



Forecasting the risk of locust outbreak using machine learning & agent-based modeling

Lucile Marescot

26 of March 2024



**Cyril Piou, Esther Diouf, Elodie Fernandez,
Mamour Touré, Arianne Cease, Lemine Hamouny**



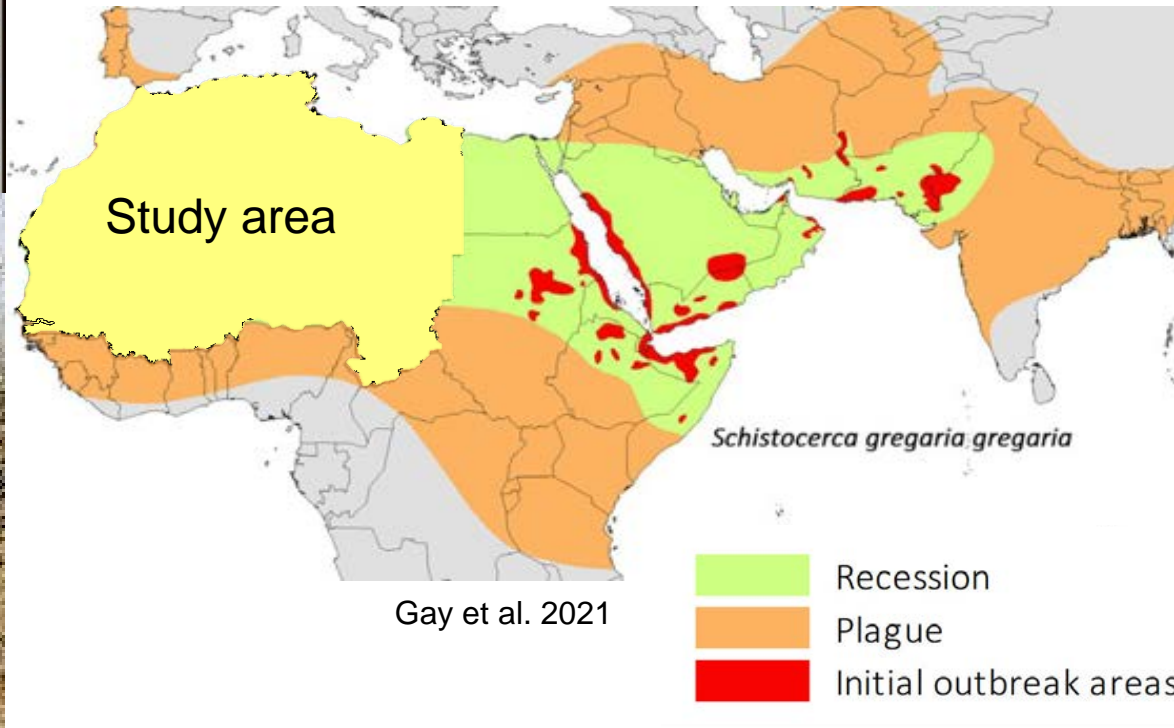
Locust : the eighth plague



2003-2005
West Africa
400 millions \$

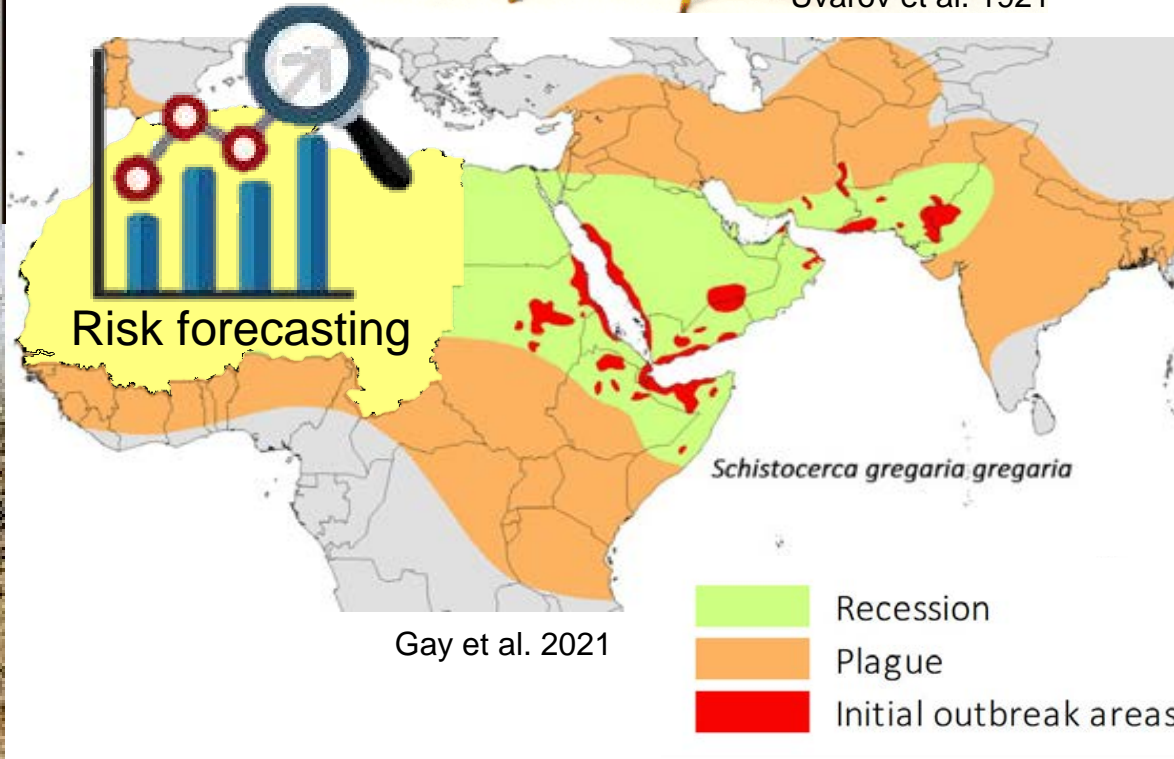


Uvarov et al. 1921



2019-2021
East Africa
2.5 billion \$

Locust : the eighth plague



Objectives

Uncertainties

Forecasted risks

(Piou & Marescot 2023)

Fundamental

Safe agricultural production

- Institutional/governance

Threats for human health
Threats for biodiversity
Economical costs

Who does it ?

Means

Protect crops/pastures from swarms

- Climate change
- Demographic stochasticity

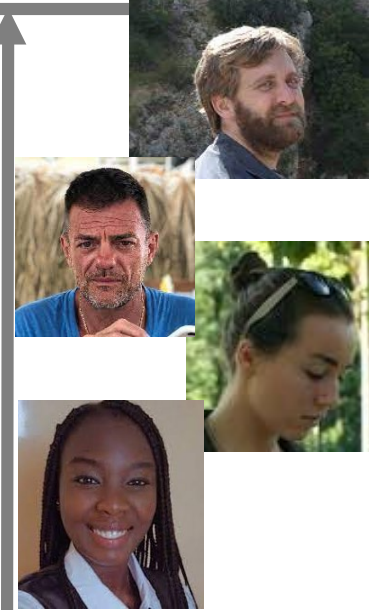
Impacts on production

Management

Global
Proactive

Control uncertainty

Outbreak



Local
Preventive

Sampling uncertainty

Gregarization Breeding

Why to do it ?

Monitoring



Structural uncertainty

Presence/absence

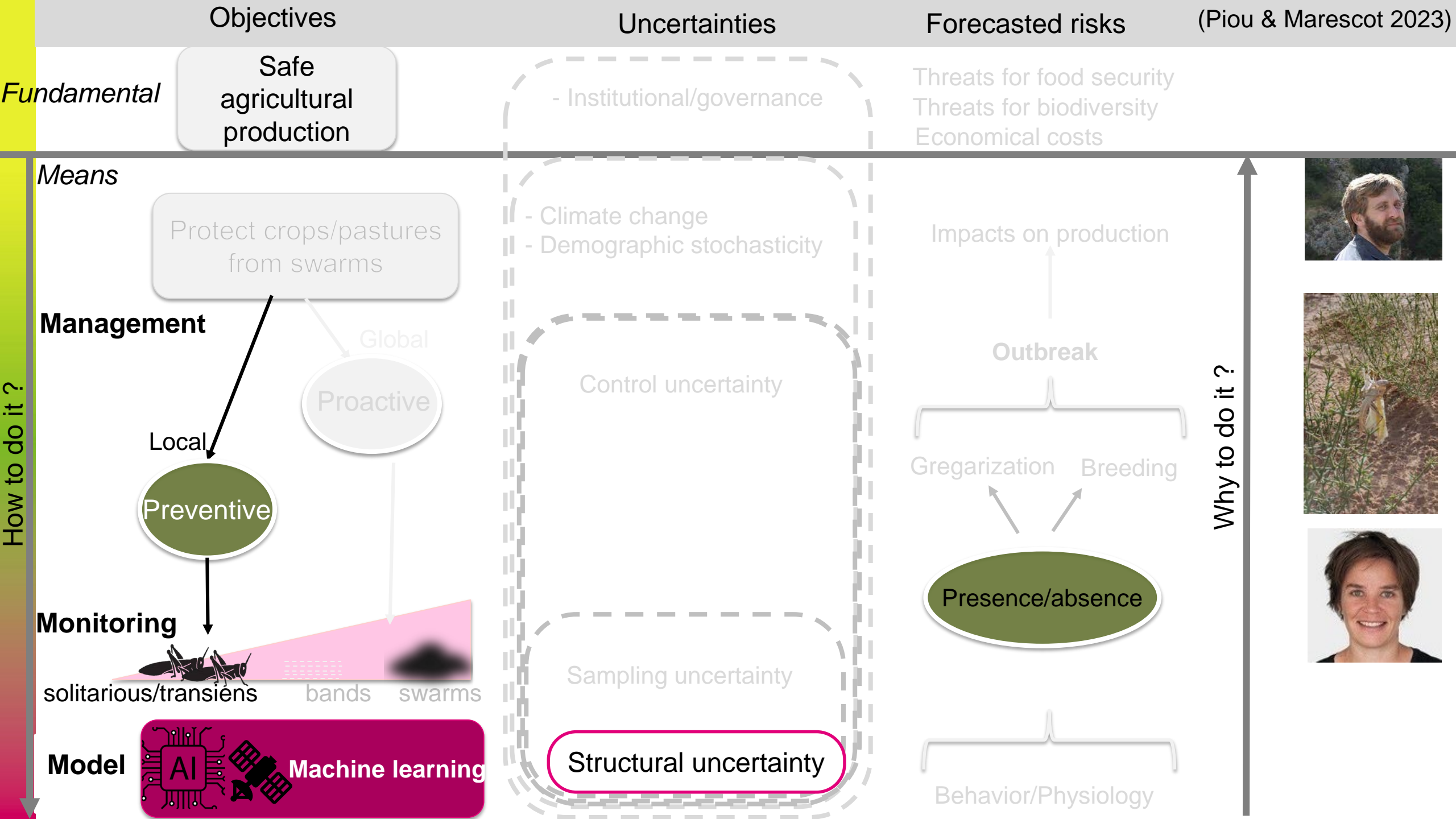


Model Lab

Mental model
Machine learning /
Agent Based Model
Experiment

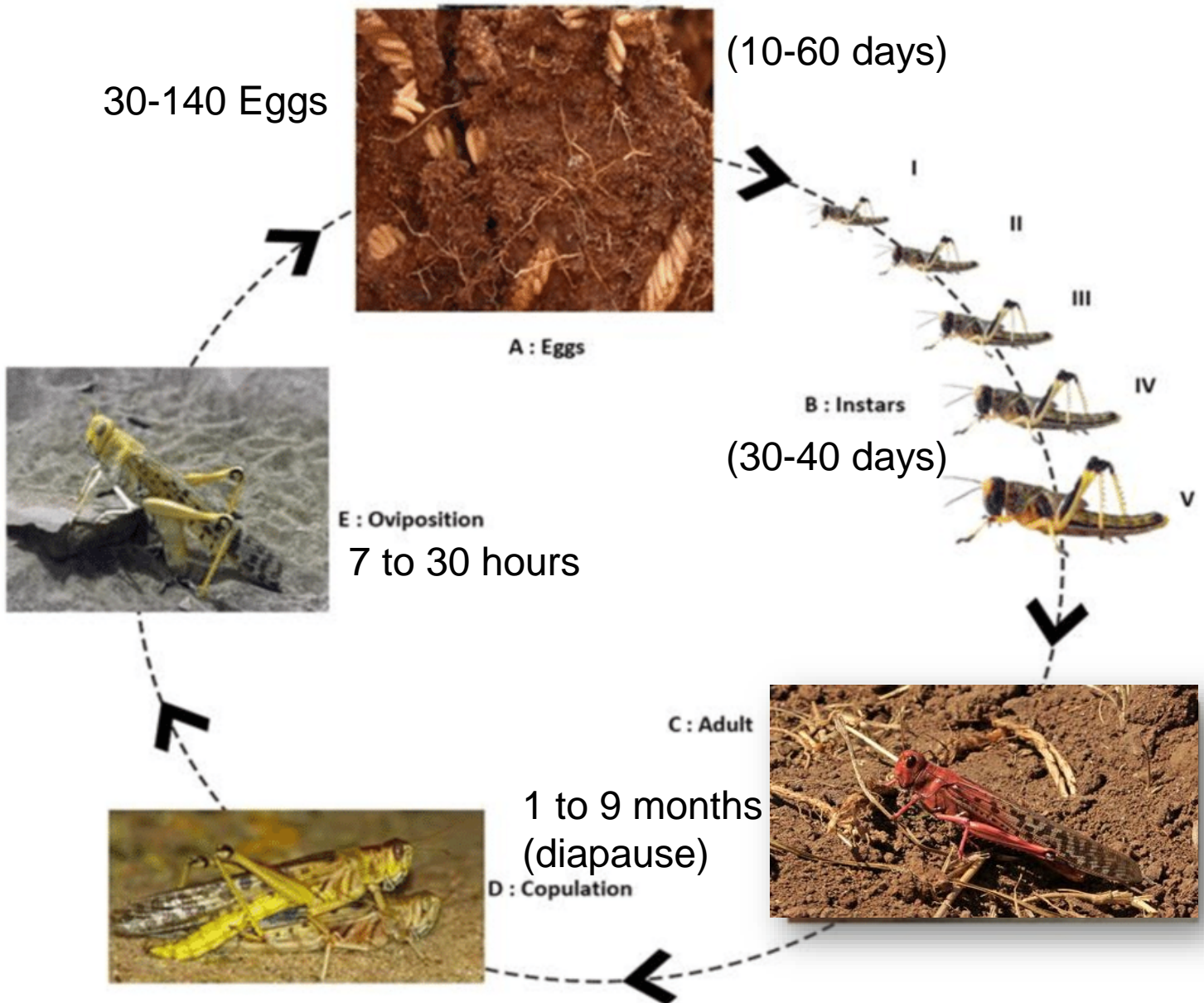
Behavior/Physiology










Desert locust life cycle

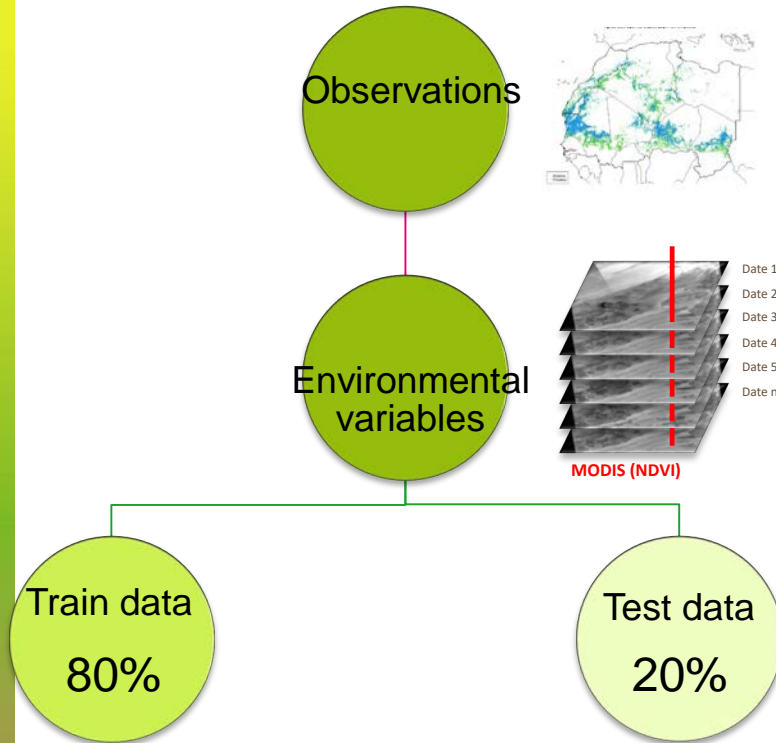
Schistocerca gregaria



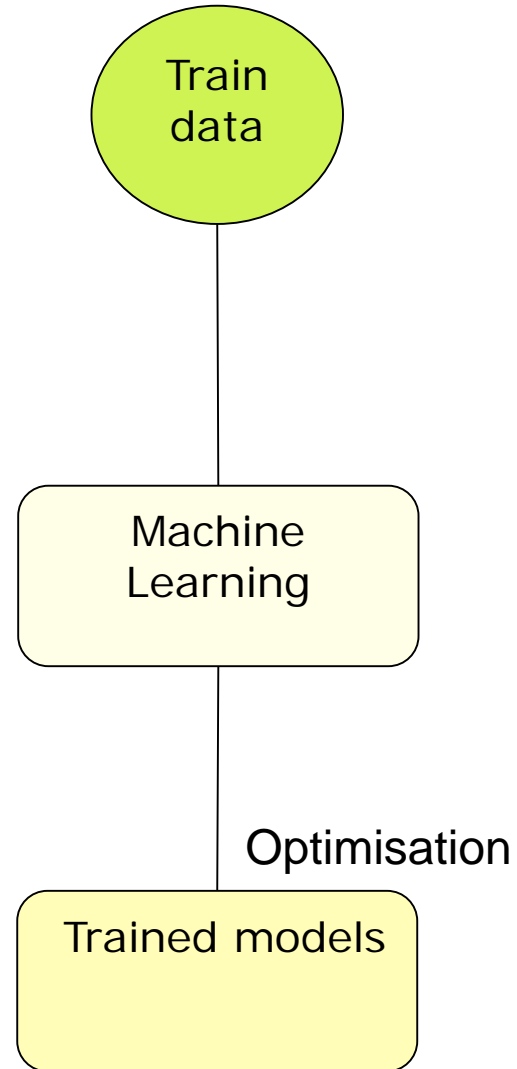
-  Temperature
-  Pluviometry
-  Photoperiod
-  Vegetation
-  Density

