmark study with reproducible code

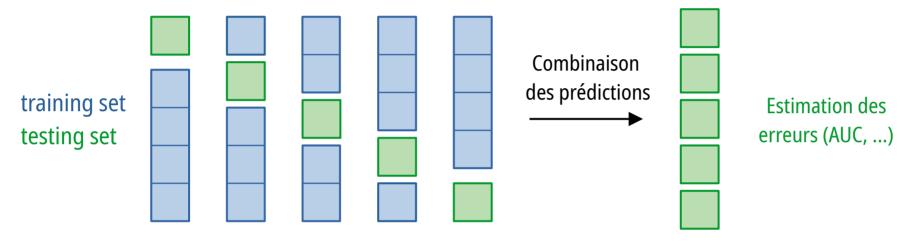
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- données d'occurrence : **EU-Forest** (+GBIF, WOODIV)
- **background points** : « *sampled irrespective of the location of species records* » 50.000 points
- modèles : GLM with lasso penalty, GAM, BRT, RandomForest (+ biomod et Maxent)



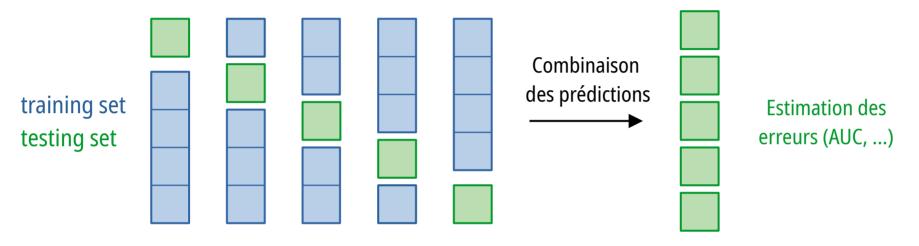


- *5-fold environmental block cross-validation* pour estimer les erreurs



- final model: « all the available training data can be used to fit a new model »
Favorise la qualité des prédictions (plutôt que l'exactitude des erreurs)

- **5-fold environmental block cross-validation** pour estimer les erreurs



- final model: « all the available training data can be used to fit a new model »
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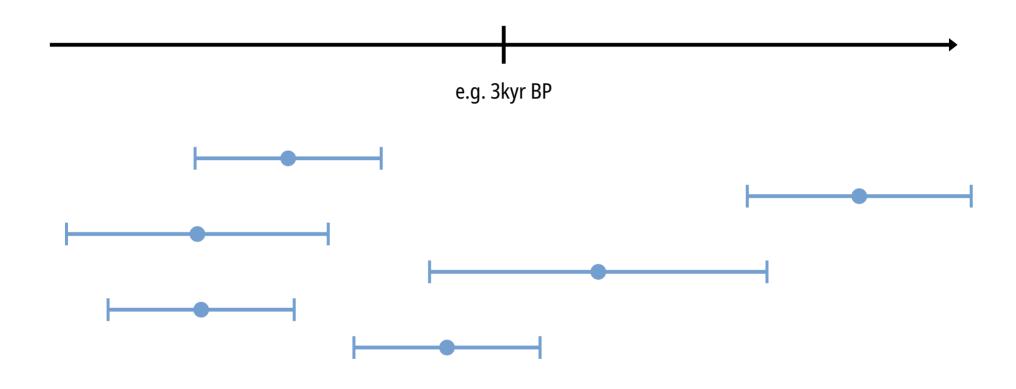
Over-estimated AUC

Exemple avec Fagus sylvatica

	Lasso GLM	GAM	Random Forest	BRT
« True » AUC	0.68	0.79	0.81	0.79
Over- estimated AUC	0.92	0.93	0.99	0.96

