

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

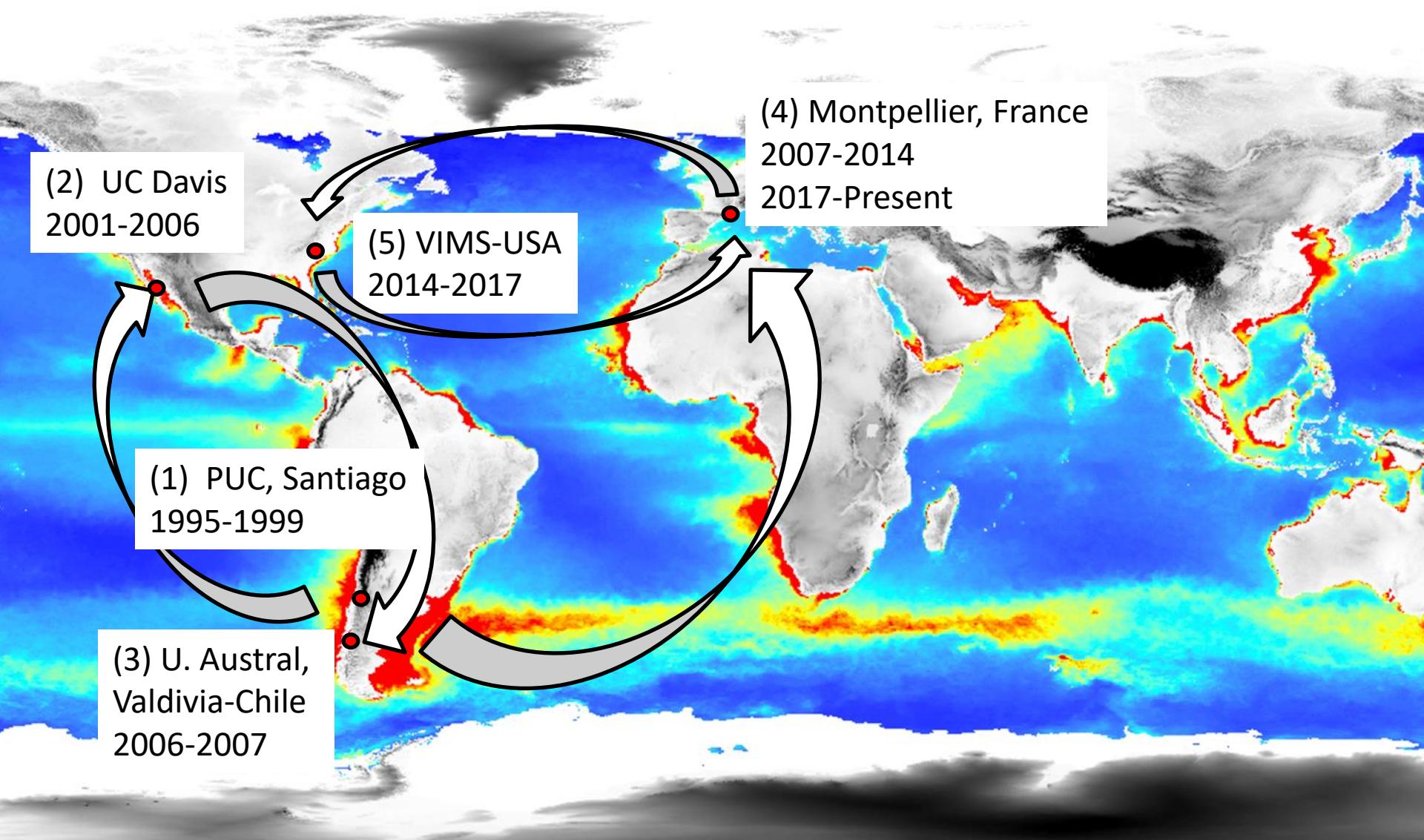
Christine N. Meynard

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

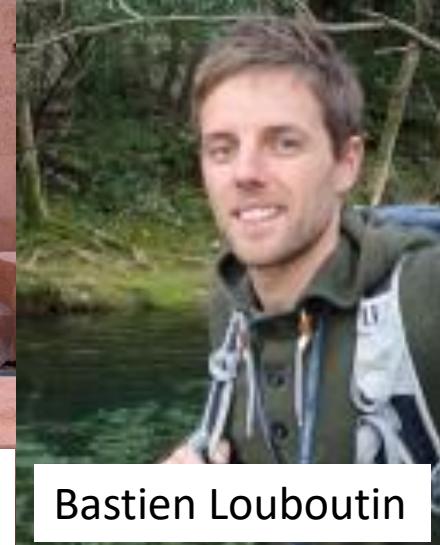
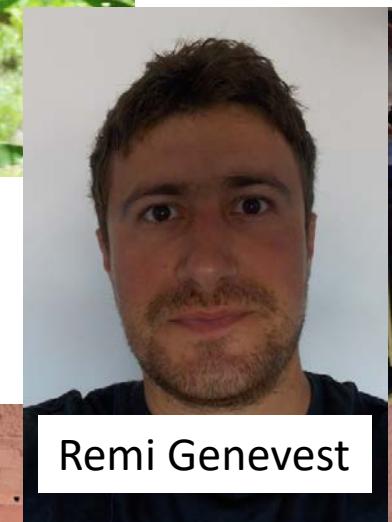
An Investigation of Local Adaptation of
Speckled Trout, *Cynoscion nebulosus*,
along the U.S. East Coast



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

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

Etudiants



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**”

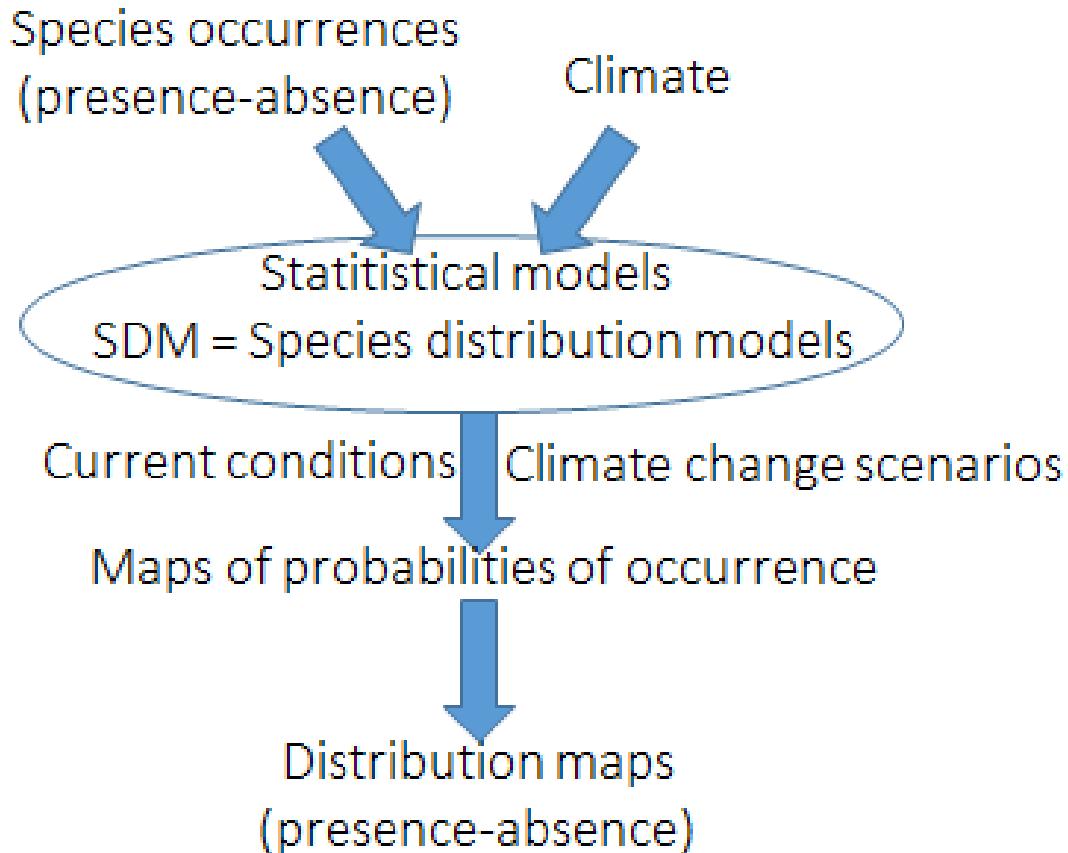
Jerome Chave: “what have we learned in 20 years?”

Chave (2013) Ecology Letters

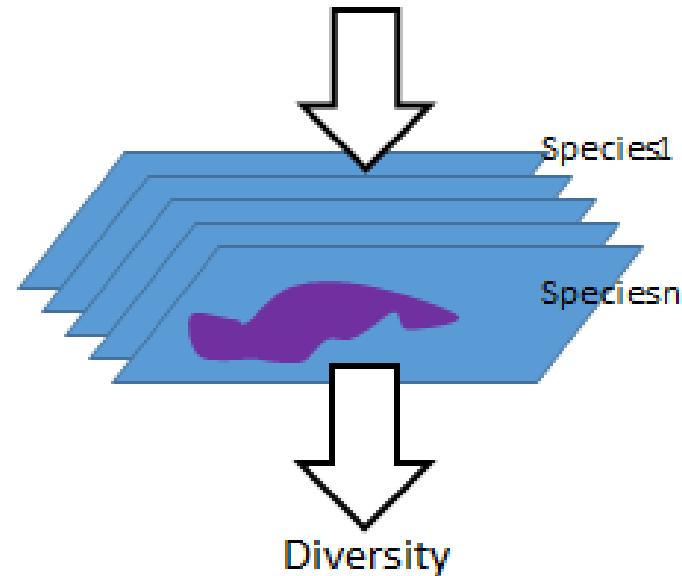
There has been **4 major technological revolutions**:

- 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 CO₂, 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

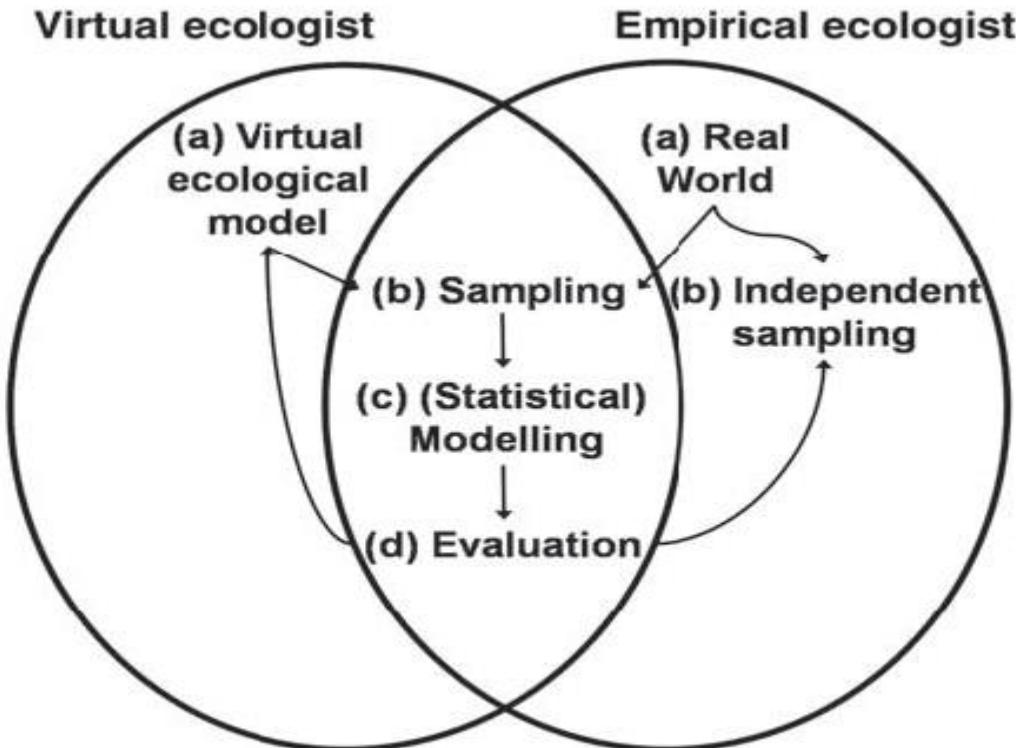


Figure 1. The elements of the virtual ecologist approach.