(Shresta et al. 2021)

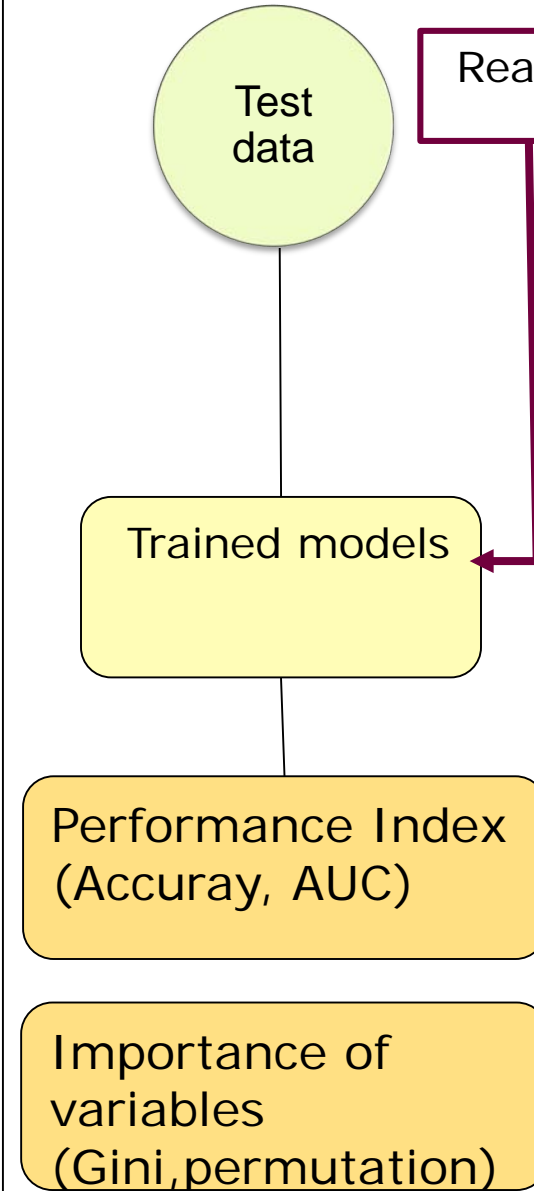
Step 1: Data processing



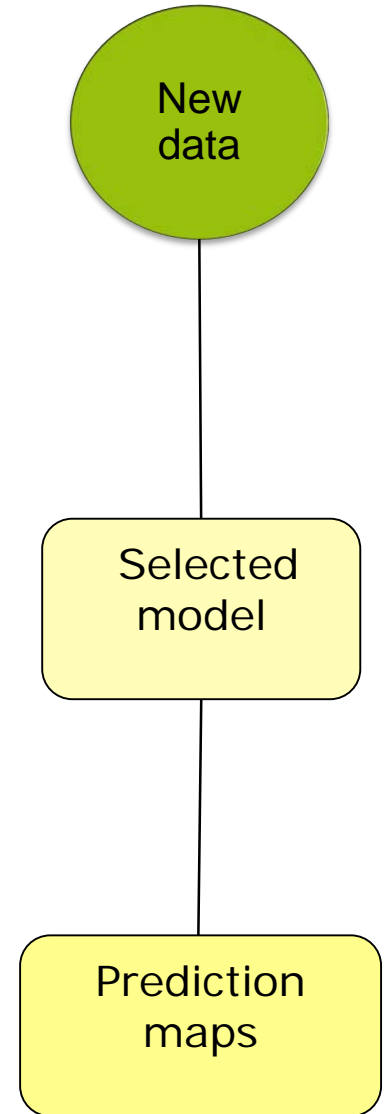
Step 2: Training



Steps 3 : Evaluation

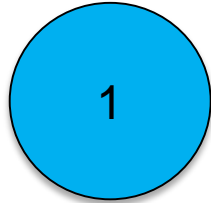


Step 4: Forecasting

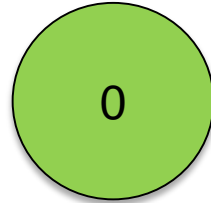


Response variable

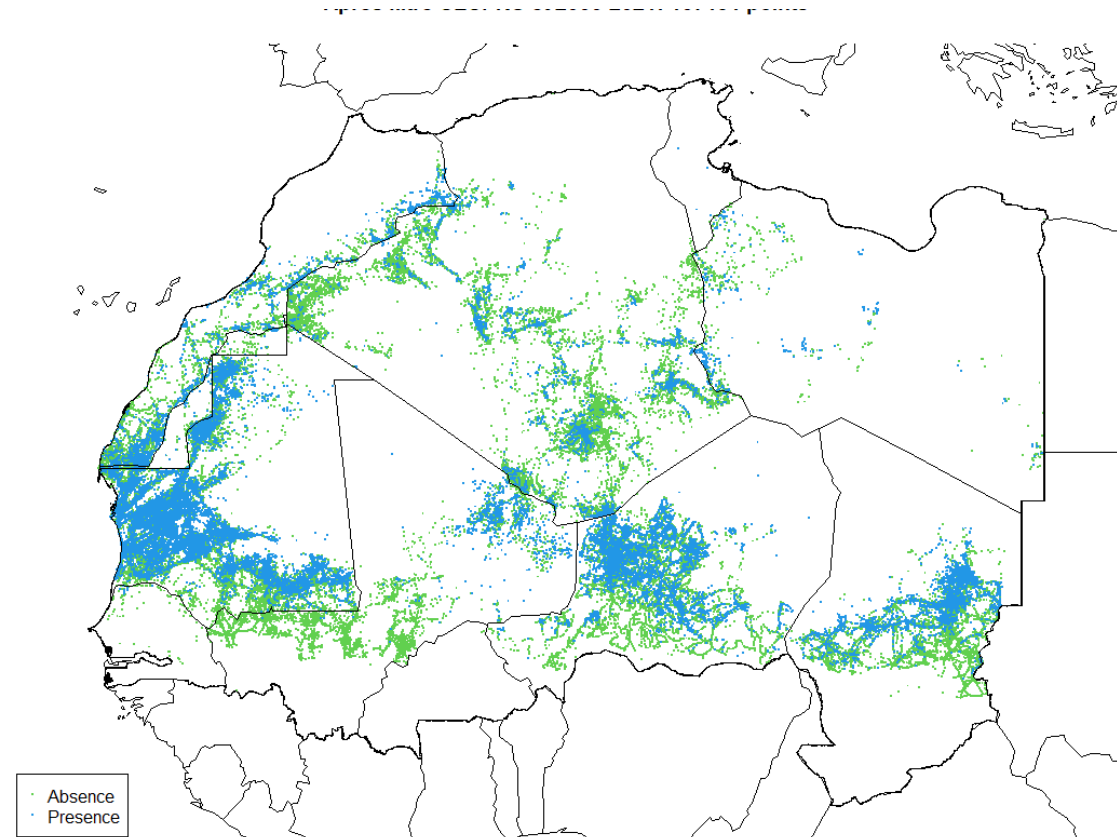
Presence



Absence

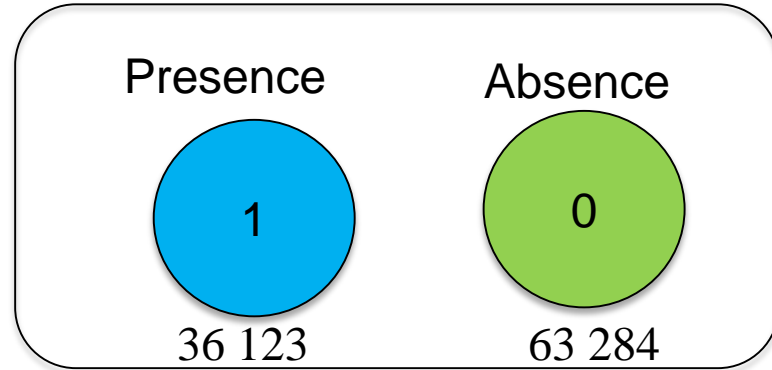


- 10 countries : Mauritania, Morocco, Algeria, Tunisia, Libya, Mali, Niger, Chad, Senegal, Burkina Fasso
- 99 407 prospection points from 2003 to 2022
- 12 861 240 sites

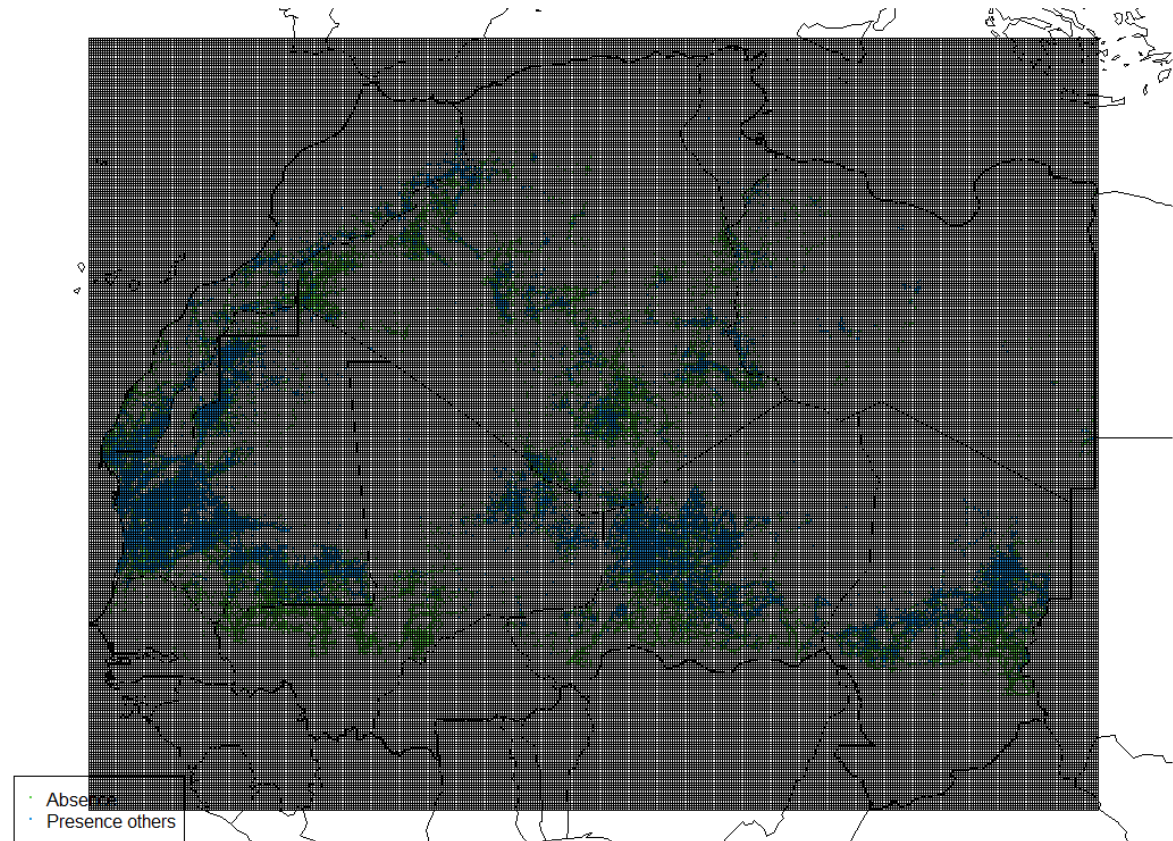


Response variable

Site of 1 km² every 10 days



- 10 countries : Mauritania, Morocco, Algeria, Tunisia, Libya, Mali, Niger, Chad, Senegal, Burkina Fasso
- 99 407 prospection points from 2003 to 2022
- 12 861 240 sites



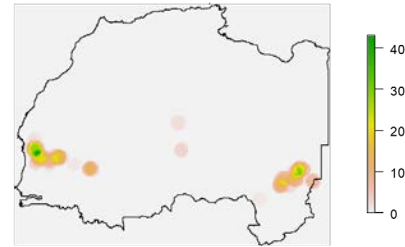
Step 1: Data processing

Explanatory variables from remote sensing

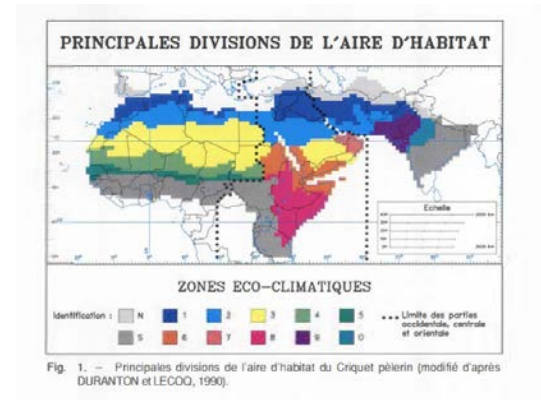
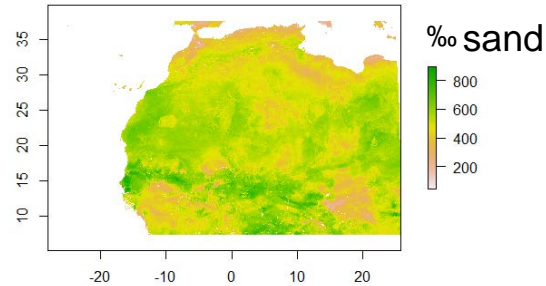
Recent observations on
100km buffer



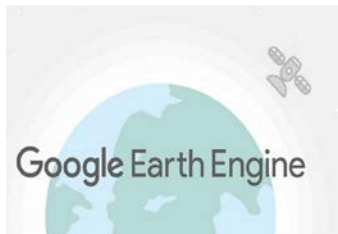
- Static Variables
- Ecological Units
 - Sand cover
 - Min and Max NDVI



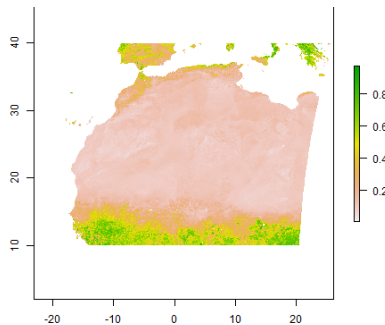
Spatial autocorrélation



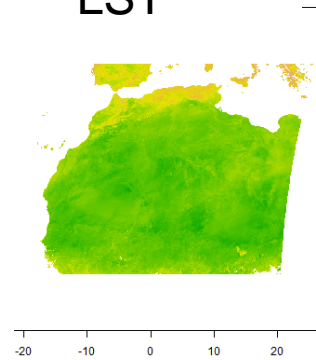
Dynamic variables extracted from MODIS and SMOS at time t, t-1, t-2 ...t-10 (ten decades = 100 days)