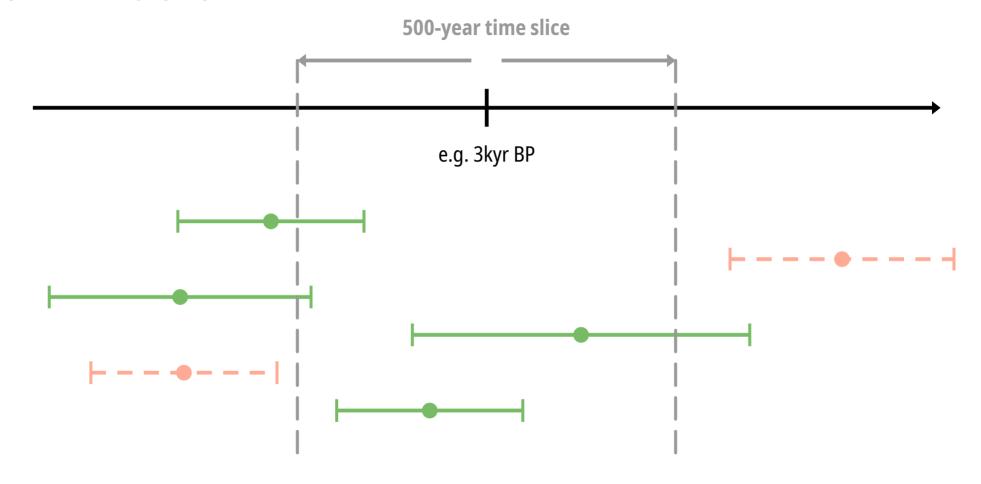


# Temporal coverage per pollen site

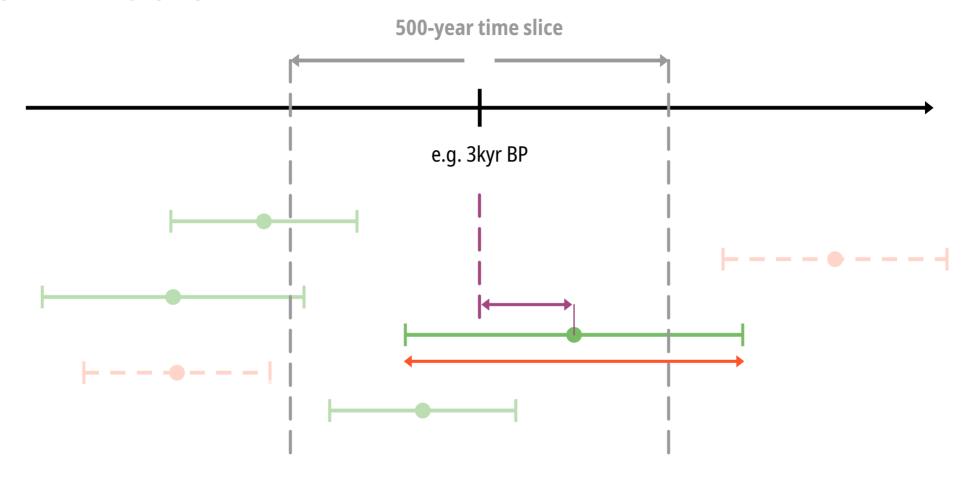


Multiple samples within the same site

# Temporal coverage per pollen site

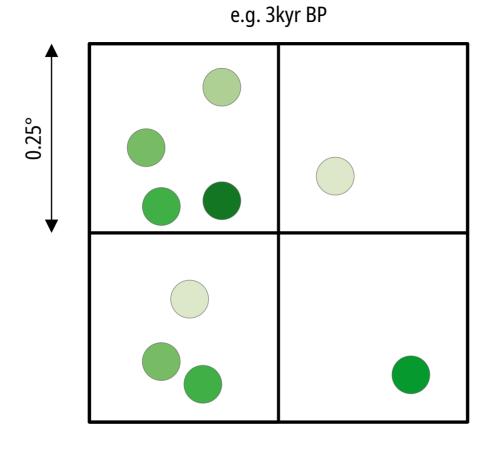


# Temporal coverage per pollen site



Weighted mean of pollen counts, taking into account both uncertainty and time distance