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

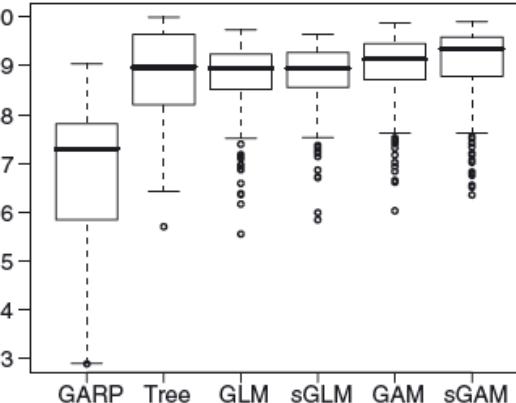
Classification rate

Types of models

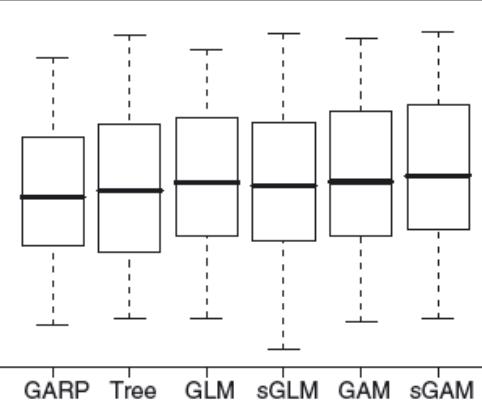
Types of species

Probabilities

Correlation

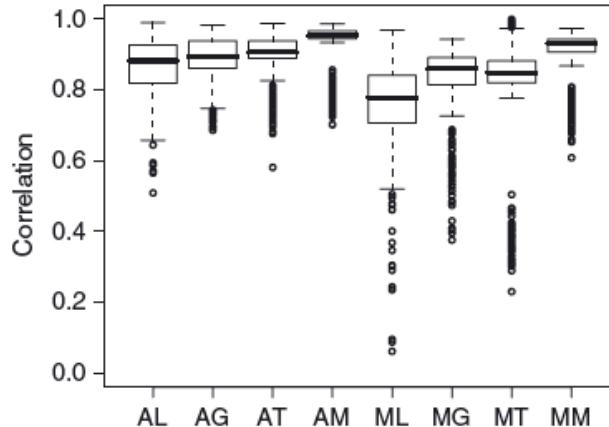


AUC



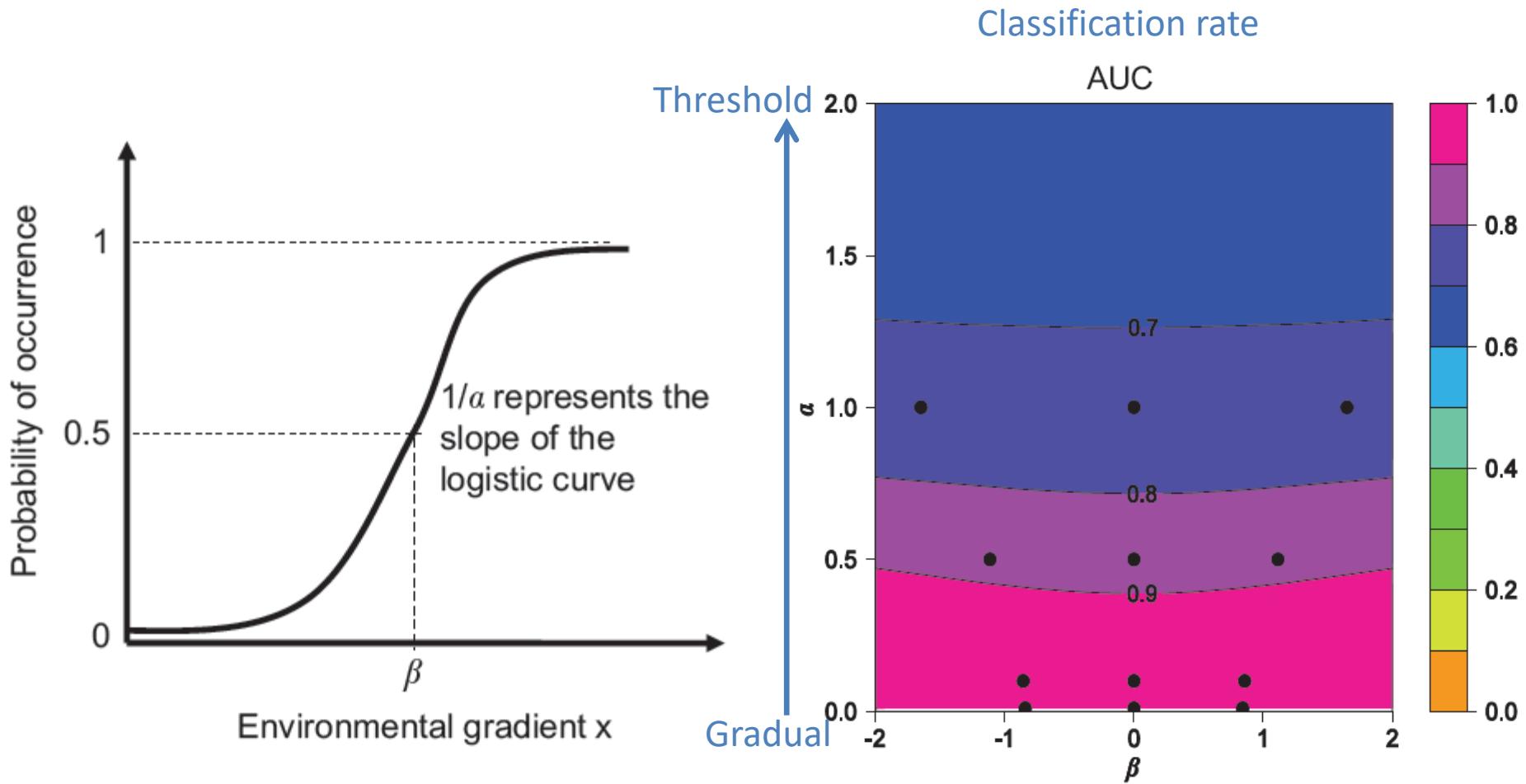
AUC

Correlation



Meynard & Quinn 2007 *Journal of Biogeography*

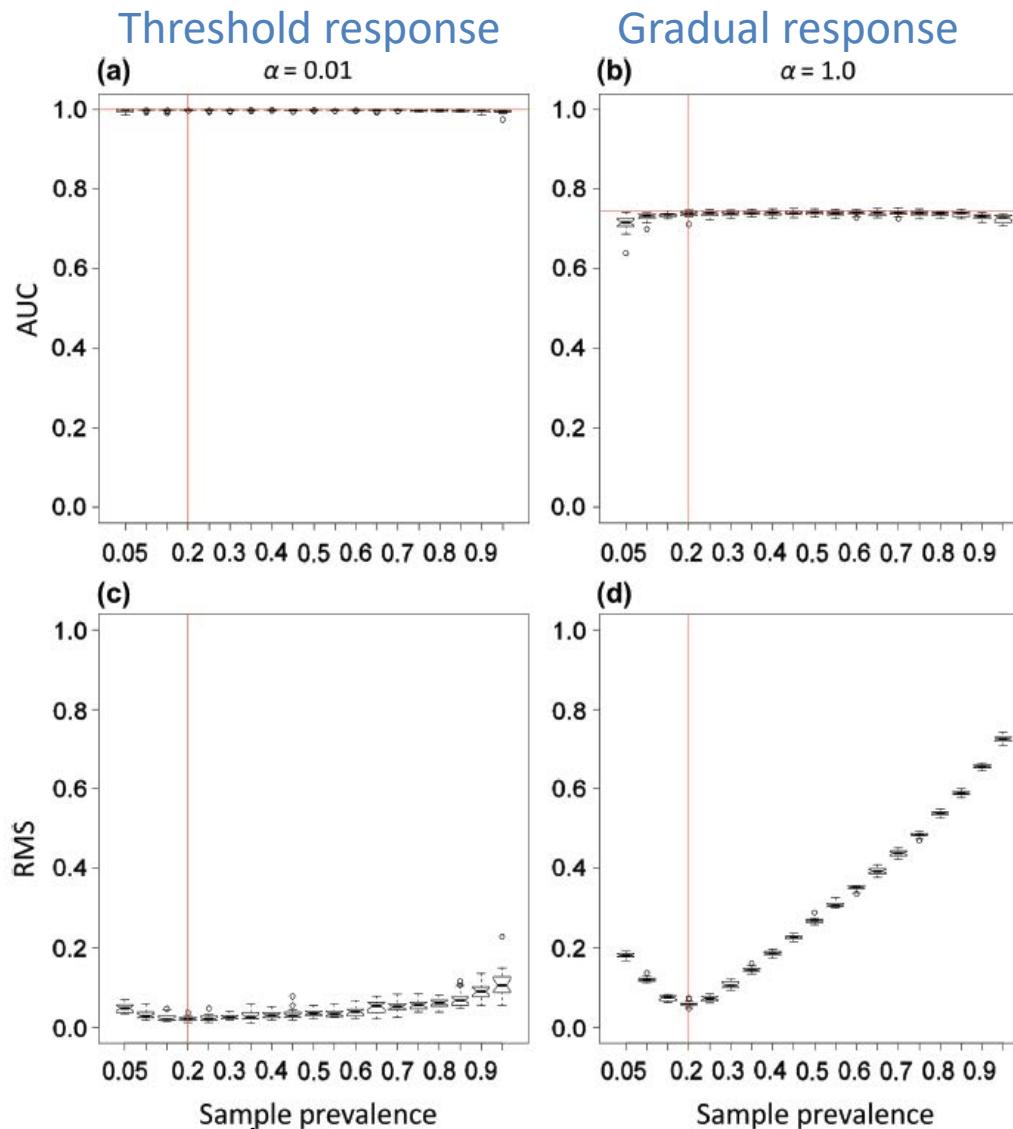
Virtual Species



Meynard & Kaplan 2012 *Ecography*

Virtual Species

Classification rate

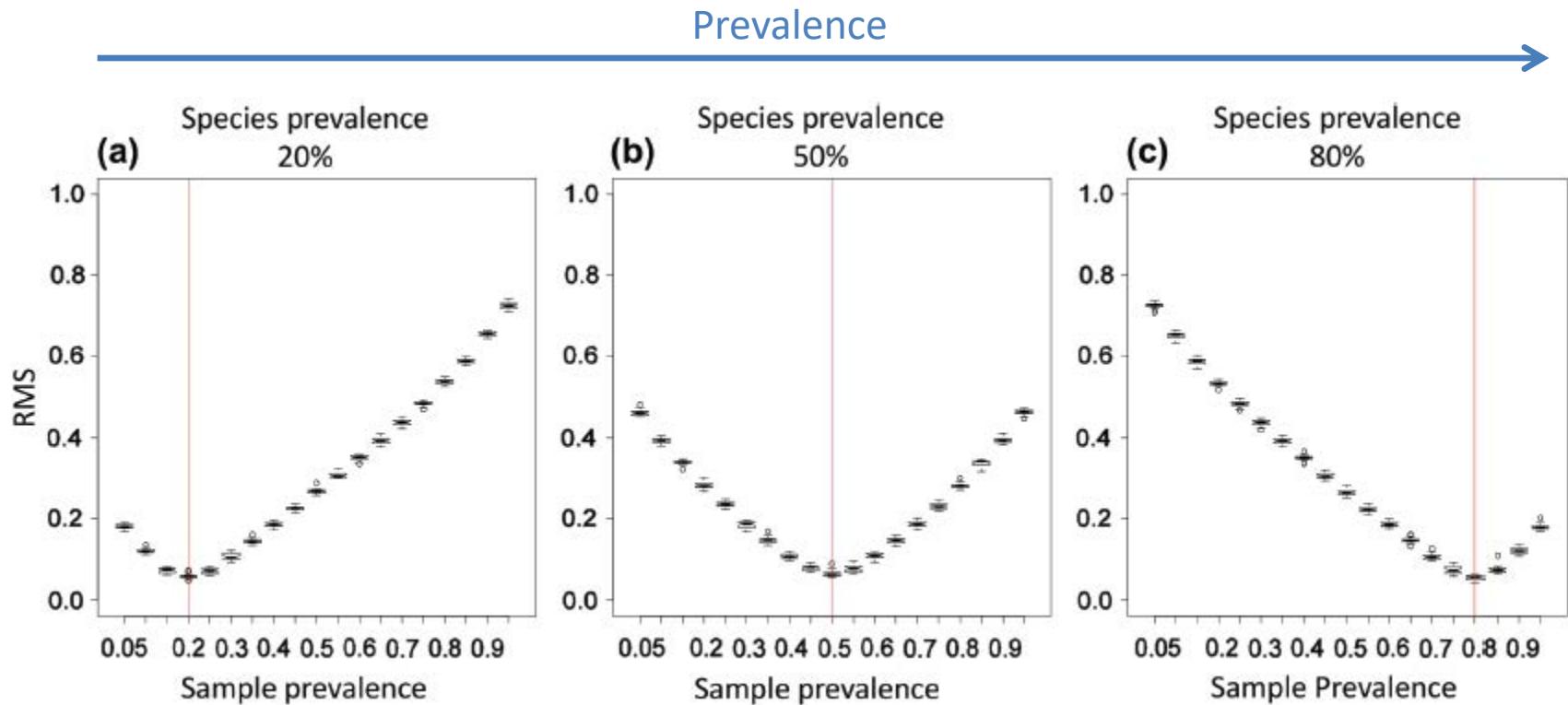


Probabilities

Meynard & Kaplan 2012 *Ecography*

Virtual Species

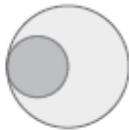
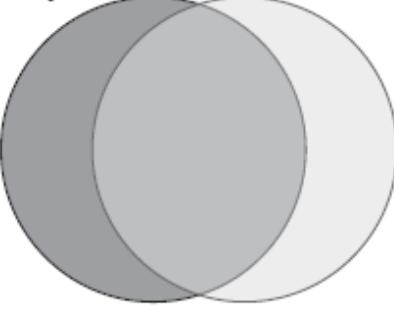
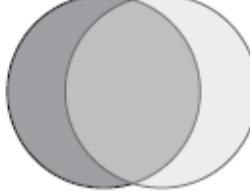
Prediction of Probabilities



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)  75% overprediction, 0% under-prediction, prevalence = 0.01	(b)  15% over-prediction, 0% under-prediction, prevalence = 0.01	(c)  0% over-prediction, 15% under-prediction, prevalence = 0.01			
TSS = 0.60 Sensitivity = 0.70 Specificity = 0.90  (d) 30% over- & under-prediction, prevalence = 0.25	Sørensen = 0.70 UPR = 0.30 OPR = 0.30	TSS = 0.67 Sensitivity = 0.70 Specificity = 0.97  (e) 30% over- & under-prediction, prevalence = 0.1	Sørensen = 0.70 UPR = 0.30 OPR = 0.30	TSS = 0.70 Sensitivity = 0.70 Specificity = 1.00  (f) 30% over- & under-prediction, prevalence = 0.01	Sørensen = 0.70 UPR = 0.30 OPR = 0.30

Leroy et al 2018 *Journal of Biogeography*

Virtual Species

Standards for virtual species studies

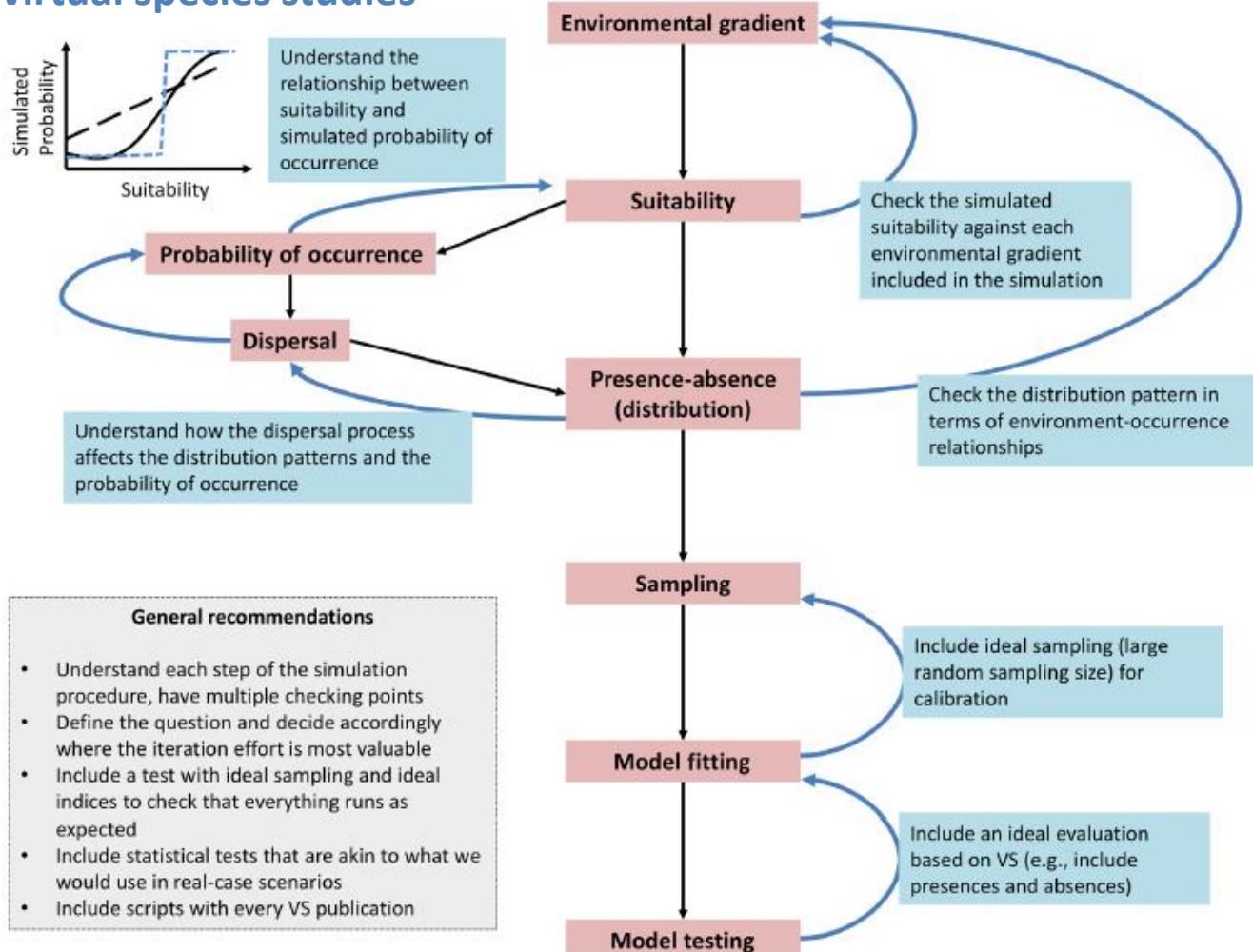
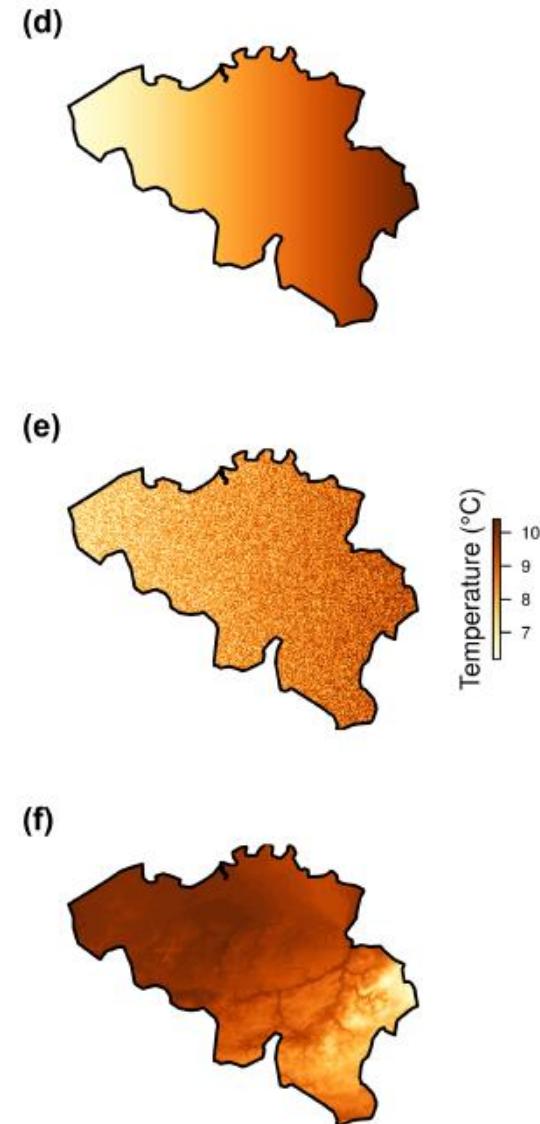
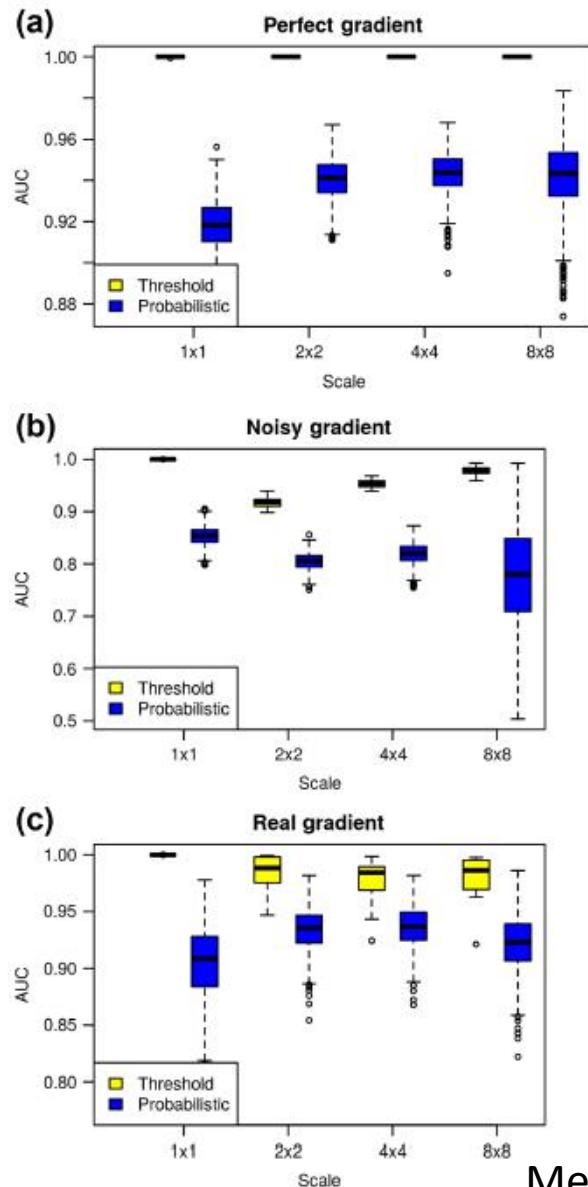
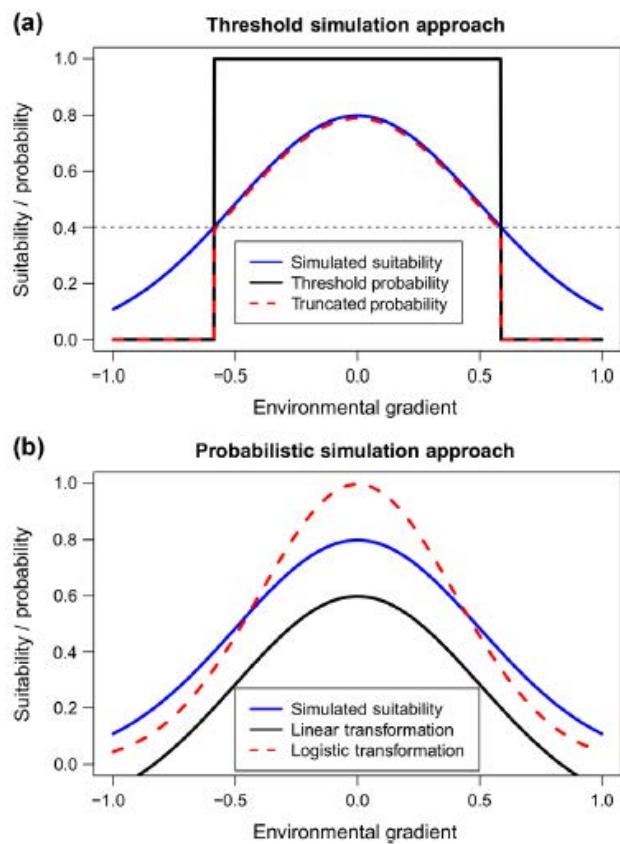
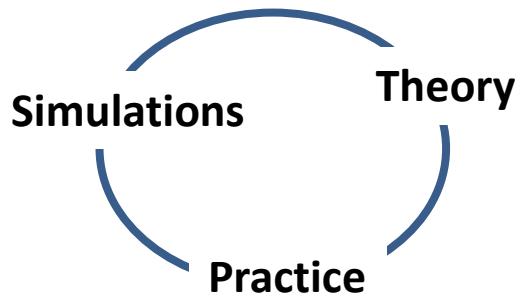


Figure 3. Graphical representation of recommendations and guidelines proposed in section 'General recommendations and guidelines' for future virtual species studies.