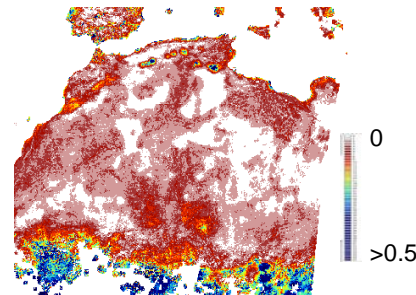
NDVI



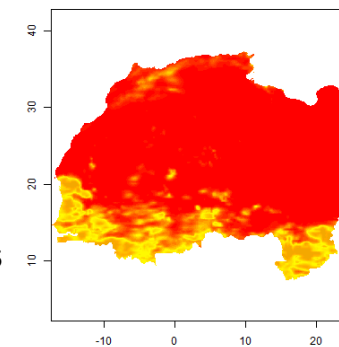
LST



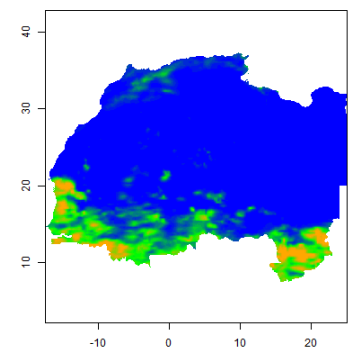
SM



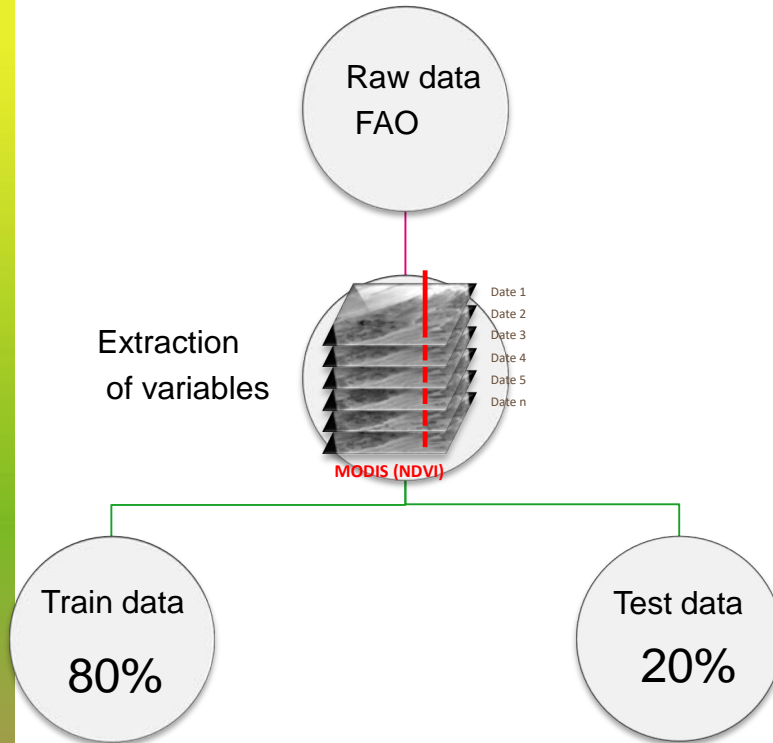
RAIN



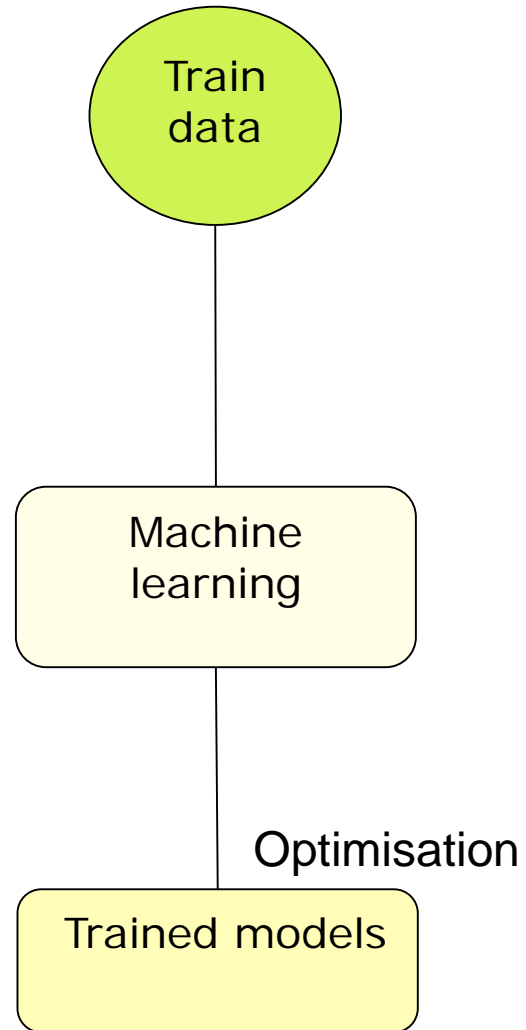
NPP



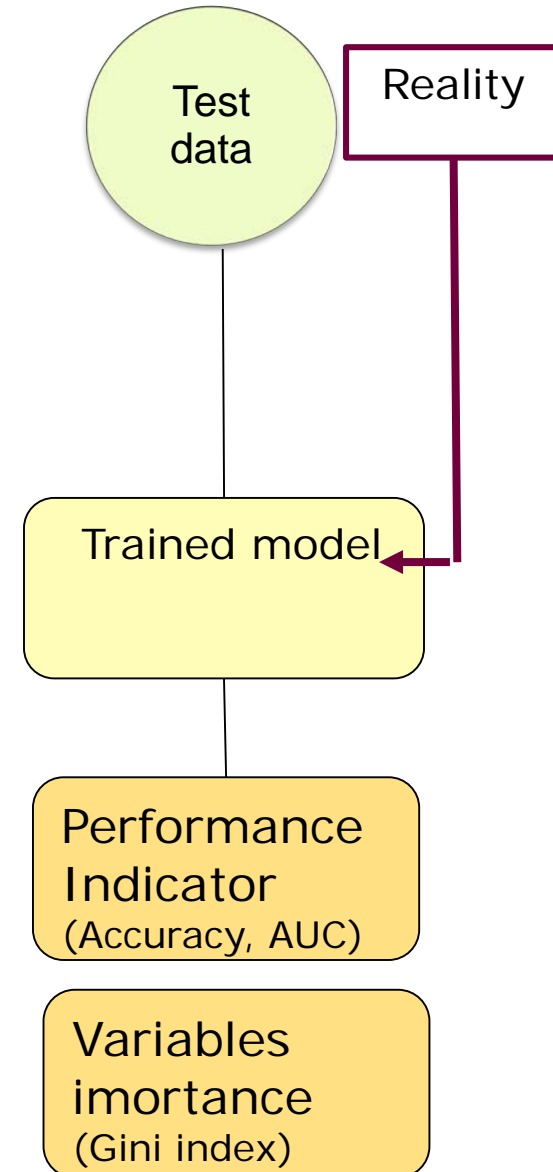
Step 1 : data processing



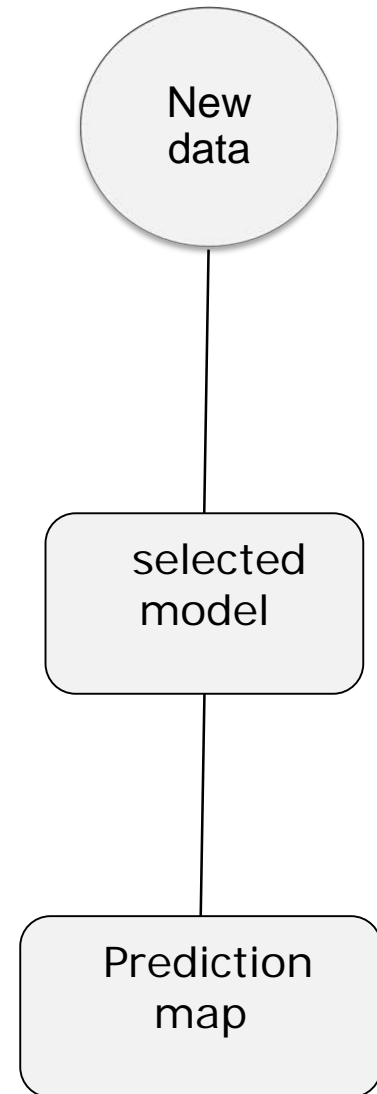
Step 2 : Training



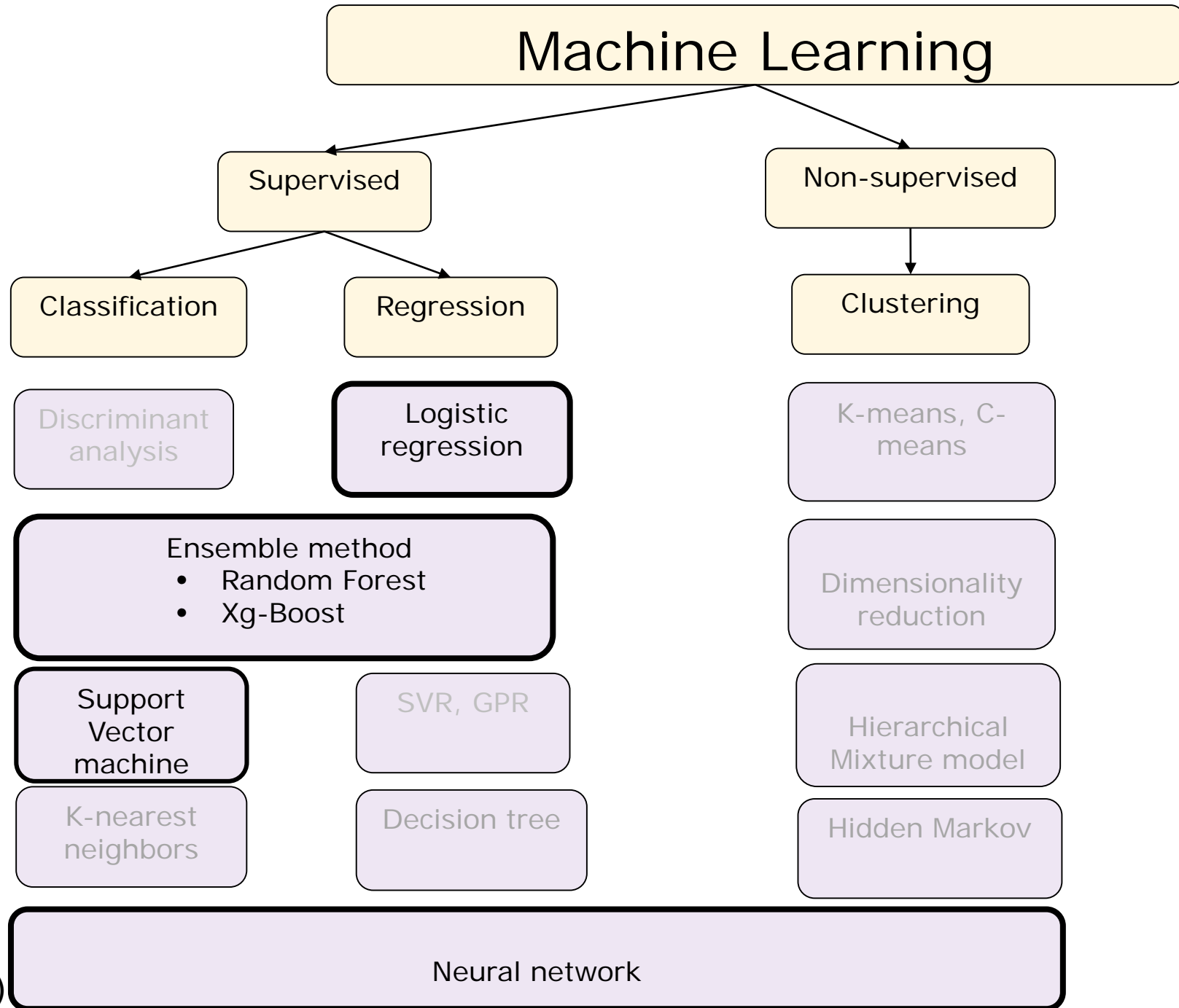
Step 3 : Evaluation



Step 4 : Prediction



Step 2 : Training

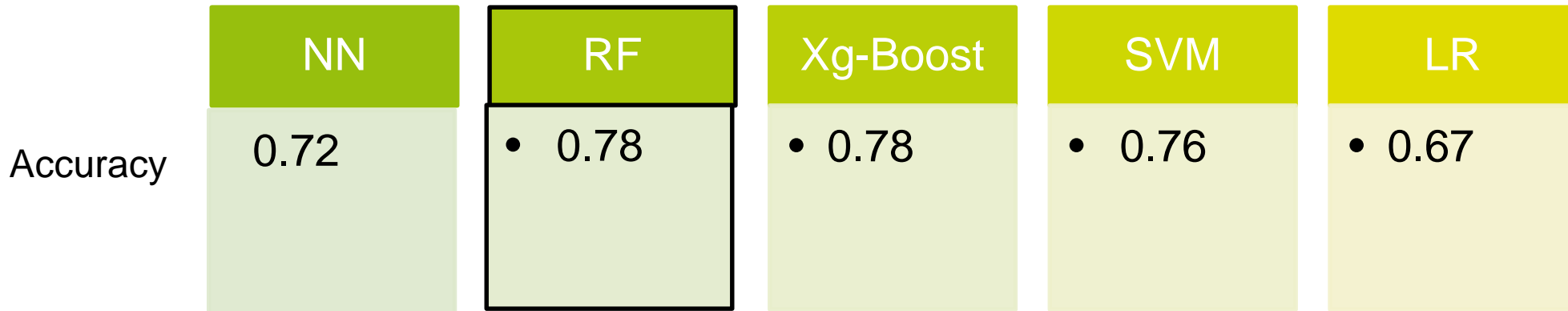


(Breiman 2001)

(McCulloch & Pitts 1943)

Choose the best algorithm

- Presence/Absence ~ 53 variables
- 5 machine learning algorithms (*Tidymodels* package)
- Cross validation and hyperparameter tuning



- Recursive feature elimination (*caret* package)

Select models for presence/absence

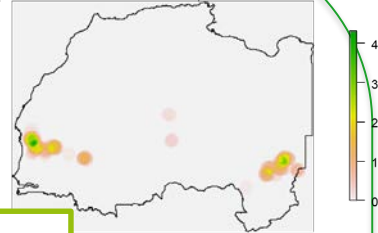
Full model AUC = 0.77

Spatial autocorrélation

+

Seasons

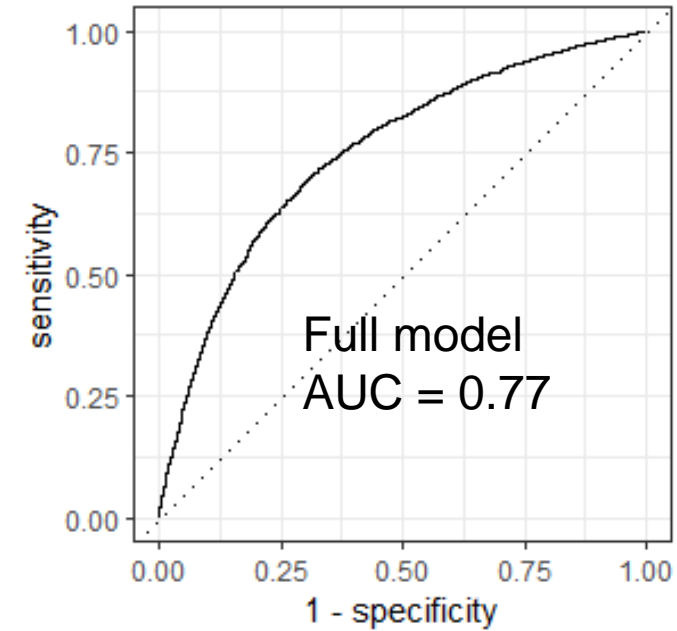
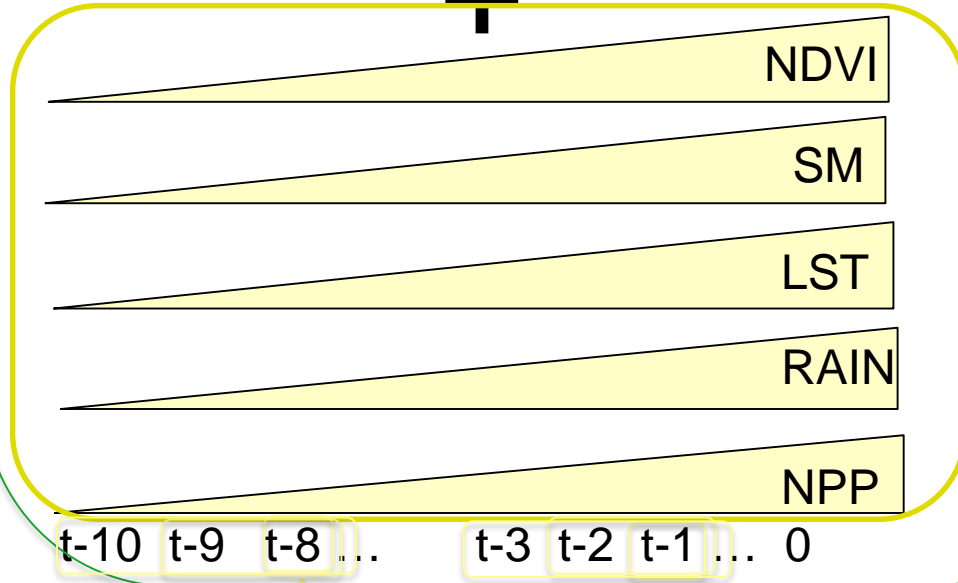
+



Static variables

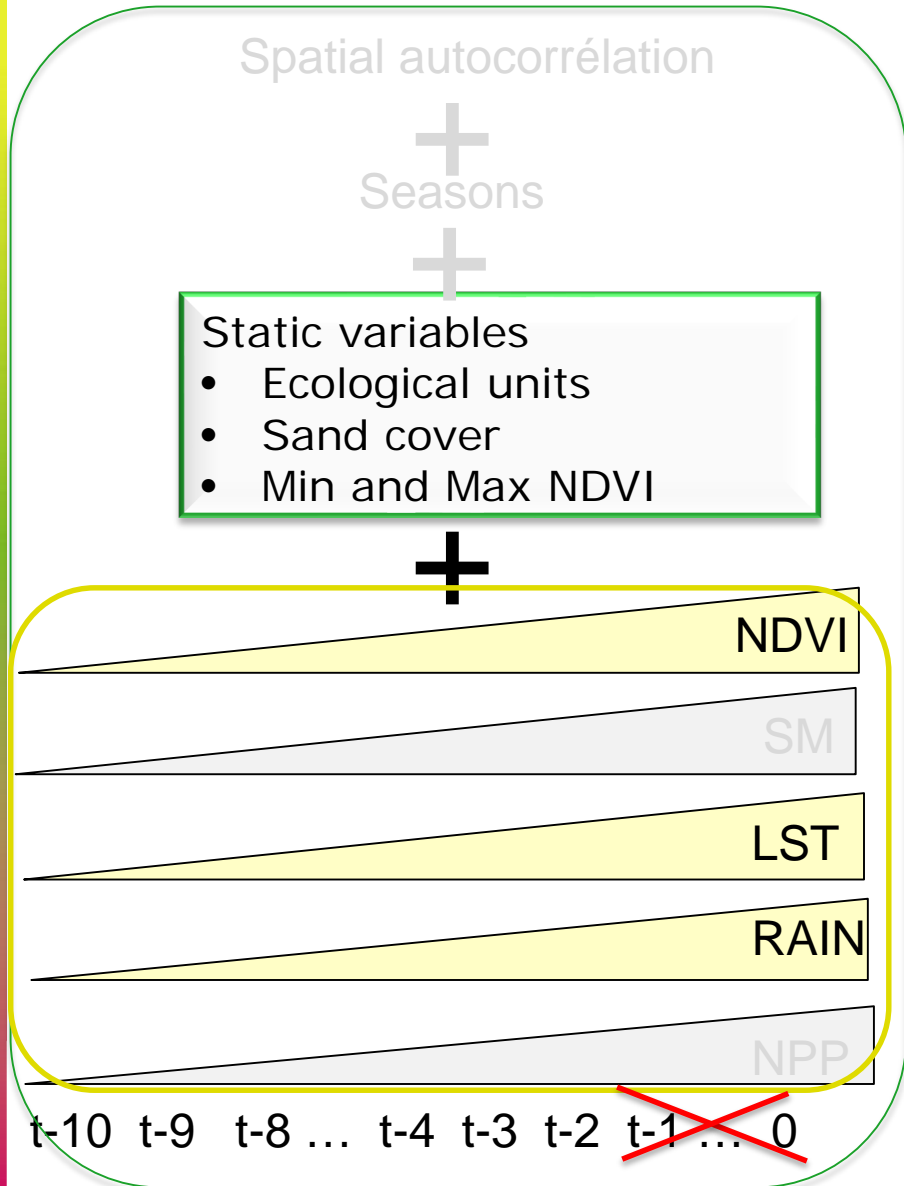
- Ecological units
- Sand cover
- Min and Max NDVI

