

# Aires de répartitions, métacommunautés et biodiversité : du théorique à l'appliqué

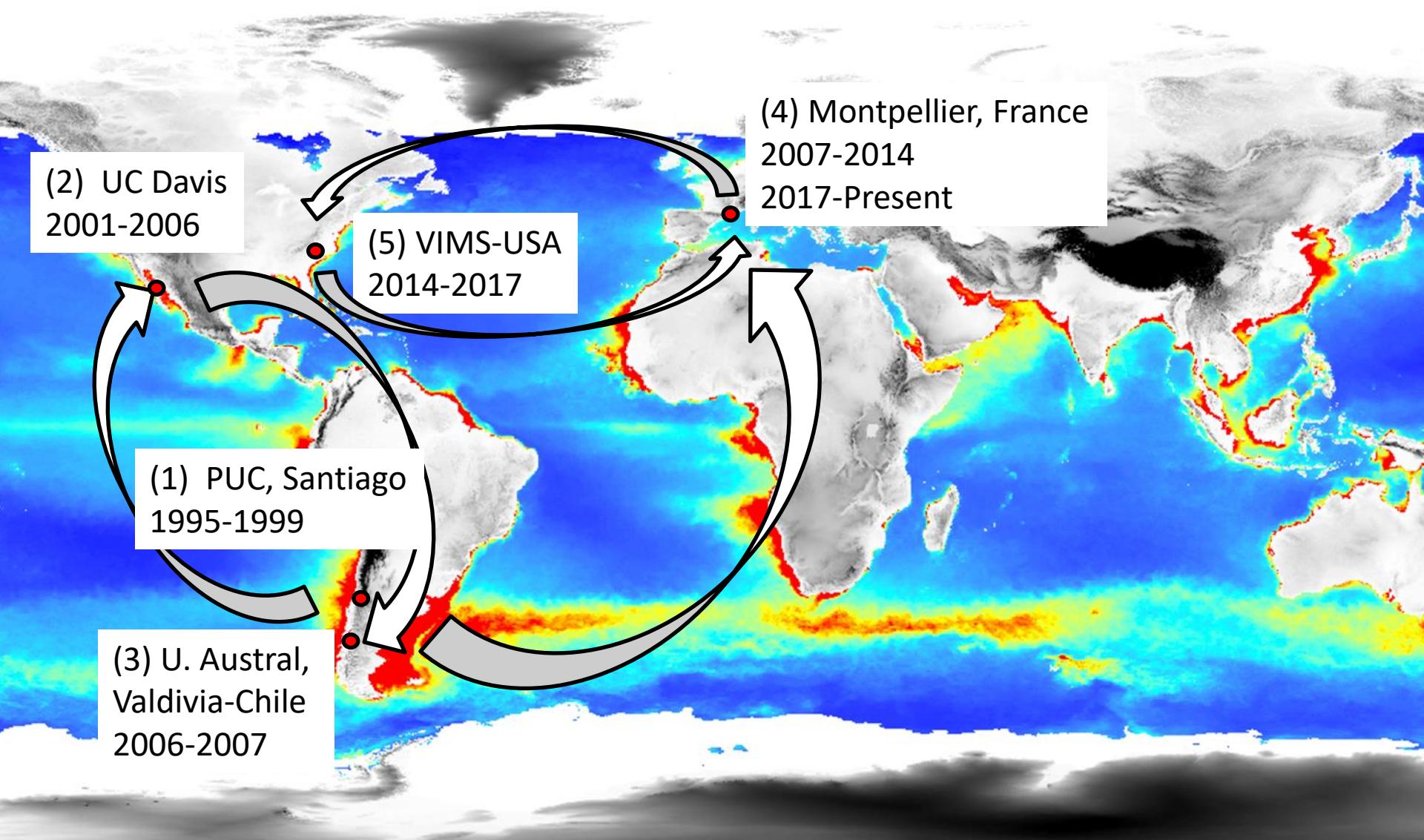
Christine N. Meynard

Candidate au diplôme d'Habilitation pour  
Diriger des Recherches (HDR)

Université de Montpellier, école doctorale GAIA



# Parcours



# Etudiants

- Encadrements direct: 2 licence, 2 M1, 8 M2, 1 masters USA, 1 doctorant
- Comités de thèse et co-publications étudiants
- Enseignement niveau master et doctorat en SIG, statistiques, biologie de populations. Au total autour de 400h.

# Etudiants actuels



Dorian Frisch  
M1, U. Rennes

Changements climatiques et zones de contact  
entre *Rhabdomys sp.* en Afrique du Sud  
Collaboration avec Guila Ganem (ISEM)  
Projets Labex, OSU, ANR en cours d'évaluation



Isis Poinas  
Doctorante

Effets non-intentionnels des pratiques  
agricoles sur la biodiversité des bordures  
Collaboration avec Guillaume Fried (ANSES)  
Projet ANSES-INRAE

# Etudiants actuels

## Comités de thèse



**Jingwei Song, PhD Candidate**  
W&M, VIMS, USA  
Supervisor: Jan McDowell

An Investigation of Local Adaptation of  
Speckled Trout, *Cynoscion nebulosus*,  
along the U.S. East Coast



**Hannah Bevan, PhD Candidate**  
U Central Florida, USA  
Supervisor: David Jenkins

Why, when, and how to best apply  
species distribution analyses

# Etudiants

Claire Dufour



Annabelle Sueur

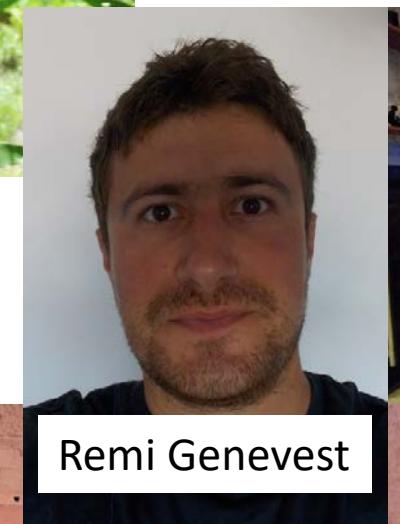


Laura Henckel

Livia Rodrigues  
De Sa



Remi Genevest



Manon Perrigault



Sophie Monsarrat



Natalia Carrasco



Bastien Louboutin

Paula Iturrealde-Polit



# Diversity and origins of life have fascinated biologists for centuries



# Simon Levin, MacArthur Award lecture in 1989

## “The problem of pattern and scale in ecology”

“Applied challenges, such as the prediction of the ecological causes and consequences of global climate change, require the **interfacing of phenomena that occur on very different scales of space, time, and ecological organization**”

“The key to prediction and understanding lies in the elucidation of **mechanisms underlying observed patterns**”

“Our efforts to develop theories of the way ecosystems or communities are organized must revolve around attempts to discover **patterns that can be quantified within systems, and compared across systems**”

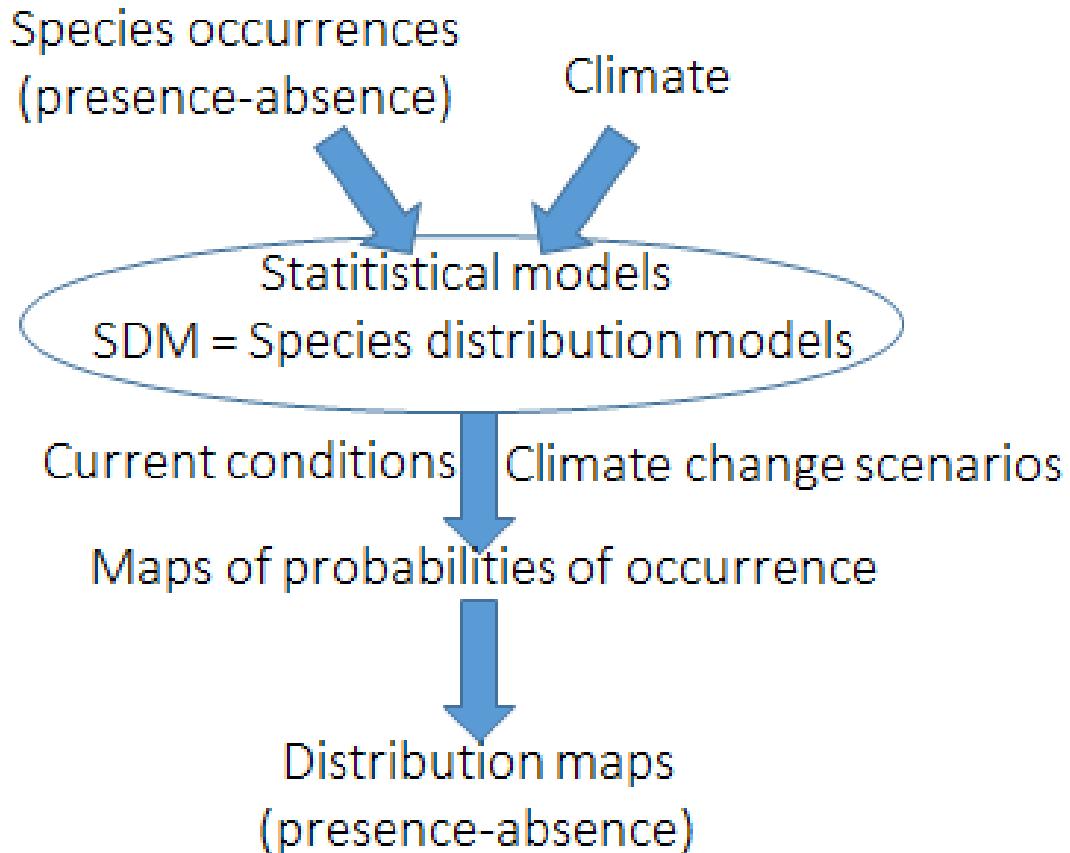
# Jerome Chave: “what have we learned in 20 years?”

Chave (2013) Ecology Letters

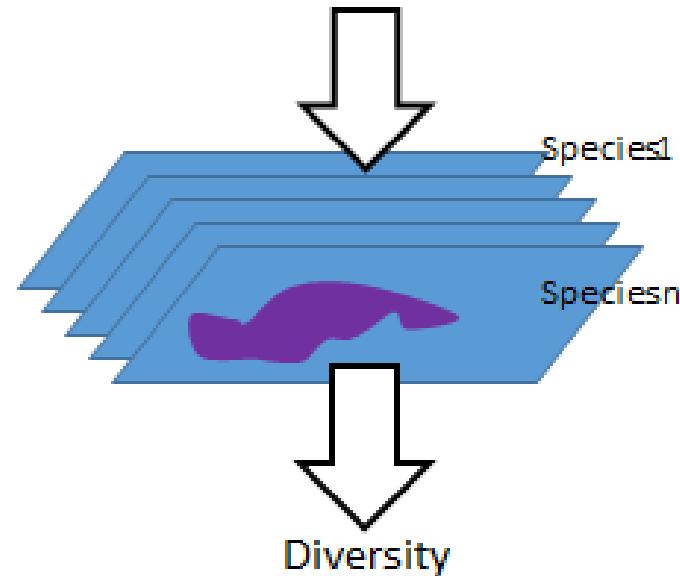
There has been **4 major technological revolutions**:

- The advent of the **numeric era**
- **Molecular biology** technological advances
- **Environmental sensing** (including environmental but also tracking movement)
- Development of **global communications**: global collaborations, citizen science

## Species distribution models (SDMs)



Stacking individual species distributions



# Agricultural challenges

- **Food security under a growing population:** we will need to produce more quickly
- **Agriculture is the main cause of habitat fragmentation and habitat loss:** need to find a way to stop converting land from natural habitat
- Important **contributor to climate change** (emission of CO<sub>2</sub>, disruption of nitrogen cycle): we need to produce differently
- **Human health concerns:** we need to consume more greens and fruits and change our production system

# SDMs in the agricultural context

- Major crops: physiological modelling (but...)
- **Insect / arthropod pests**
- **Insects as vectors of disease**

# SDMs in the agricultural context

- Usually opportunistic presence data
- Systematic surveys are rare
- Pests are often missed if they are not in the mist of an outbreak
- Absence data is often under-reported or not recorded at all

# Virtual Ecology

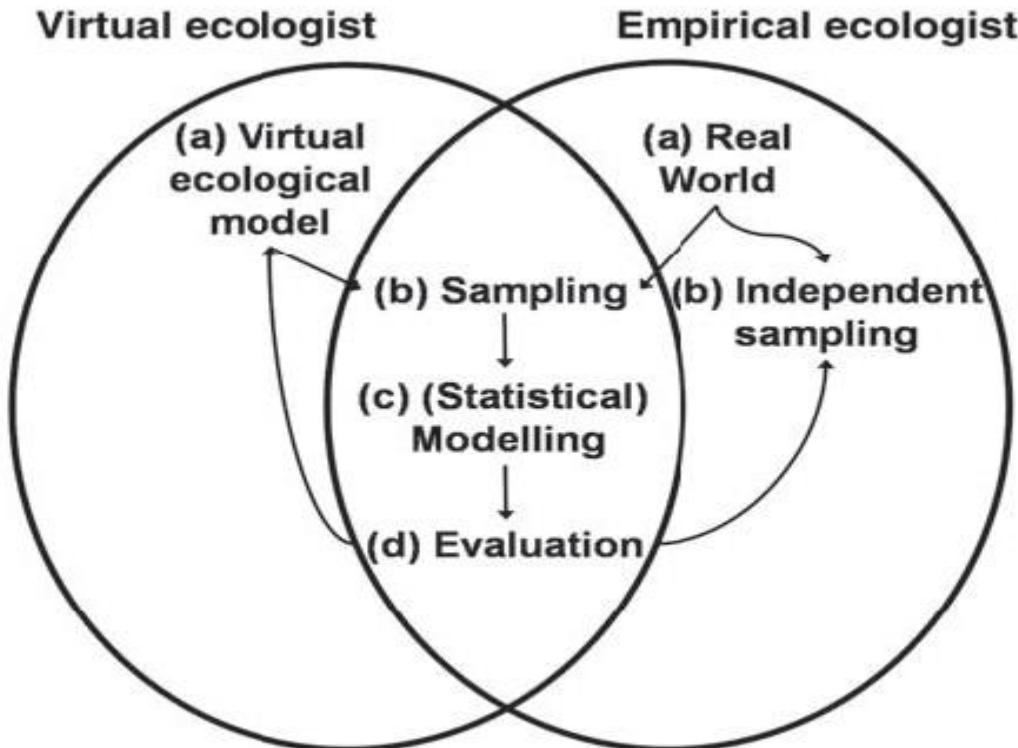


Figure 1. The elements of the virtual ecologist approach.

Zurrell et al 2010 *Oikos*

# Outline

## I- Species distribution models

- Virtual species: what have we learnt?
- The role of uncertainty in predicting species distributions

## II- Metacommunities and multiple facets of diversity

# I - Species distribution models

Virtual species: what have I learnt?