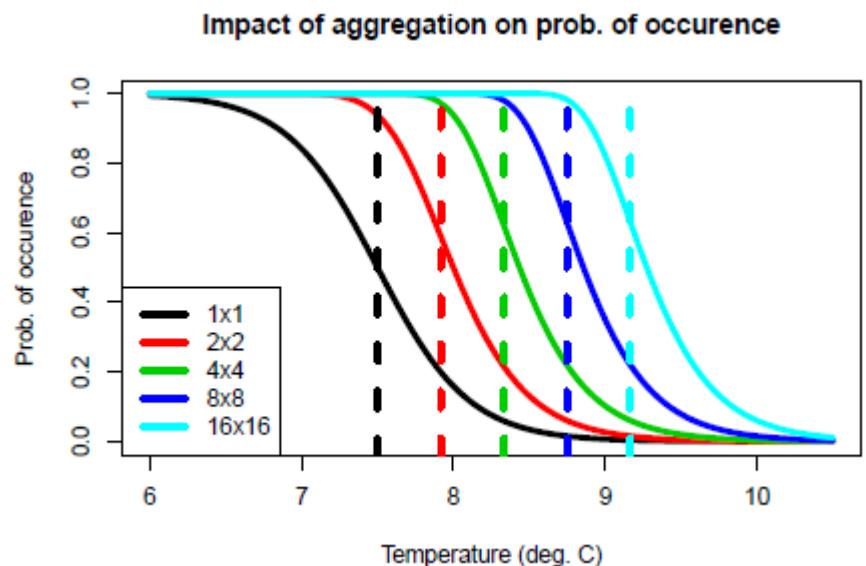
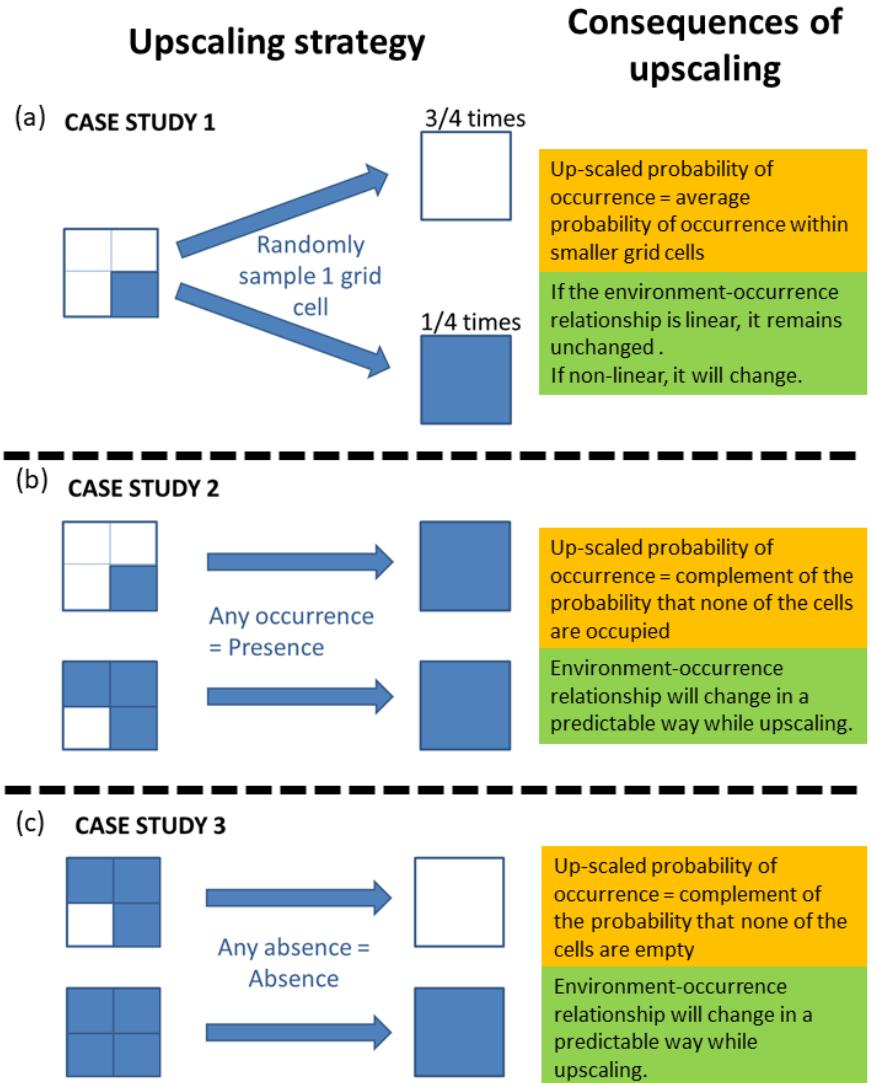
Meynard et al 2019 *Ecography*

Virtual Species



Meynard et al 2019 *Ecography*

A theoretical framework of upscaling for species distribution models



Meynard & Kaplan (*In Prep*)

Virtual Species

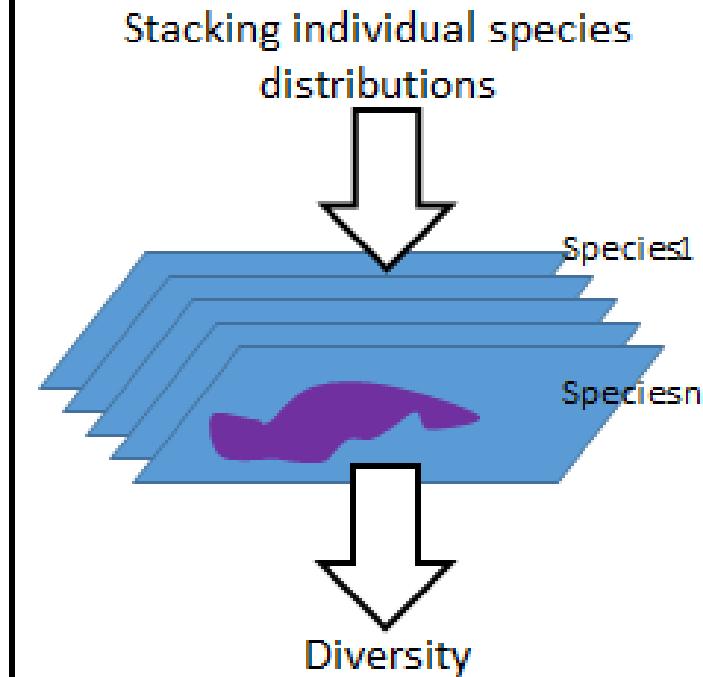
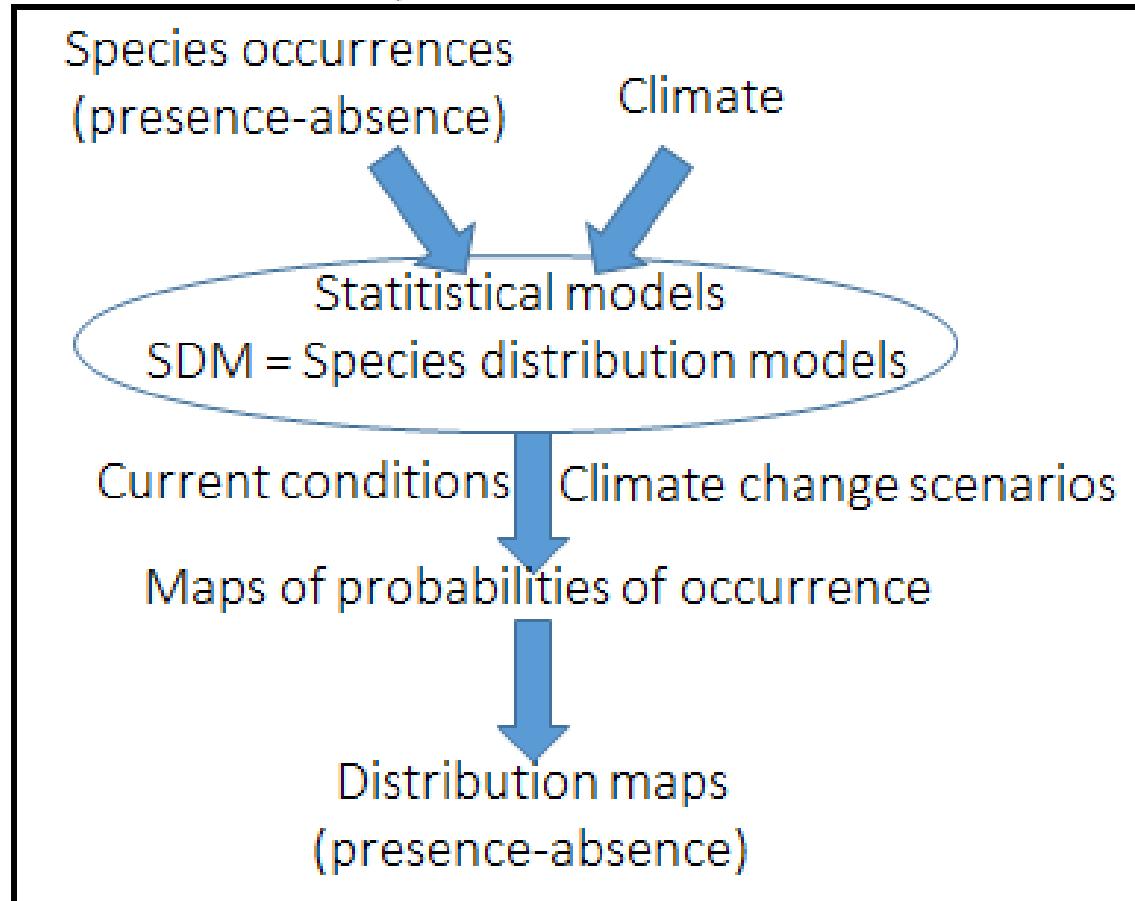
- Predictive ability in terms of probabilities versus presence-absence classifications are not equivalent
- Upscaling data usually changes BOTH the probability of occurrence AND the shape of the environment-occurrence relationship in a predictable way

I - Species distribution models

Uncertainties

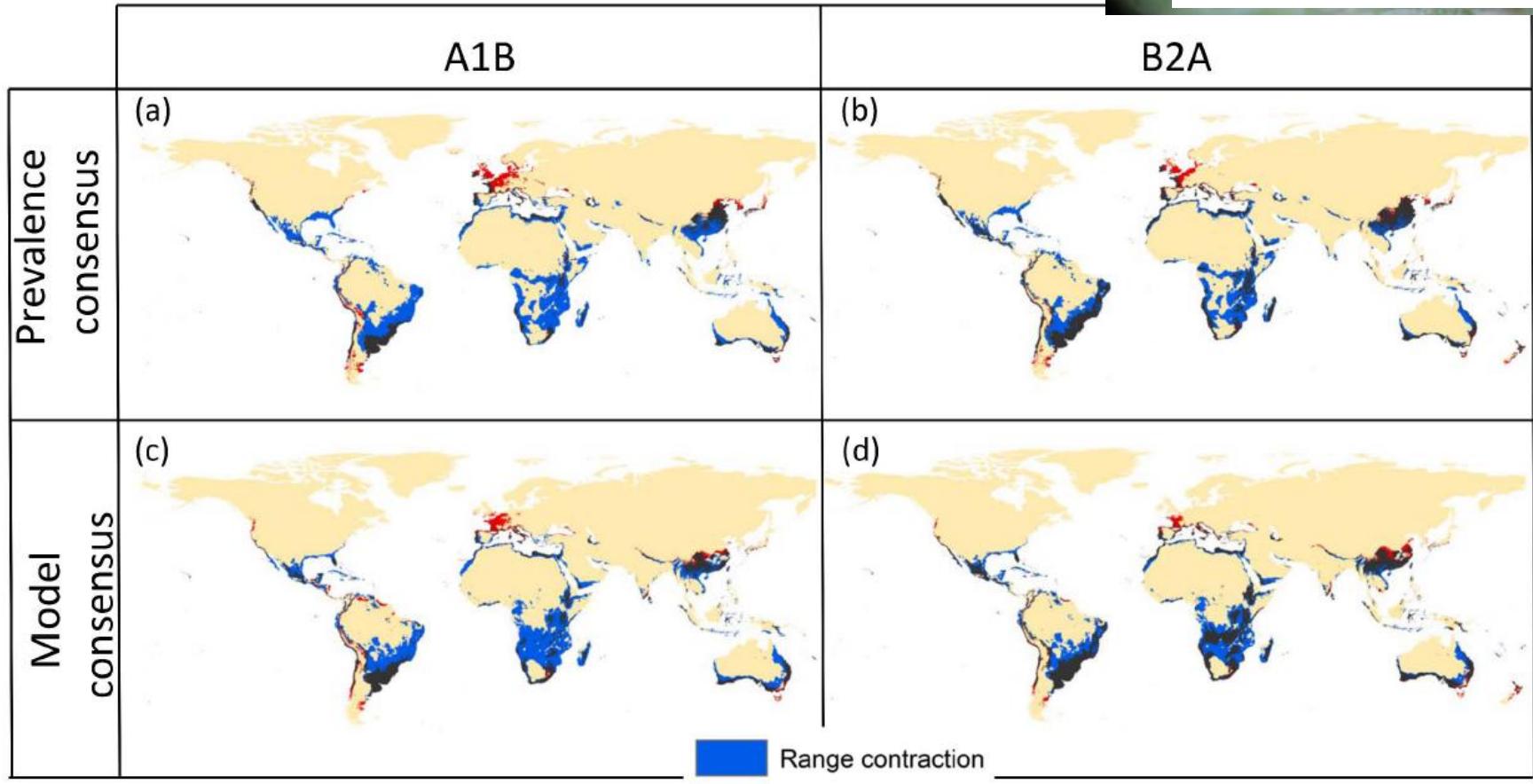
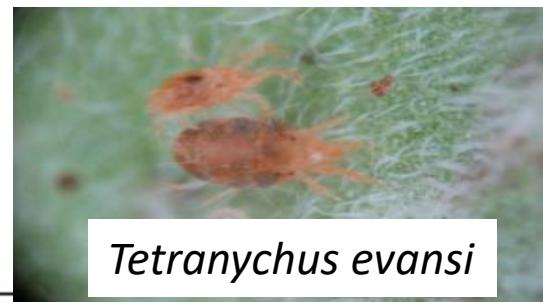
Uncertainties

Species distribution models (SDMs)