+

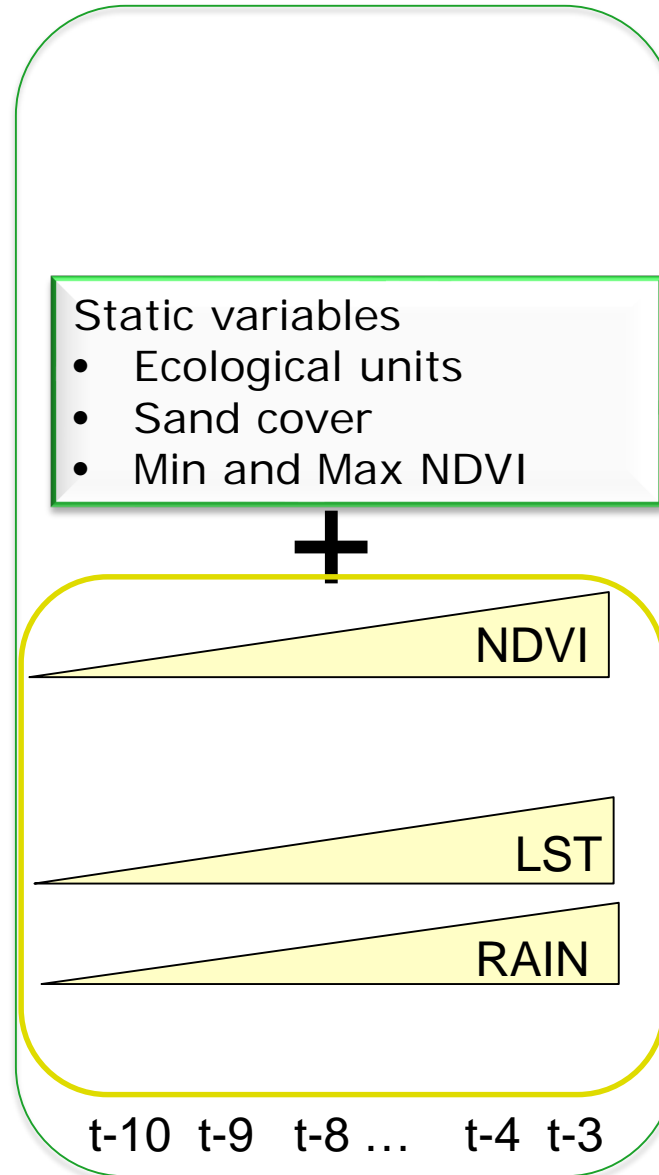


Select operational models

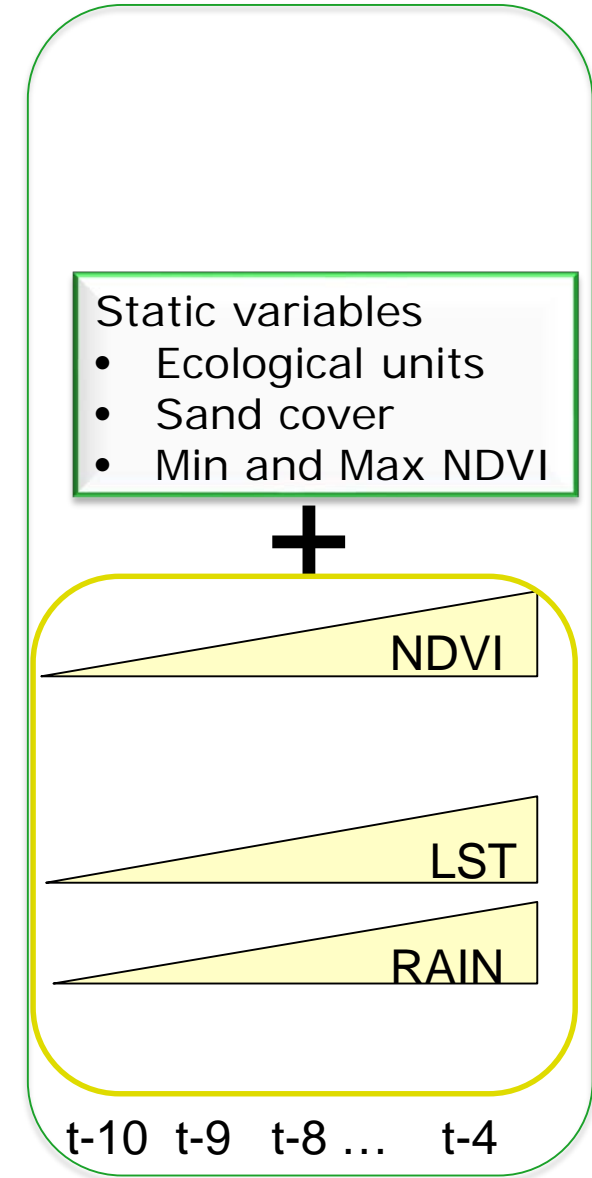
Model 1 AUC = 0.741



Model 2 AUC = 0.734

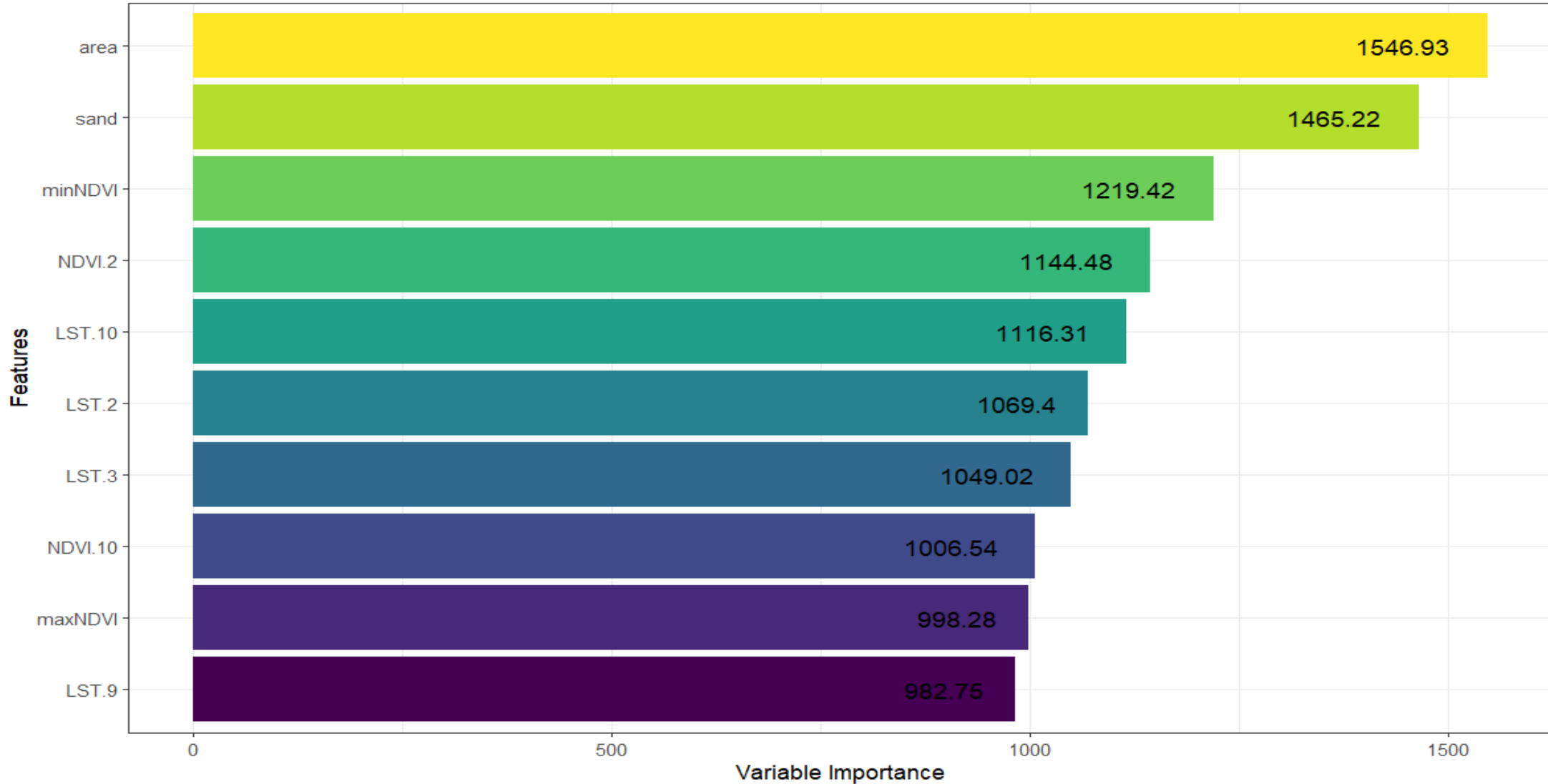


Model 3 AUC = 0.73

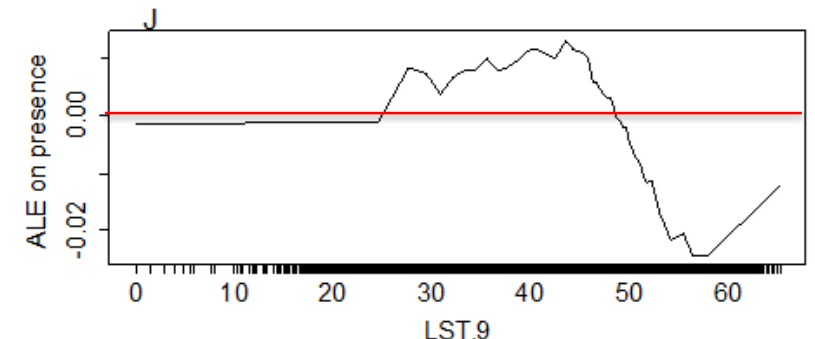
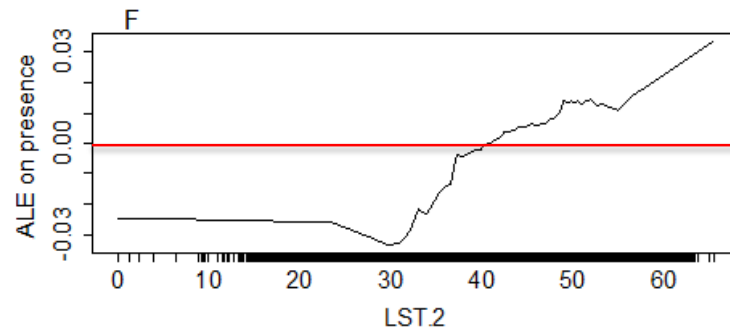
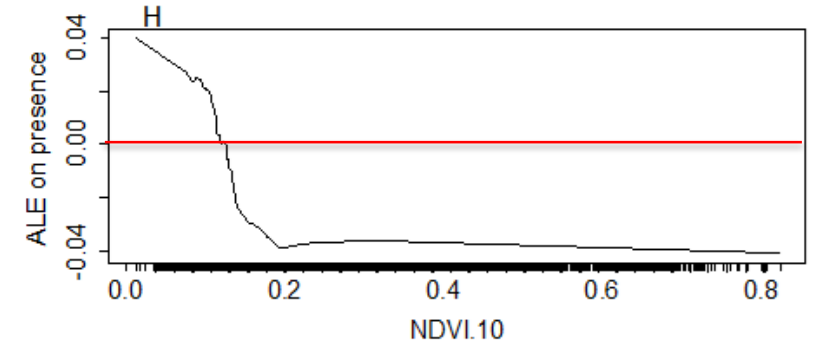
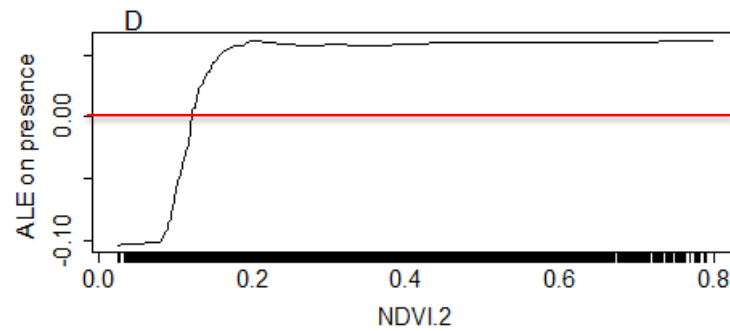
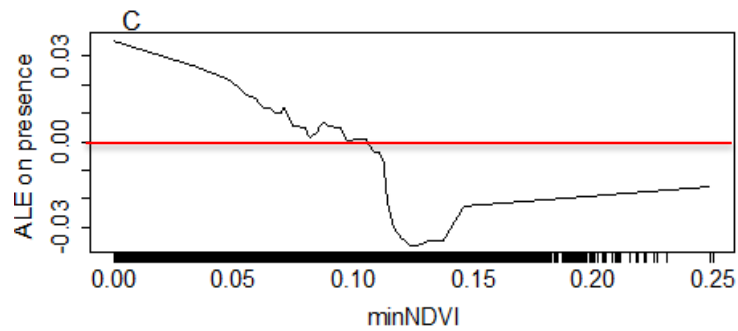
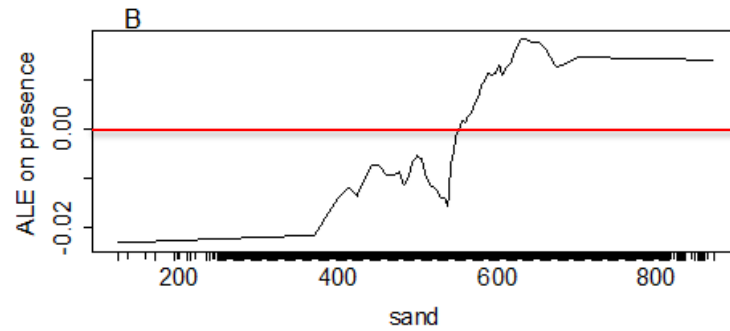
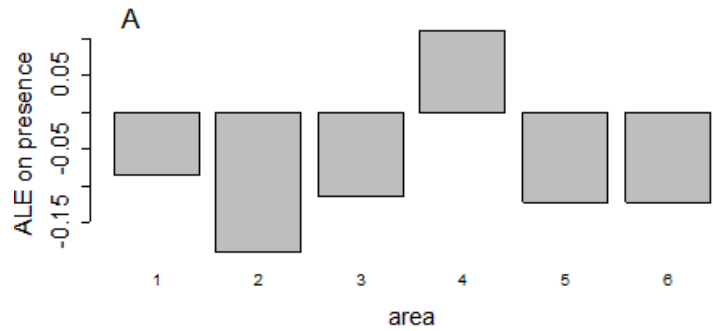