# Virtual Species

(a) Artificial Species Probability of Occurrence Map

(b) Presence/Absence Map

(c) Sample Presence/Absence and Environmental Layers  
50 times

(d) Use 80% of the Data to Build the Statistical Model

(e) Use the Other 20% of the Data to Test the Model

(f) Compare Real vs. Predicted Presence/Absence and Probability of Occurrence

Classification rate

Types of models

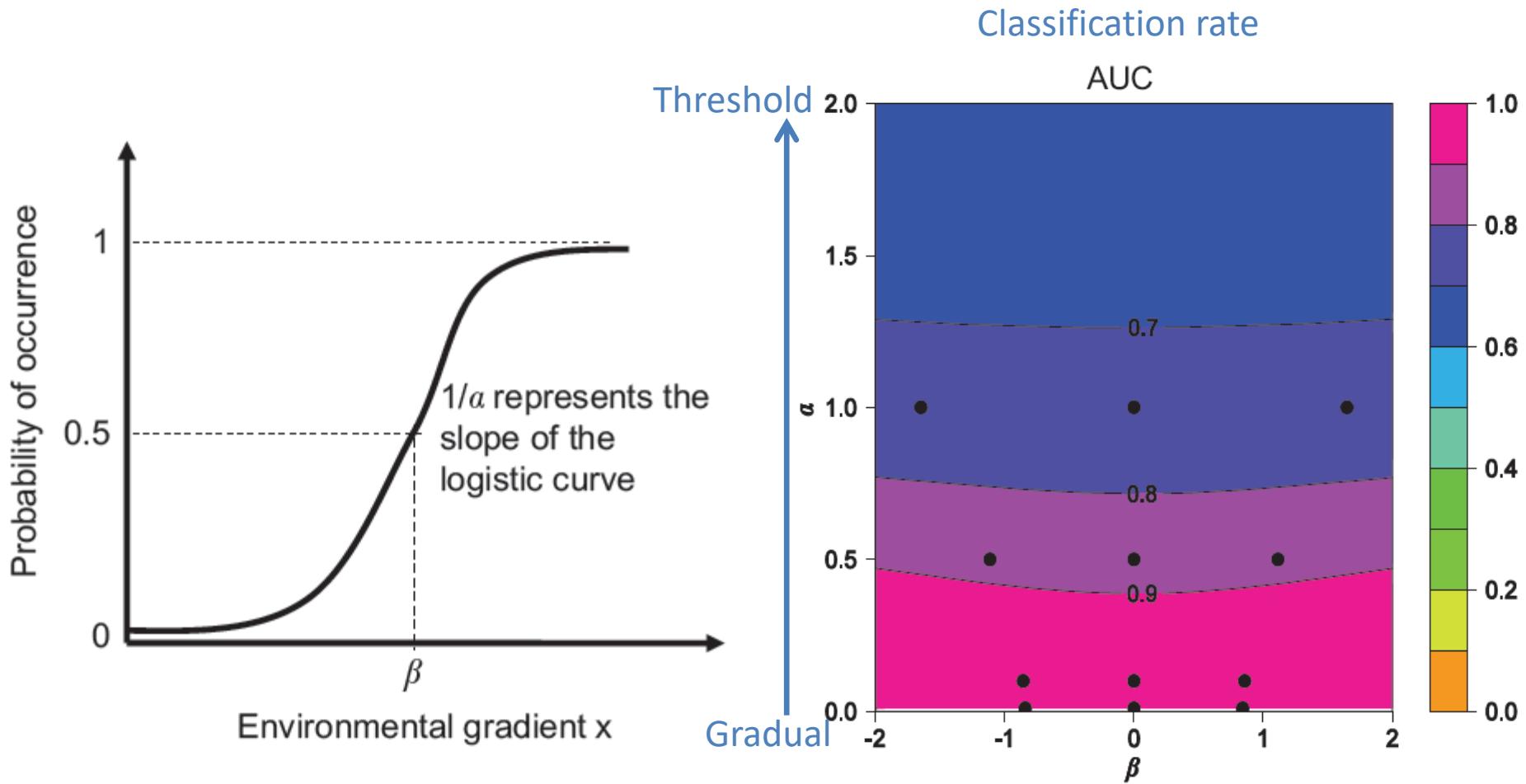
Types of species

Probabilities

Correlation

Meynard & Quinn 2007 *Journal of Biogeography*

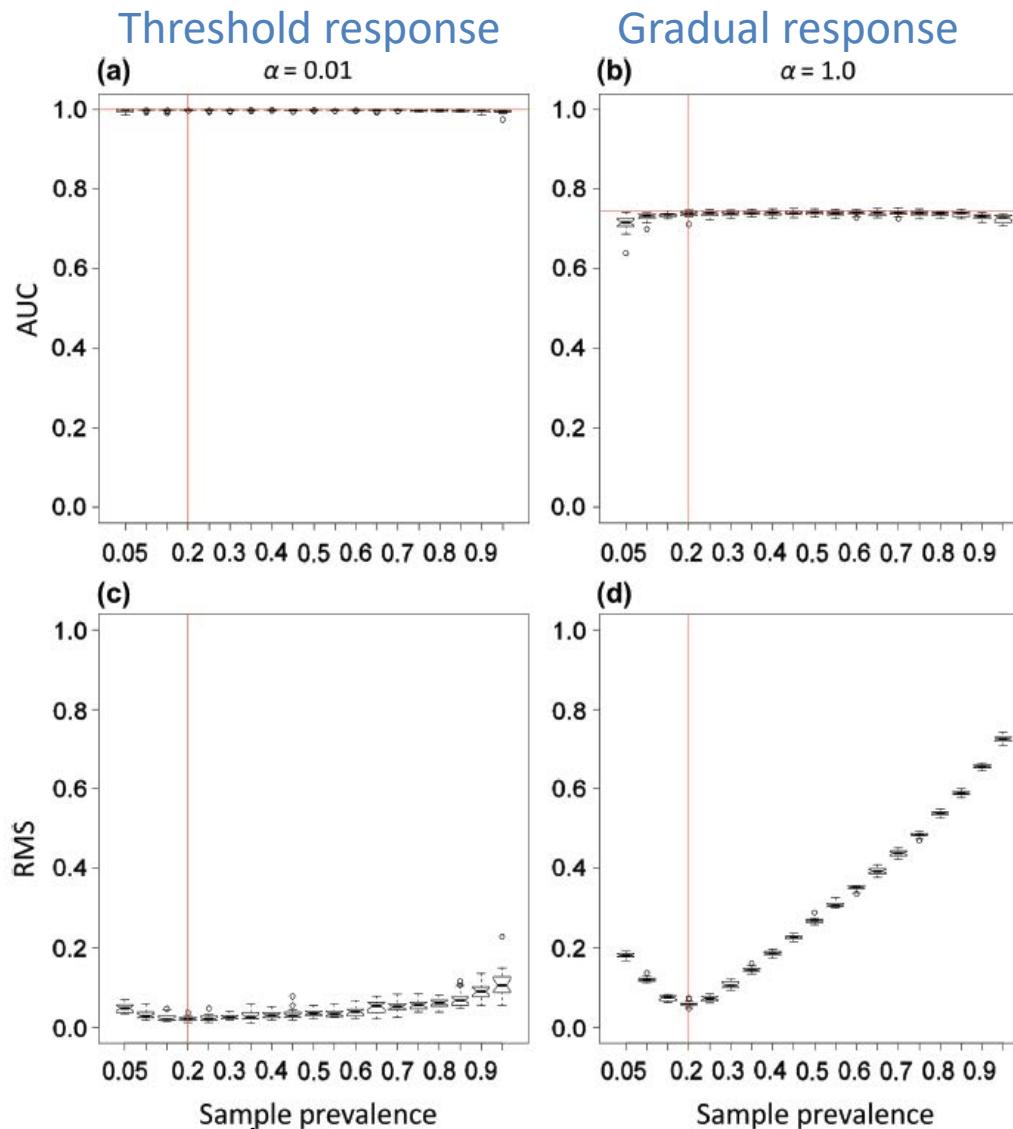
# Virtual Species



Meynard & Kaplan 2012 *Ecography*

# Virtual Species

Classification rate

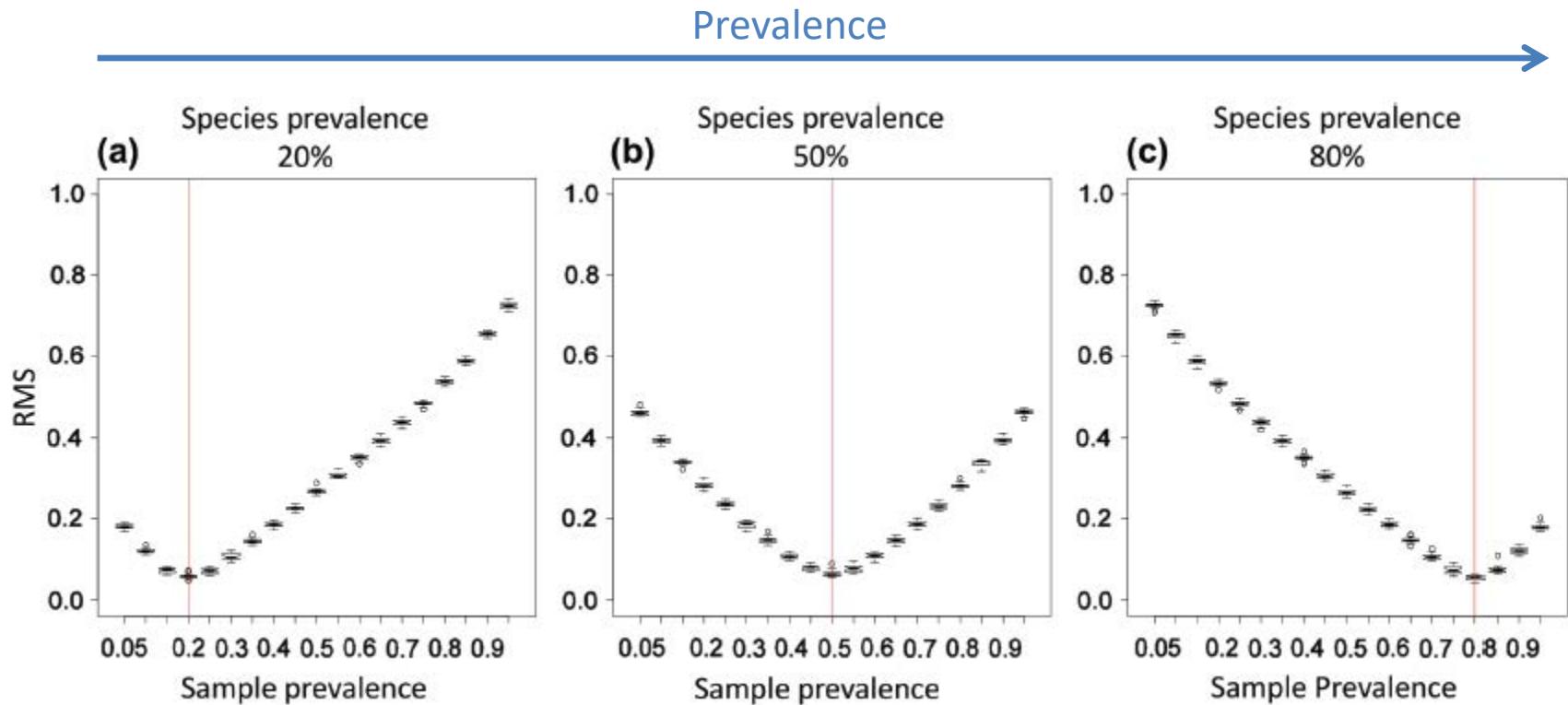


Probabilities

Meynard & Kaplan 2012 *Ecography*

# Virtual Species

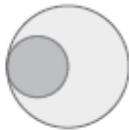
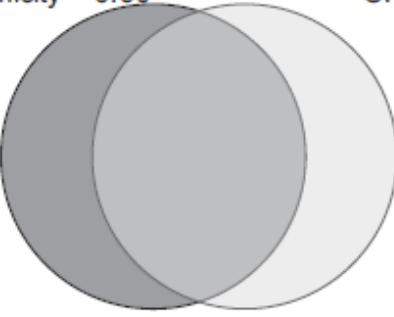
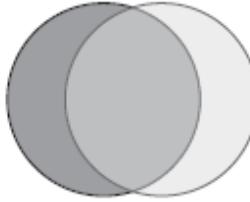
## Prediction of Probabilities



Meynard & Kaplan 2012 *Ecography*

# Virtual Species

## Prediction of the range

TSS = 0.97 Sensitivity = 1.00 Specificity = 0.97	Sørensen = 0.40 UPR = 0.00 OPR = 0.75	TSS = 1.00 Sensitivity = 1.00 Specificity = 1.00	Sørensen = 0.92 UPR = 0.00 OPR = 0.15	TSS = 0.85 Sensitivity = 0.85 Specificity = 1.00	Sørensen = 0.92 UPR = 0.15 OPR = 0.00
(a) 	(b) 	(c) 			
TSS = 0.60 Sensitivity = 0.70 Specificity = 0.90	Sørensen = 0.70 UPR = 0.30 OPR = 0.30	TSS = 0.67 Sensitivity = 0.70 Specificity = 0.97	Sørensen = 0.70 UPR = 0.30 OPR = 0.30	TSS = 0.70 Sensitivity = 0.70 Specificity = 1.00	Sørensen = 0.70 UPR = 0.30 OPR = 0.30
(d) 	(e) 	(f) 			
75% overprediction, 0% under-prediction, prevalence = 0.01	15% over-prediction, 0% under-prediction, prevalence = 0.01	0% over-prediction, 15% under-prediction, prevalence = 0.01			

Leroy et al 2018 *Journal of Biogeography*

# Virtual Species

## Standards for virtual species studies

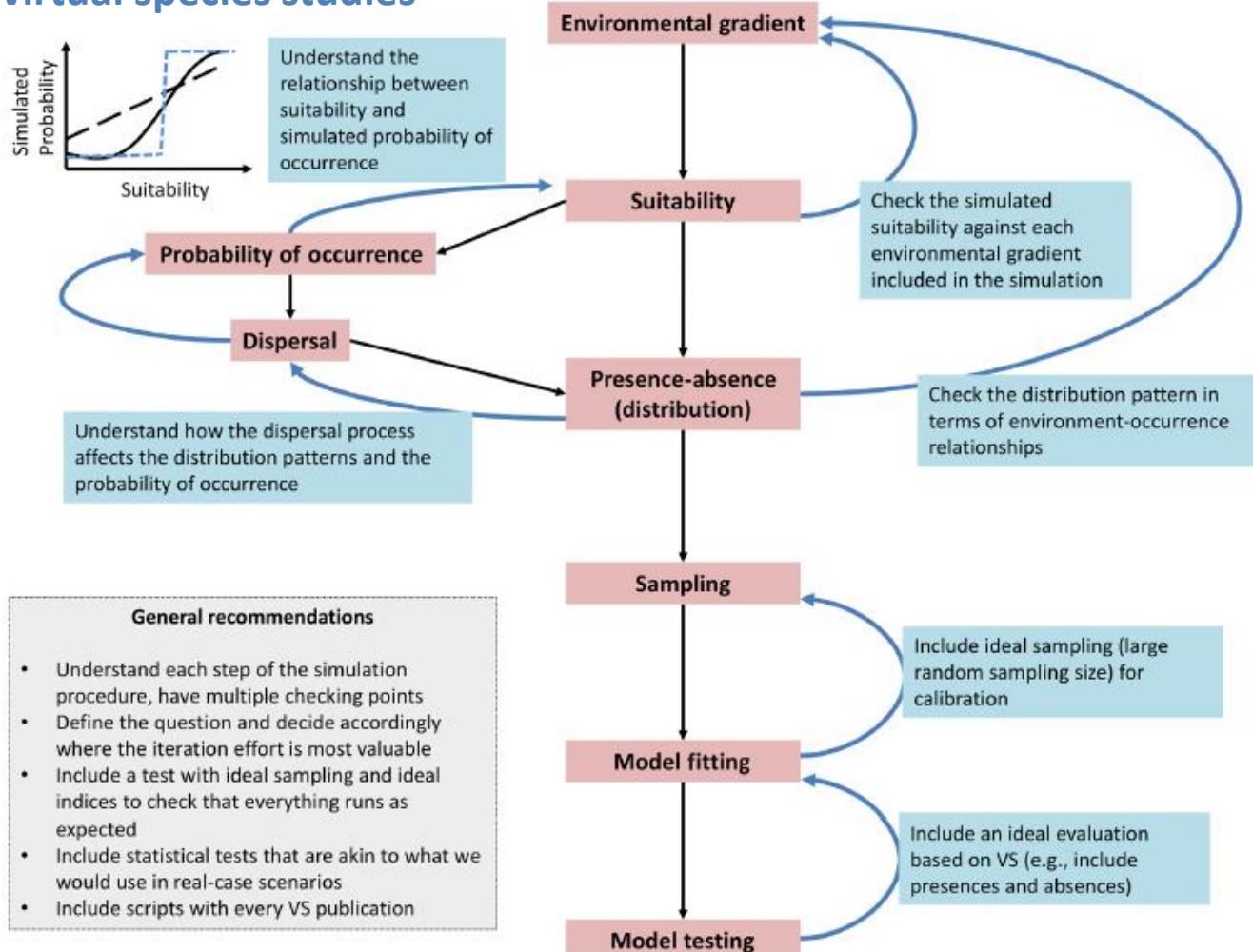
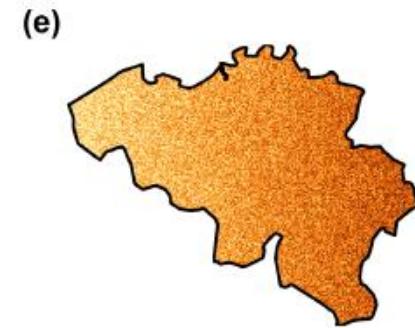
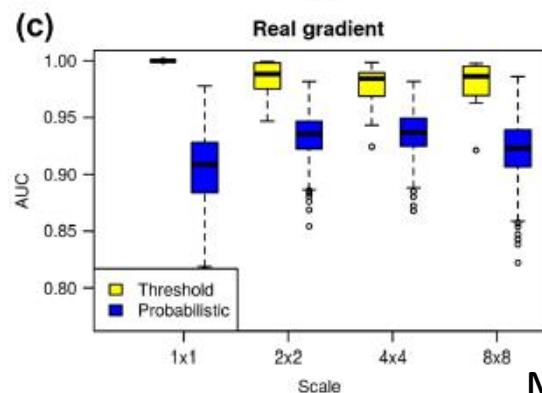
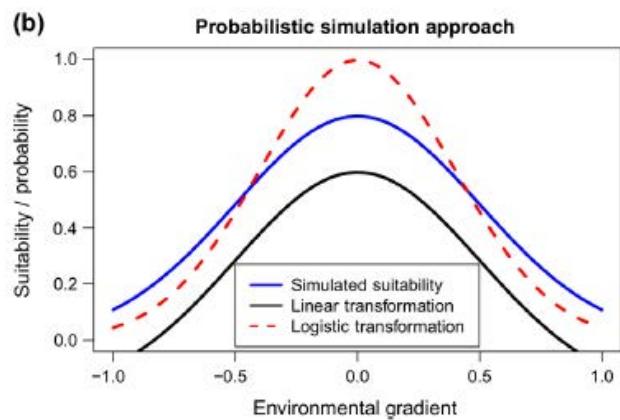
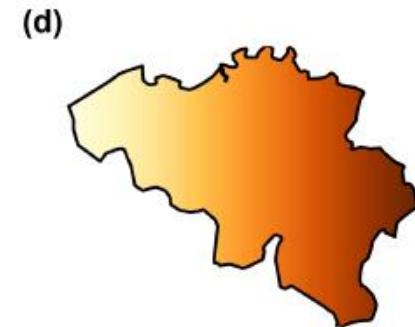
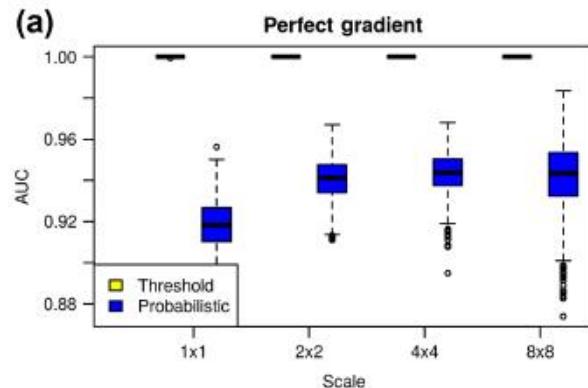
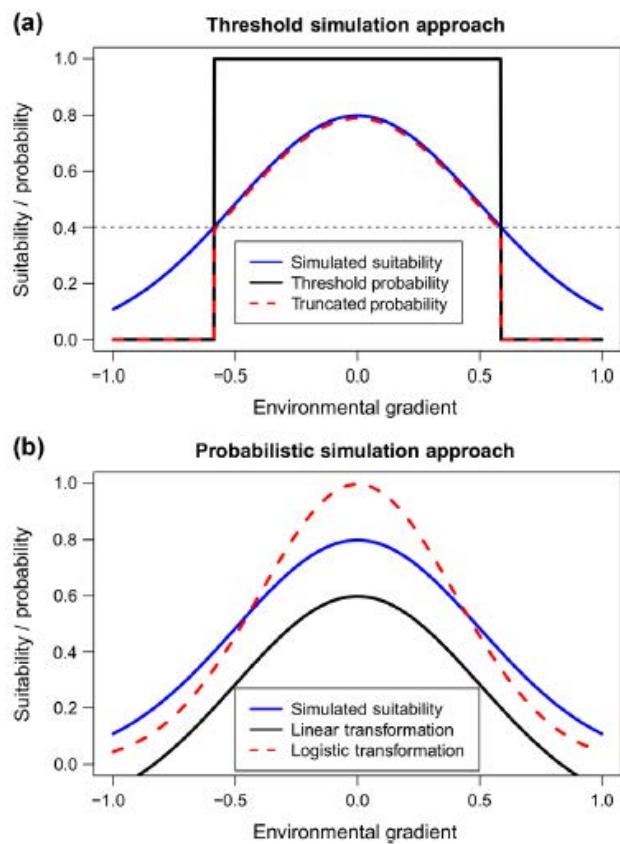
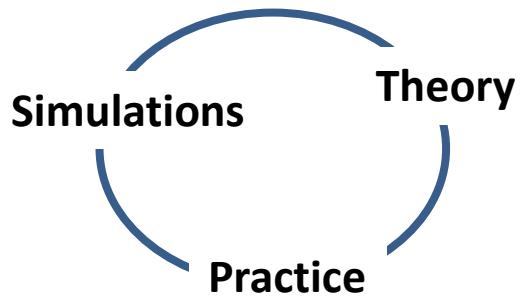


Figure 3. Graphical representation of recommendations and guidelines proposed in section 'General recommendations and guidelines' for future virtual species studies.