Uncertainties

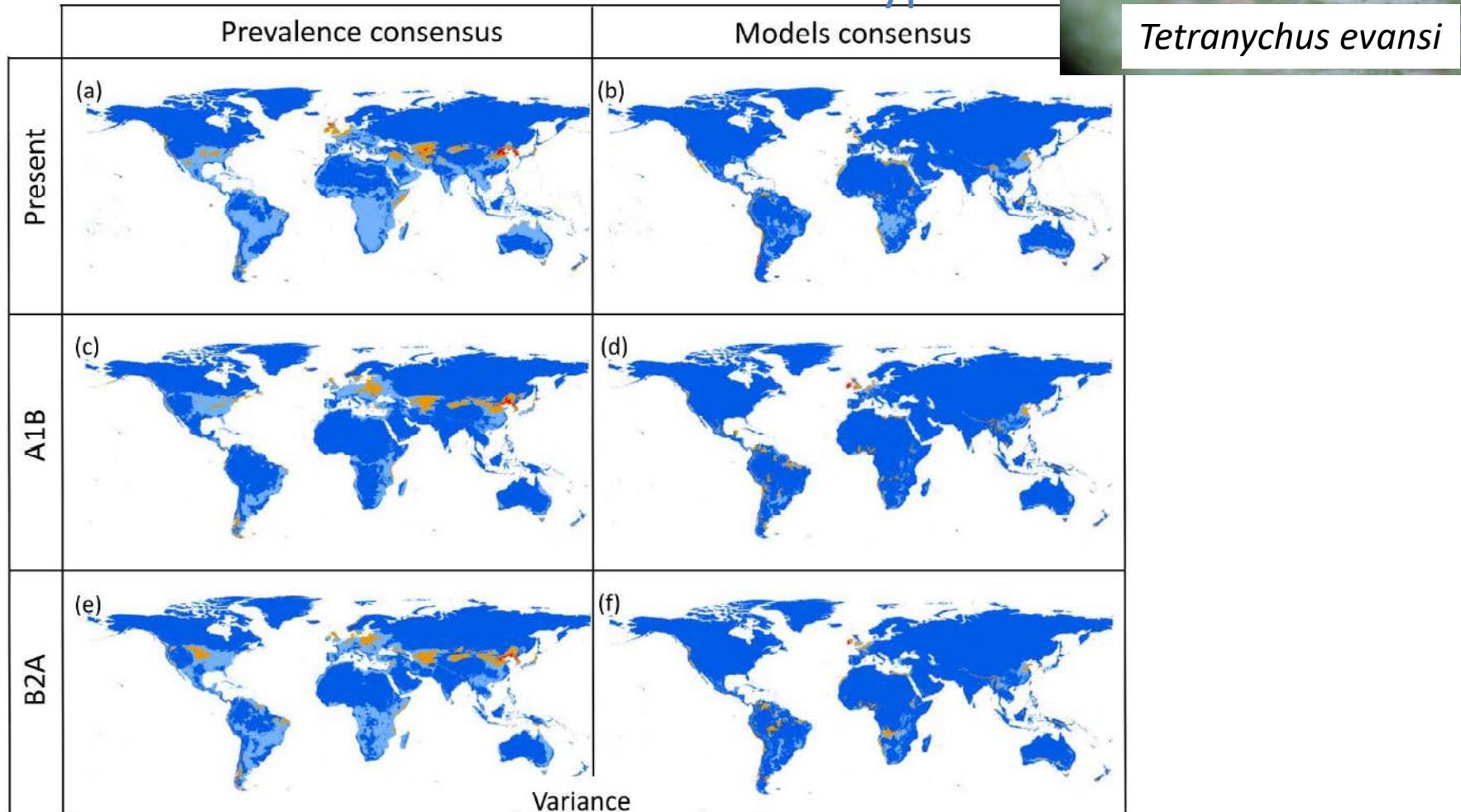
Prevalence and model types



Meynard et al 2013 *PlosOne*

Uncertainties

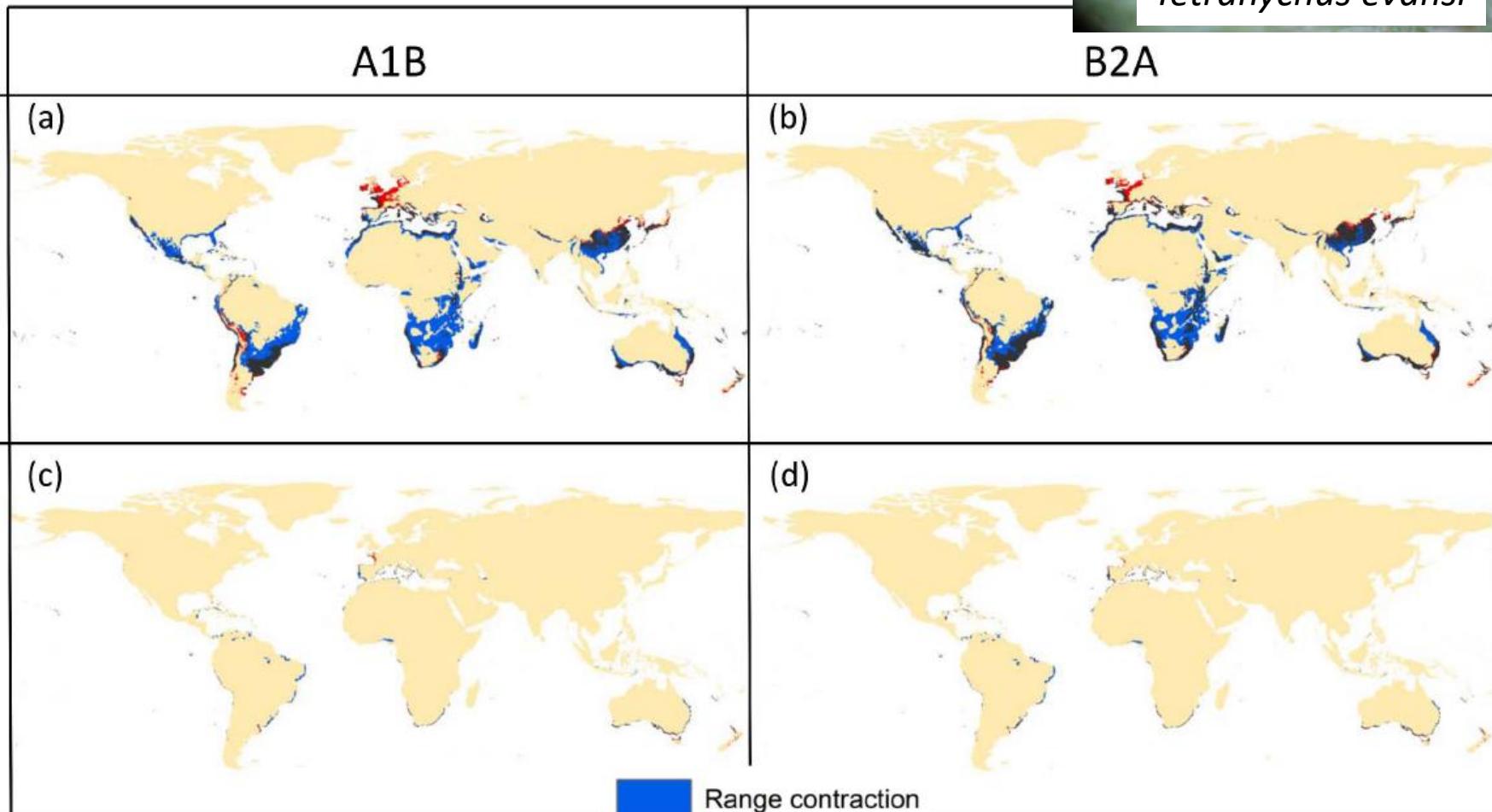
Prevalence and model types



Meynard et al 2013 *PlosOne*

Uncertainties

Species vs sub-species



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

Meynard et al 2013 *PlosOne*

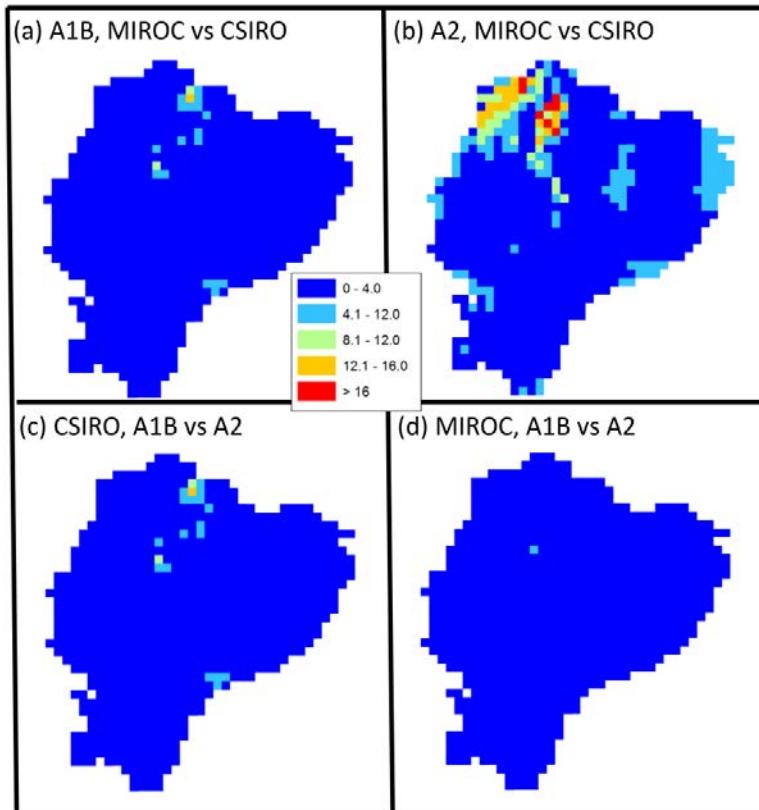
Uncertainties



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

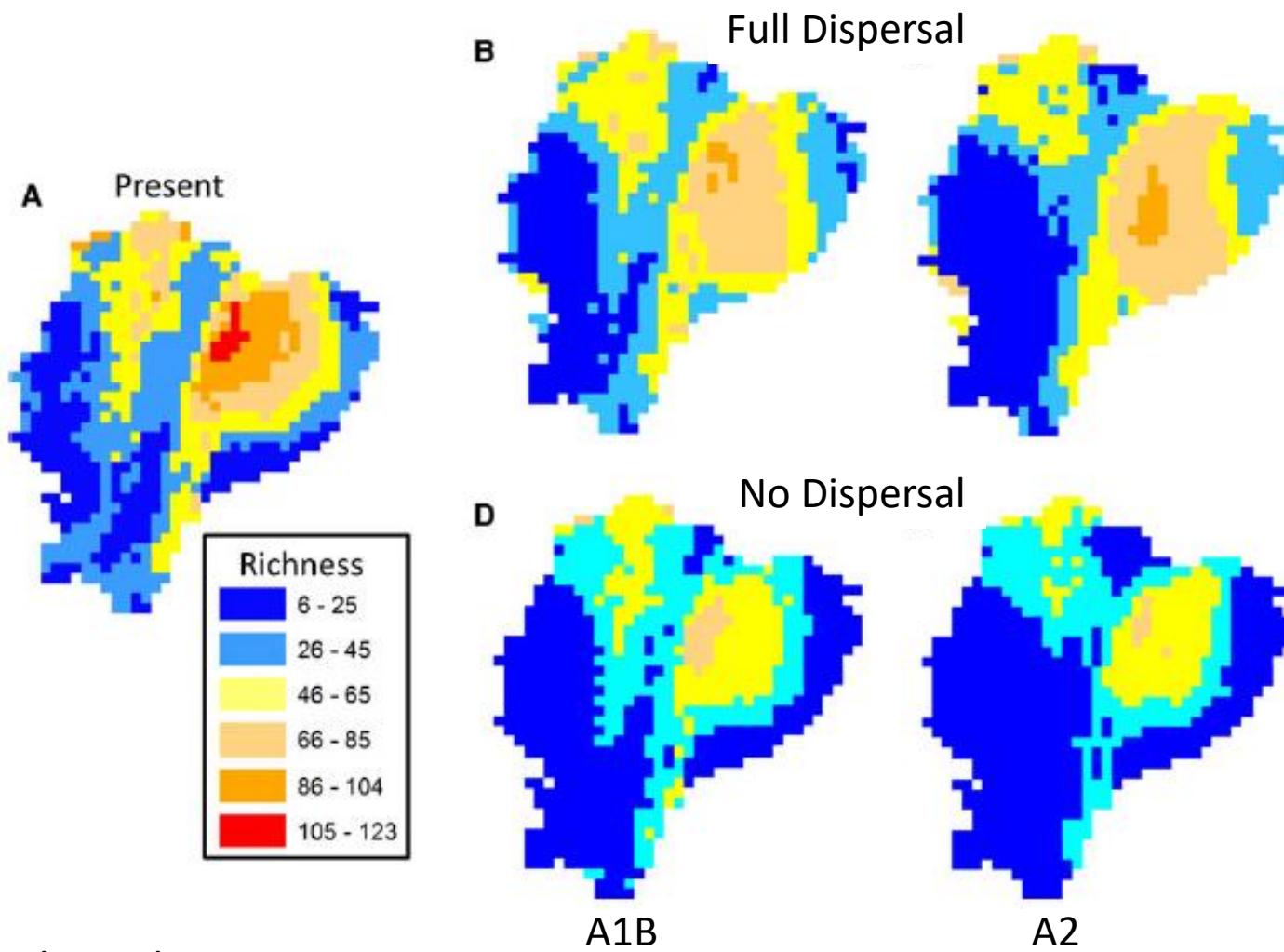
Between scenarios

Coefficient of variation between richness estimates in 2080.

Iturralde-Polit et al 2017 *Biotropica*

Uncertainties

Dispersal scenario



Iturralde-Polit et al 2017 *Biotropica*

Uncertainties

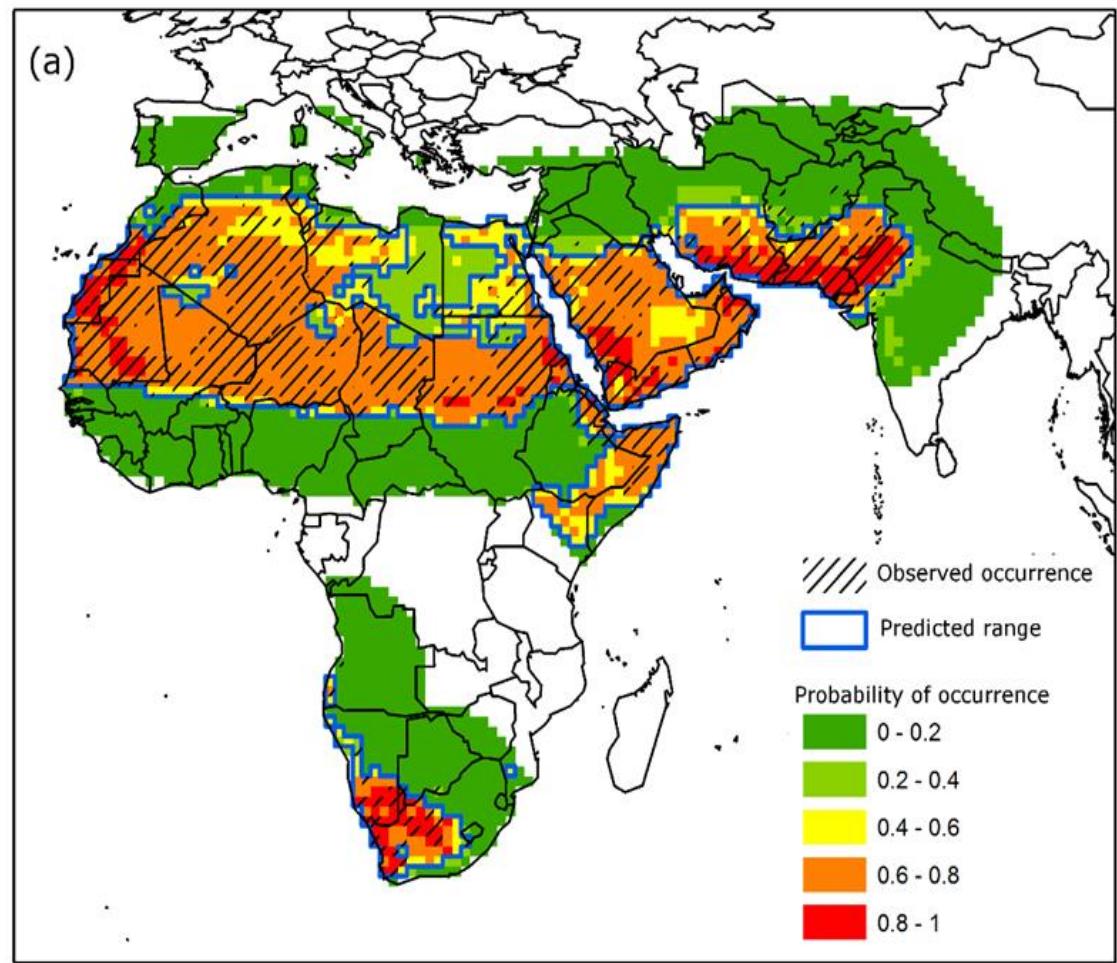
Types of models

Desert Locust



Ensemble forecasting
with 8 different models

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



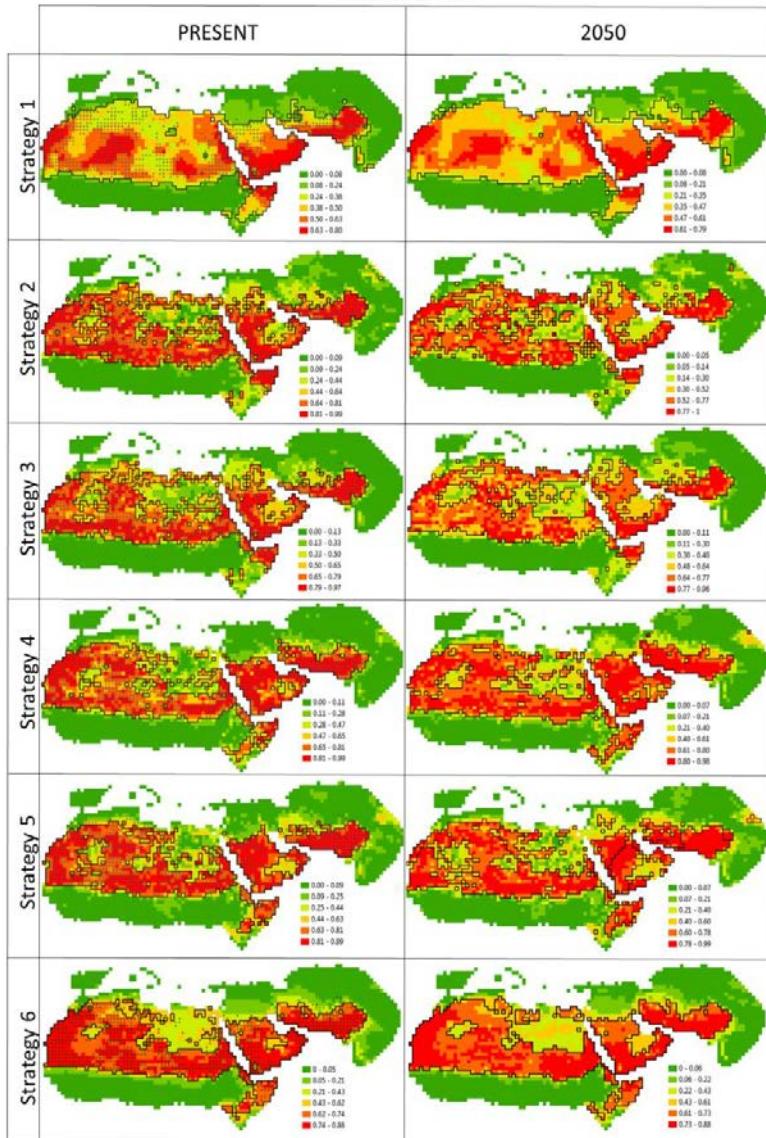
Meynard et al 2017
Global Change Biology