Step 3 : Evaluation

Variable importance



Response curve : accumulated local effect

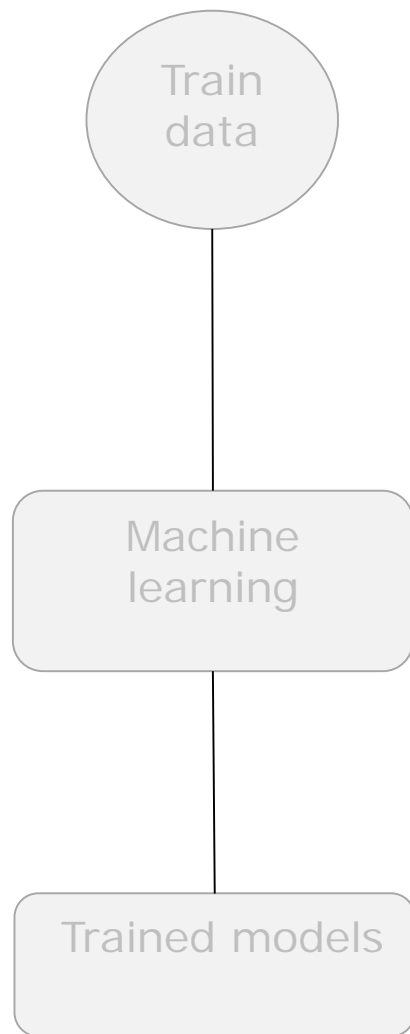


Adaptation to dry habitat
with fast and patchy vegetation
growth

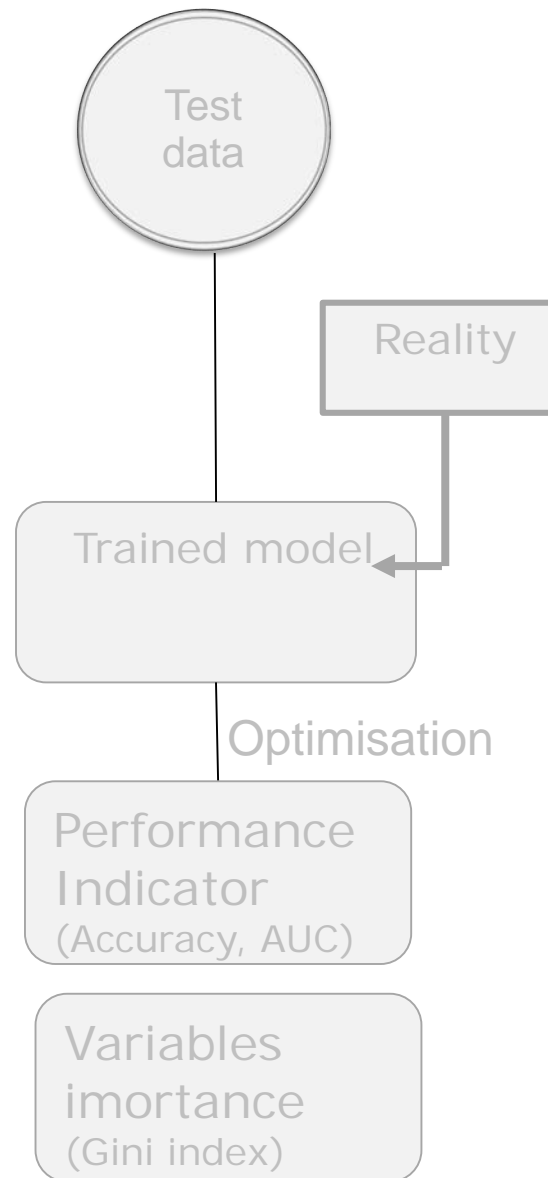
Etape 1 : traitement des données



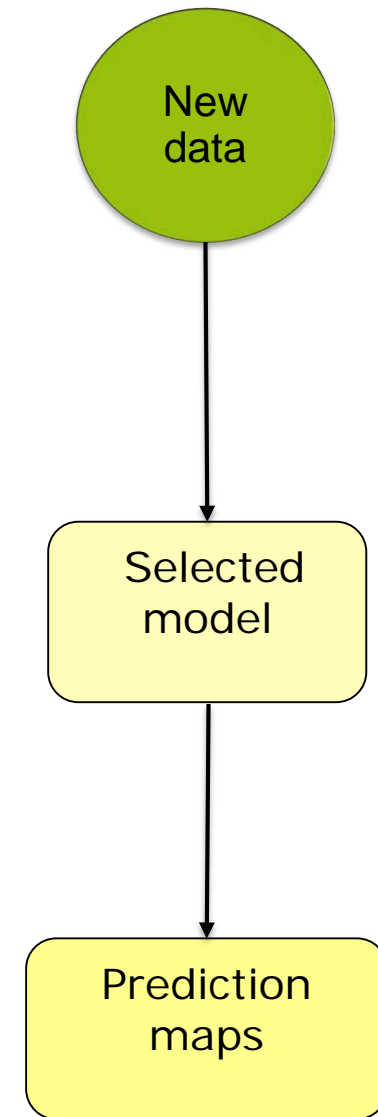
Step 2 : Training



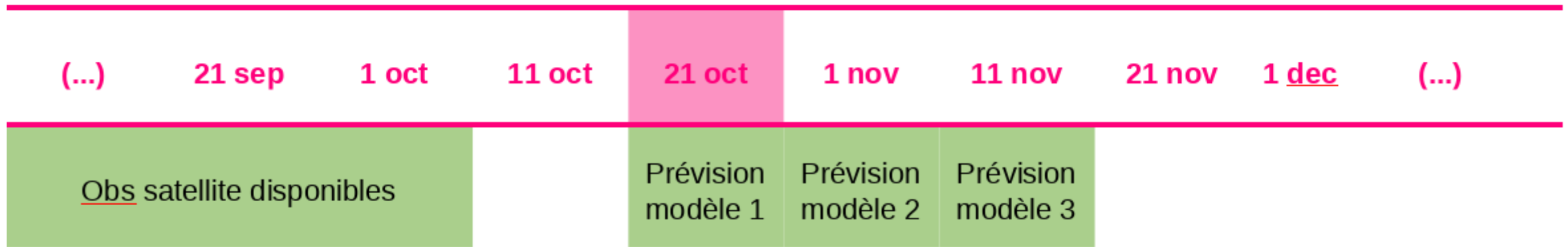
Step 3 : Evaluation



Step 4: Forecasting



Operational forecast



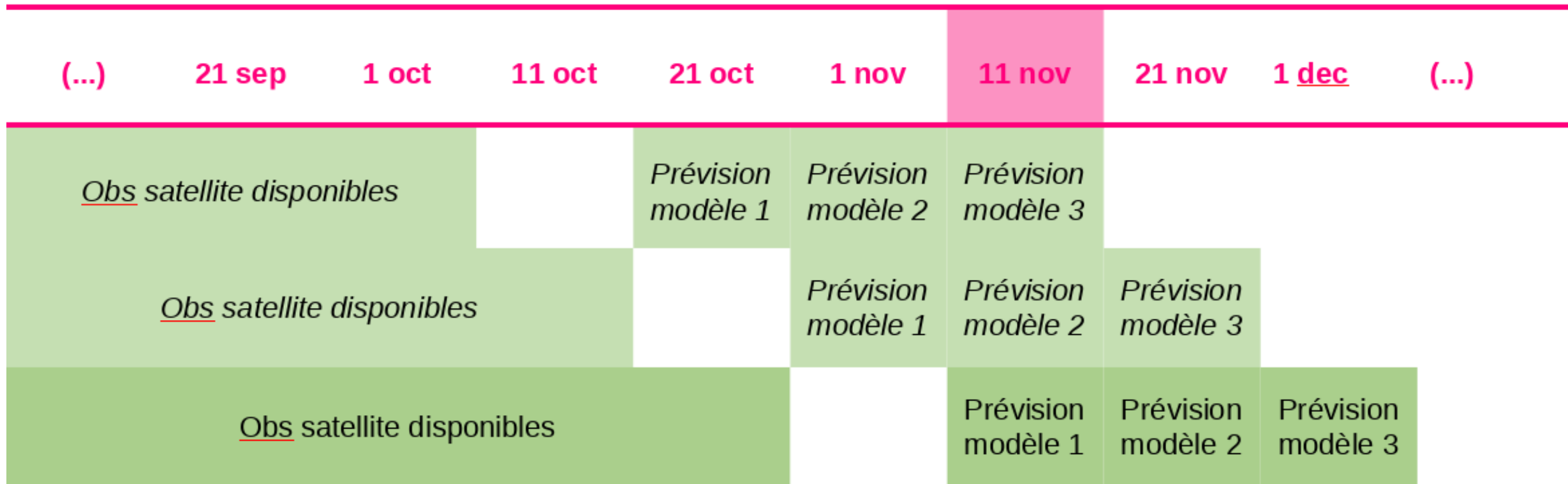
Model 1: No SAT data for the last decade
Model 2: No SAT data for the last 2 decades
Model 3: No SAT data for the last 3 decades

Operational forecast



(...)	21 sep	1 oct	11 oct	21 oct	1 nov	11 nov	21 nov	1 <u>dec</u>	(...)
<u>Obs</u> satellite disponibles				Prévision modèle 1	Prévision modèle 2	Prévision modèle 3			
<u>Obs</u> satellite disponibles					Prévision modèle 1	Prévision modèle 2	Prévision modèle 3		

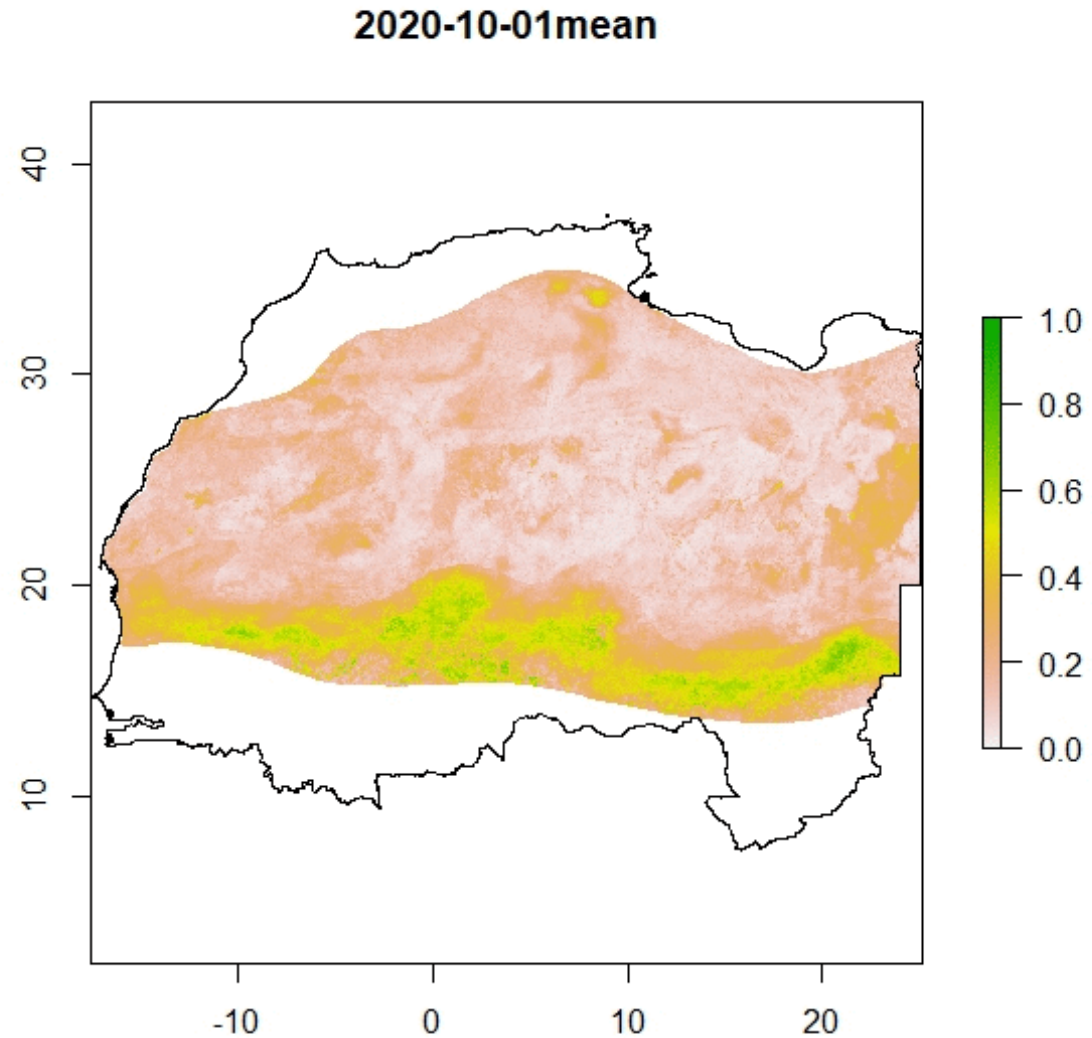
Operational forecast



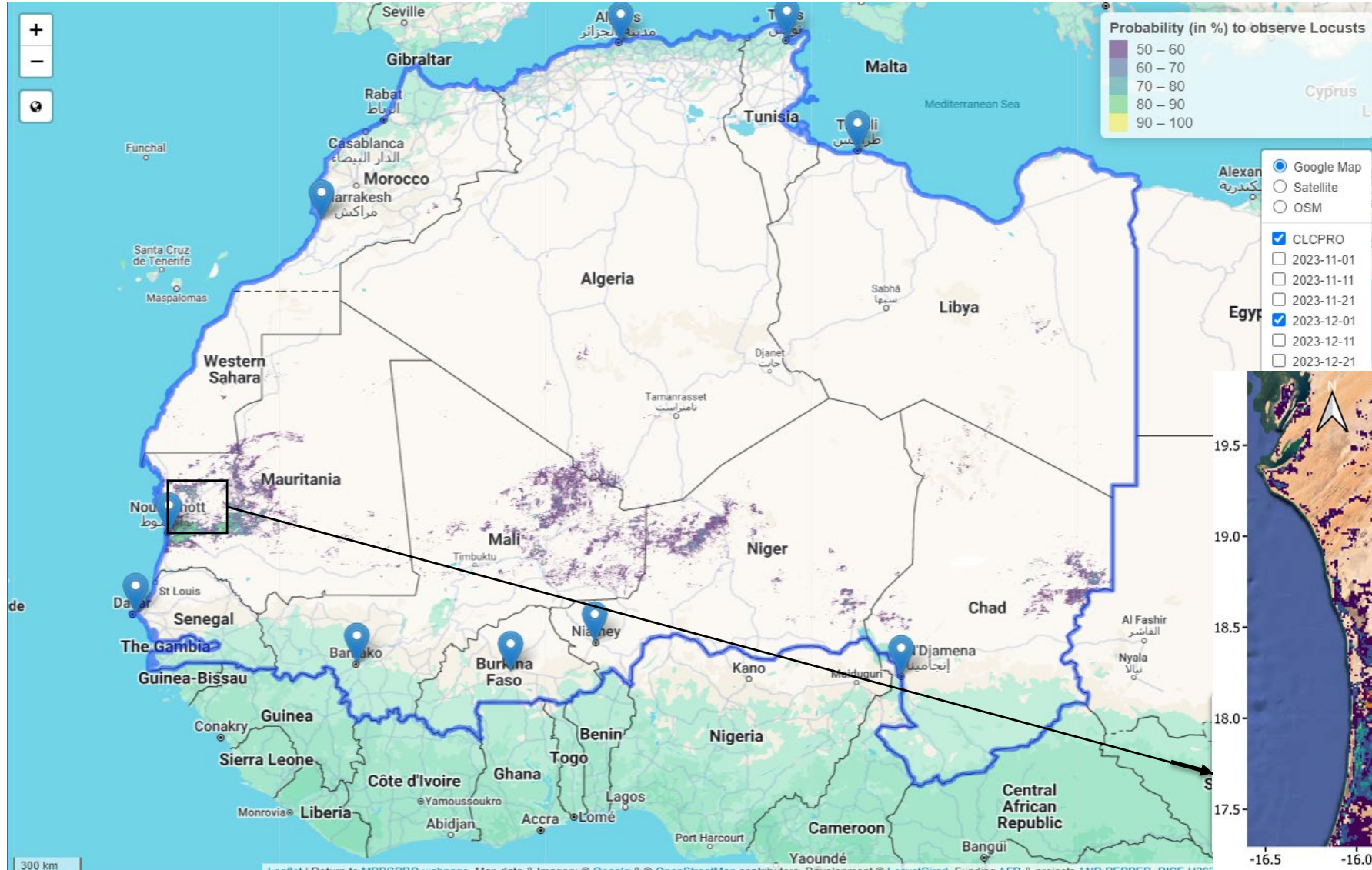
Retrospective validation

Where are Desert locust most likely to be found ?

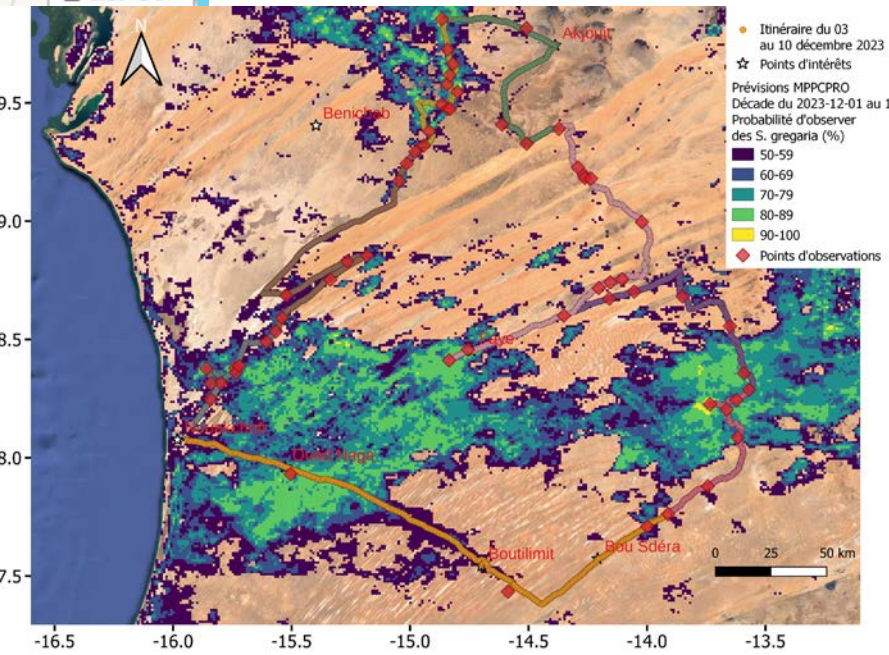
- Mostly present in the saharo-sahalian zone
- Less than 5 % of false negative