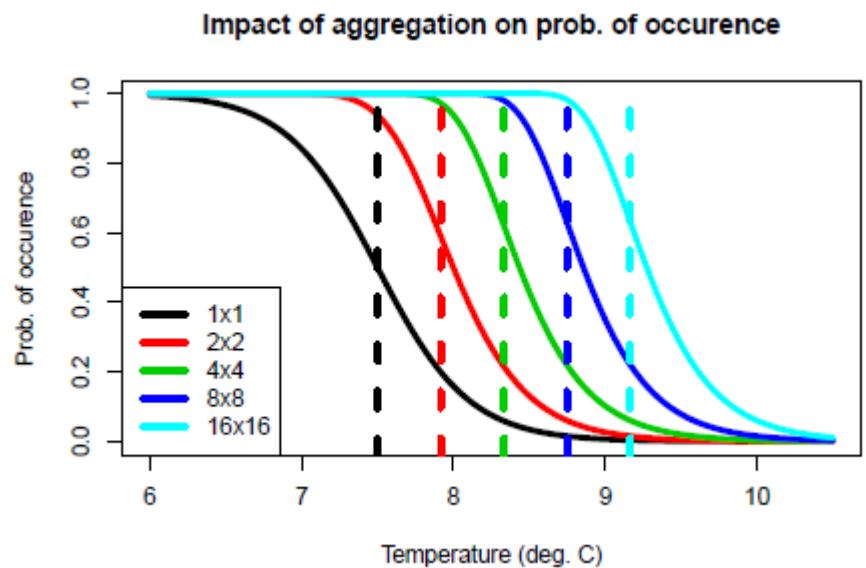
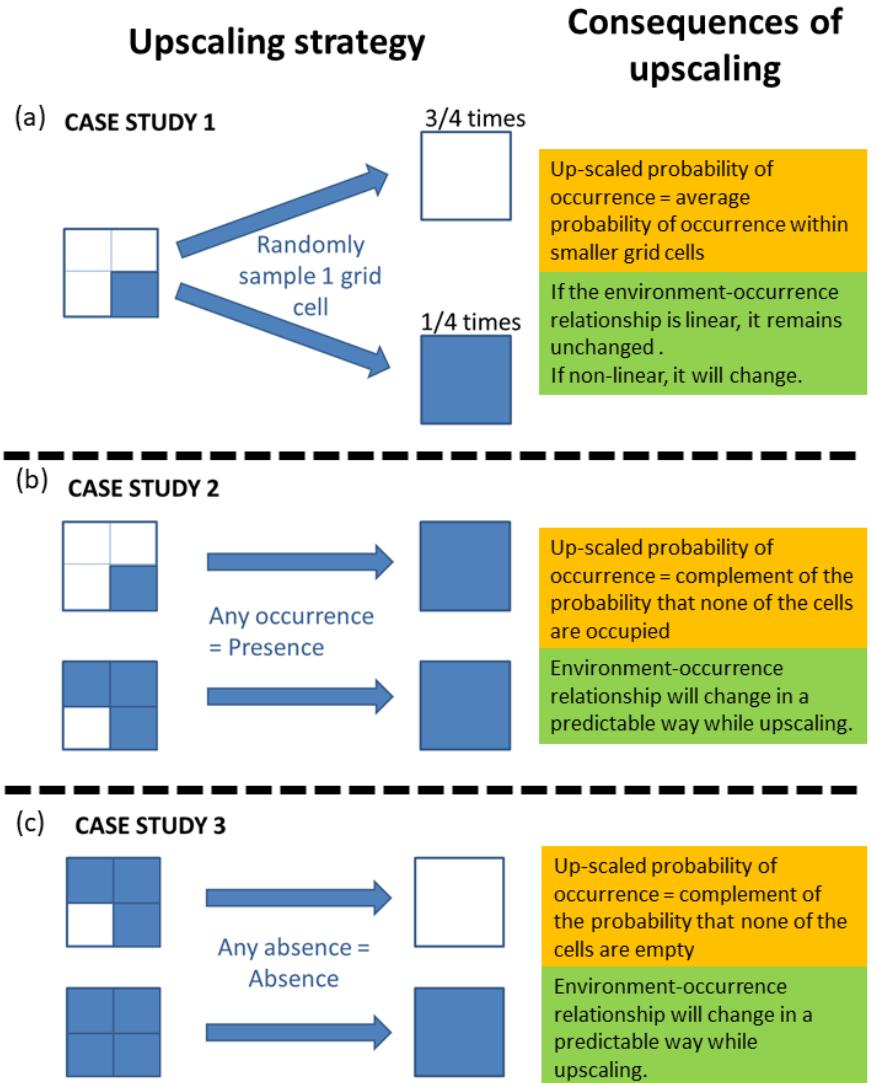
Meynard et al 2019 *Ecography*

# Virtual Species



Meynard et al 2019 *Ecography*

# A theoretical framework of upscaling for species distribution models



Meynard & Kaplan (*In Prep*)

# Virtual Species

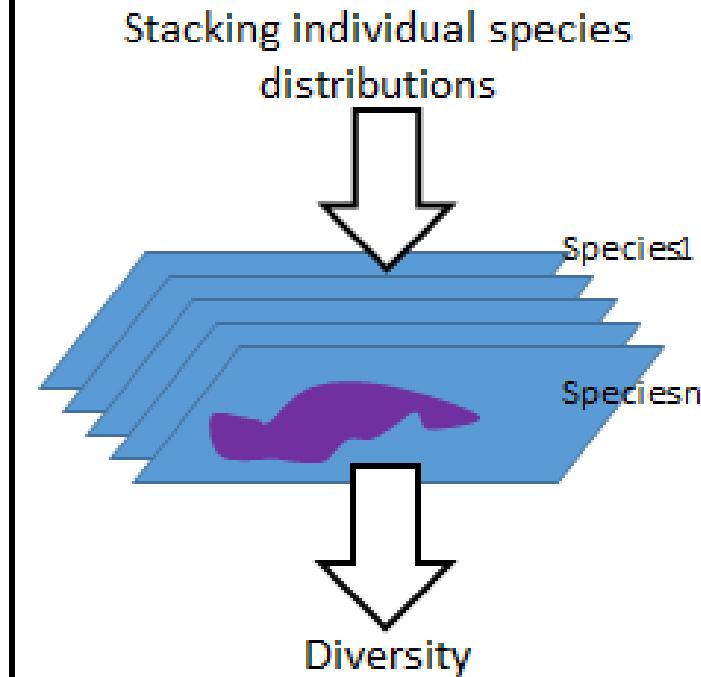
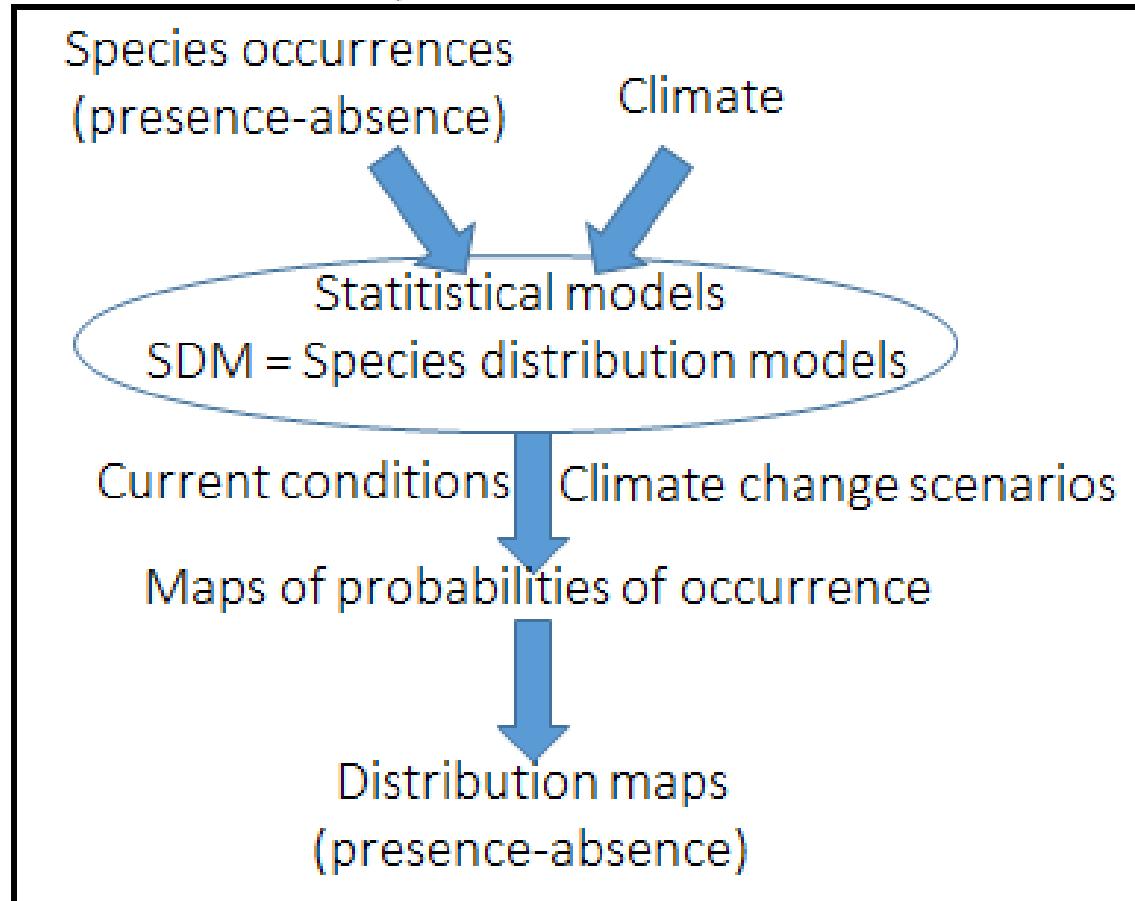
- Predictive ability in terms of probabilities versus presence-absence classifications are not equivalent
- Upscaling data usually changes BOTH the probability of occurrence AND the shape of the environment-occurrence relationship in a predictable way

# I - Species distribution models

Uncertainties

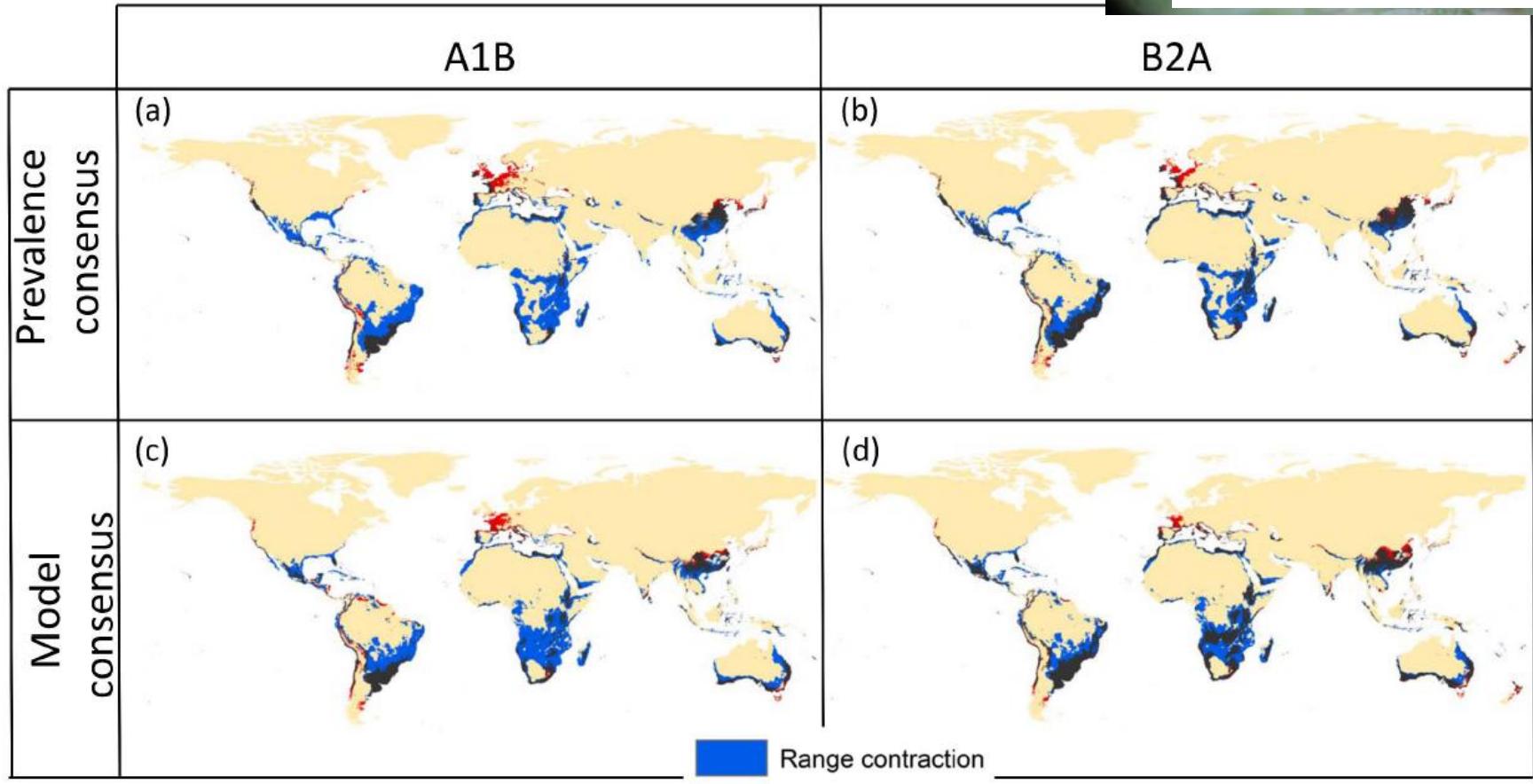
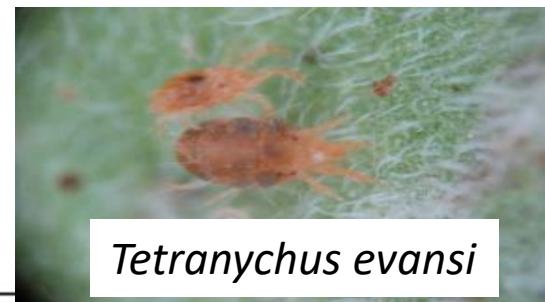
# Uncertainties