Uncertainties

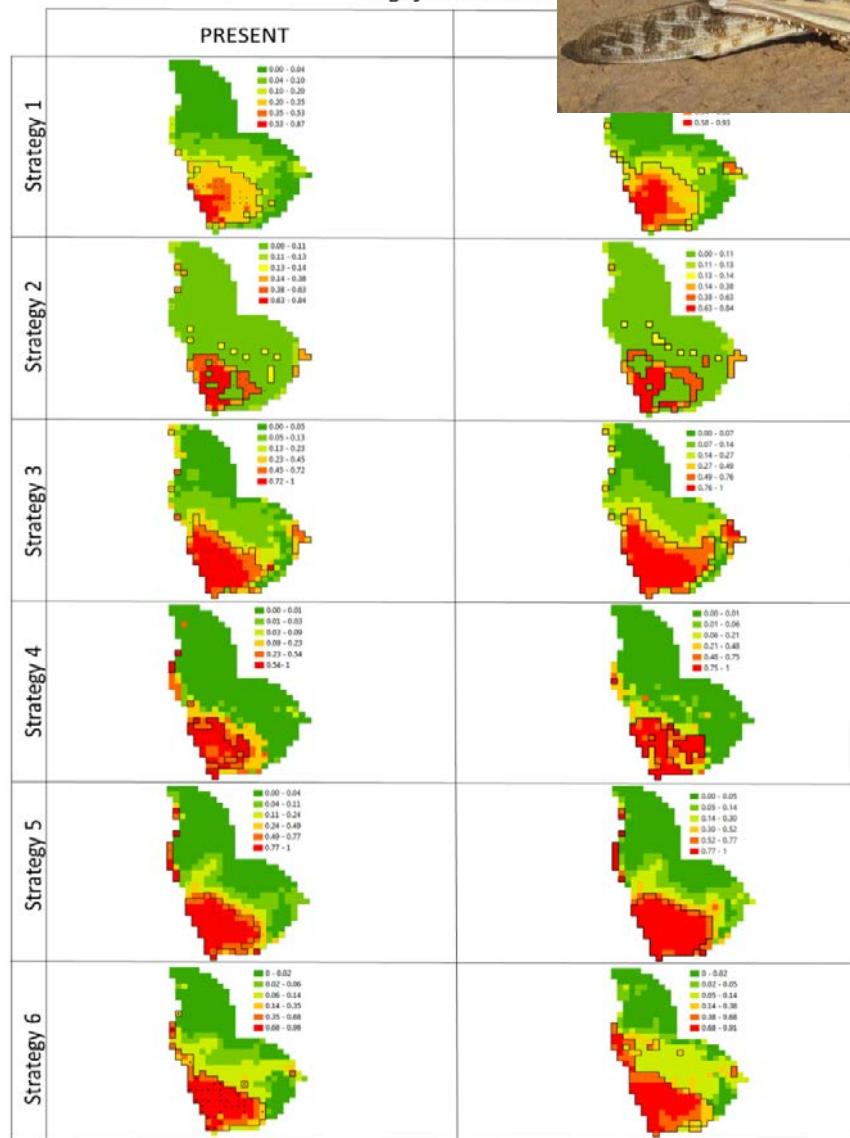
Desert Locust

Variable selection

S. g. gregaria



S. g. flaviventris



Uncertainties

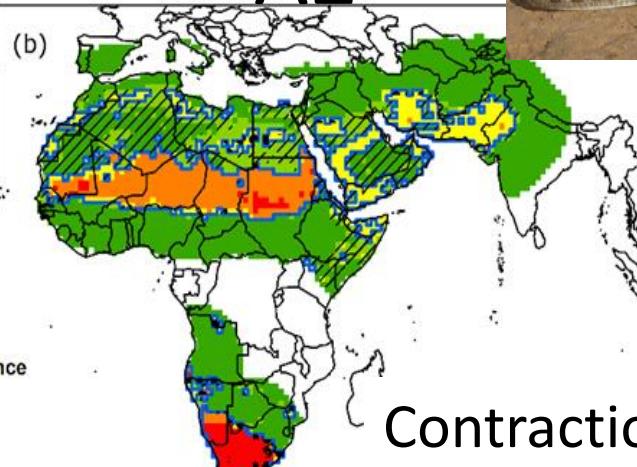
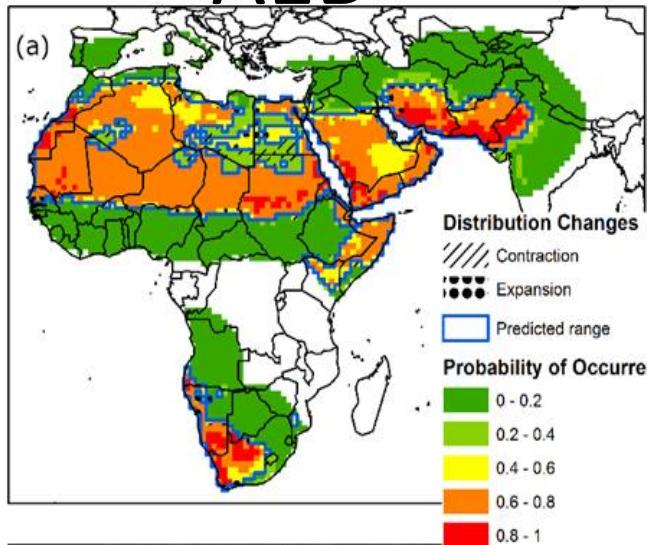
Climatic scenarios

Desert Locust

A1B

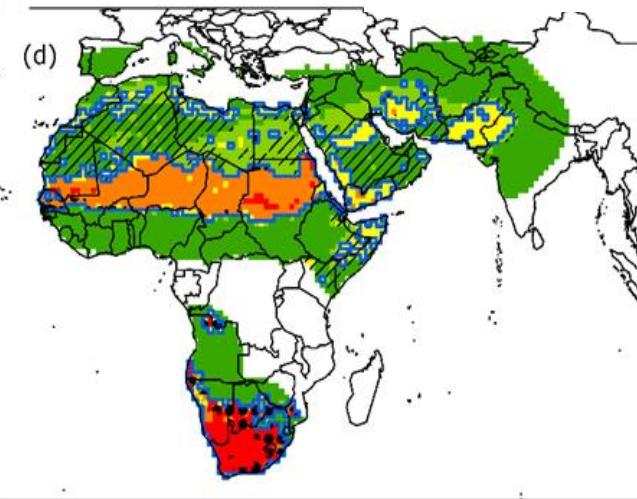
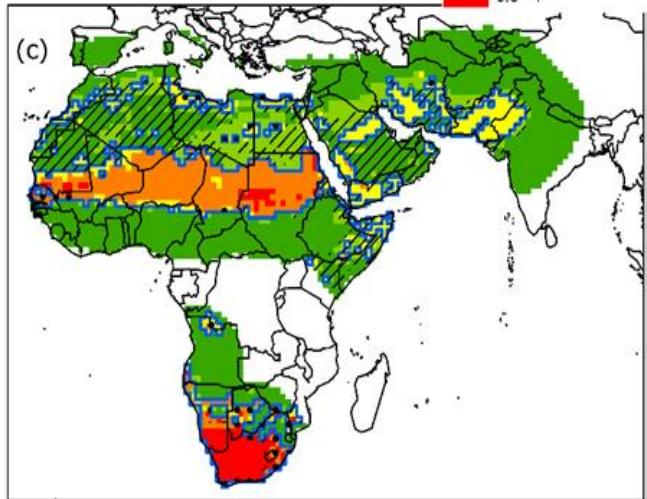
A2

2050



Contraction in the north
Expansion in the south

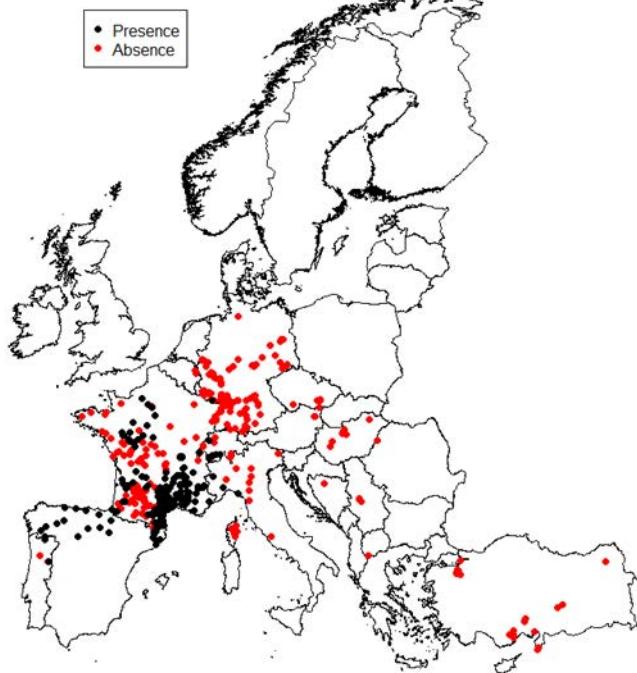
2090



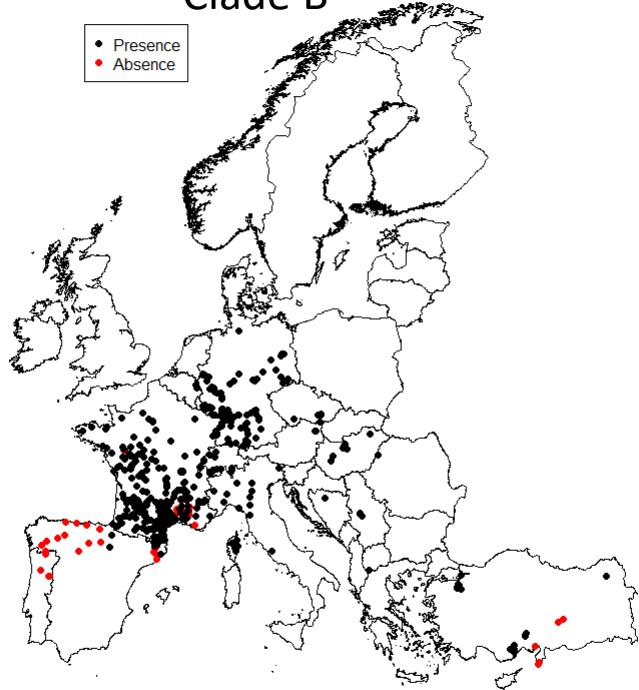


Occurrence data

Clade A



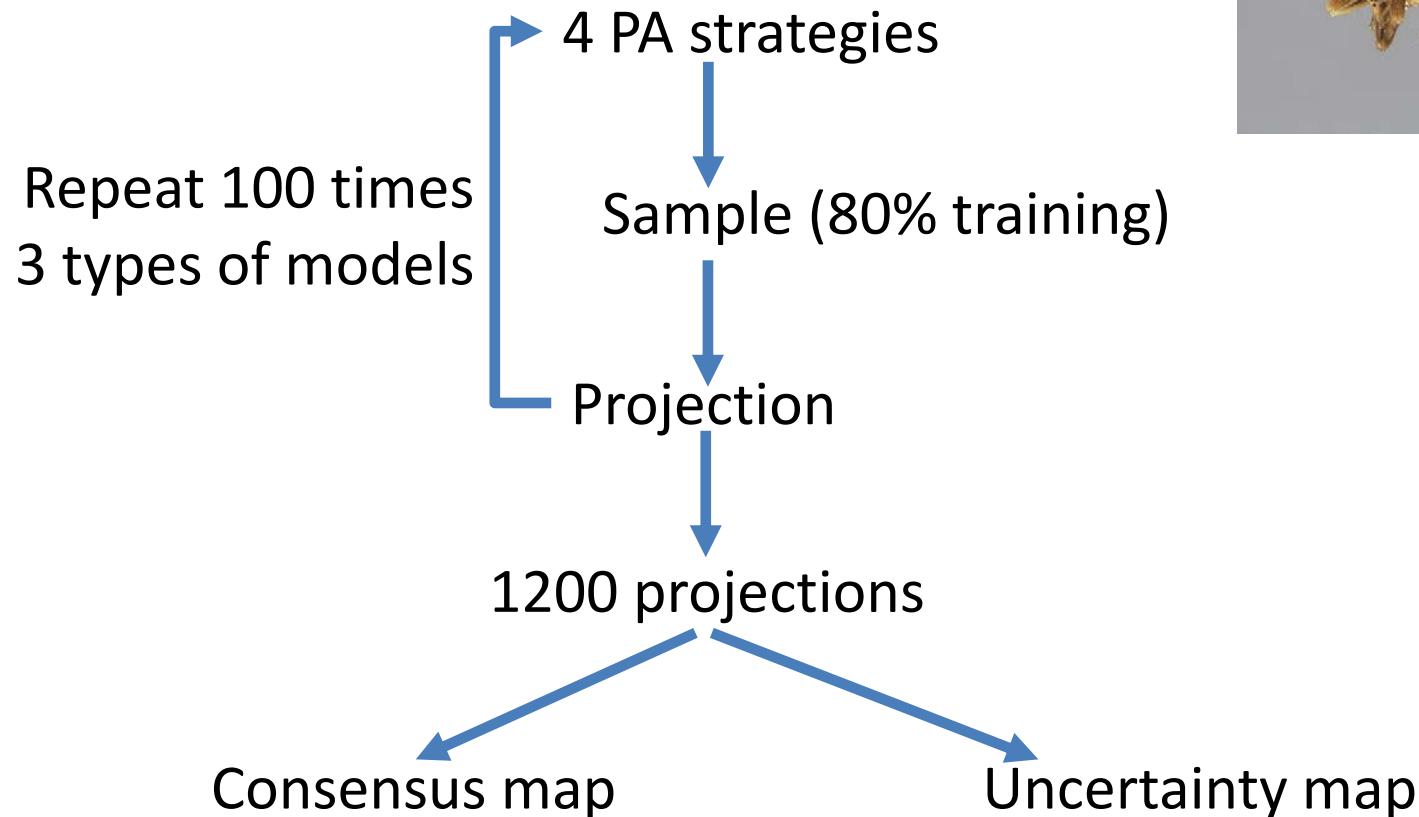
Clade B



Presence-only data modelling

Adding Pseudo-Absences (PA)

Iterative approach

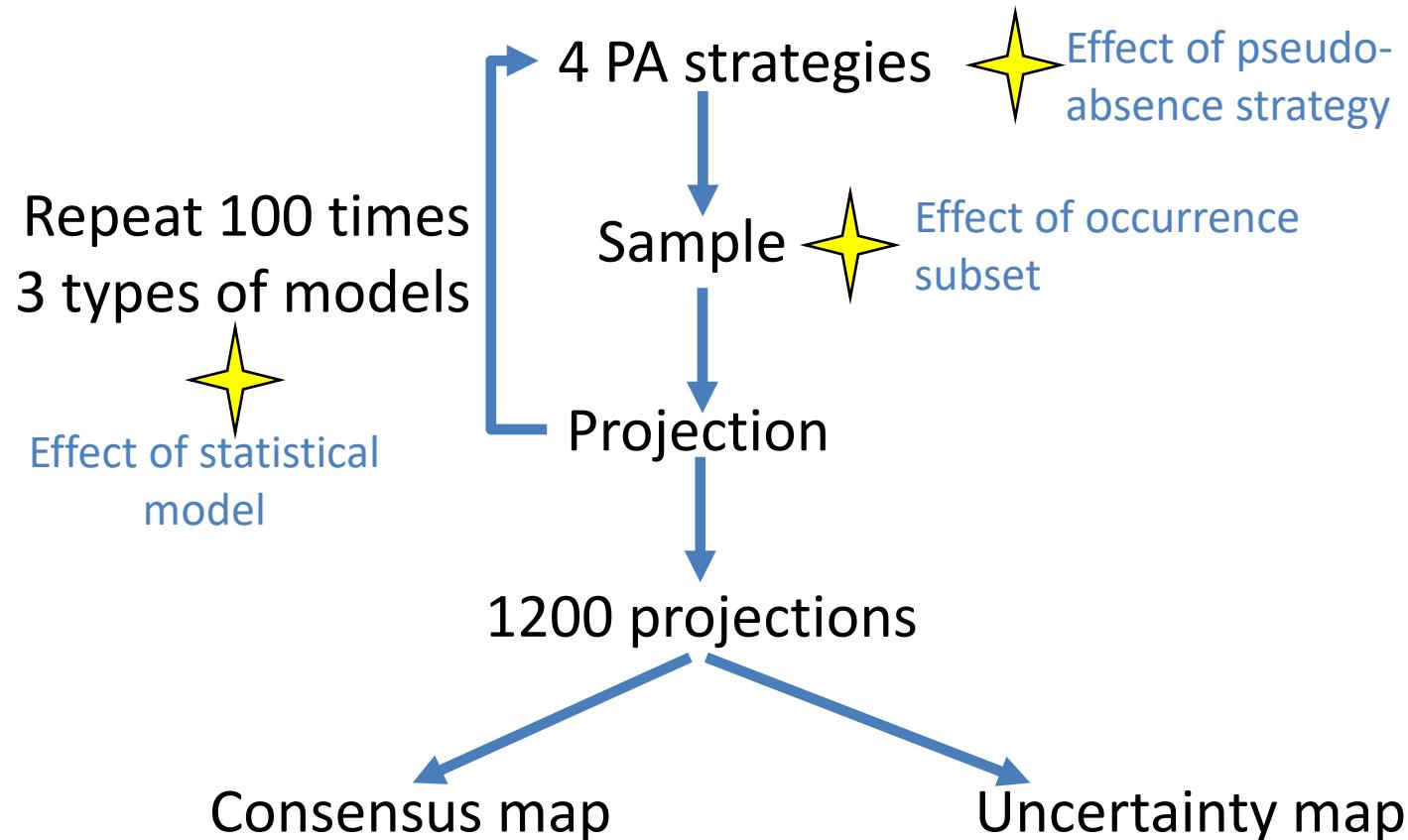


Presence-only data modelling

Adding Pseudo-Absences (PA)