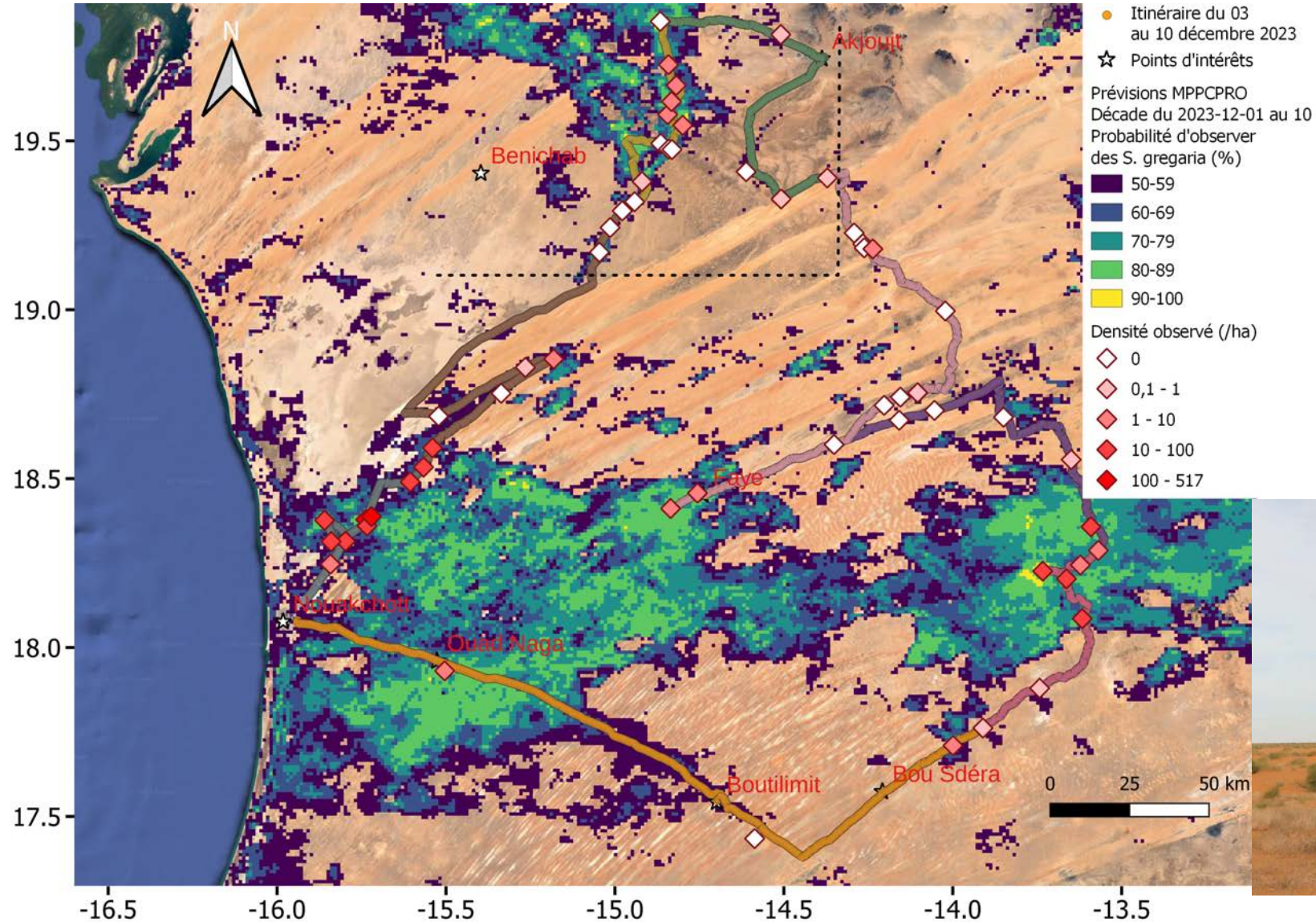
Forecasting map for field prospection



- 1500km drive
- Locust density with 3 prospectors
- 58 sites

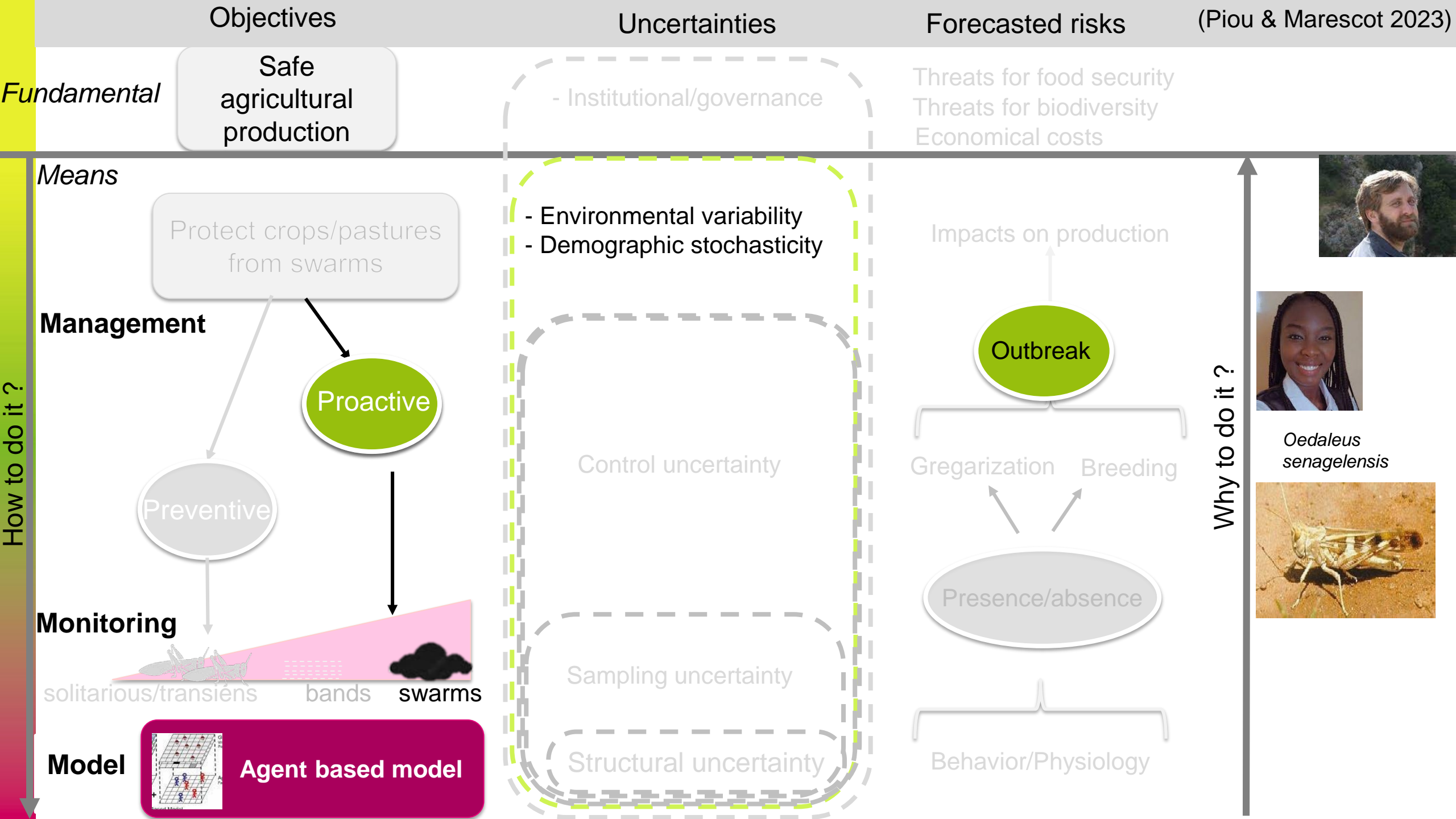


Field validation



- 81 % of correct predictions
- Correlation Classes probability of presence - Density (Kendall tau= 0.64)





Fundamental

Safe agricultural production

- Institutional/governance

Threats for food security
Threats for biodiversity
Economical costs

Means

Protect crops/pastures from swarms

- Environmental variability
- Demographic stochasticity

Impacts on production

Management

Preventive

Proactive

Outbreak

How to do it?

Monitoring

solitarious/transiens bands swarms

Control uncertainty

Gregarization Breeding

Why to do it?

Model

Agent based model

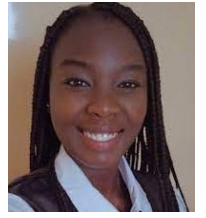
Sampling uncertainty

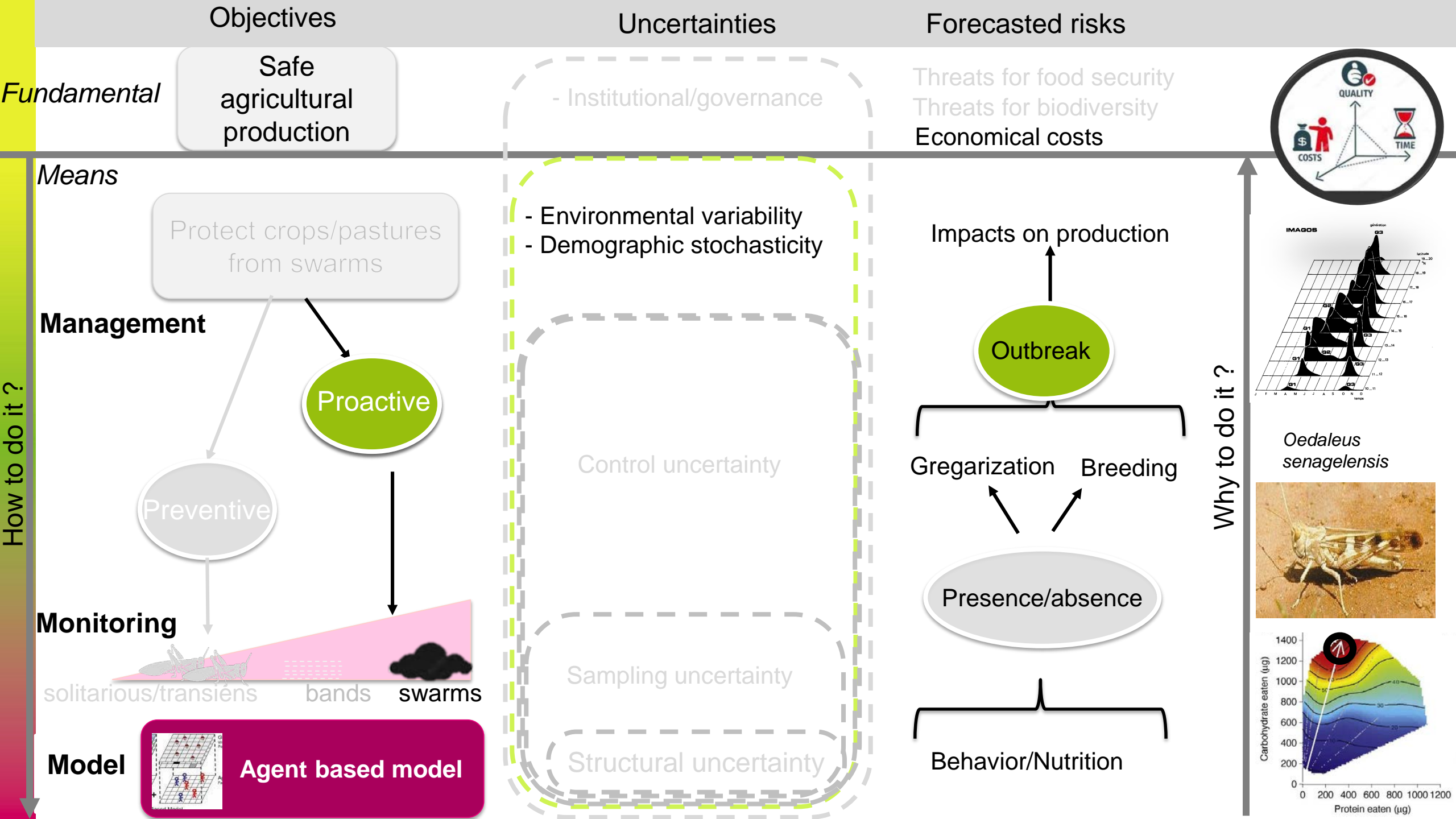
Presence/absence

Oedaleus senagelensis

Structural uncertainty

Behavior/Physiology





Objectives

Uncertainties

Forecasted risks

Fundamental

Safe agricultural production

- Institutional/governance

Threats for food security
Threats for biodiversity
Economical costs



Means

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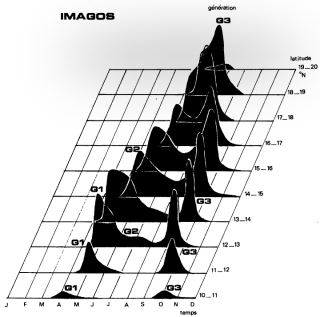
Management

Preventive

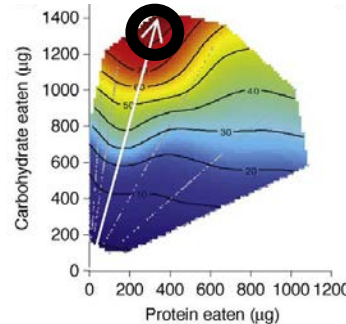
Proactive

Impacts on production

Outbreak



Oedaleus senagelensis



Monitoring

solitaries/transients bands swarms

Gregarization Breeding

Presence/absence

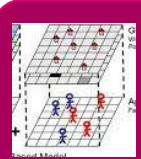
Behavior/Nutrition

Control uncertainty

Sampling uncertainty

Structural uncertainty

Model

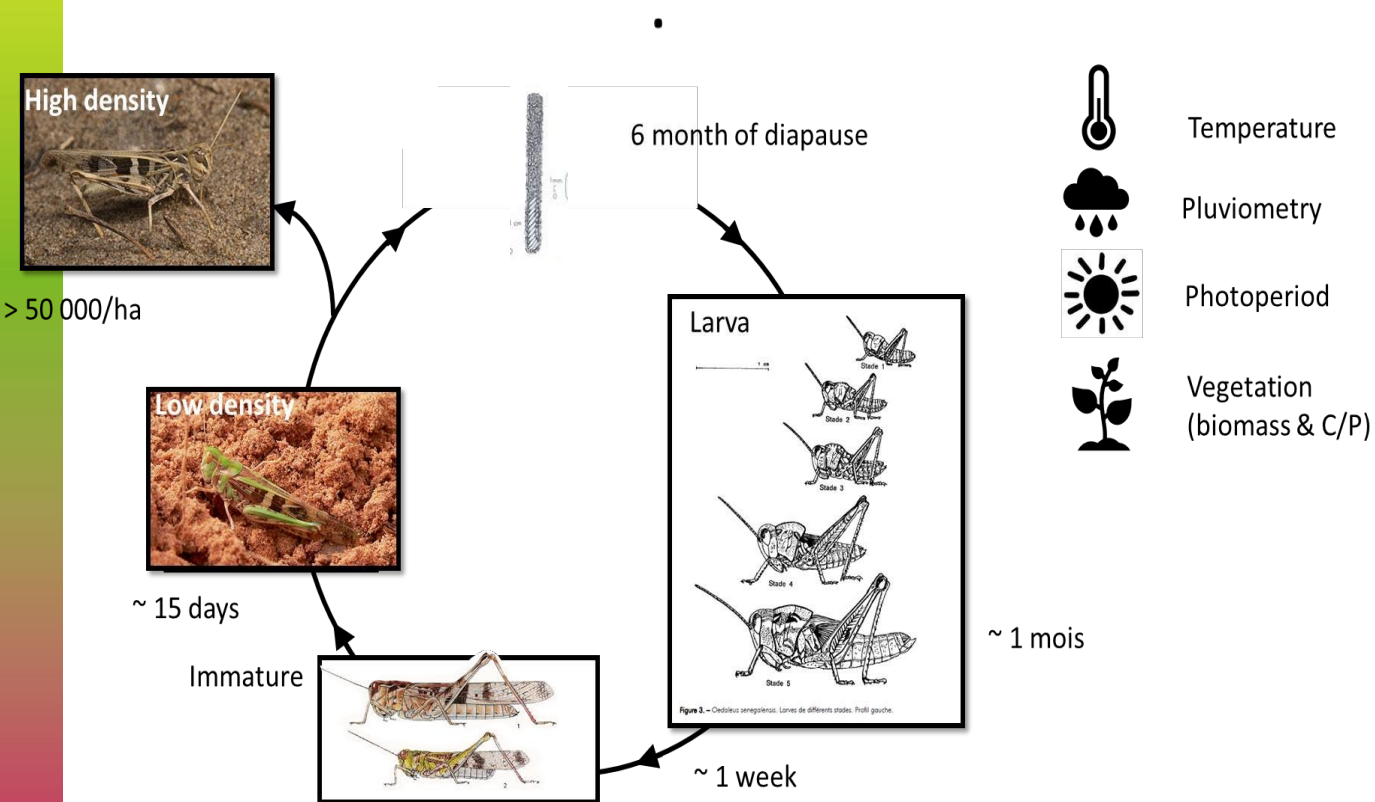


Agent based model

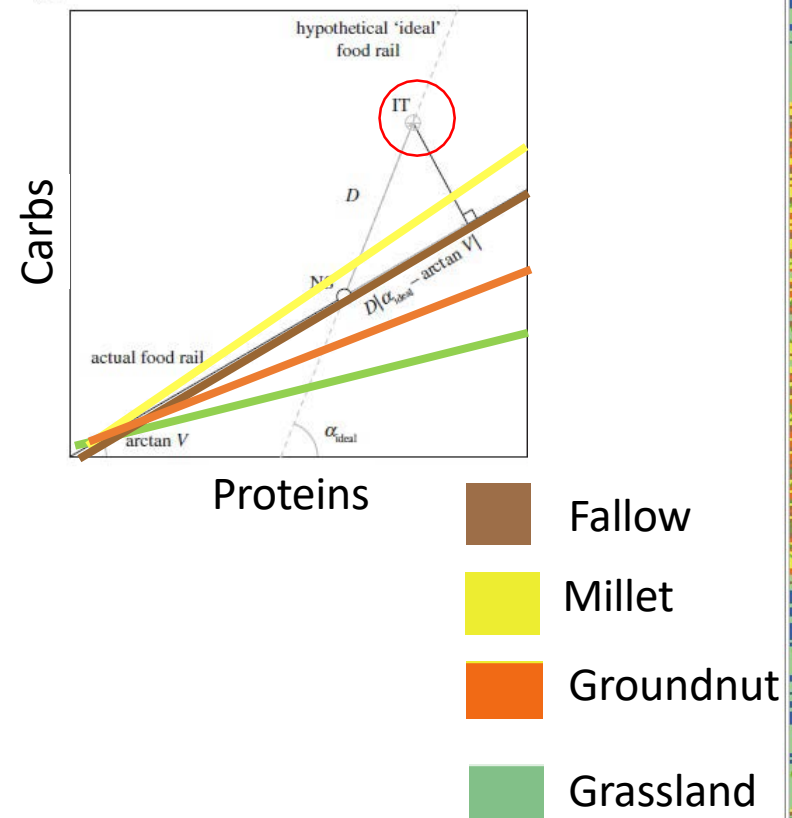
Why to do it ?