## Species distribution models (SDMs)



# Uncertainties

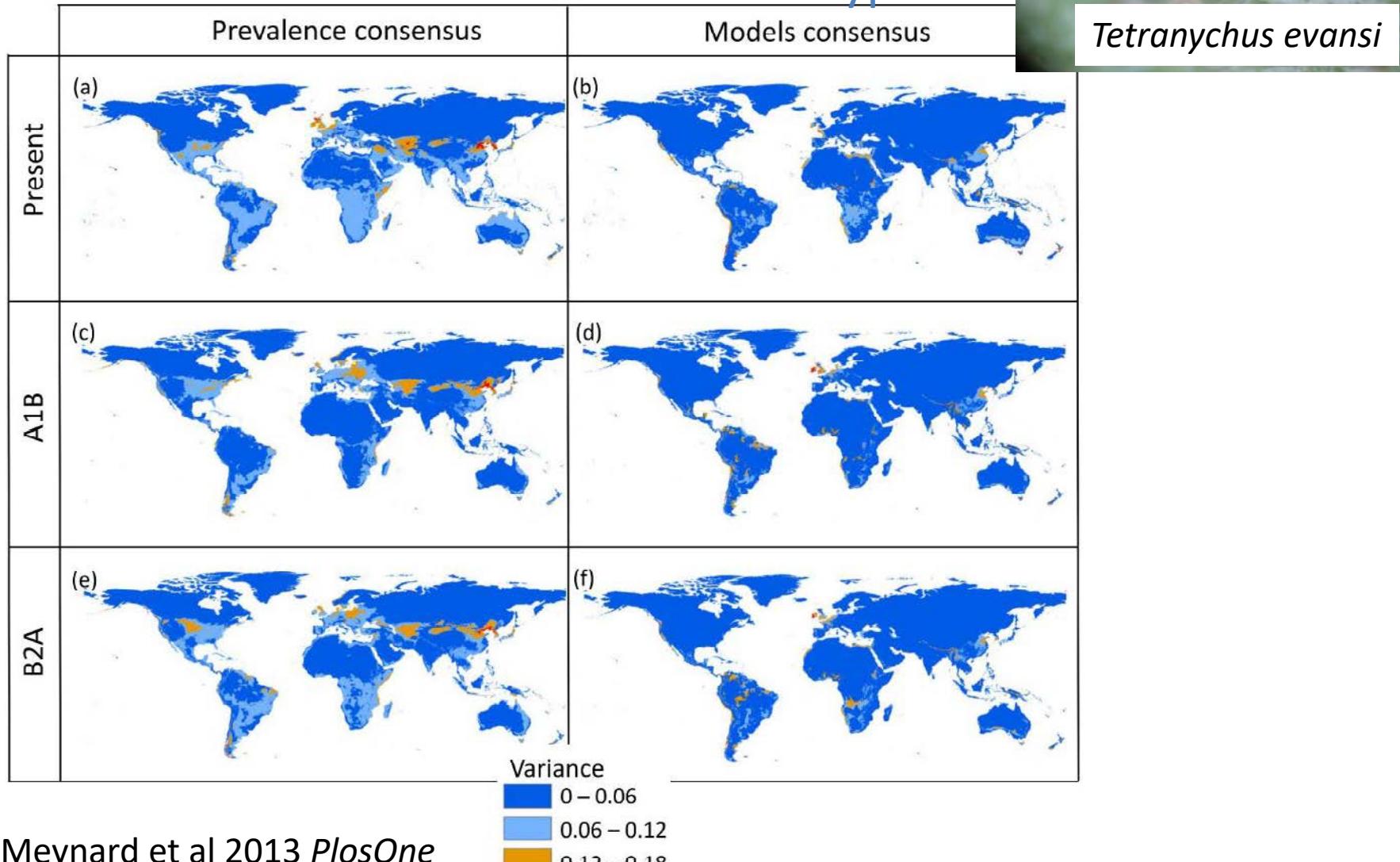
## Prevalence and model types



Meynard et al 2013 *PlosOne*

# Uncertainties

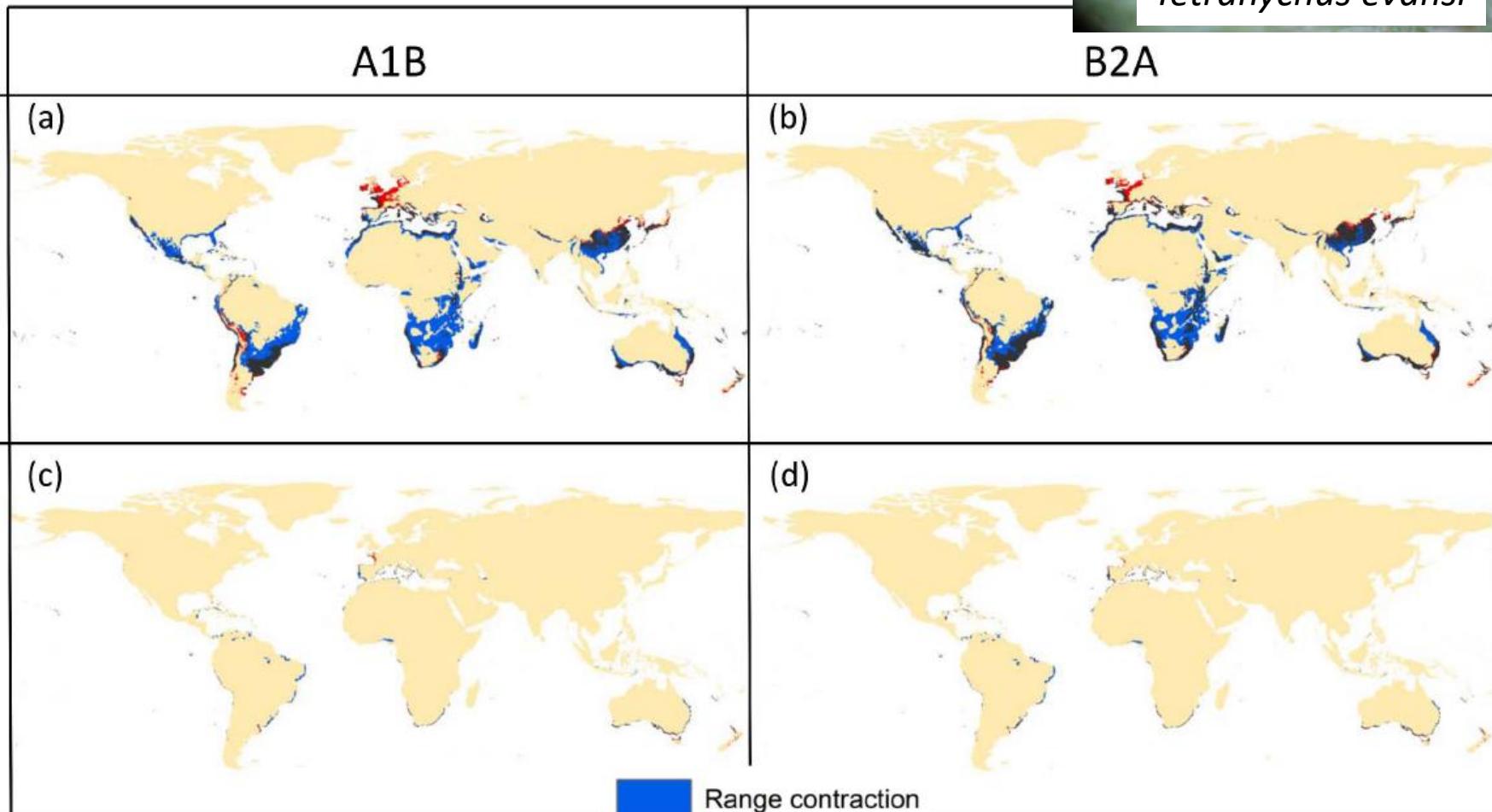
## Prevalence and model types



Meynard et al 2013 *PlosOne*

# Uncertainties

## Species vs sub-species



- Range contraction
- Outside range
- Range expansion
- Unchanged range

Meynard et al 2013 *PlosOne*

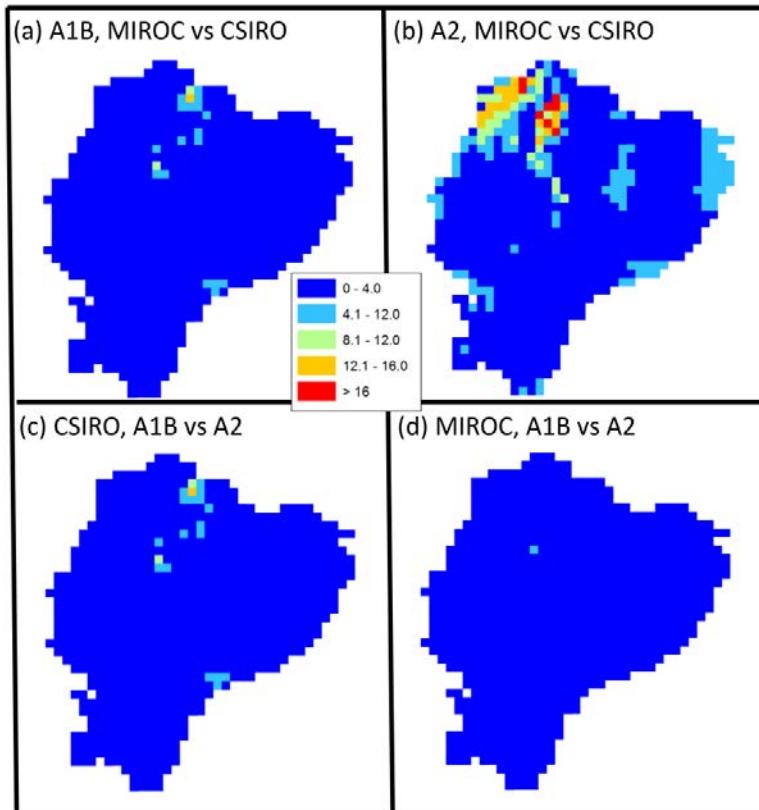
# Uncertainties



Climate model vs socio-economic scenario

200 mammals in Ecuador 1) 2) 3) 4) 5) 6)

Between GCMs



For A2, there are great differences between GCMs

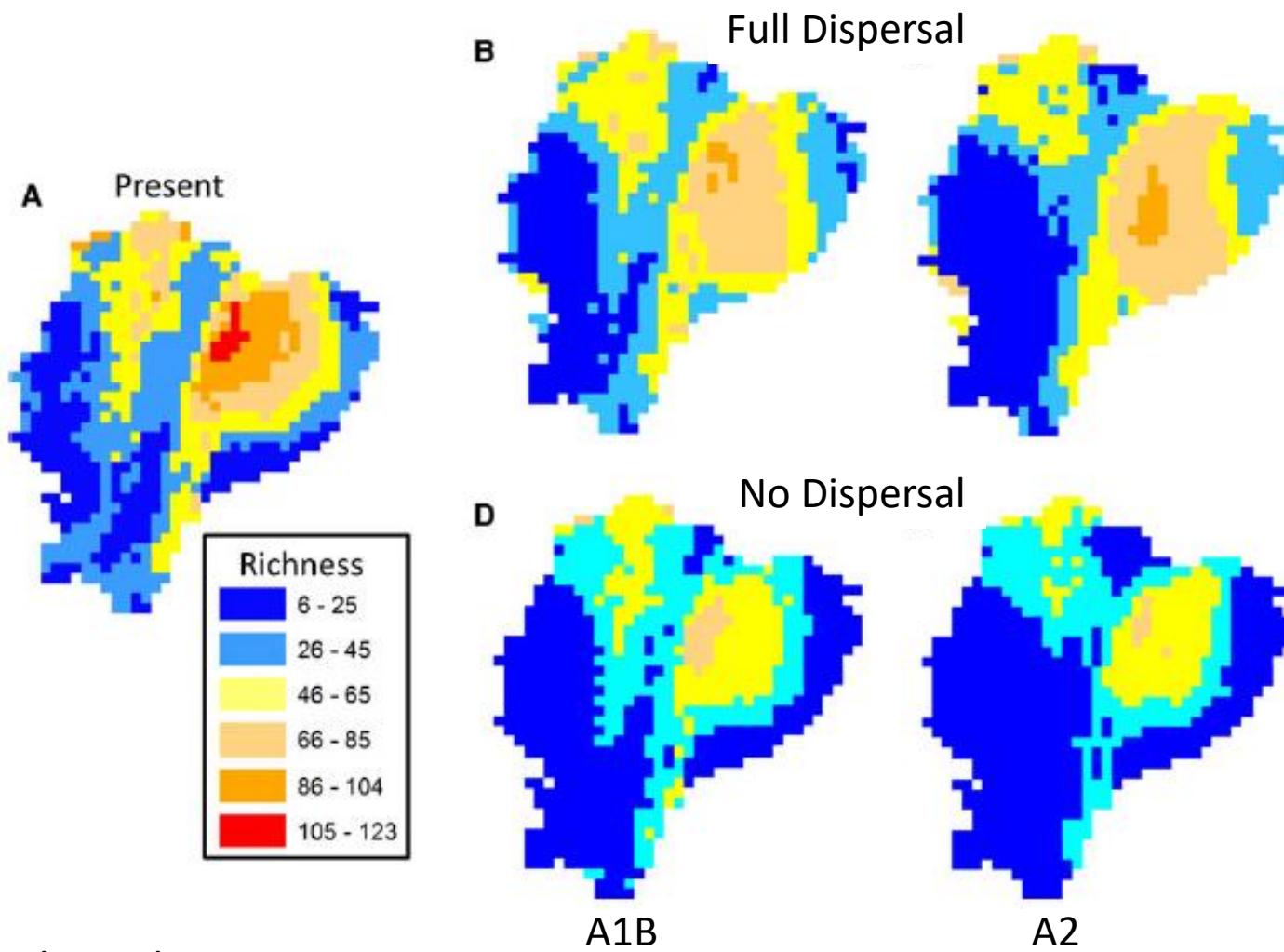
Between scenarios

Coefficient of variation between richness estimates in 2080.

Iturralde-Polit et al 2017 *Biotropica*

# Uncertainties

## Dispersal scenario



Iturralde-Polit et al 2017 *Biotropica*

# Uncertainties

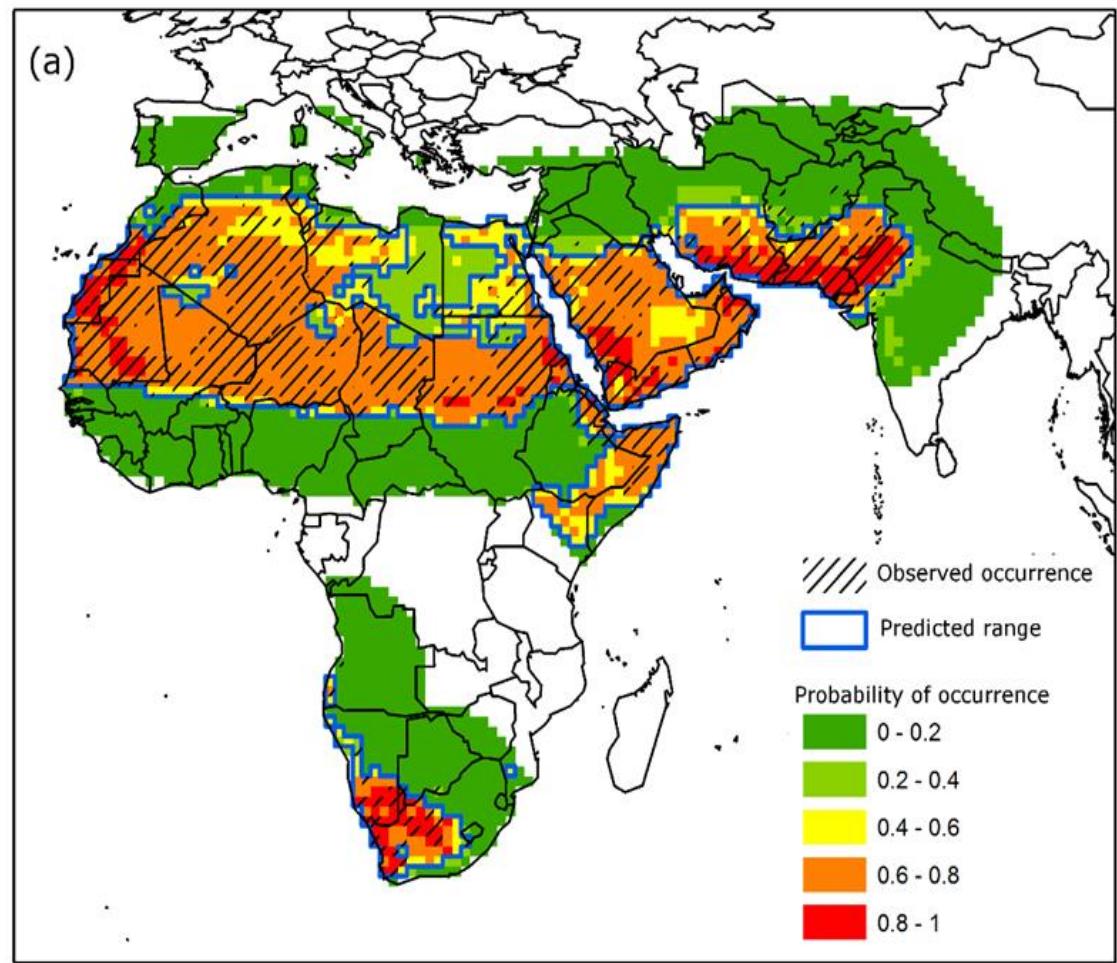
## Types of models

Desert Locust



**Ensemble forecasting**  
with 8 different models

- Maxent
- Bioclim
- Domain
- GAM
- GLM
- BRT
- RandomForest
- Classification trees



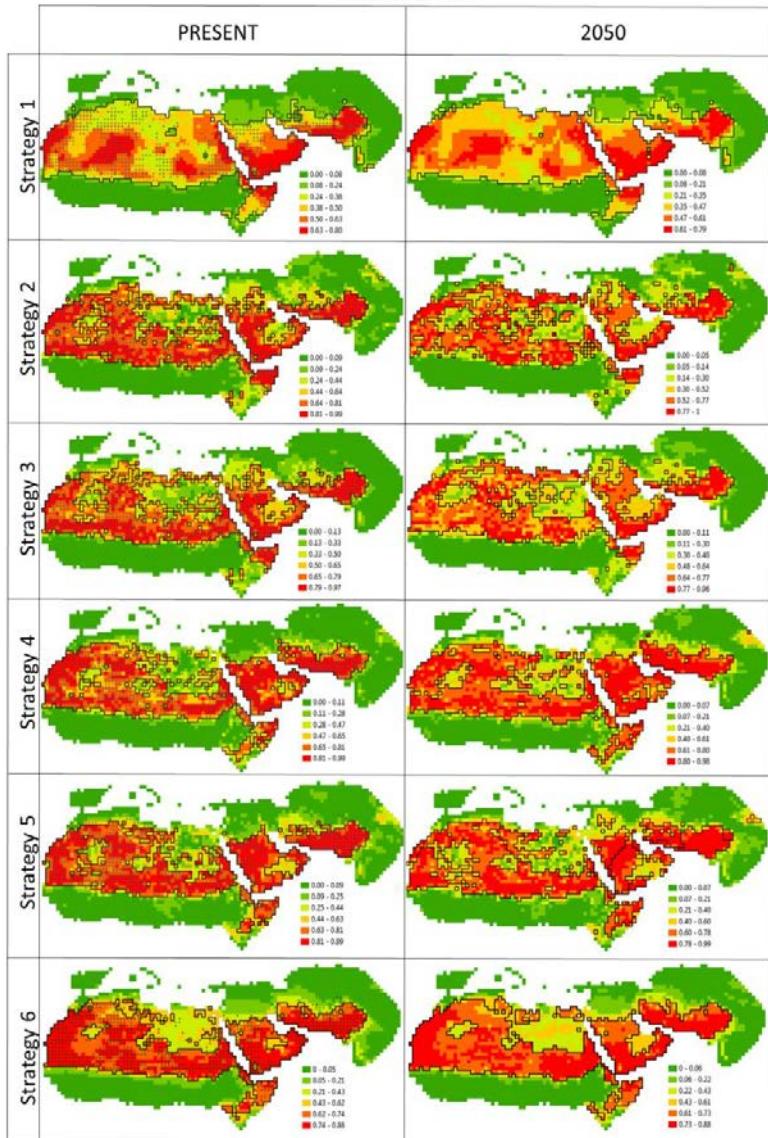
Meynard et al 2017  
*Global Change Biology*