Iterative approach

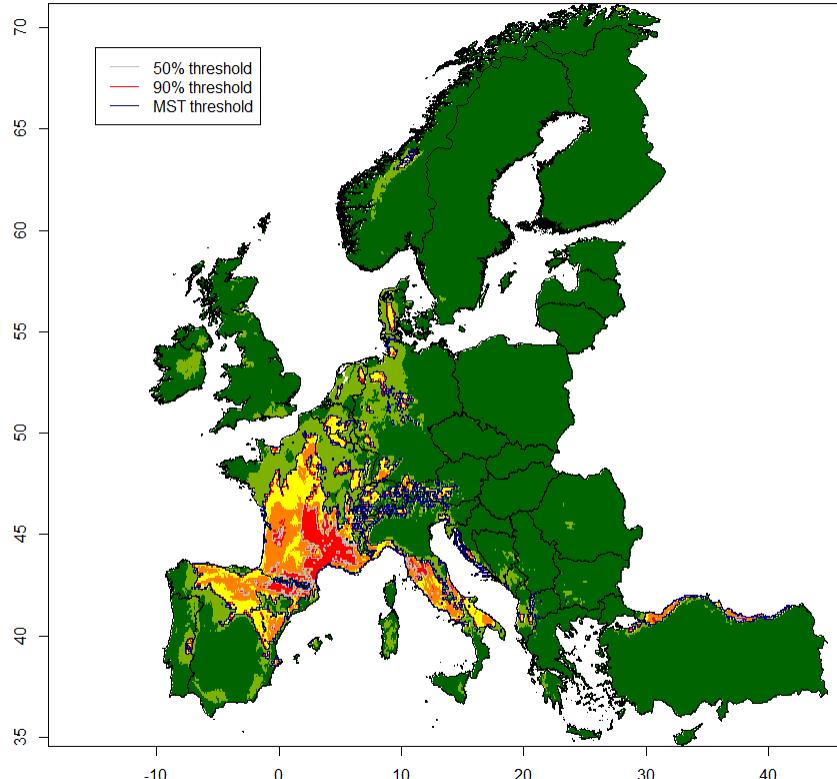


Cacopsylla pruni

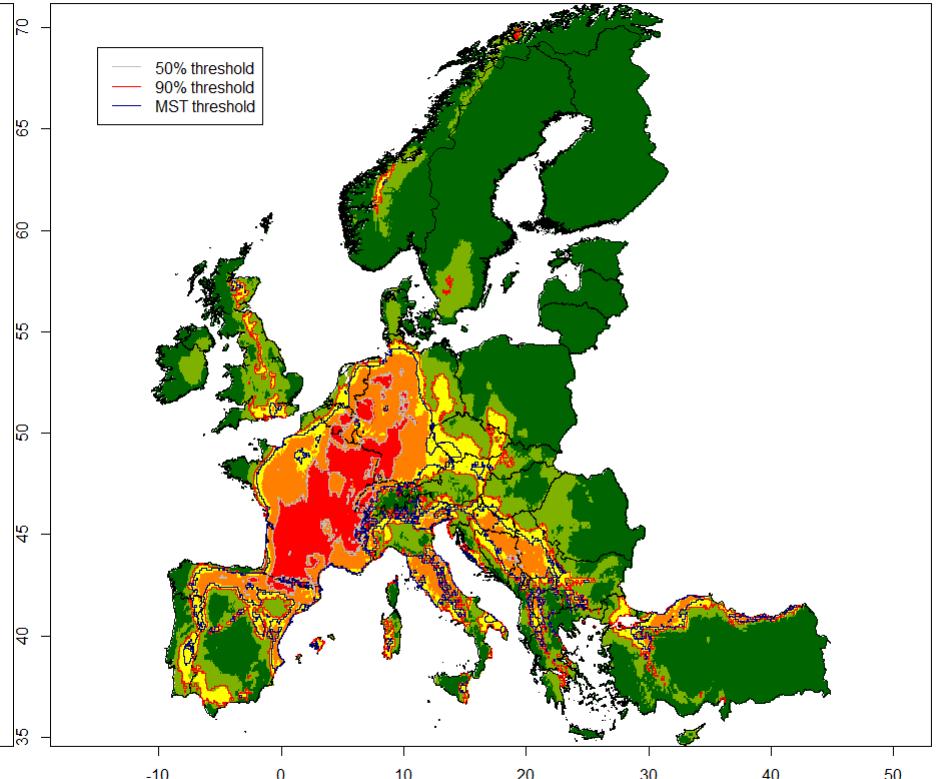


Psyllid current distributions

Clade A



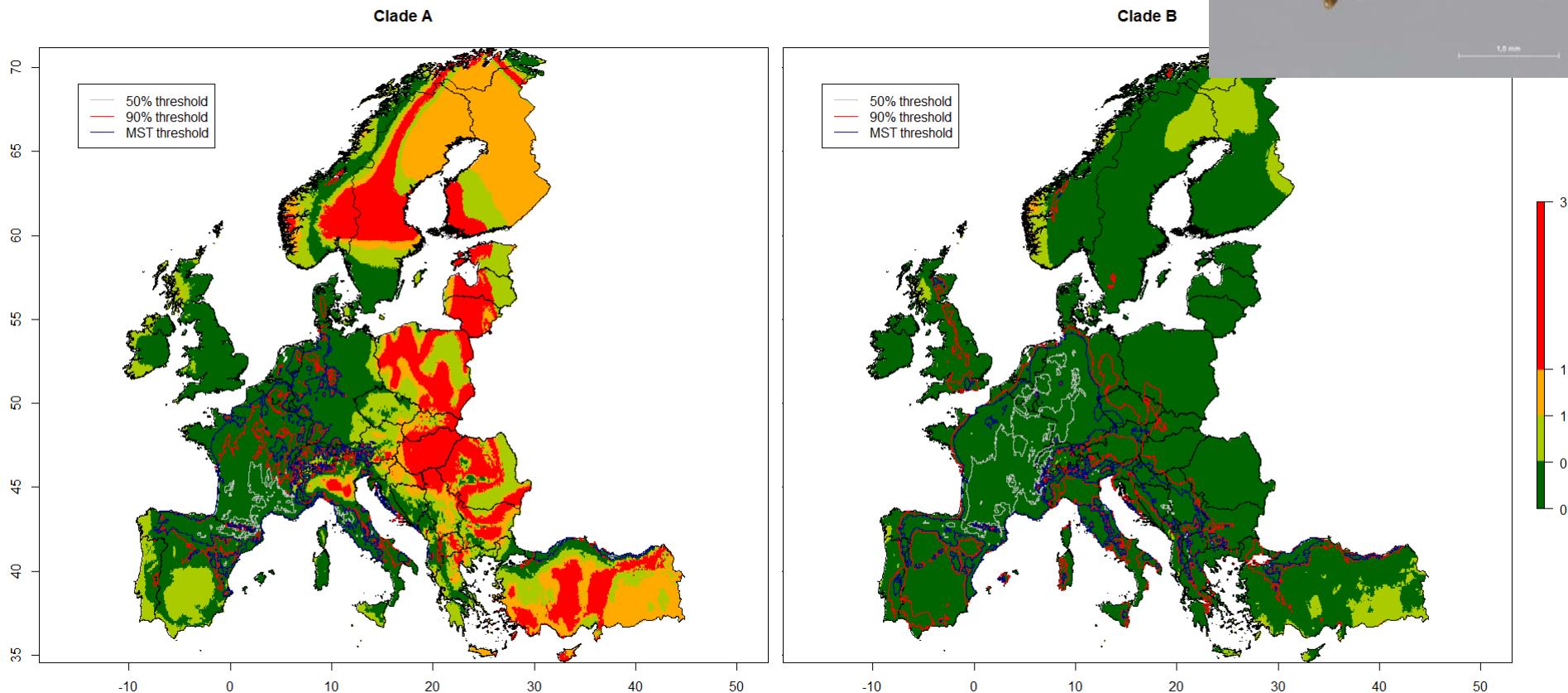
Clade B



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



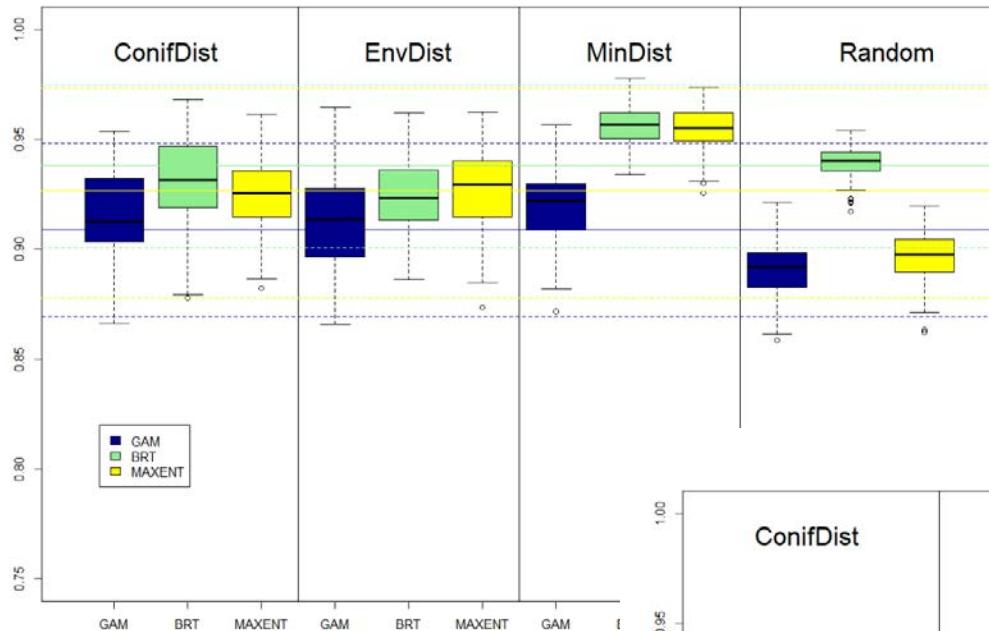
Psyllid current distributions



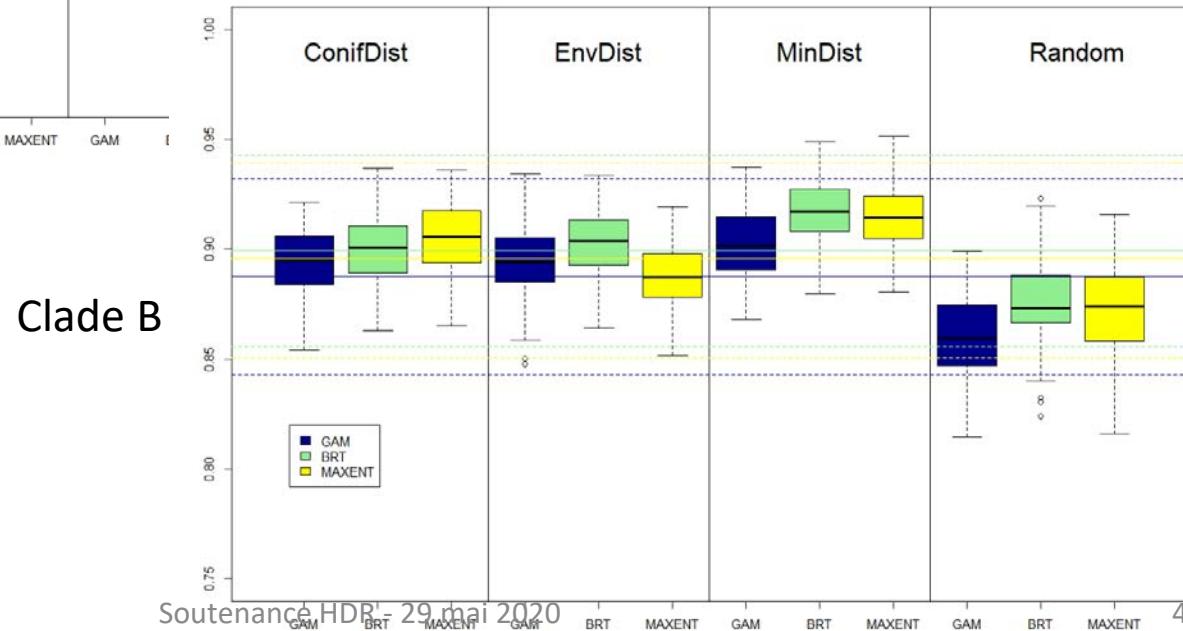
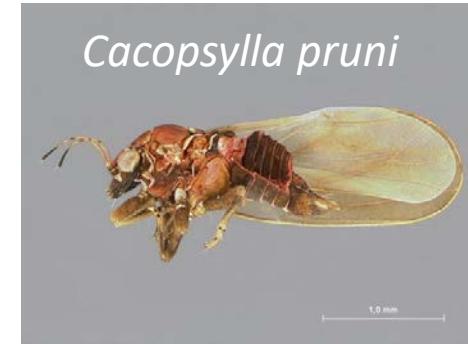
CV between 1,200 maps (standard deviation / mean):
3 models x 4 PA methods x 100 iterations each

Uncertainties

Types of pseudo-absences



Clade A



Clade B

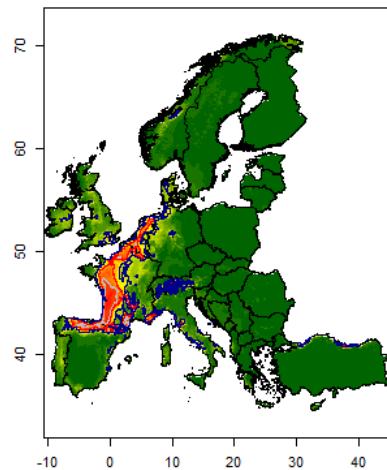
Uncertainties

Climate models x SE scenario

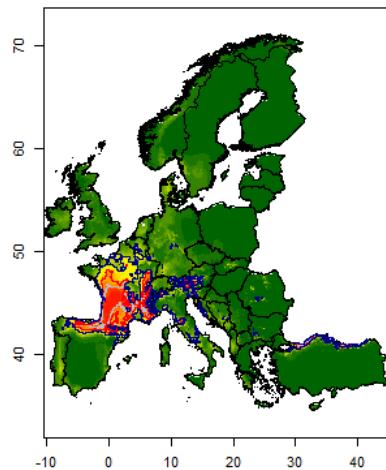
Cacopsylla pruni



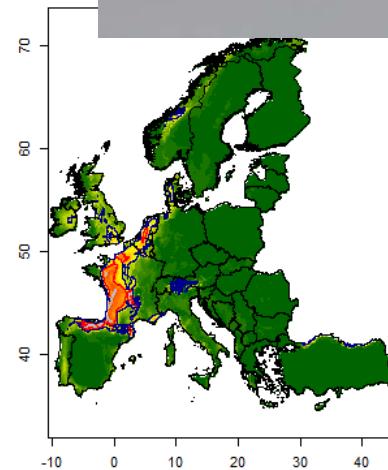
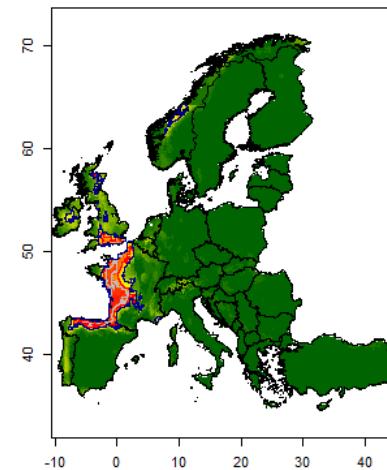
Clade A, RCP4.5-CC-2070