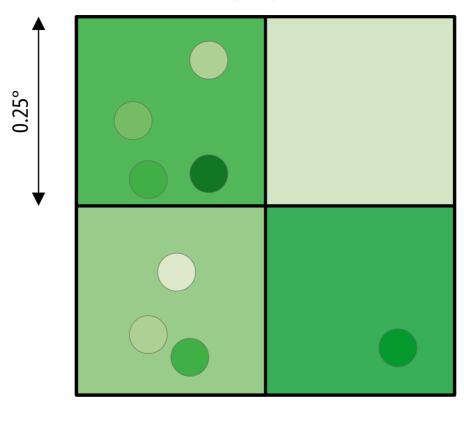
# **Spatial upscaling**



Species pollen relative abundance

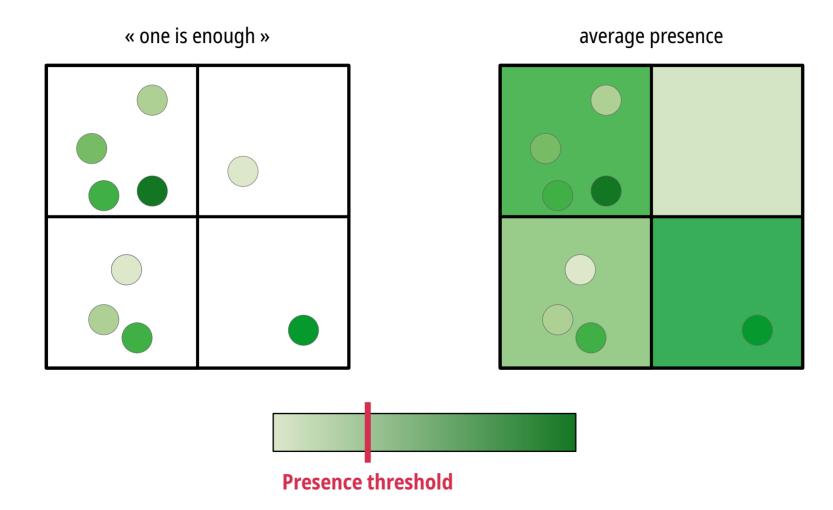
# **Spatial upscaling: relative abundance**