# Uncertainties

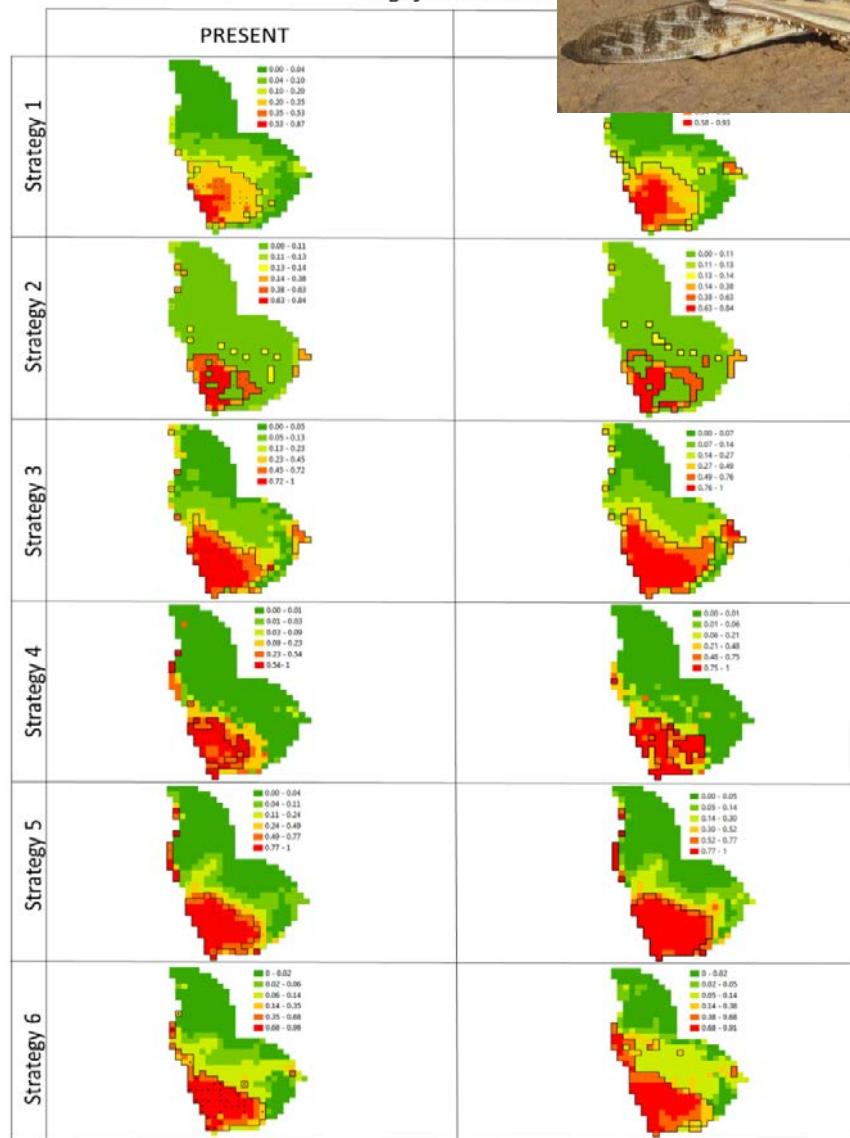
Desert Locust

## Variable selection

*S. g. gregaria*



*S. g. flaviventris*



# Uncertainties

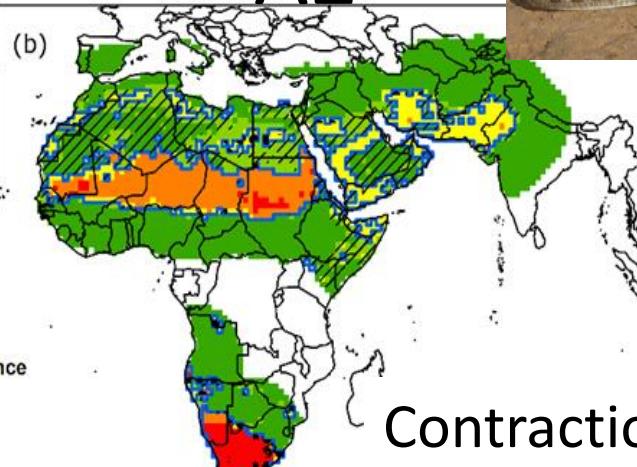
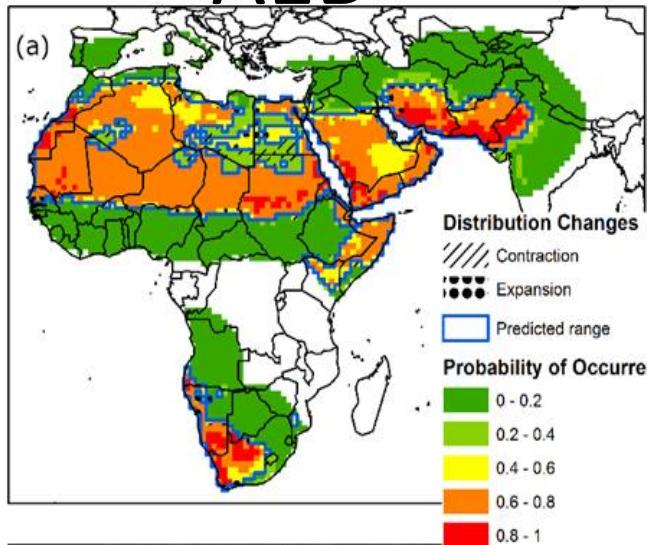
## Climatic scenarios

Desert Locust

A1B

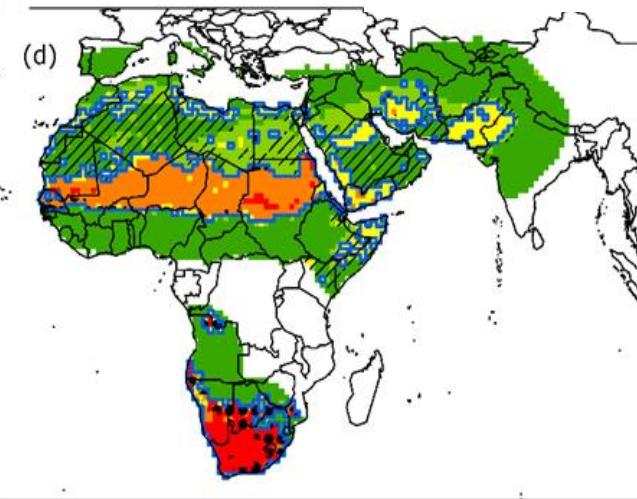
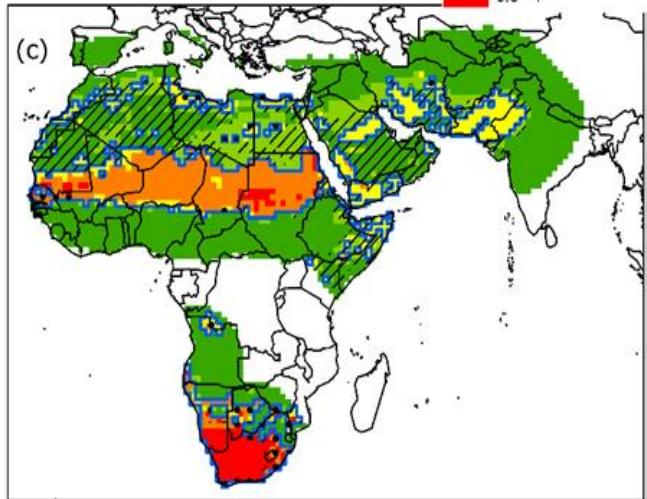
A2