Agent-based model linking nutritional ecology and population dynamics

Can soil amendment affect population dynamics and behaviour and reduce crop damage ?

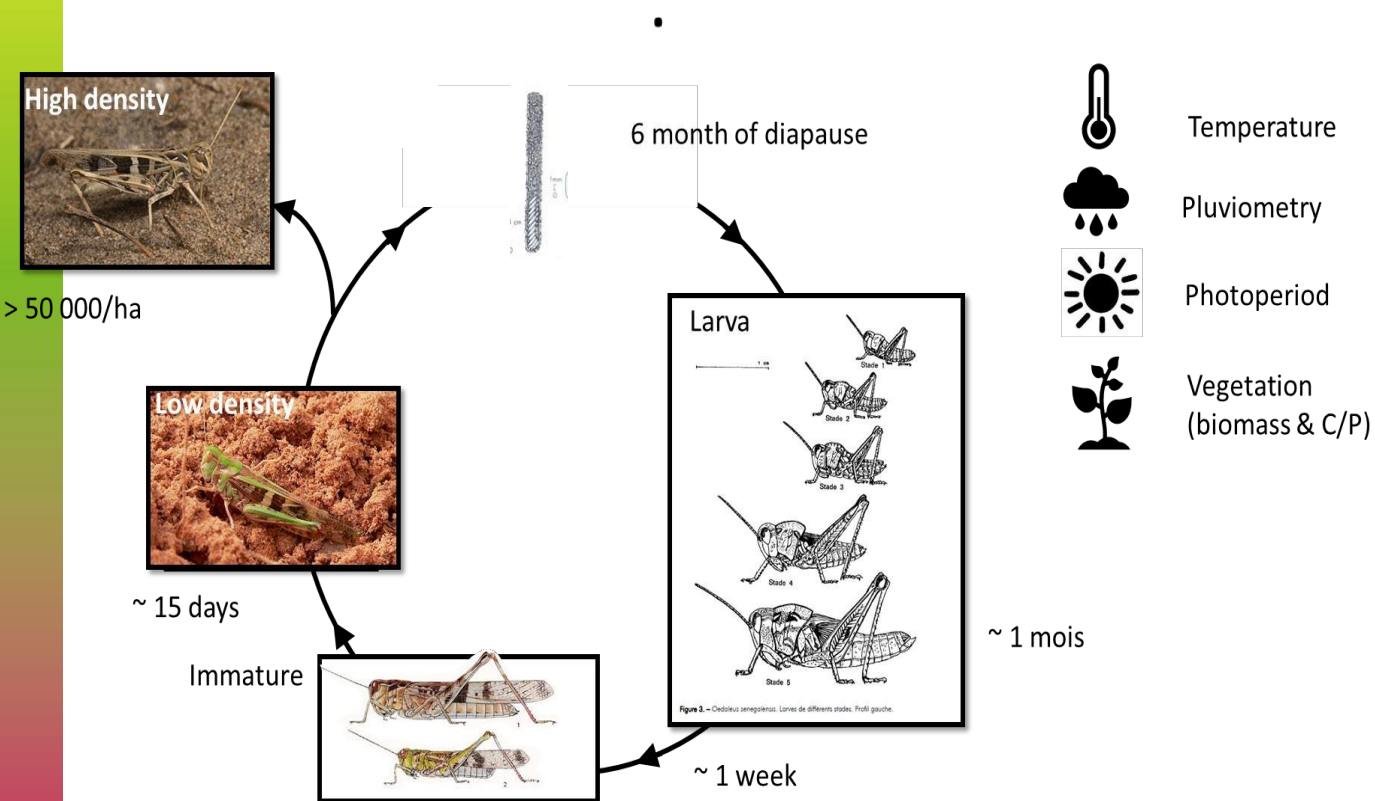


Nutritional geometric framework

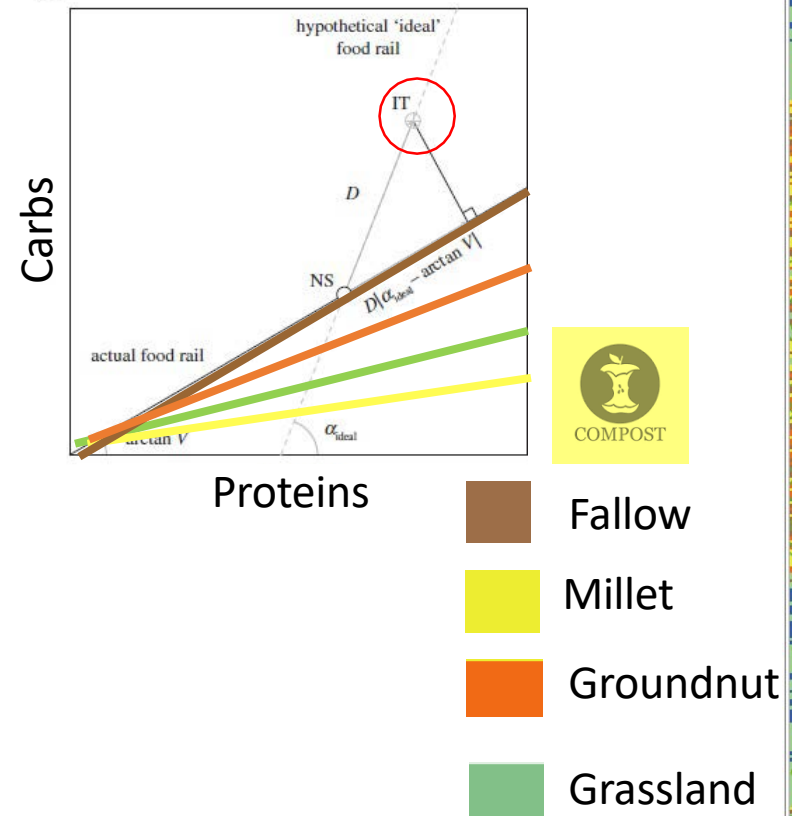


Agent-based model linking nutritional ecology and population dynamics

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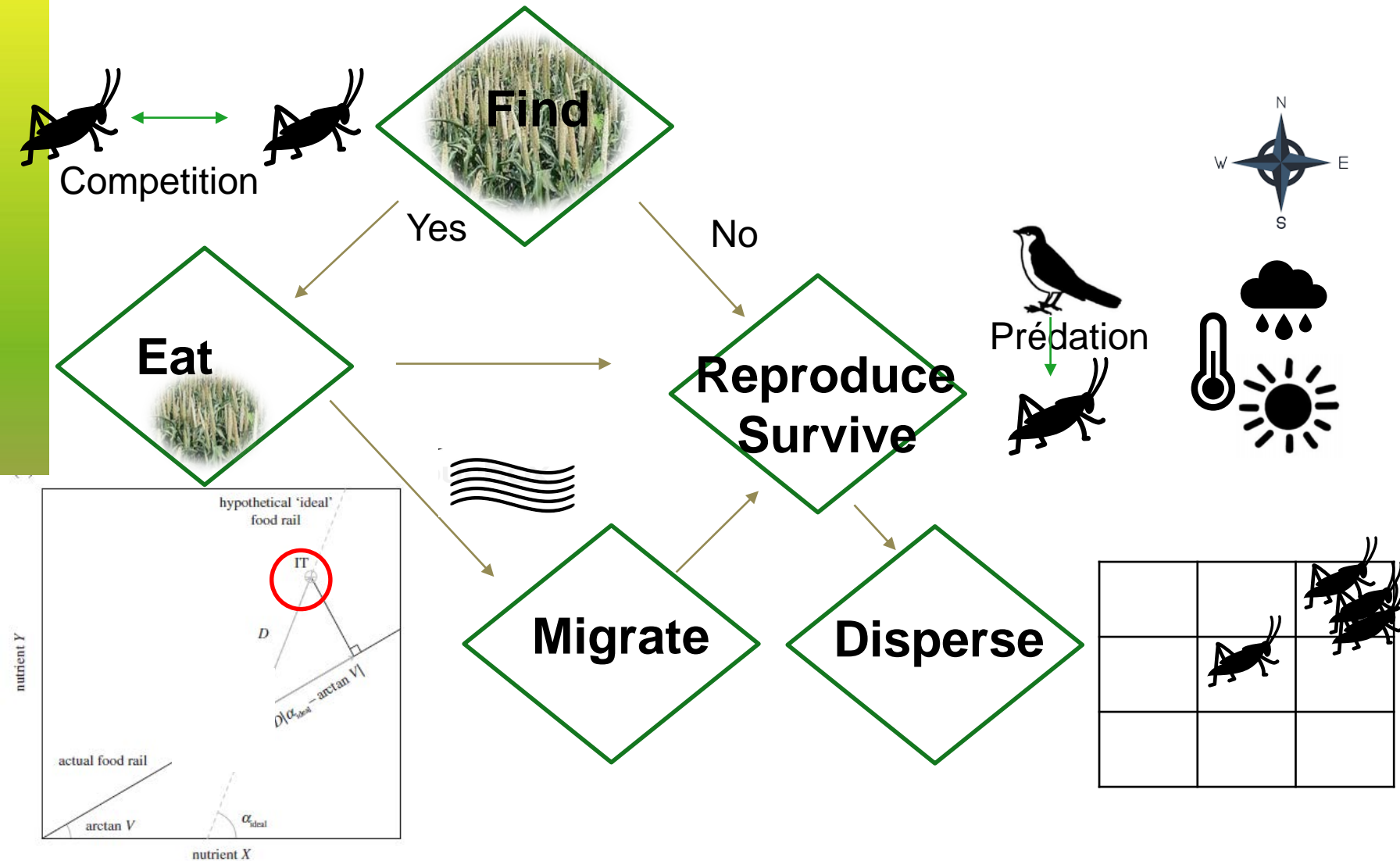


Nutritional geometric framework



(Launois 1979, Simpson et al. 2010, Legall et al. 2019, 2021)

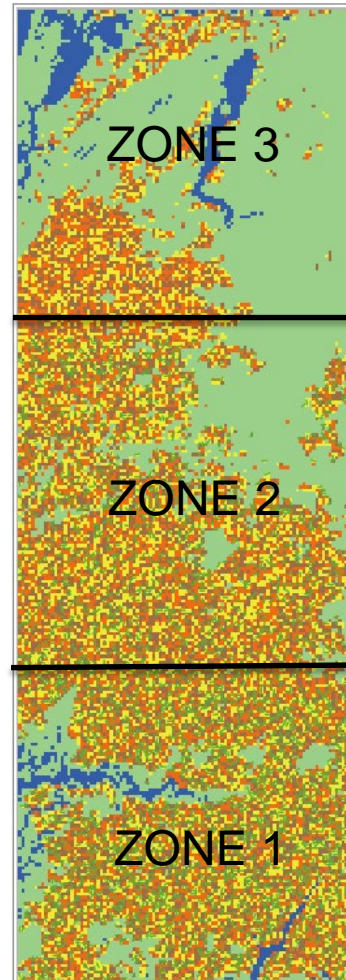
Model structure



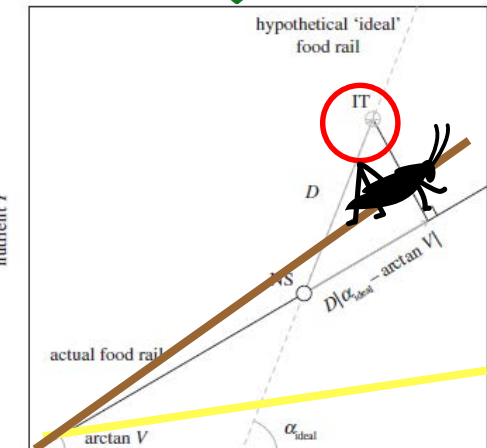
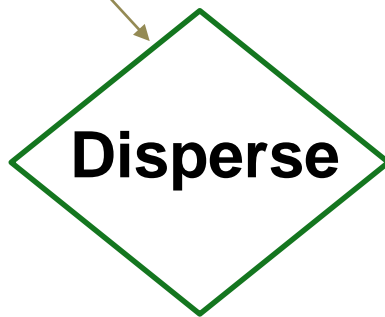
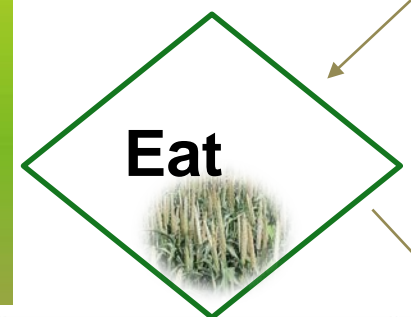
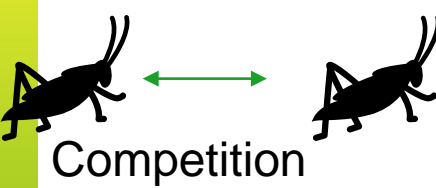
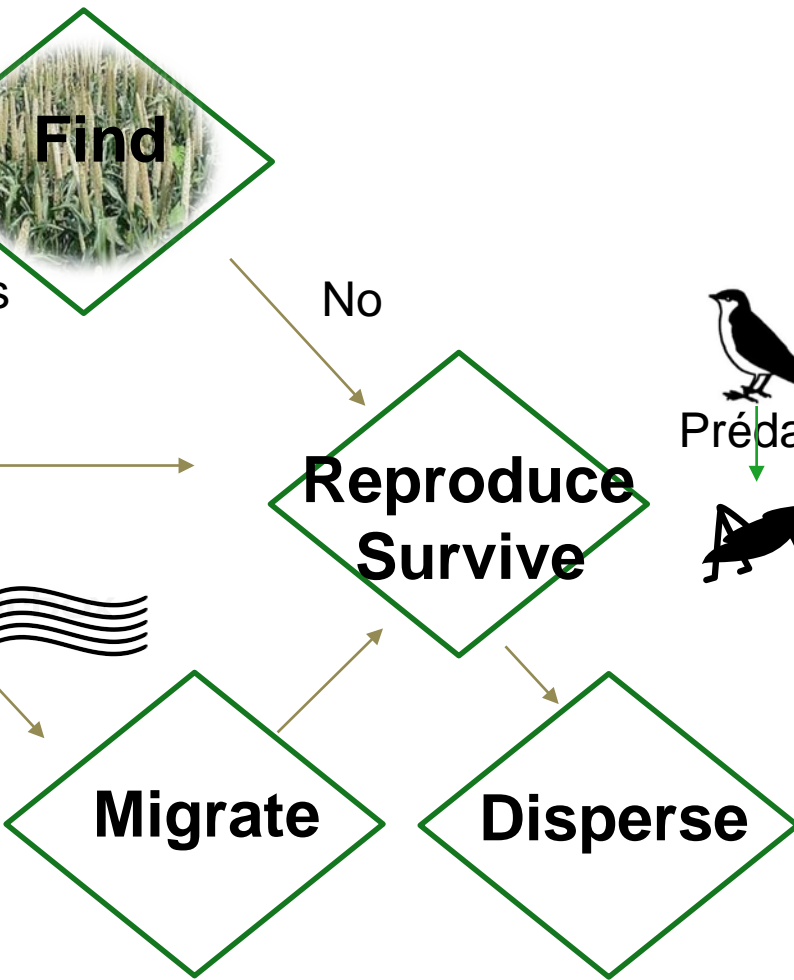
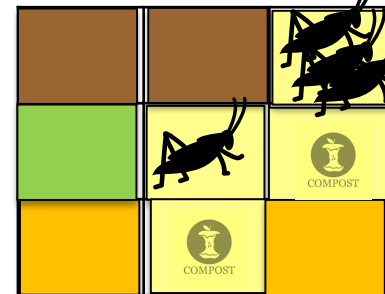
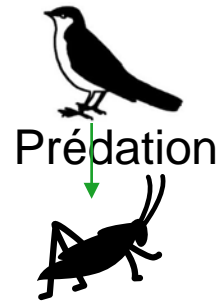
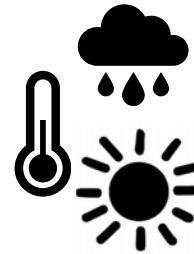
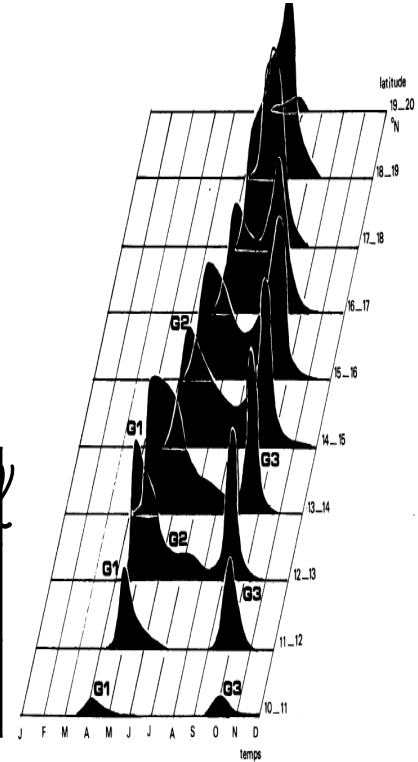
Model structure

Above-ground biomass and yield for millet (Lavarenne J., model SARRAH)