e.g. 3kyr BP

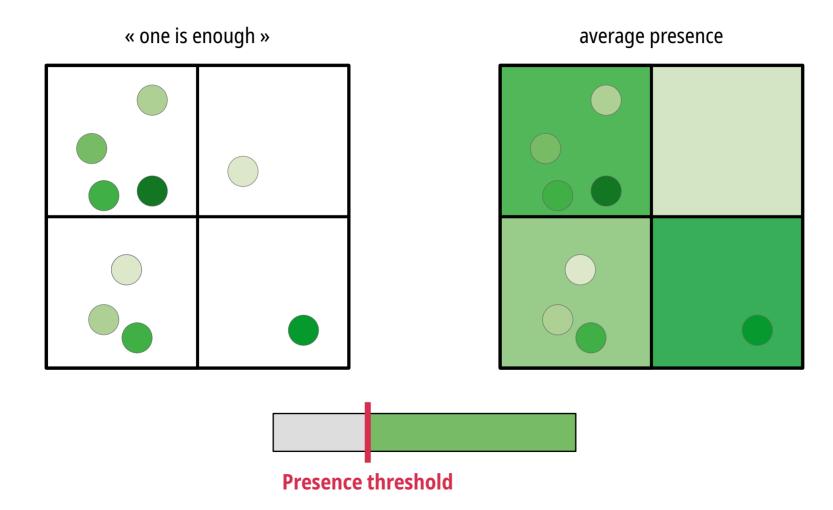


**Average** species relative abundance

# Spatial upscaling: presence/absence



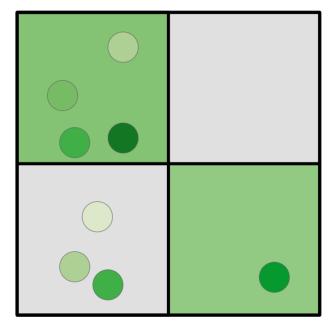
# Spatial upscaling: presence/absence



# Spatial upscaling: presence/absence

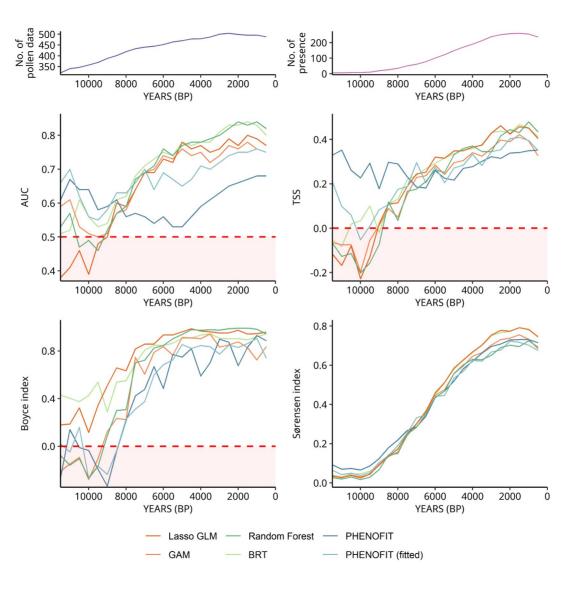
« one is enough »

average presence



absence

presence



#### **CLIMATE APPROACH: Climatic distance**

# **PAST CLIMATE** e.g. 9000 years BP N predictors

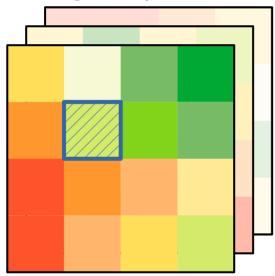
# **BASELINE CLIMATE** 1970 - 2000

N predictors

#### **CLIMATE APPROACH: Climatic distance**

#### **PAST CLIMATE**

e.g. 9000 years BP

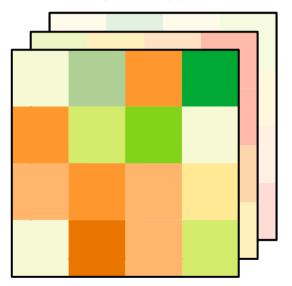


From each cell:

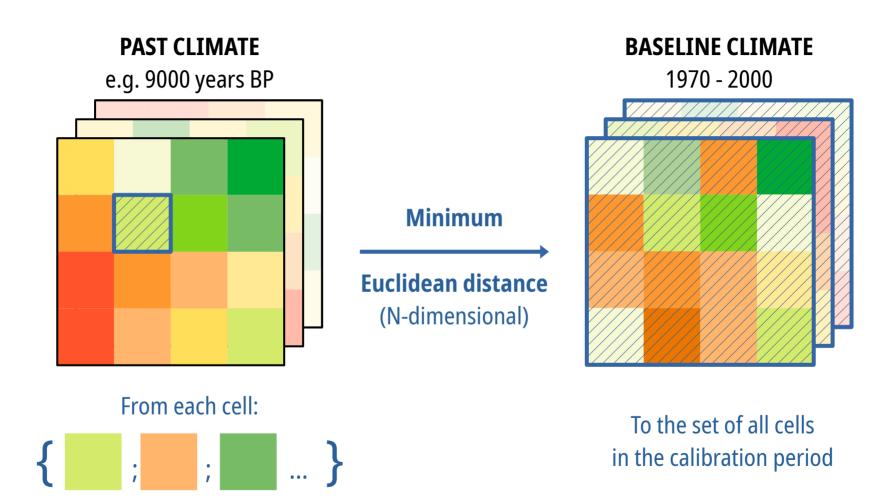


#### **BASELINE CLIMATE**

1970 - 2000

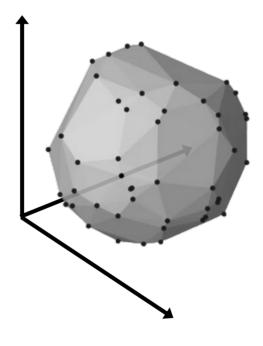


#### **CLIMATE APPROACH: Climatic distance**



# **CLIMATE APPROACH: Hypervolume similarity**

**PAST CLIMATE** e.g. 9000 years BP

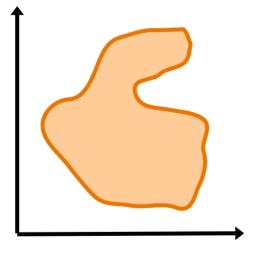


N-dimensional hypervolume

# **CLIMATE APPROACH: Hypervolume similarity**

**PAST CLIMATE** 

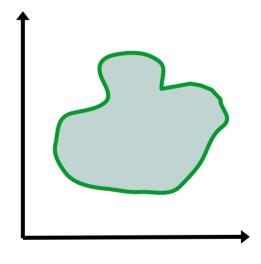
e.g. 9000 years BP



N-dimensional hypervolume

#### **BASELINE CLIMATE**

1970 - 2000



N-dimensional hypervolume

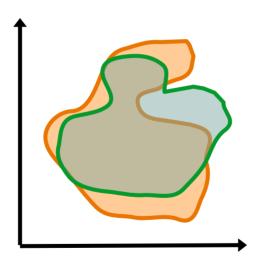
# **CLIMATE APPROACH: Hypervolume similarity**