2050



Contraction in the north  
Expansion in the south

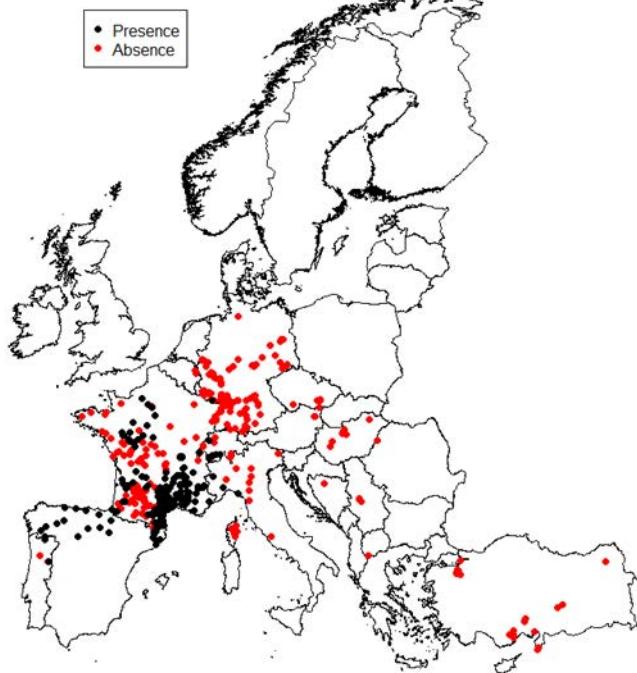
2090



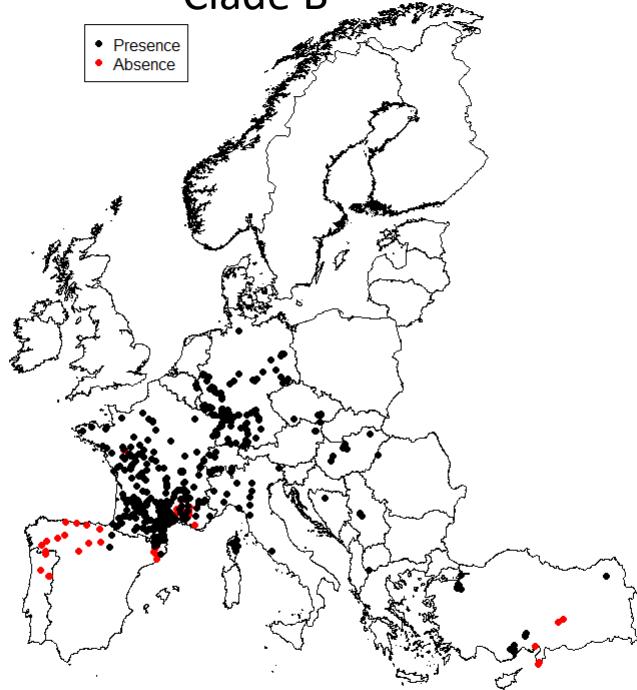


# Occurrence data

Clade A



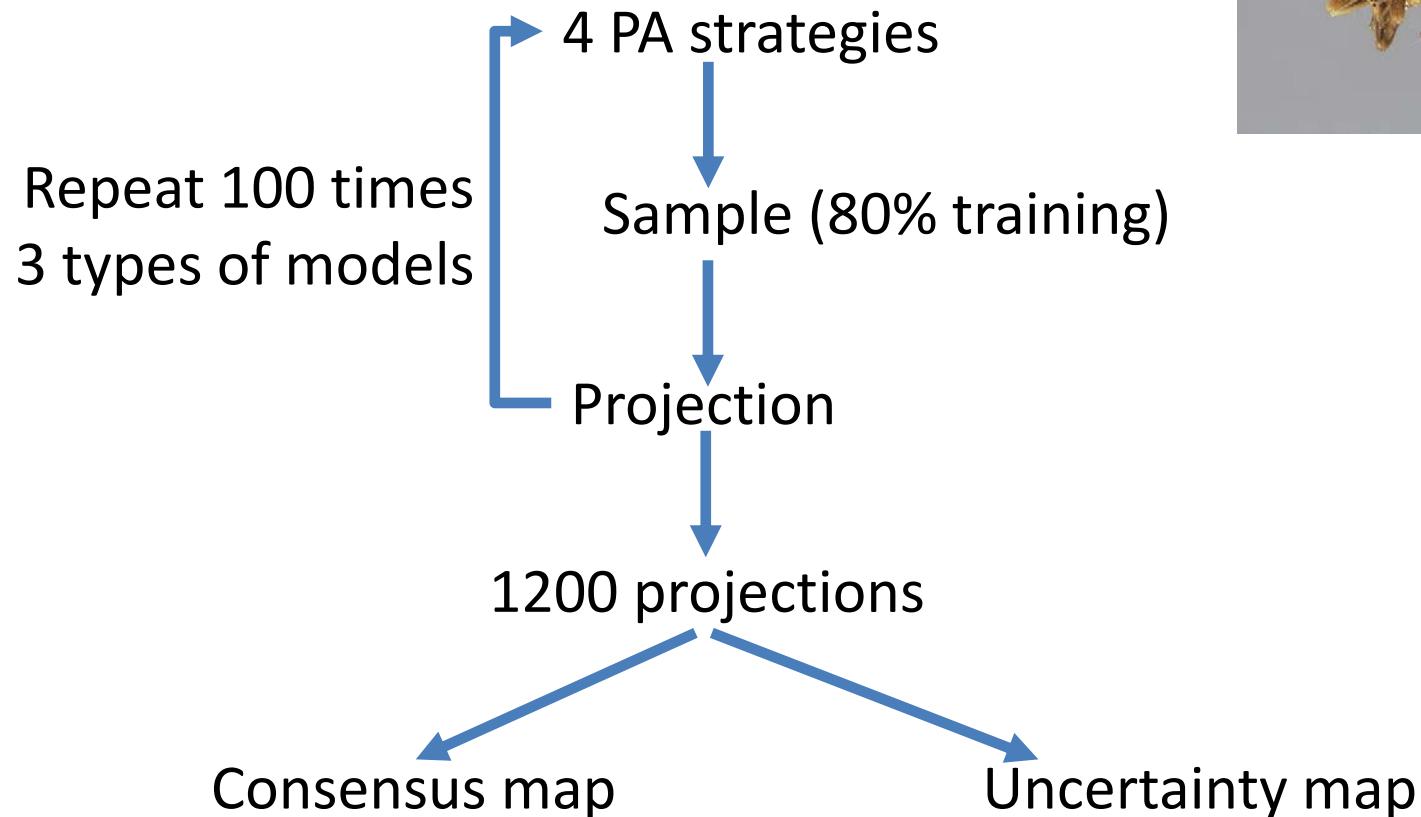
Clade B



# Presence-only data modelling

## Adding Pseudo-Absences (PA)

### Iterative approach

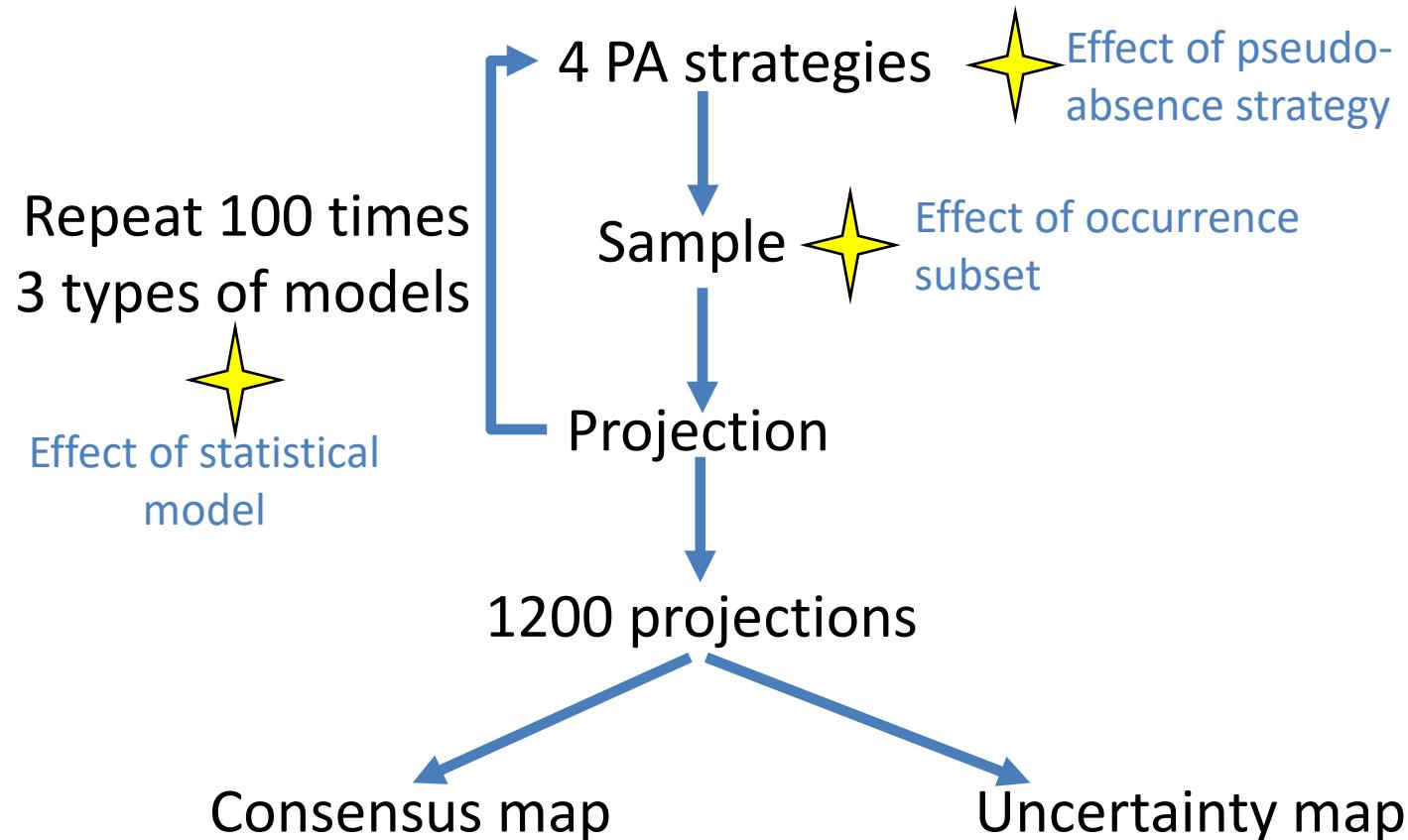


# Presence-only data modelling

## Adding Pseudo-Absences (PA)



### Iterative approach

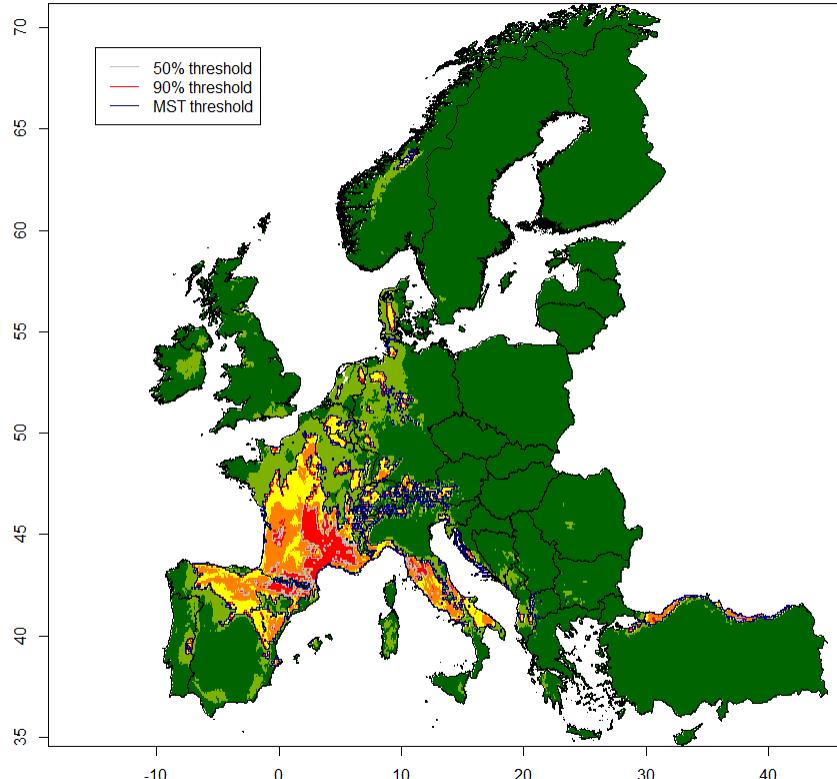


*Cacopsylla pruni*

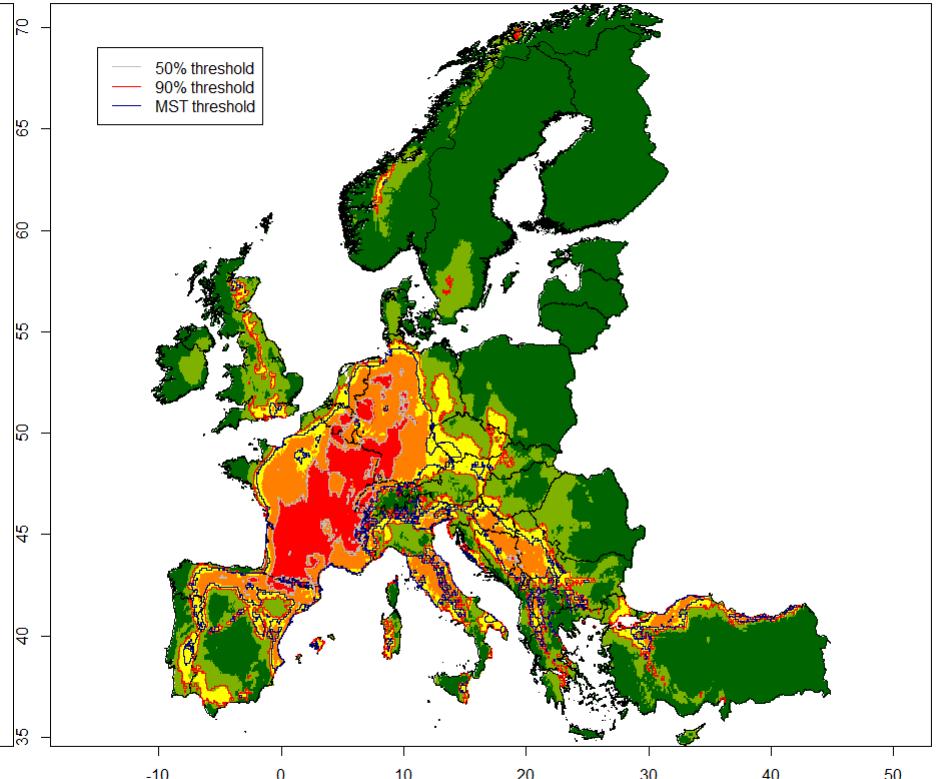


# Psyllid current distributions

Clade A



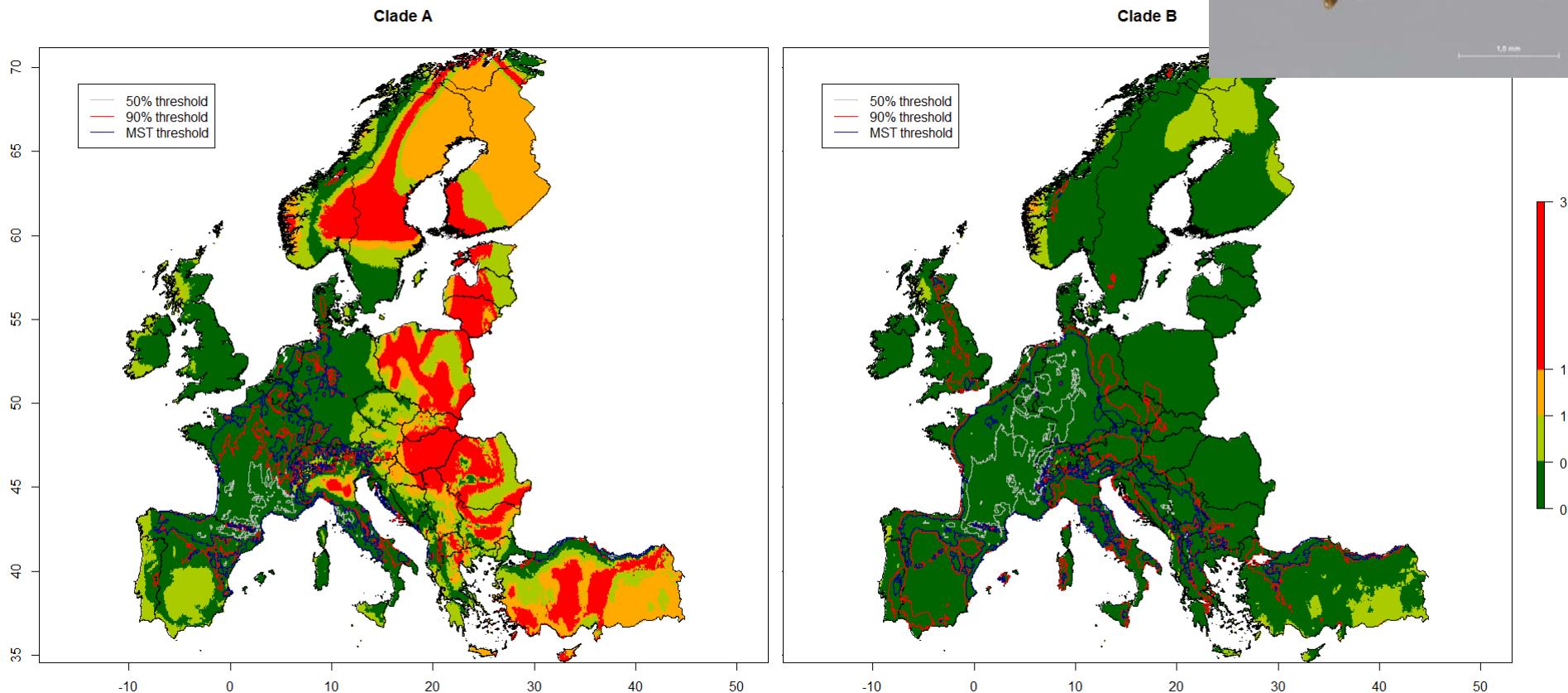
Clade B



Consensus from 1,200 maps:  
3 models x 4 PA methods x 100 iterations each



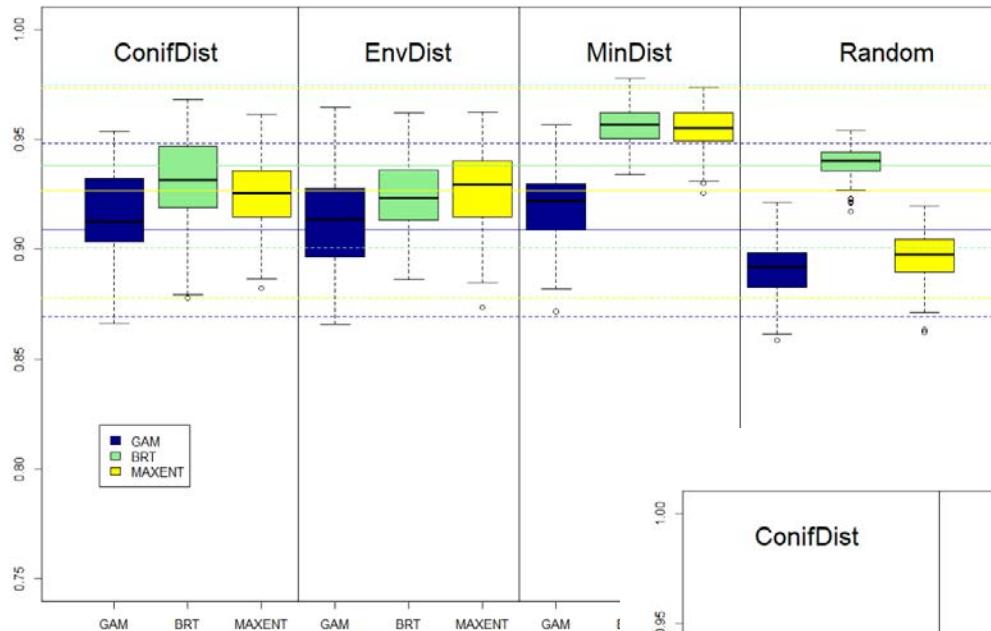
# Psyllid current distributions



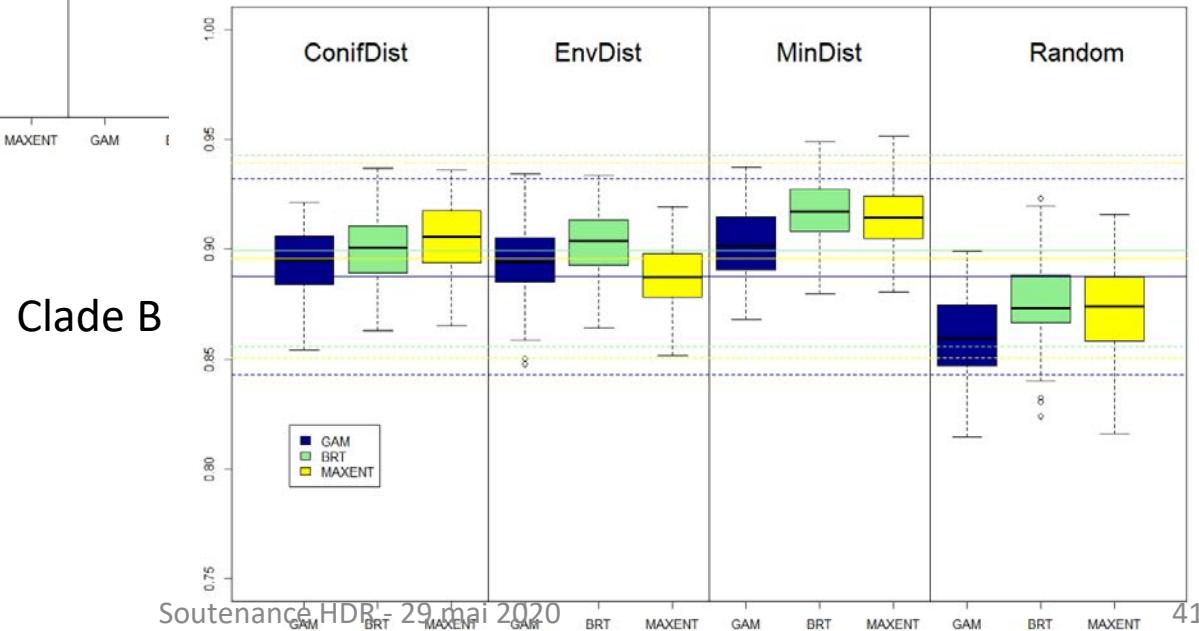
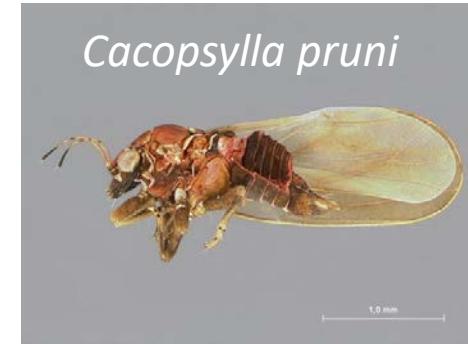
CV between 1,200 maps (standard deviation / mean):  
3 models x 4 PA methods x 100 iterations each

# Uncertainties

## Types of pseudo-absences



Clade A



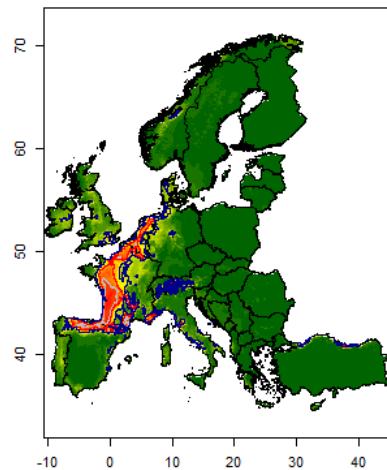
# Uncertainties

## Climate models x SE scenario

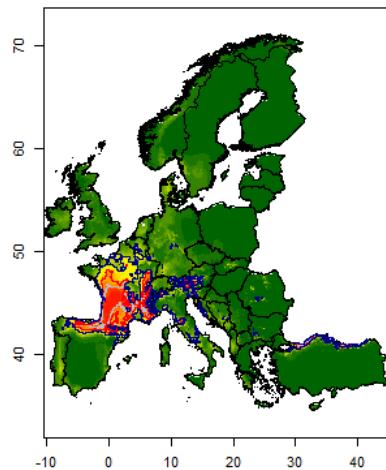
*Cacopsylla pruni*



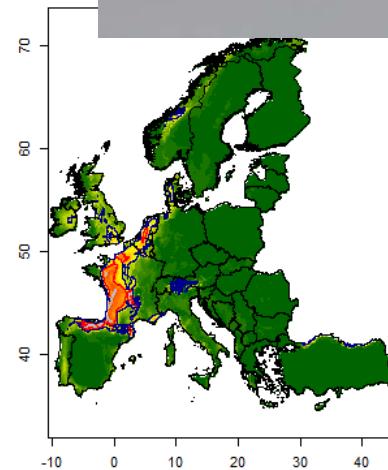
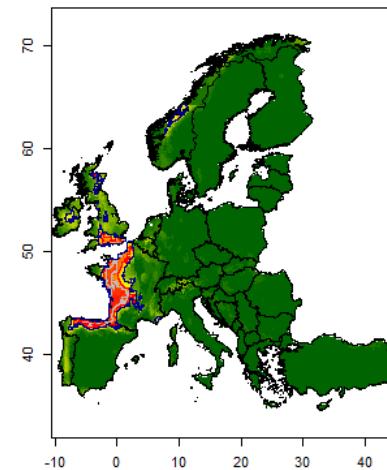
Clade A, RCP4.5-CC-2070