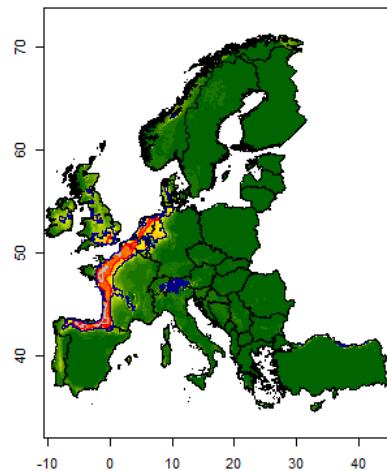
Clade A, RCP4.5-IP-2070



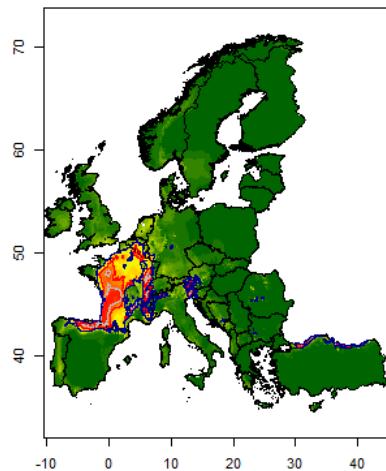
Clade A, RCP4.5-MR-2070



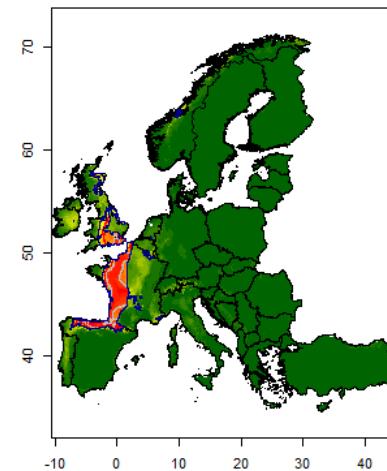
Clade A, RCP8.5-CC-2070



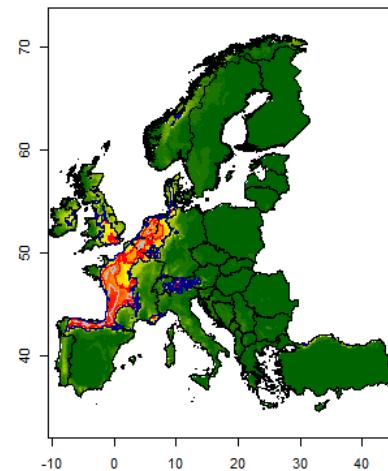
Clade A, RCP8.5-IP-2070



Clade A, RCP8.5-MR-2070

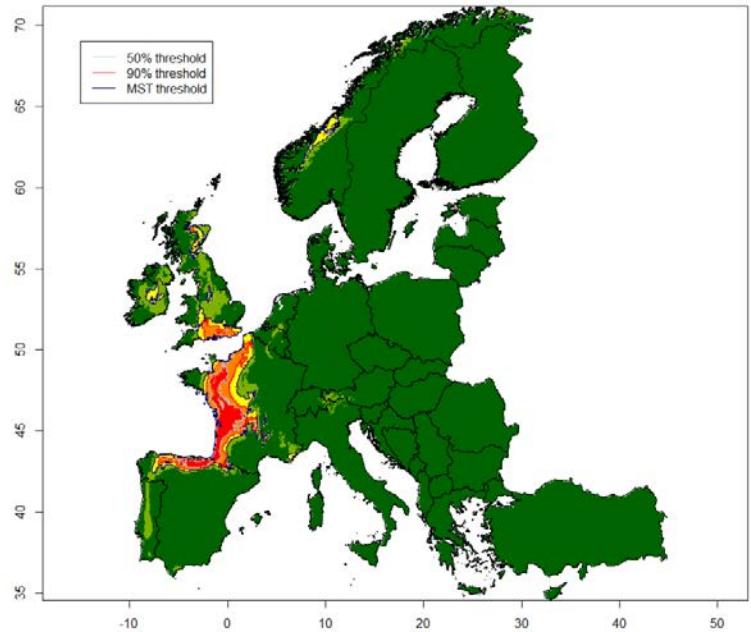


Clade A, RCP8.5-NO-2070



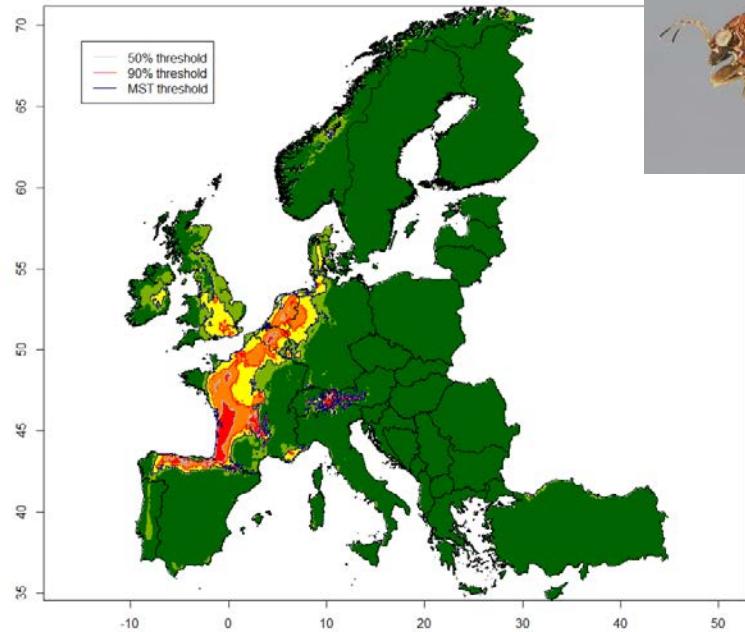
RCP 4.5 - MR

Clade A

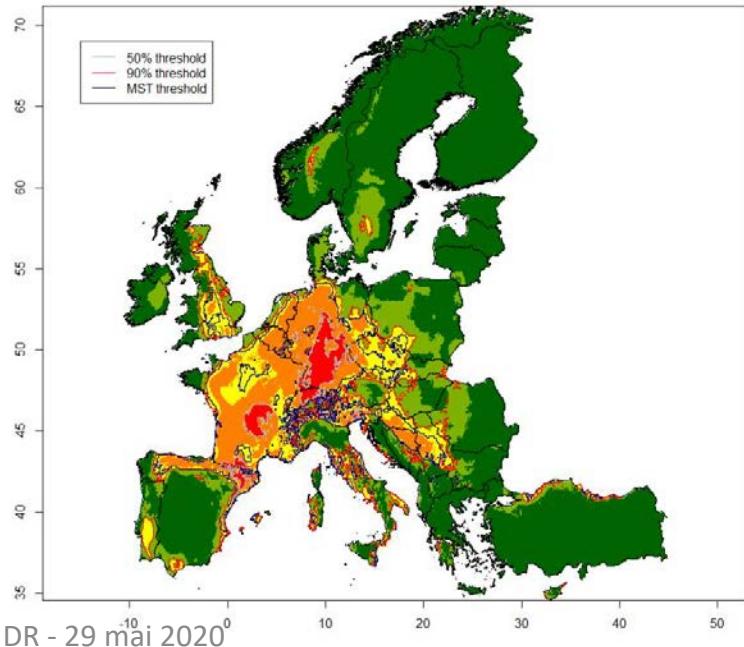
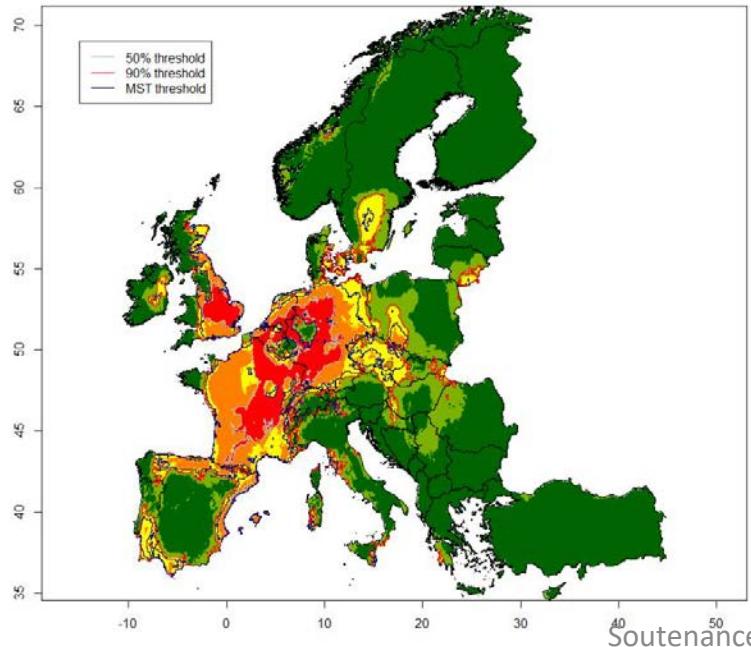


RCP 8.5 - NO

Clade A



Clade B

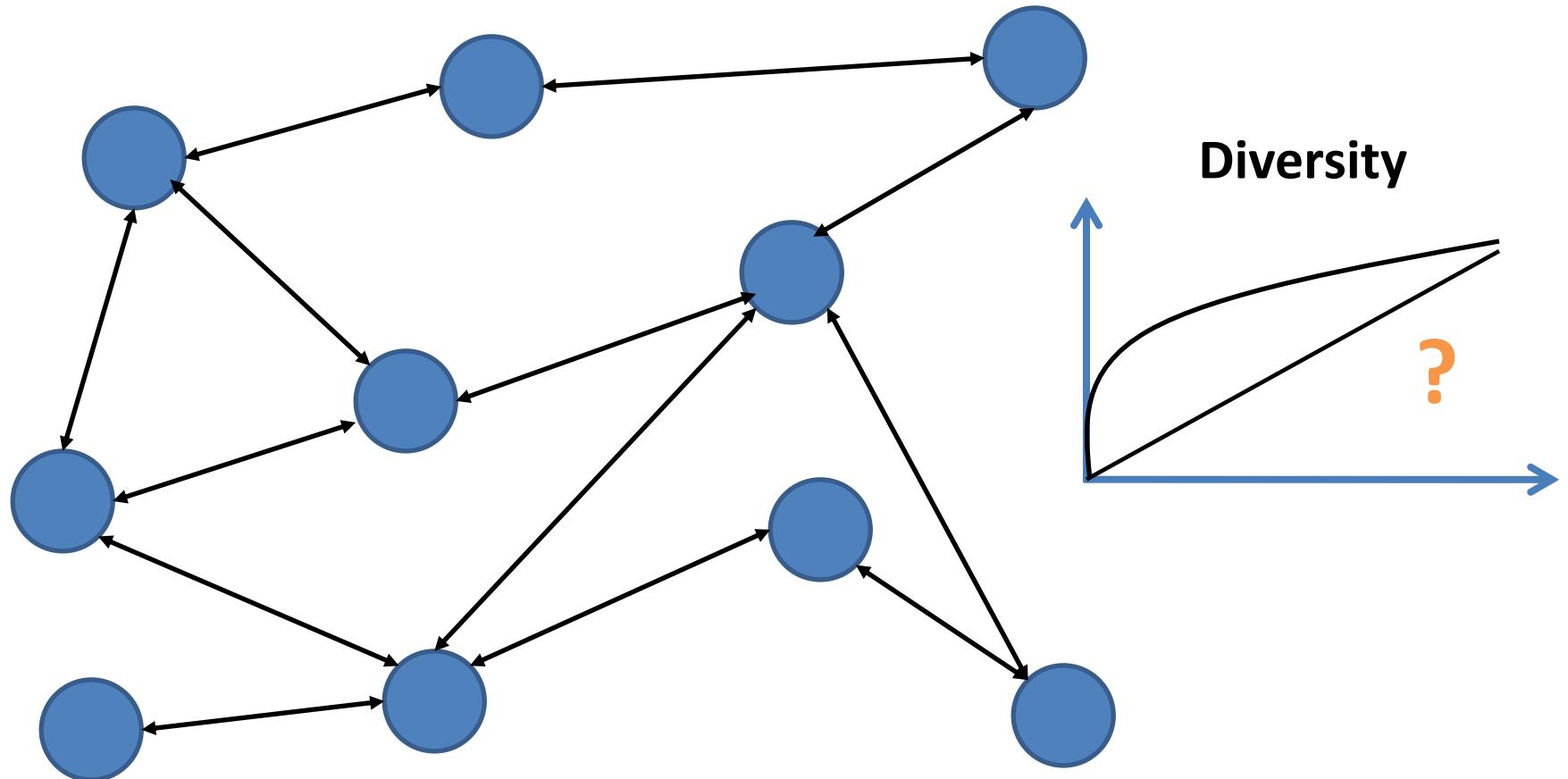


Uncertainties

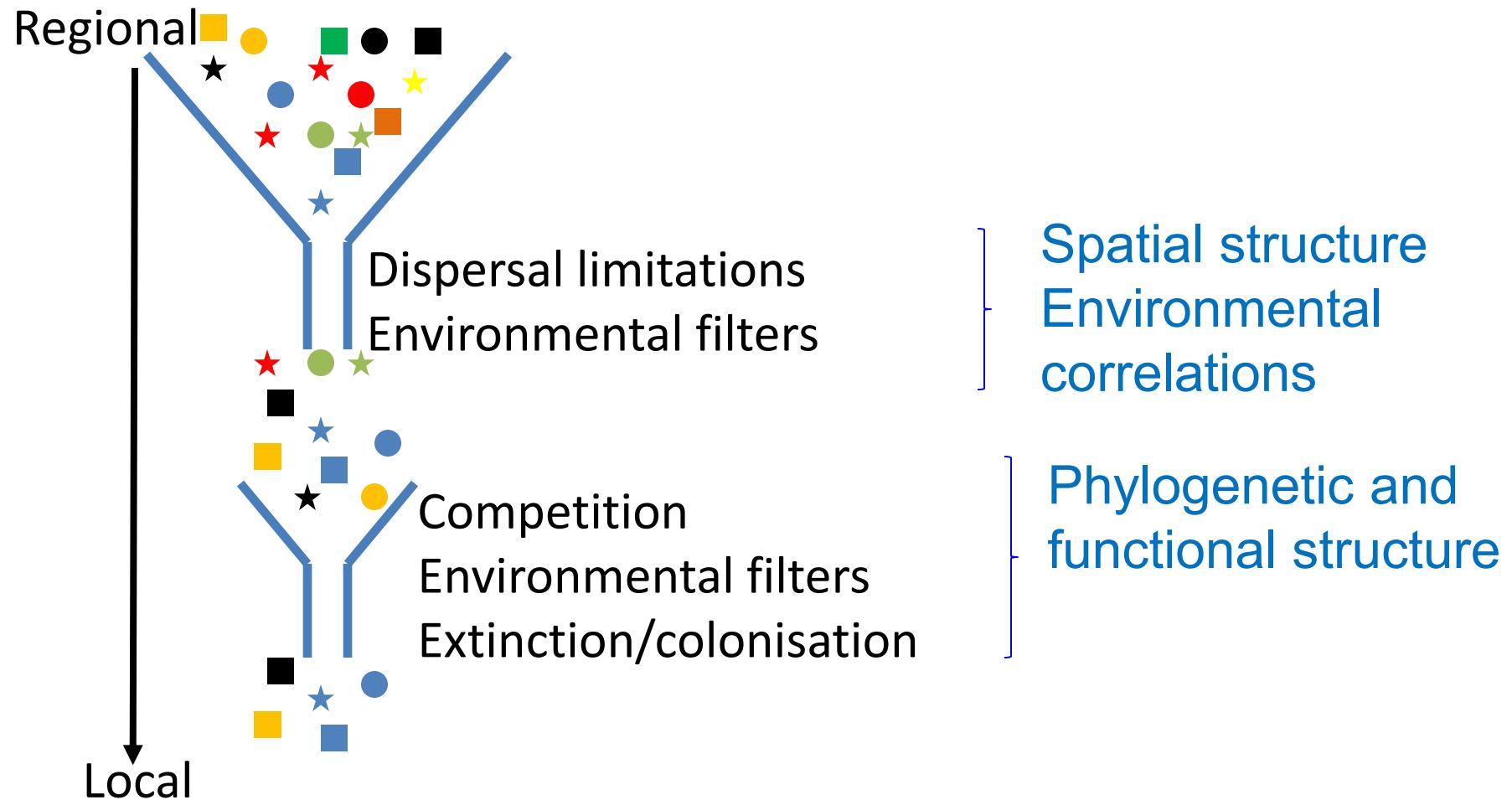
- There is no solution for SDMs with presence-only data that will fit all cases
- Combining modelling with iterations to integrate sources of uncertainty
- Ensemble forecasting has been advocated for model uncertainty: can be extended to other sources of uncertainty
- Another alternative: use virtual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

II – Metacommunities and multiple facets of diversity

Metacommunities

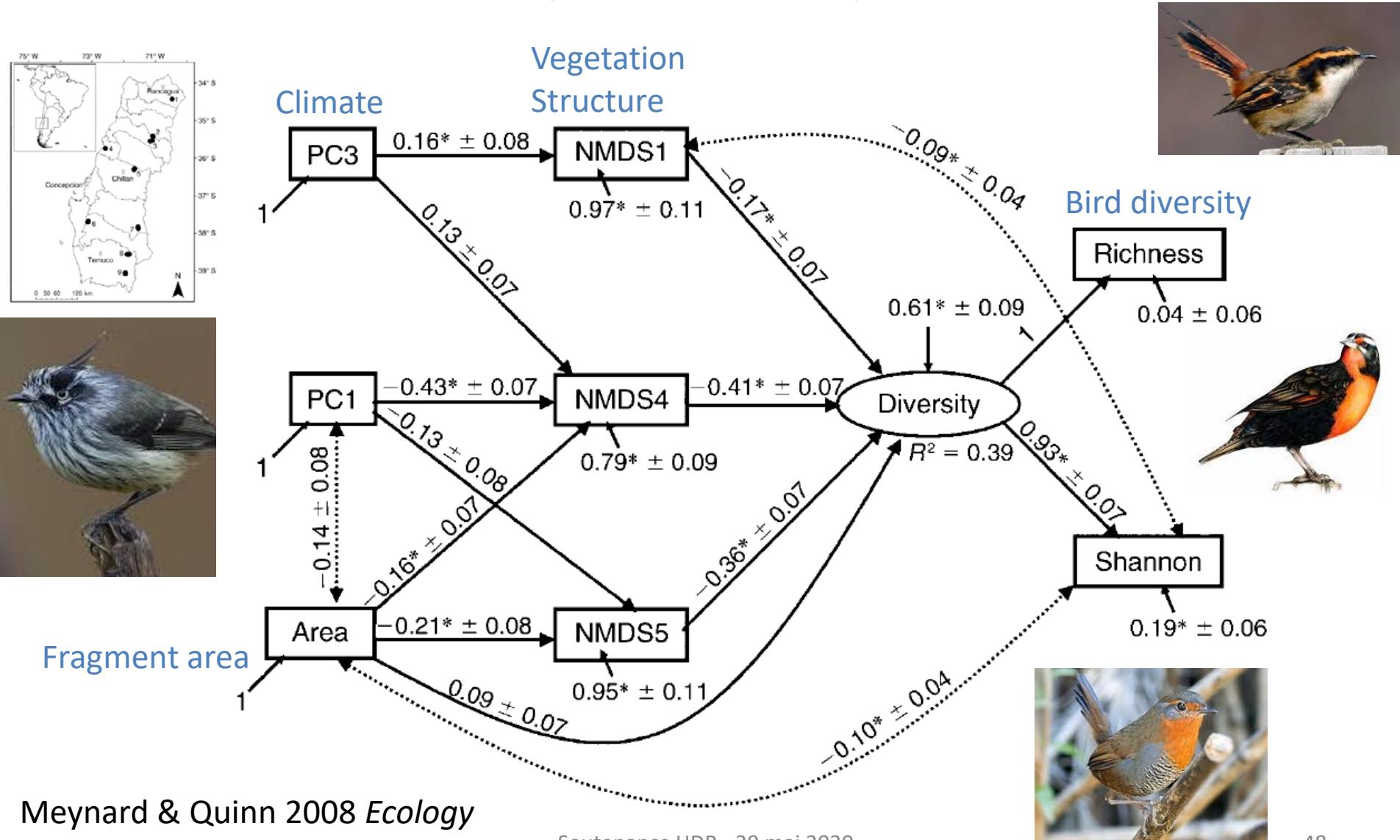


Processes driving diversity at different scales



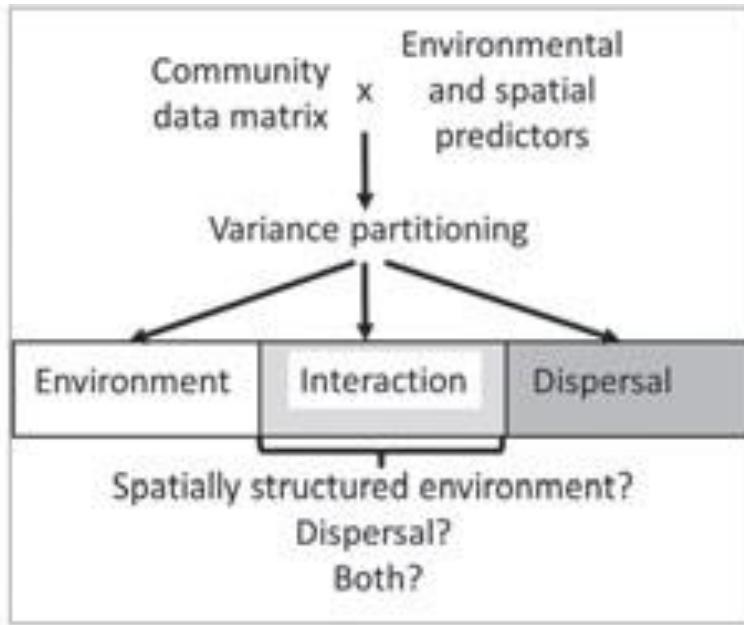
Processes driving diversity at different scales