#### **PAST CLIMATE**

e.g. 9000 years BP

#### **BASELINE CLIMATE**

1970 - 2000



# **CLIMATE APPROACH:** Hypervolume similarity

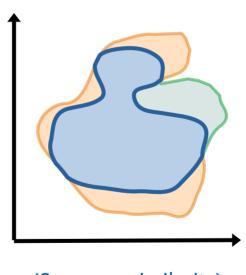
#### **PAST CLIMATE**

e.g. 9000 years BP

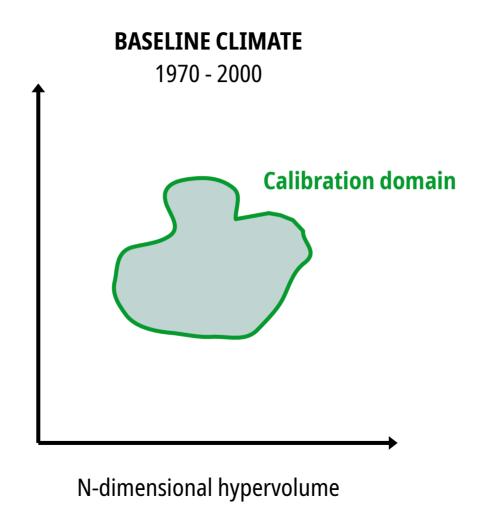
#### **BASELINE CLIMATE**

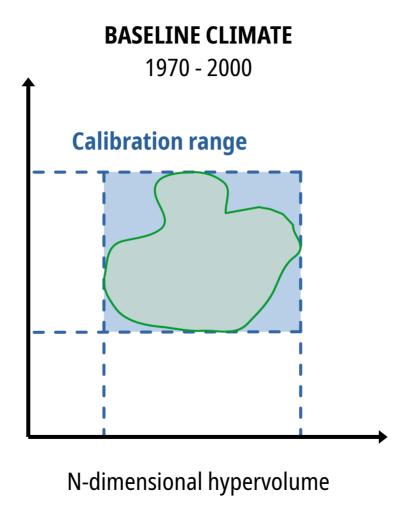
1970 - 2000

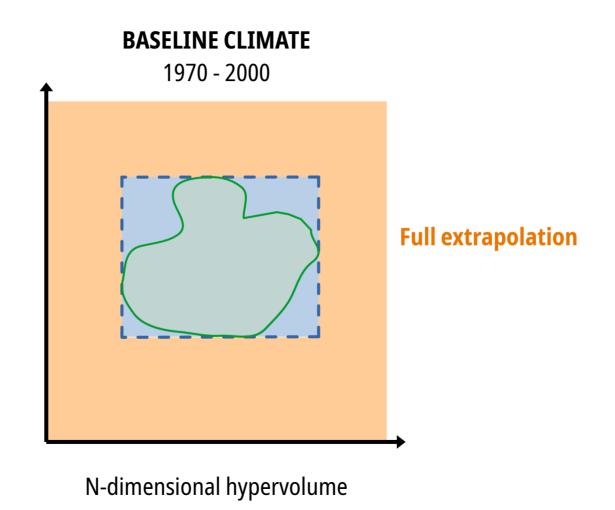
## **Hypervolume overlap**

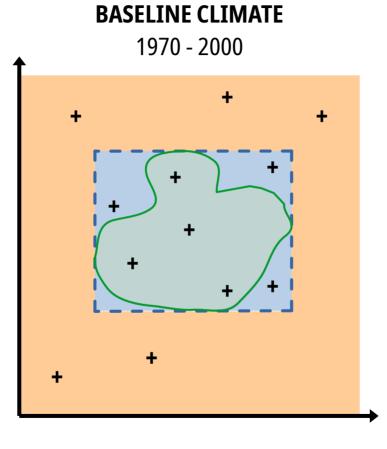


(Sørensen similarity)