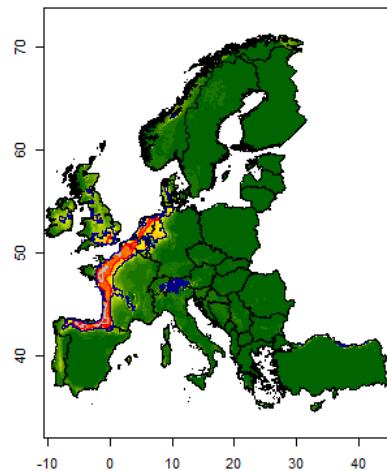
Clade A, RCP4.5-IP-2070



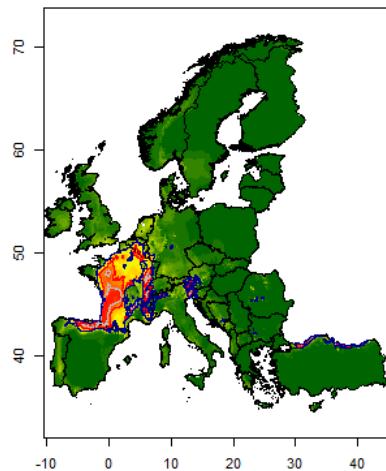
Clade A, RCP4.5-MR-2070



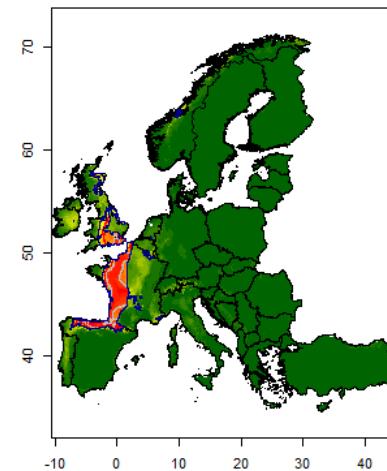
Clade A, RCP8.5-IP-2070



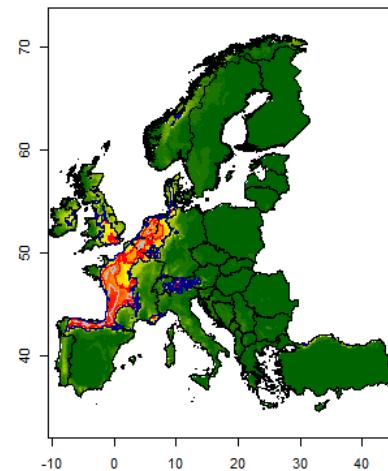
Clade A, RCP8.5-MR-2070



Clade A, RCP8.5-NO-2070

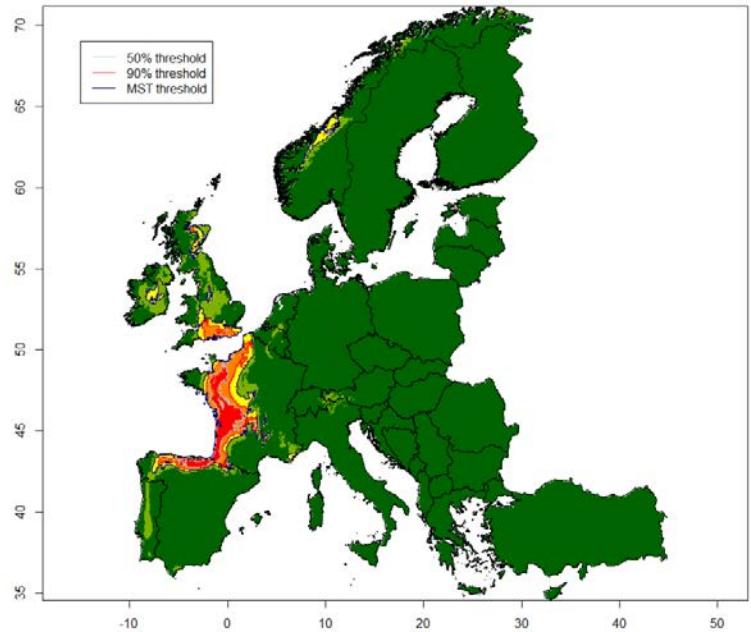


Clade A, RCP8.5-NO-2070



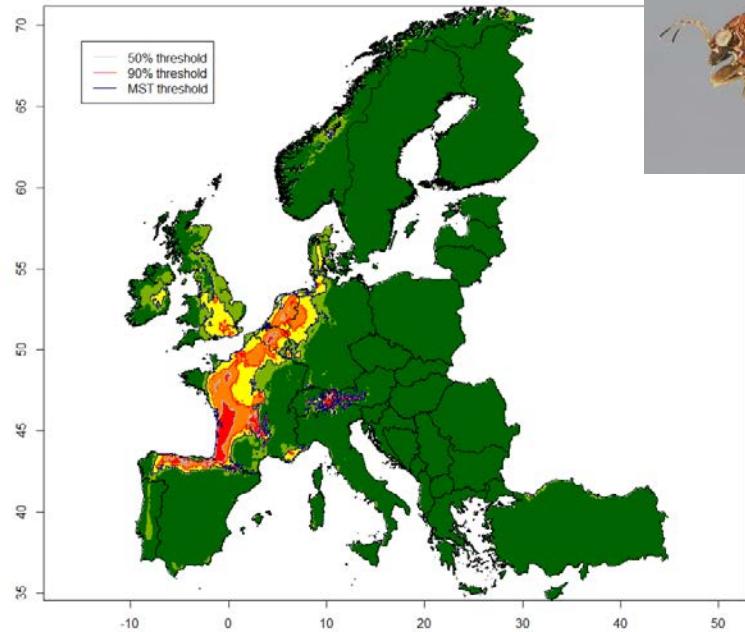
RCP 4.5 - MR

Clade A

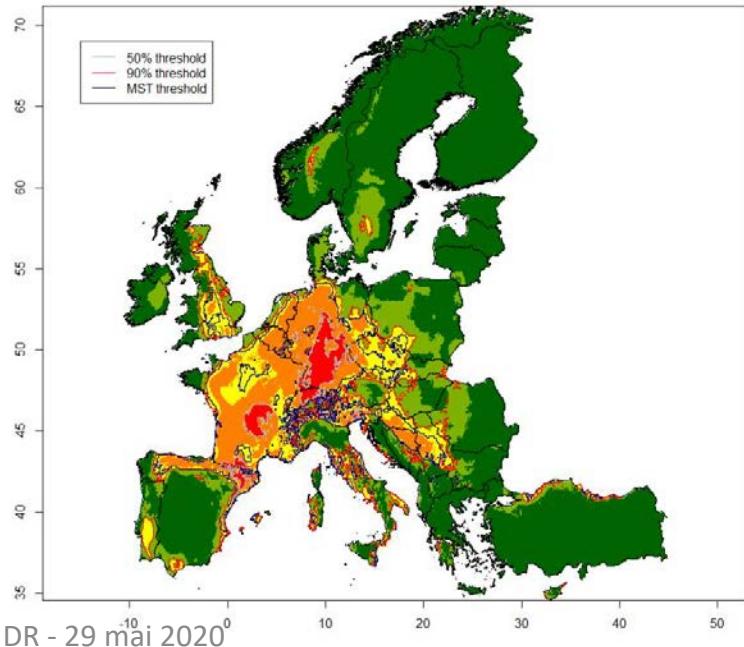
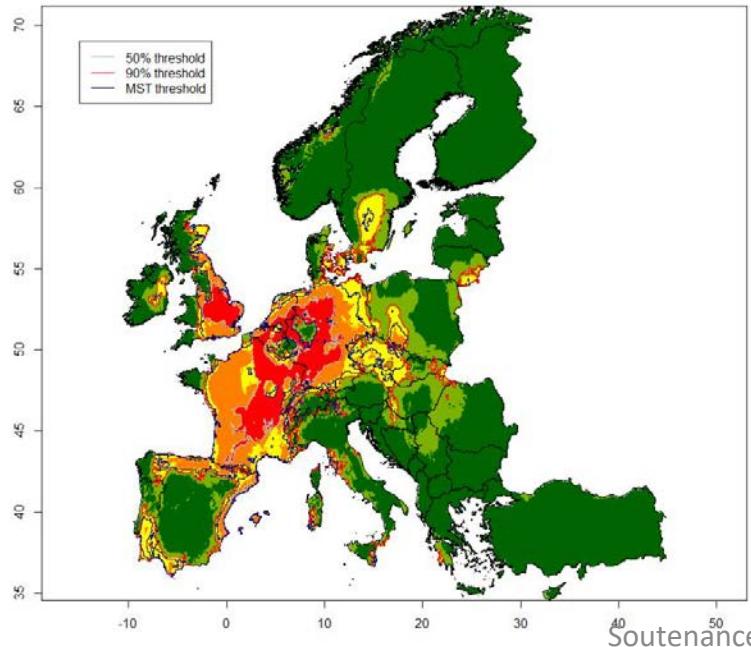


RCP 8.5 - NO

Clade A



Clade B

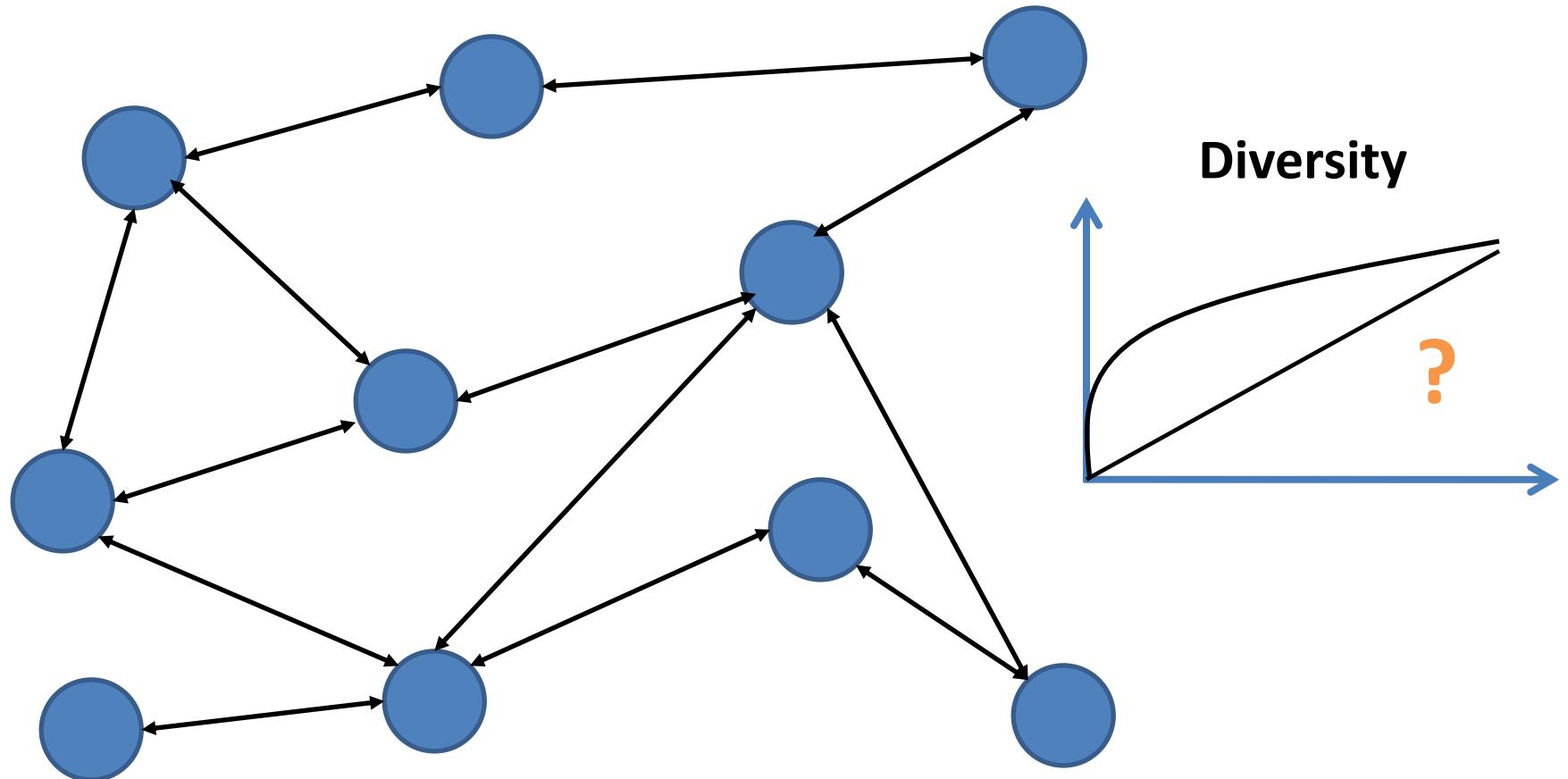


# Uncertainties

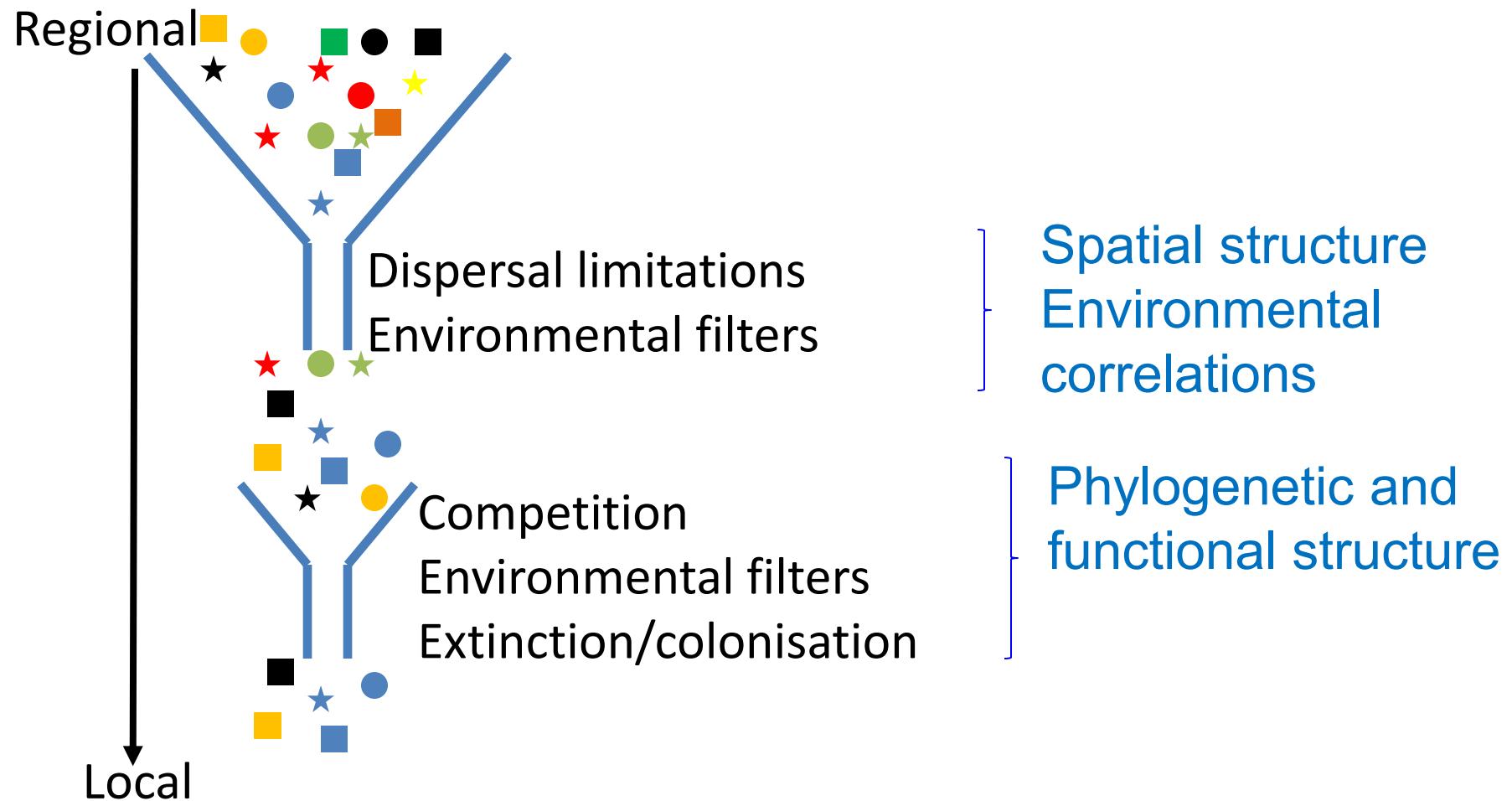
- There is no solution for SDMs with presence-only data that will fit all cases
- Combining modelling with iterations to integrate sources of uncertainty
- Ensemble forecasting has been advocated for model uncertainty: can be extended to other sources of uncertainty
- Another alternative: use virtual species to do sensitivity analyses

## II – Metacommunities and multiple facets of diversity

# Metacommunities

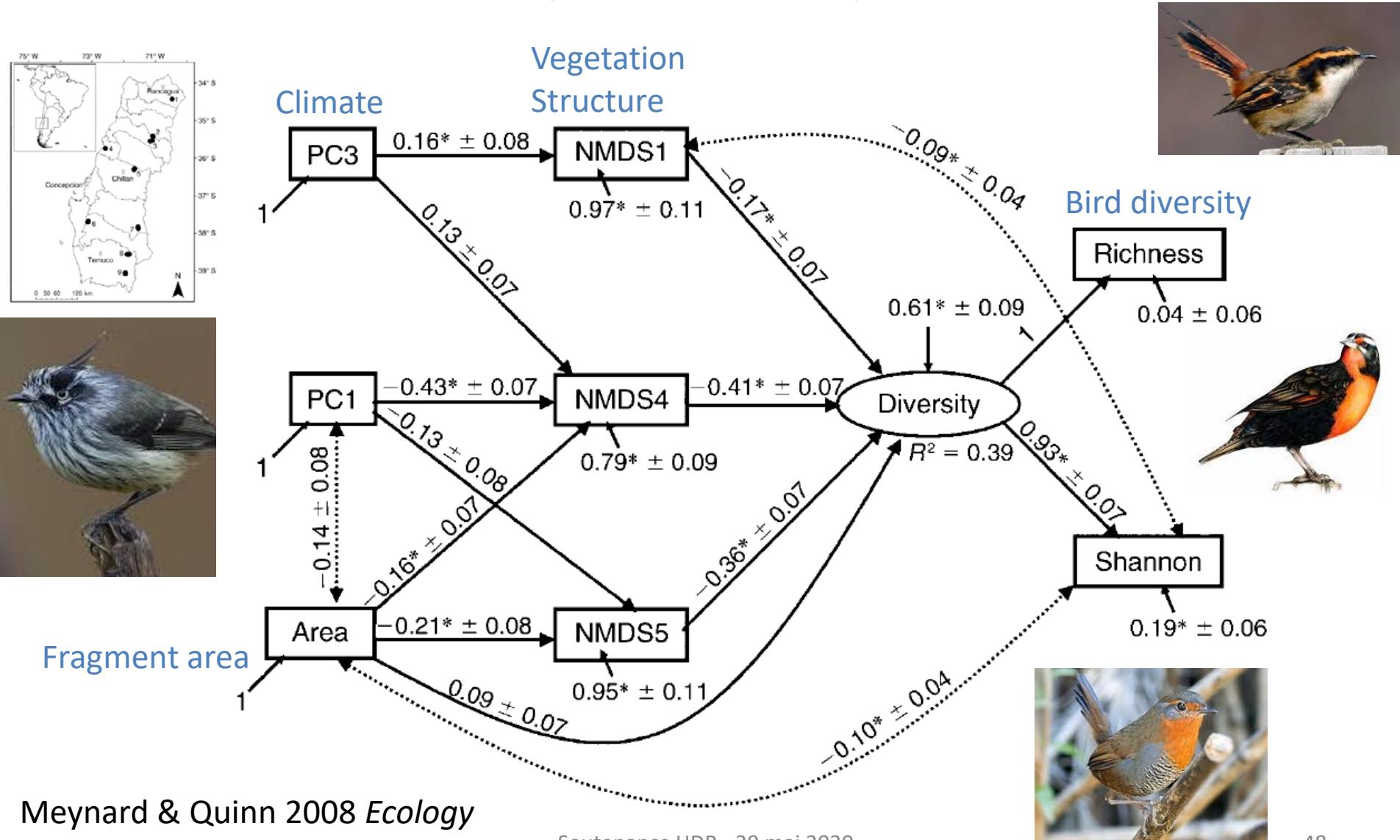


# Processes driving diversity at different scales



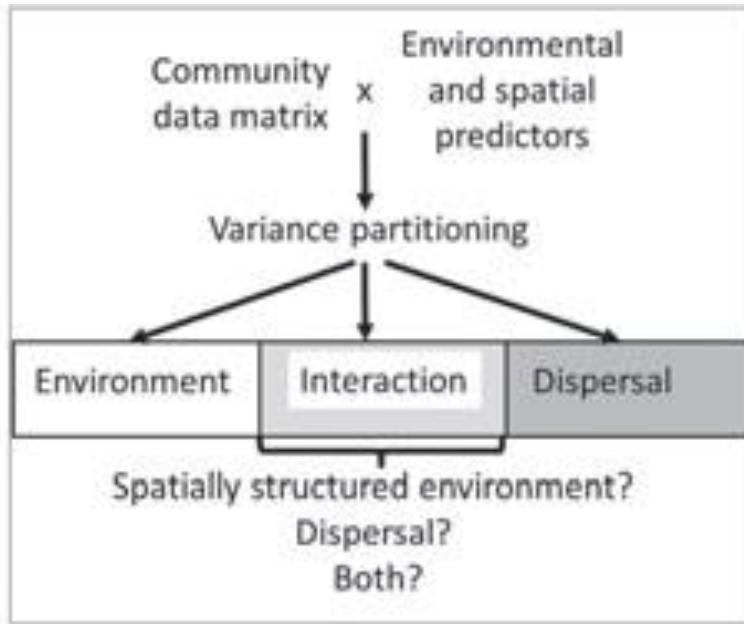
# Processes driving diversity at different scales