Bird diversity in Chilean temperate forests

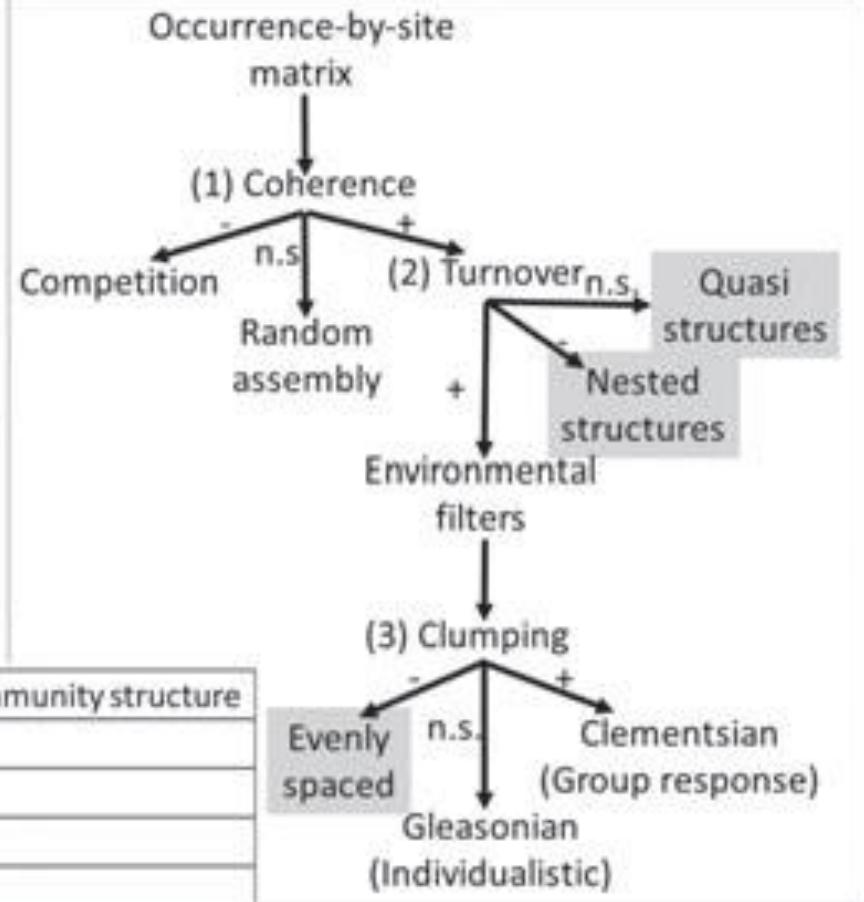


Metacommunities

Variance partitioning



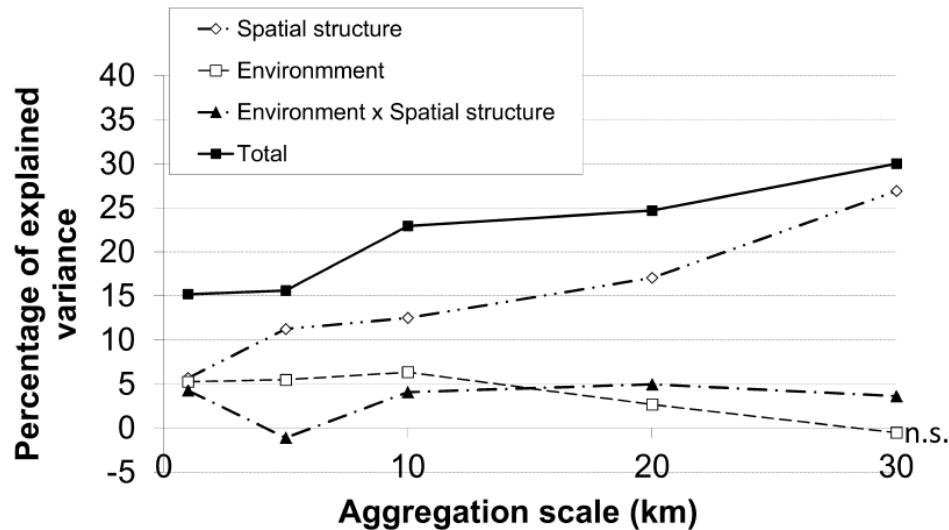
Metacommunity structure approach



Meynard et al 2013 *Journal of Biogeography*

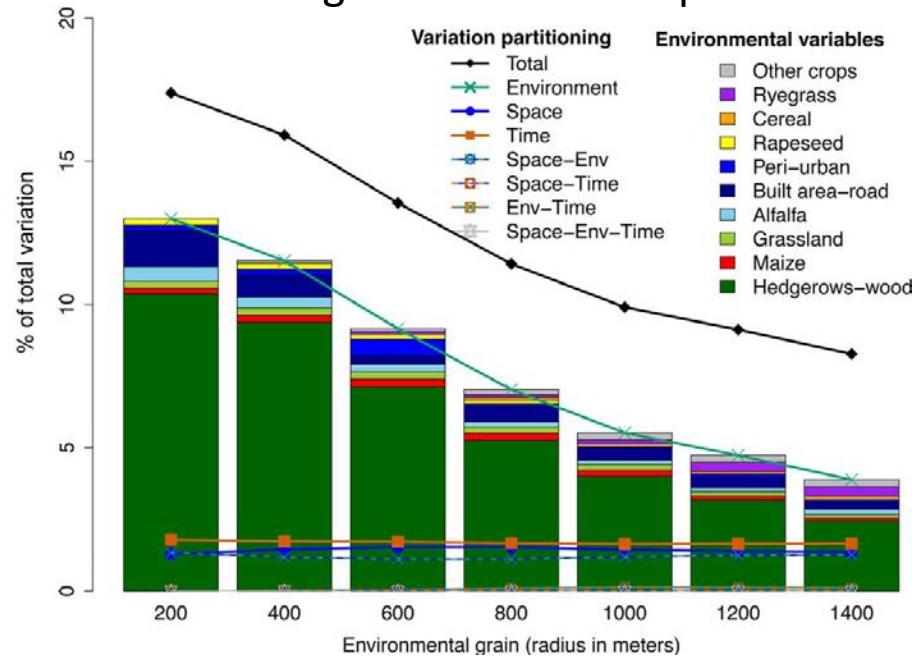
Metacommunities

Alpine grasslands



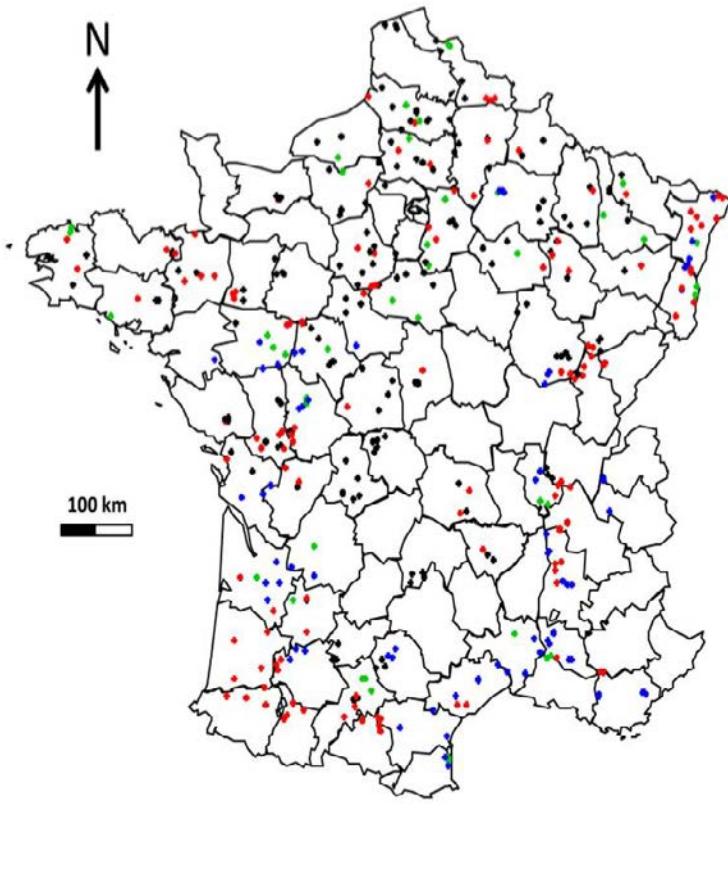
Meynard et al 2013 *Journal of Biogeography*

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

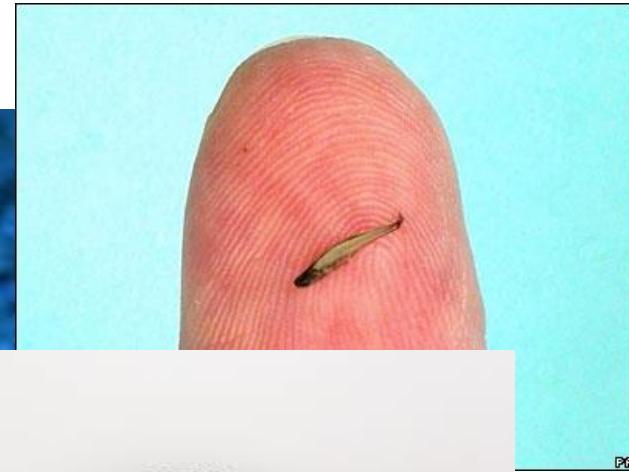
In collaboration with Guillaume Fried (ANSES)

Diversity and origins of life have fascinated biologists for centuries



Functional diversity:

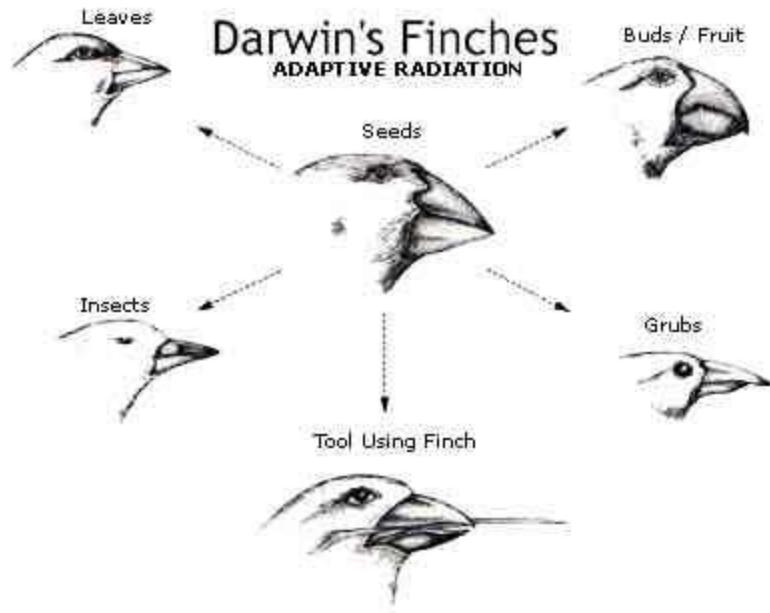
diversity of functions or life forms accumulated in an area or community



- 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

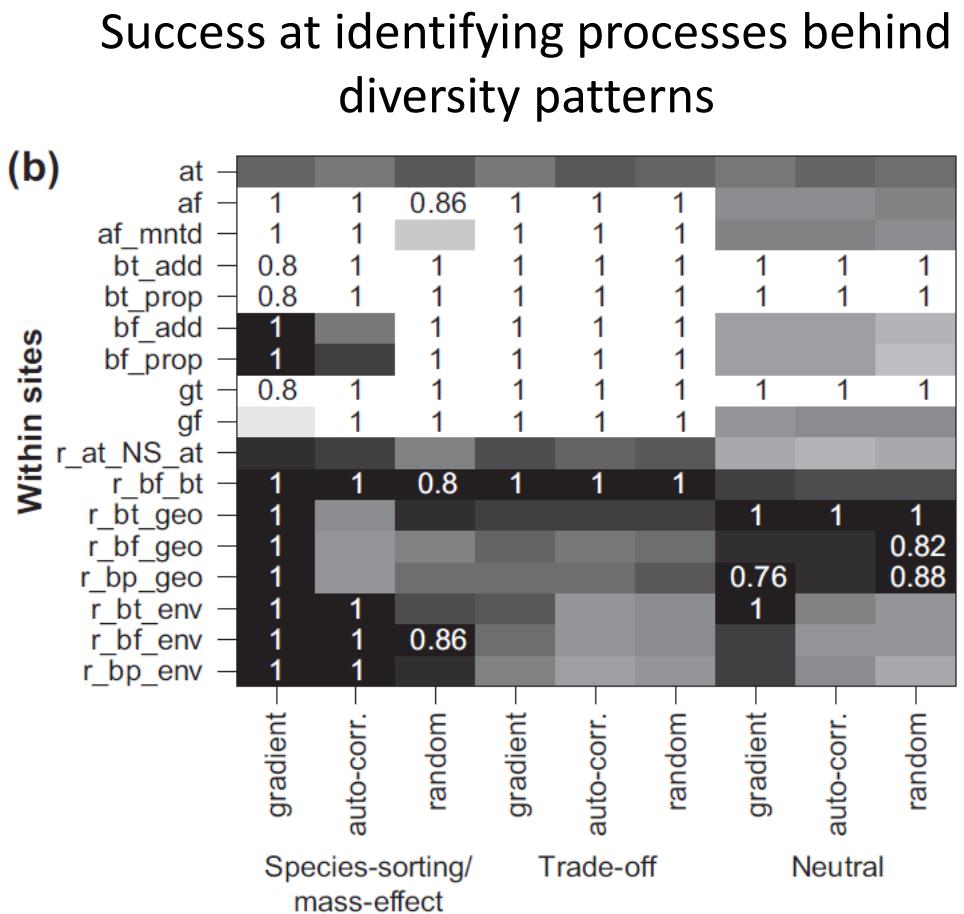
THEORY	PROCESS
Neutral	Random dispersal-competition
Environmental filtering	Environment filters out species
Colonization / extinction	Dispersal X competition
Mass effect	Dispersal X environment

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

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

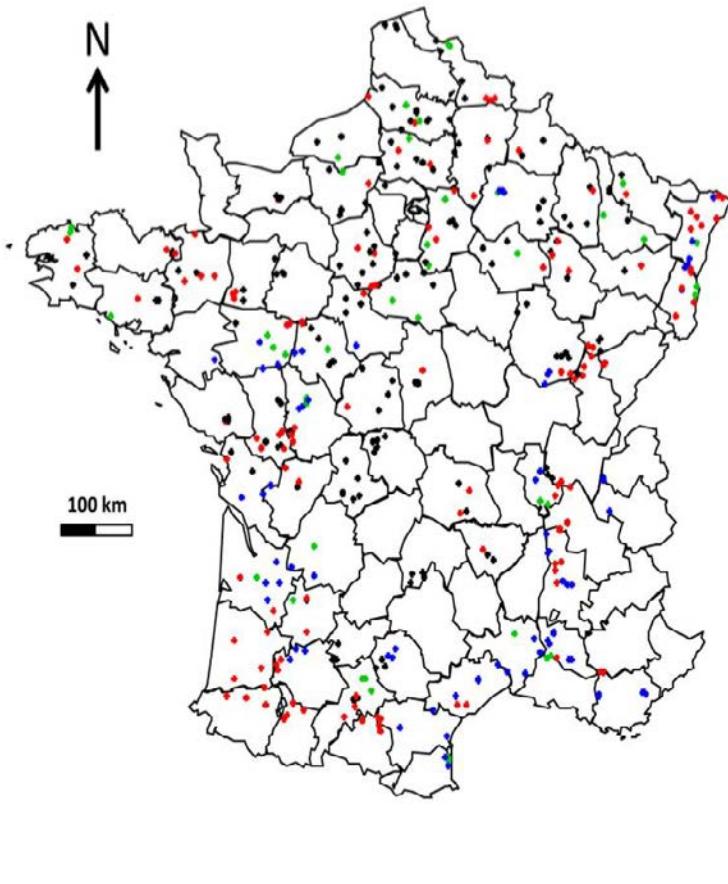
The virtual ecologist in diversity studies

- **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