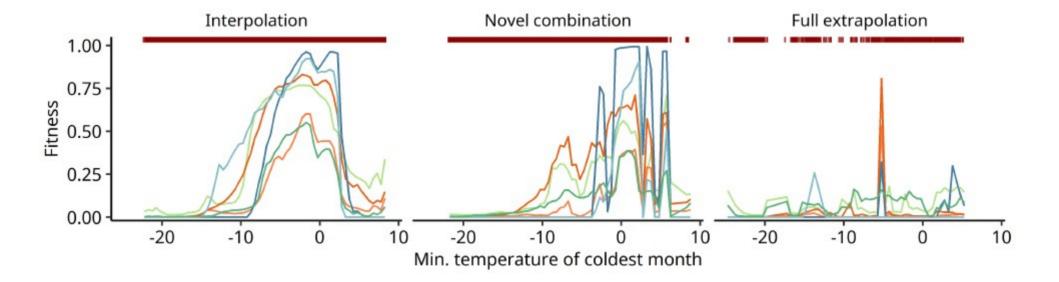


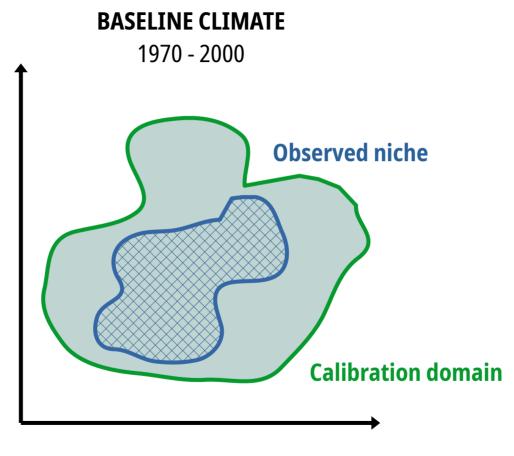




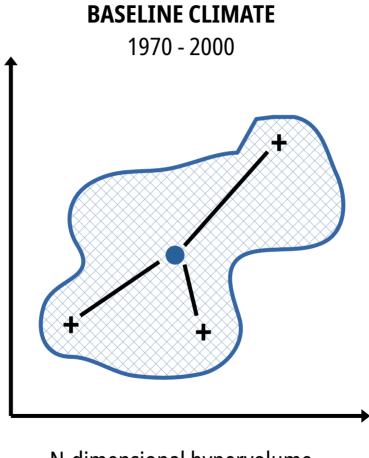
N-dimensional hypervolume

#### **CLIMATE APPROACH: Categorical differenciation and model response**

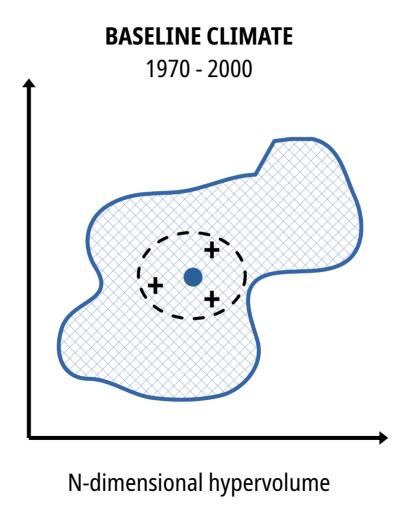


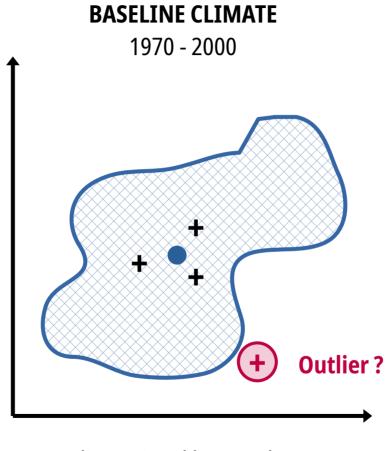


N-dimensional hypervolume



N-dimensional hypervolume





N-dimensional hypervolume