## Bird diversity in Chilean temperate forests

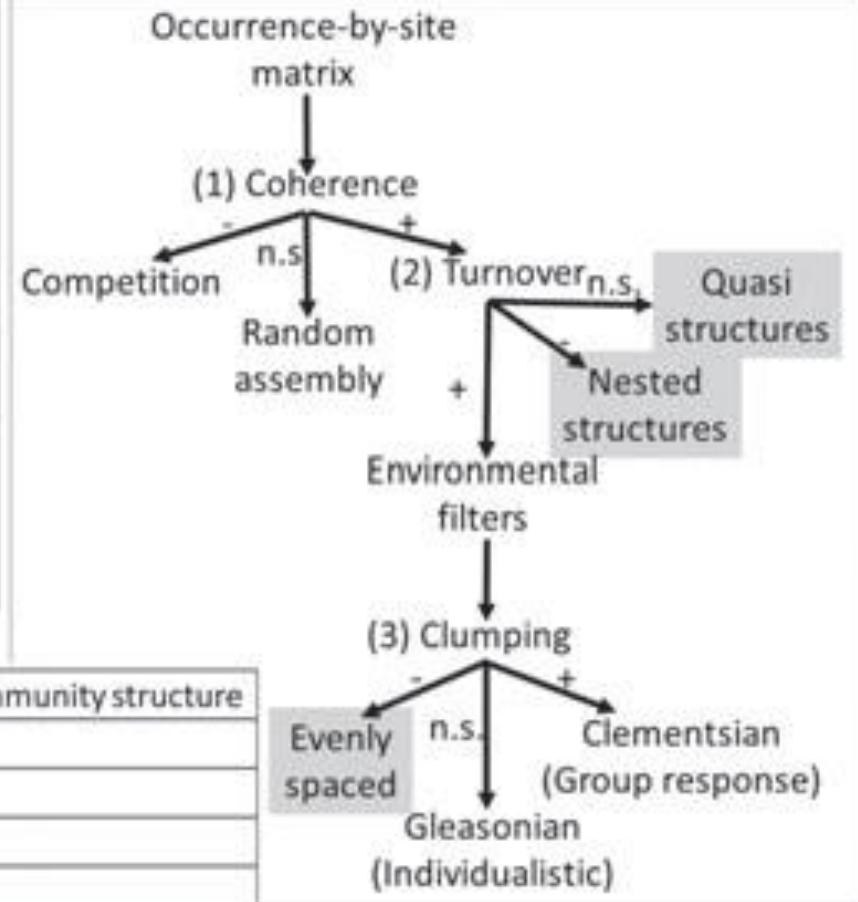


# Metacommunities

## Variance partitioning



## Metacommunity structure approach

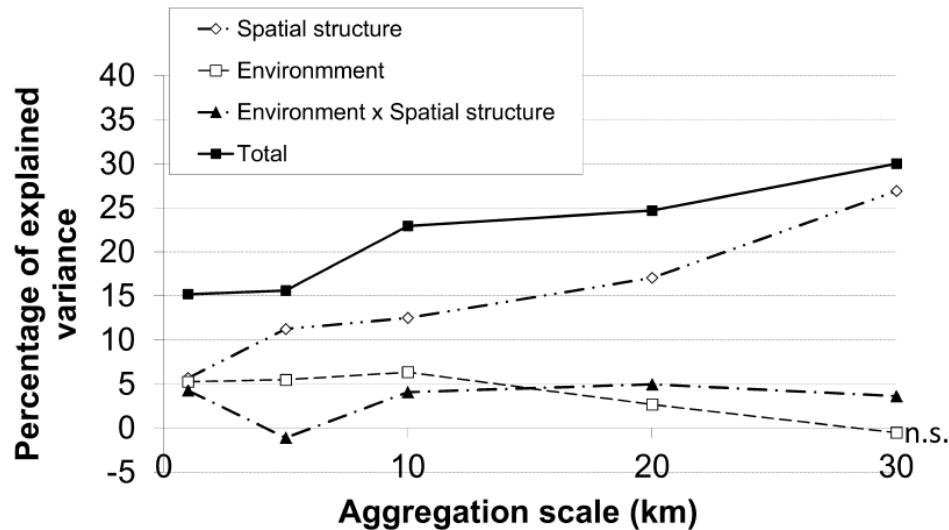


Ecological process	Variance partitioning	Metacommunity structure
Random assembly	No	Yes
Competition	No	Yes
Environmental filtering	Yes	Yes
Dispersal	Yes	No

Meynard et al 2013 *Journal of Biogeography*

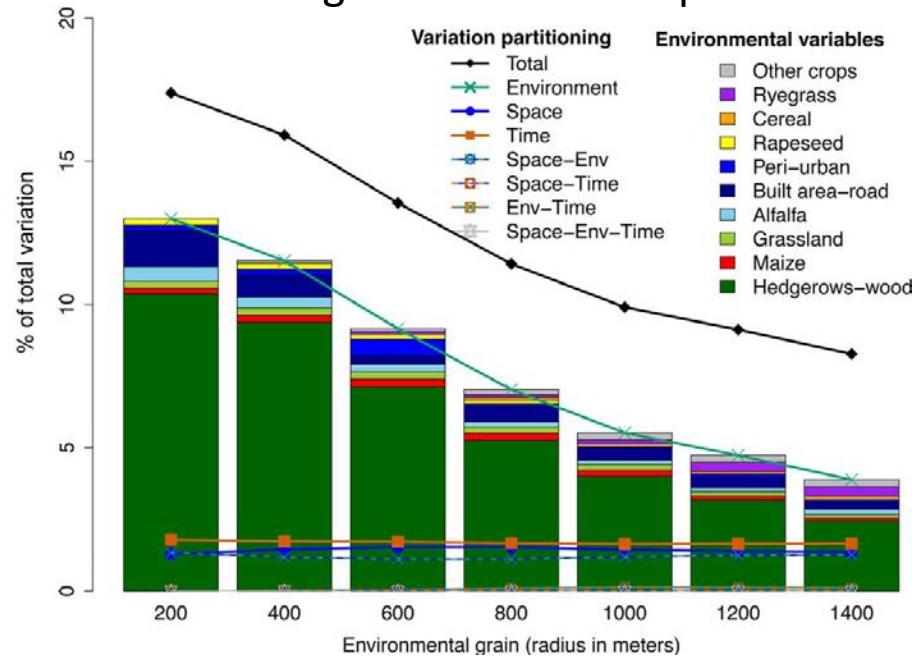
# Metacommunities

Alpine grasslands



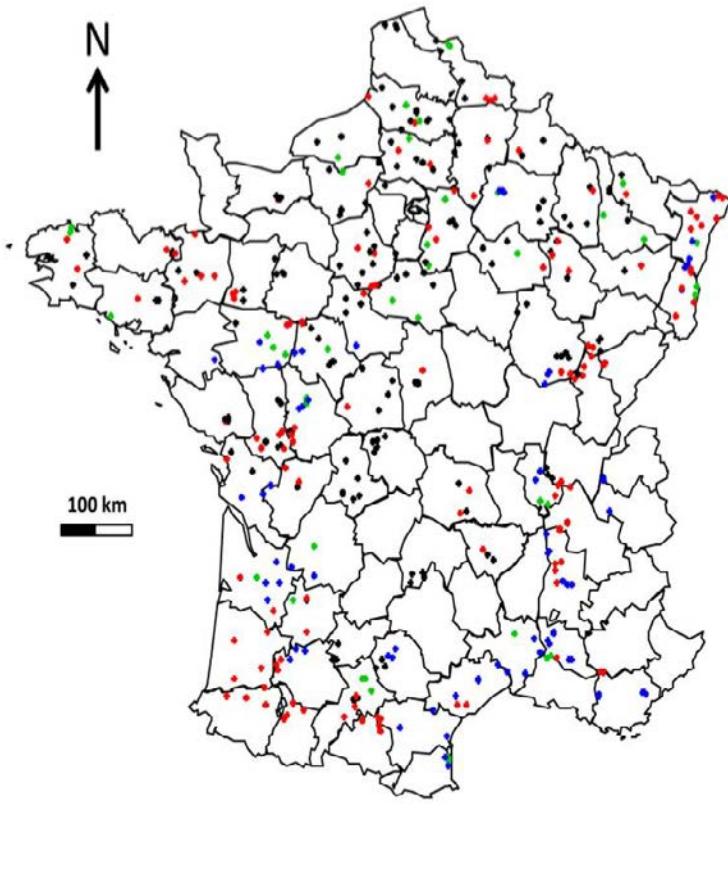
Meynard et al 2013 *Journal of Biogeography*

Birds in agricultural landscapes



Henckel et al 2019 *PlosOne*

# « Le réseau 500 ENI »



500 plots across France in field margins  
Plants, coleoptera, birds, worms  
3 main types of cultures

Thesis project (Isis Poinas):  
**Non-intentional consequences of agricultural practices on biodiversity in field margins: metacommunity perspective**

In collaboration with Guillaume Fried (ANSES)

# Diversity and origins of life have fascinated biologists for centuries



# Functional diversity:

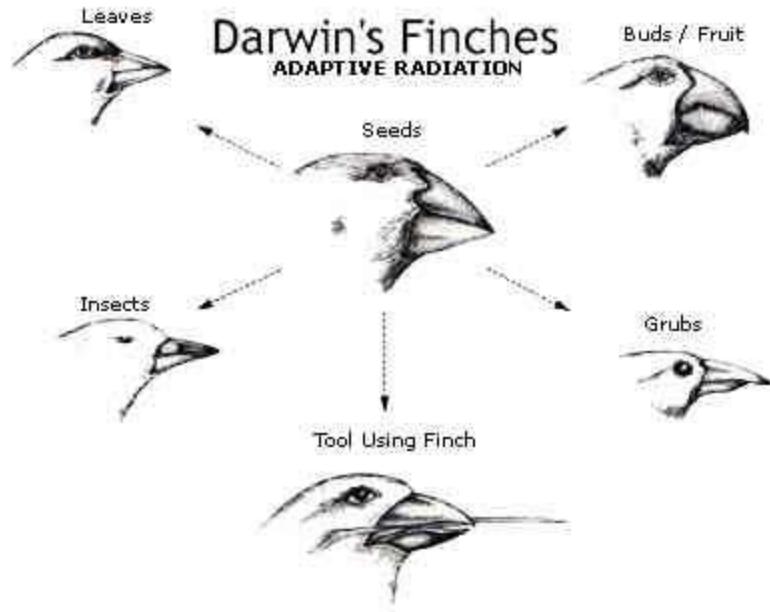
diversity of functions or life forms accumulated in an area or community



- Conserve diversity of life forms and shapes
- Ecosystem functioning and services

# Phylogenetic diversity:

## evolutionary history accumulated in an area or community



- Proxy for functional diversity
- Evolutionary history
- Evolutionary potential

# The virtual ecologist in diversity studies

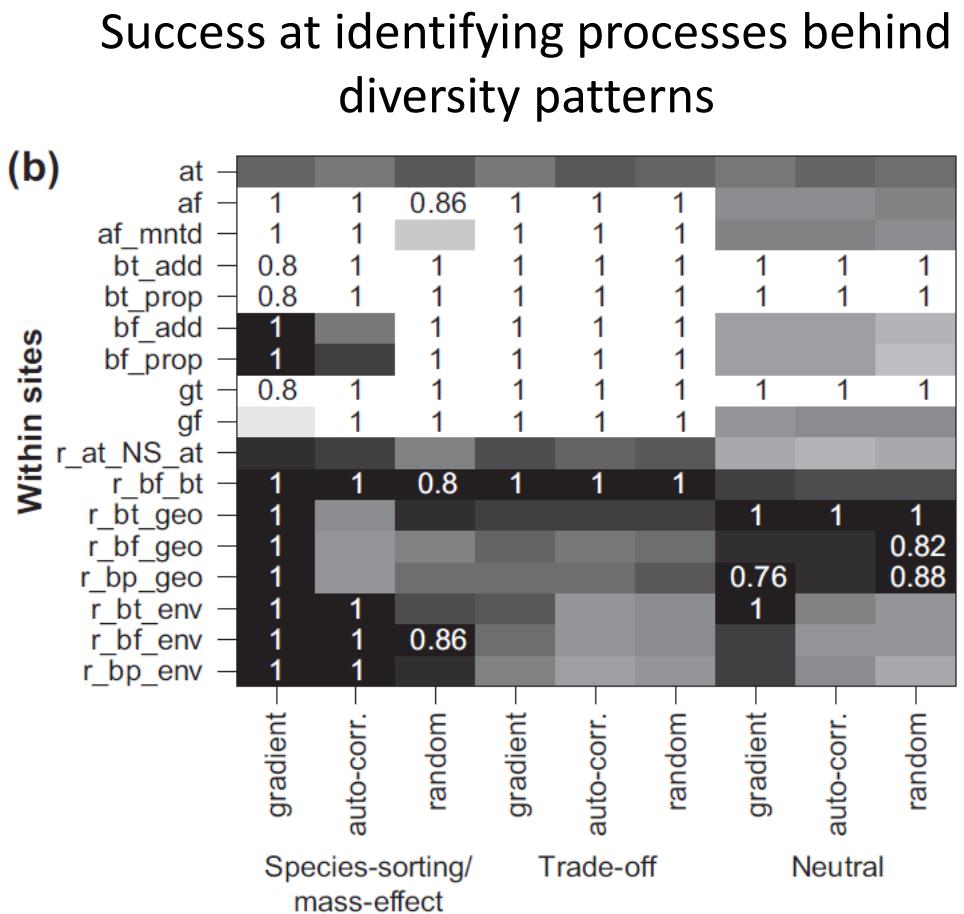
<b>THEORY</b>	<b>PROCESS</b>
Neutral	Random dispersal-competition
Environmental filtering	Environment filters out species
Colonization / extinction	Dispersal X competition
Mass effect	Dispersal X environment

- Can we distinguish different theories from the empirical patterns of diversity?

Munkemuller et al (2012) *Ecography* 35: 468-480

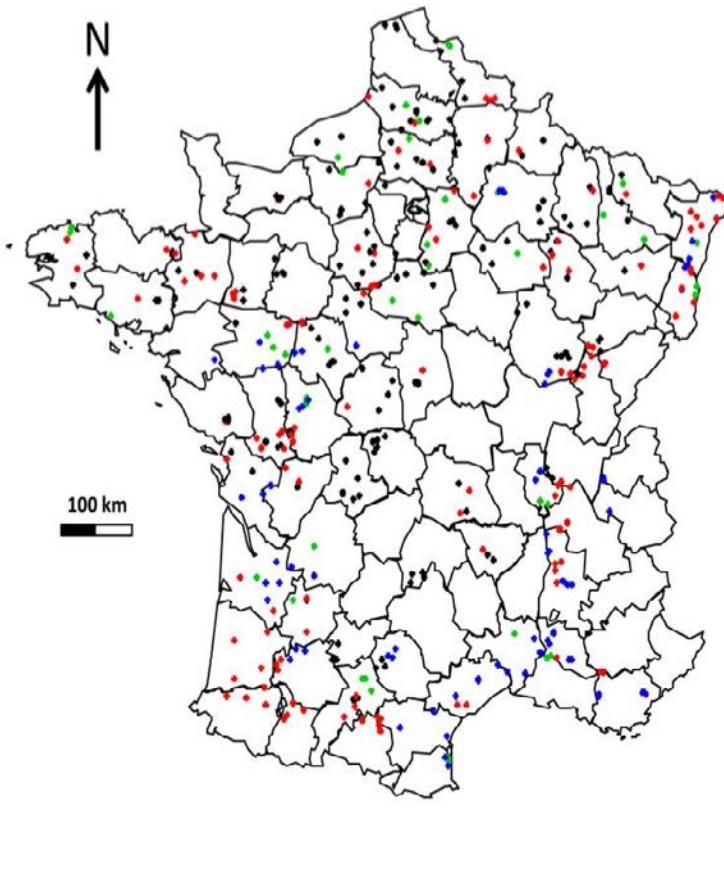
# The virtual ecologist in diversity studies

- **No single index on itself could identify the correct process, but some combinations of indices, especially those involving correlations between **beta diversity and environmental turnover**, and diversity indices including **functional traits**, did better**



Munkemuller et al (2012) *Ecography* 35: 468-480

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# Metacommunities in agricultural landscapes

- Unintended consequences of agricultural practices
- Regulatory action at national level, management practices at the local level
- Scale vs processes
- Multiple facets of diversity
- Synchrony between taxonomic groups (plants and coleoptera)

## **Virtual Species**



**David M Kaplan**  
IRD-MARBEC

**Boris Leroy**  
MNHN

## **Spider Mites**



**Maria Navajas**  
INRAE-CBGP

**Alain Migeon**  
INRAE-CBGP



**Reseau ENI**  
Guillaume Fried  
ANSES-CBGP

# **Collaborations**

## **Desert Locusts**



**MP Chapuis**  
Cirad-CBGP

**Michel Lecoq**  
Cirad



**Cyril Piou**  
Cirad-CBGP



**Nicolas Sauvion**  
INRAE- BGPI

## **Insect macroecology**



**Gael Kergoat**  
INRAE-CBGP

**Bruno Le Ru**  
IRD



**Valerie Poncet**  
IRD-DIADE

**Stephanie Manel**  
EPHE- CFE



**Agricultural Pests  
In Africa**  
Nathalie Gauthier  
ANSES-CBGP