100 km



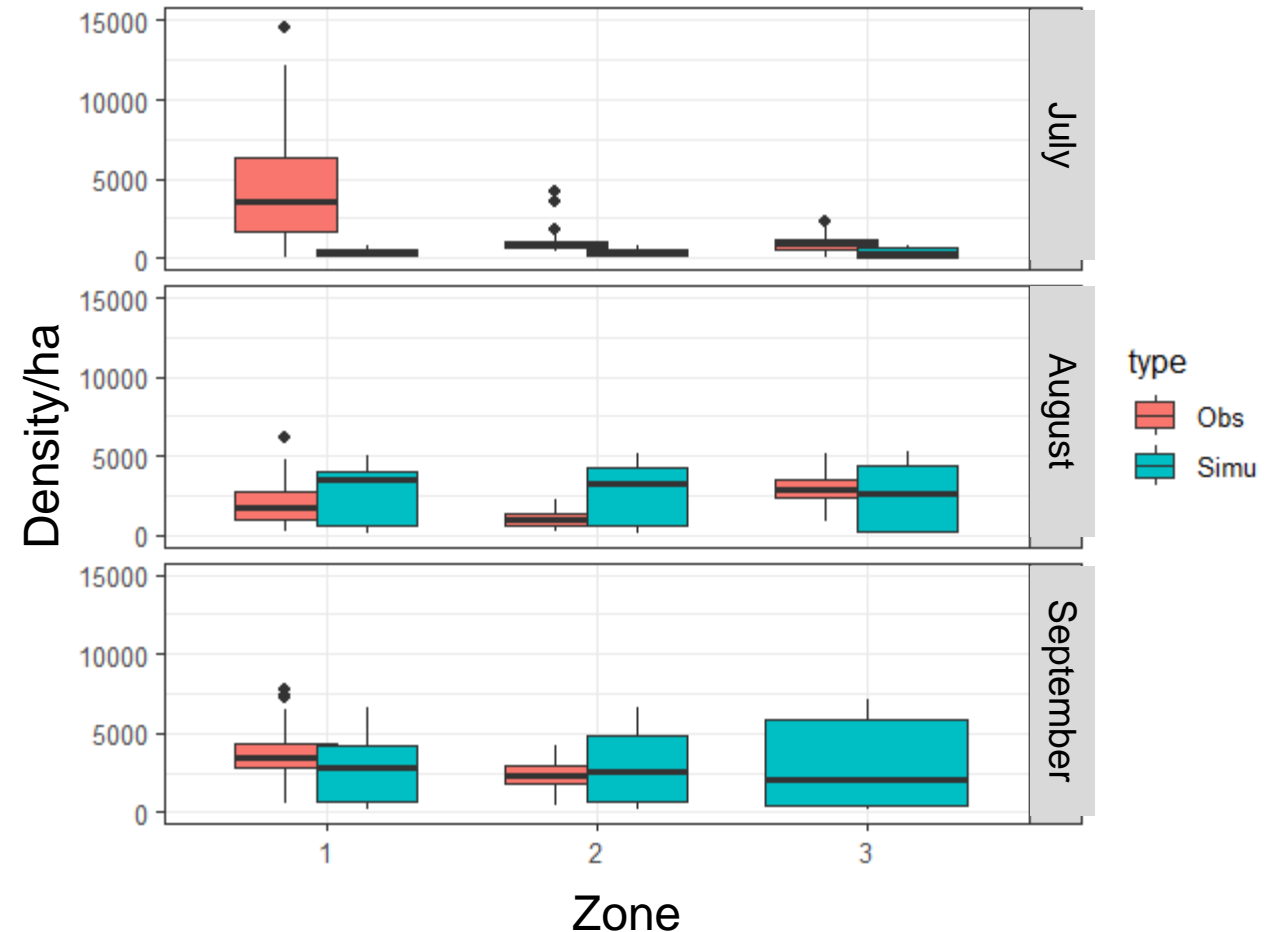
Launois et al. 1979



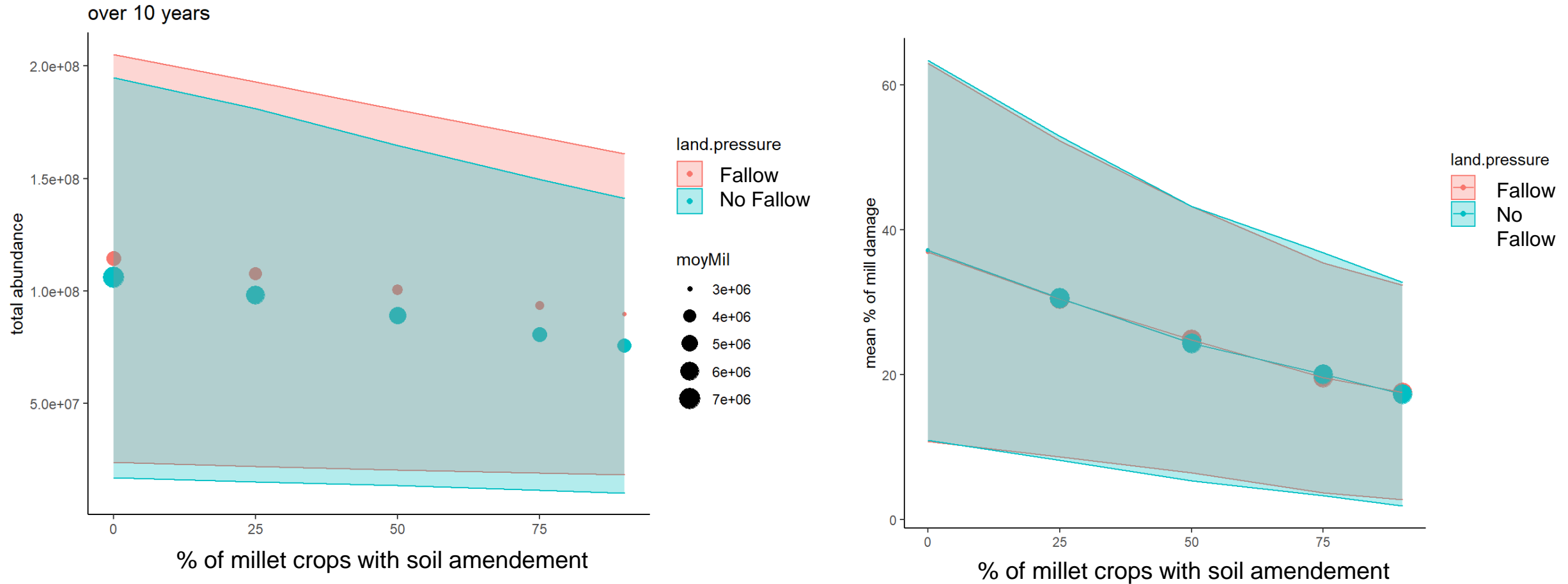
Model Calibration using pattern-oriented modelling

- Sensitive to initial population size
- Late migration date
- Small migration capacity
- Narrow orientation

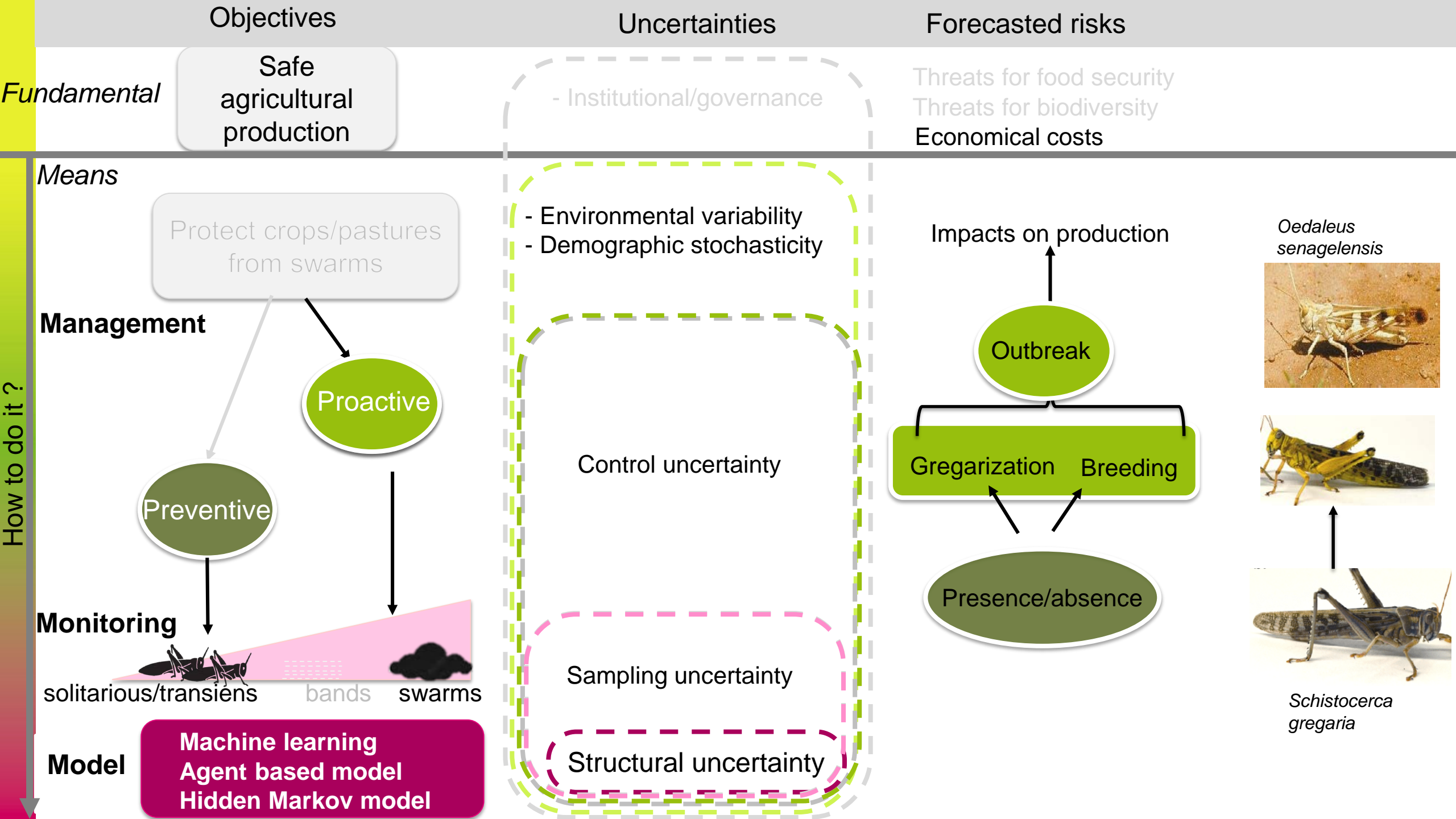
- Robust to cohesion factor and competition



Preliminary results



- Soil amendements : no thresholds to prevent outbreaks
- But it can reduce impact on crops by changing grasshopper diet



Conclusion : which model for which risk level ?

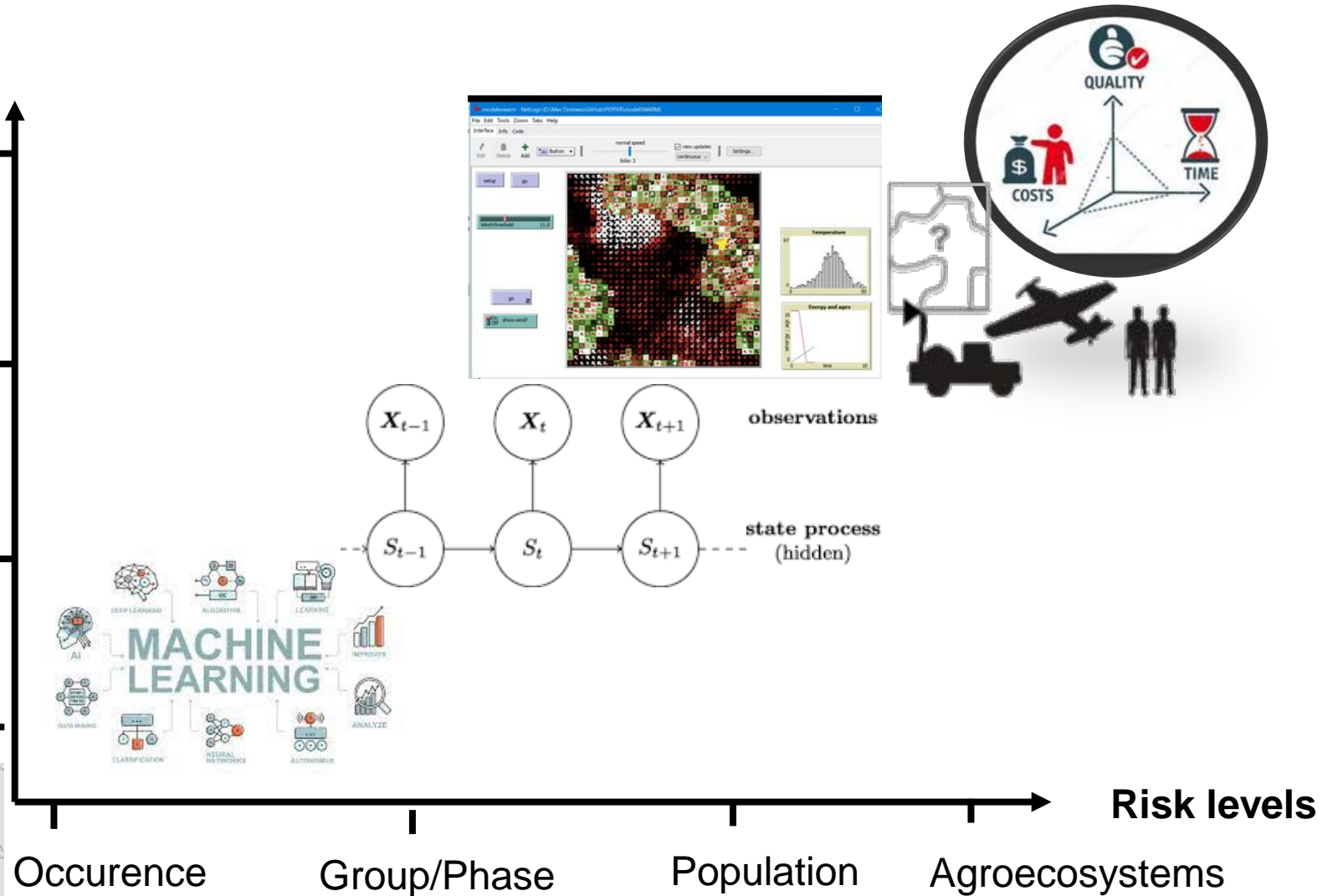
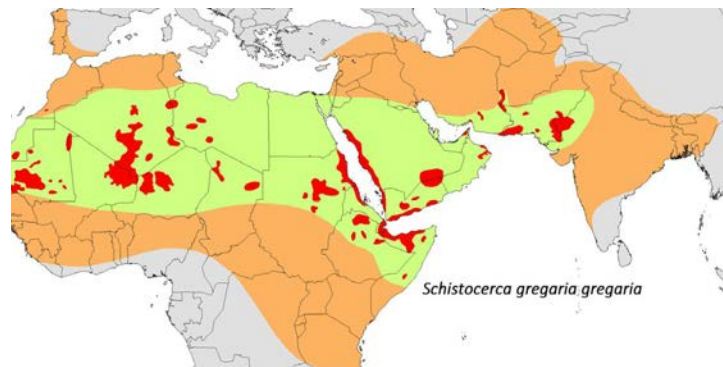
Acridian situation

Invasion

Resurgence: proaction

Remission : prevention

Recession : surveillance



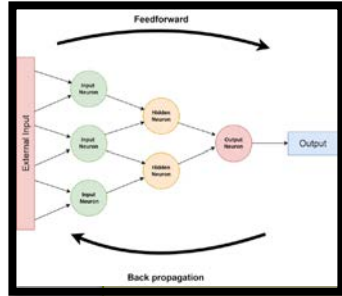
Conclusions and perspectives

- Remote sensing and machine learning are powerful tools to forecast distribution (Piou et al. 2013, 2019, Sun et al. 2021)
- Provide maps with probability of detecting the species
- Used to orientate team in the field and strengthen preventive management
- Less powerful to forecast risk of gregarization (Véran et al. 2015, Lawton et al. 2021)
- Accounting for imperfect and heterogenous detection using occupancy models
- More mechanistic approaches are necessary to predict outbreak (Gay et al. 2021)
- Need for pattern oriented modeling to become operational (Grimm et Railsback 2005)
- Ecology and human forces must stay at the heart of locust management

Thanks for your attention

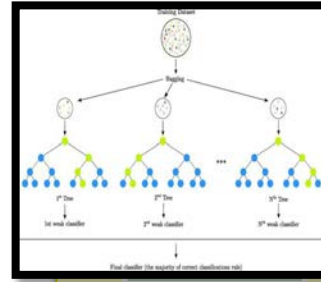


Pros and Cons between algorithms



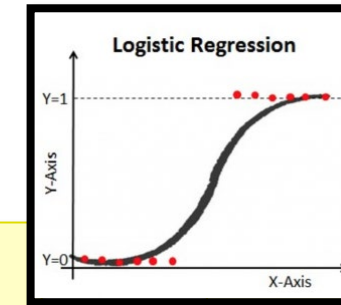
NN

- No hypothesis
- Large data sets
- Any data type
- Adapt to new data/ update prediction
- Long to train
- Difficult interpret
- Overfitting



RF

- No hypothesis
- Large data sets
- Unbalanced data
- Many variables correlated, lagged
- Few preprocessing
- Parallelized
- Robust to outliers
- Overfitting
- Interpretation



LR

- Fast to train
- Deal with NA
- Response Curve
- Effect size
- Linearity