

# PRactical Implementation of Coexistence in Europe

## Estimating Adventitious Presence and its Uncertainty

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# EU coexistence framework

- In principle, farmers should be able to cultivate the types of agricultural crops they choose - be it GM crops, conventional or organic crops;
- The adventitious presence of GMOs above 0.9% triggers labelling « as containing GMOs »;
- The presence of traces of GMOs in particular food crops - **even at a level below 0,9%** - may cause economic damages to operators;
- Many of the influencing factors are specific to national, regional and local conditions;

→ Principles of subsidiarity and of proportionality

# Various sources of adventitious presence

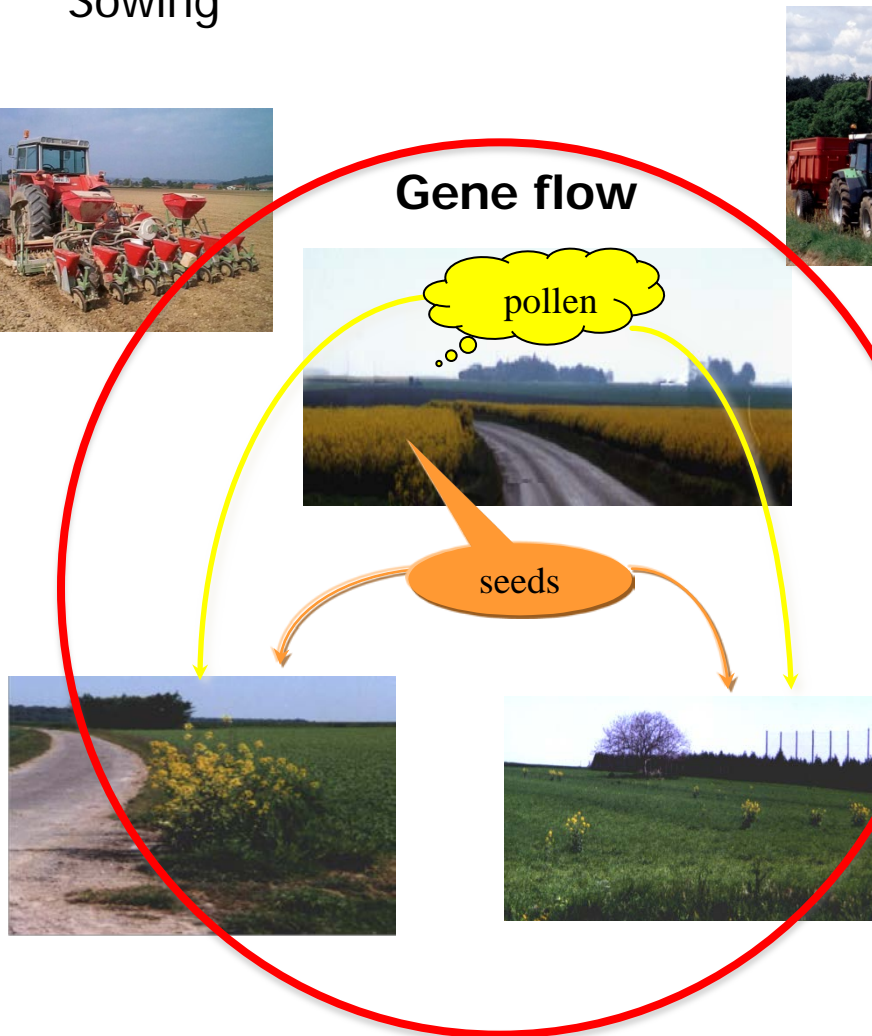
Seeds



Sowing



Gene flow

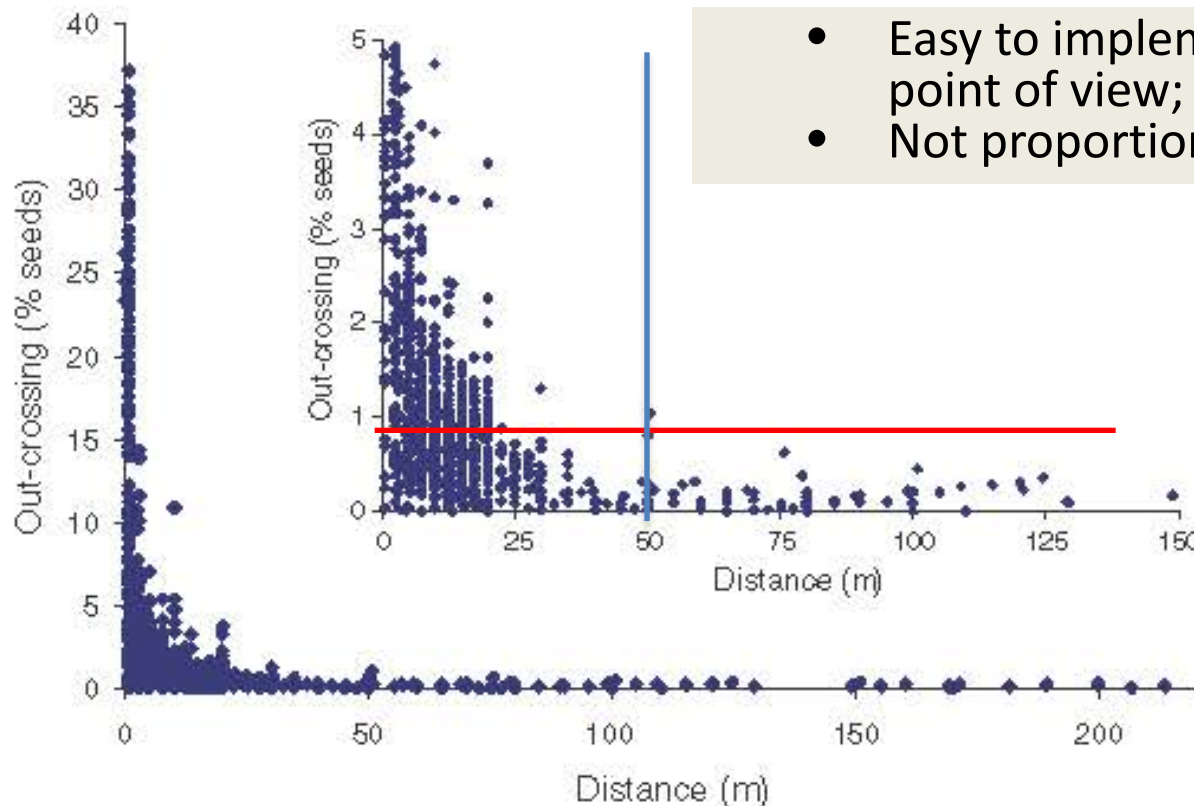


## Factors affecting gene flow

- Landscape patterns
  - Distance between donor and recipient fields
  - Field sizes and shapes
  - Topology
- Wind direction and speed
- Flowering time-lag
  - Sowing dates, earliness of varieties and climate
- Crop management practices
- Genetics
  - Pollen, persistence

# Uniform coexistence measures are not uniform

A statistical approach which takes advantage of gene flow studies and which determines adequate isolation distances



- Easy to implement from a regulatory point of view;
- Not proportional to the actual risk

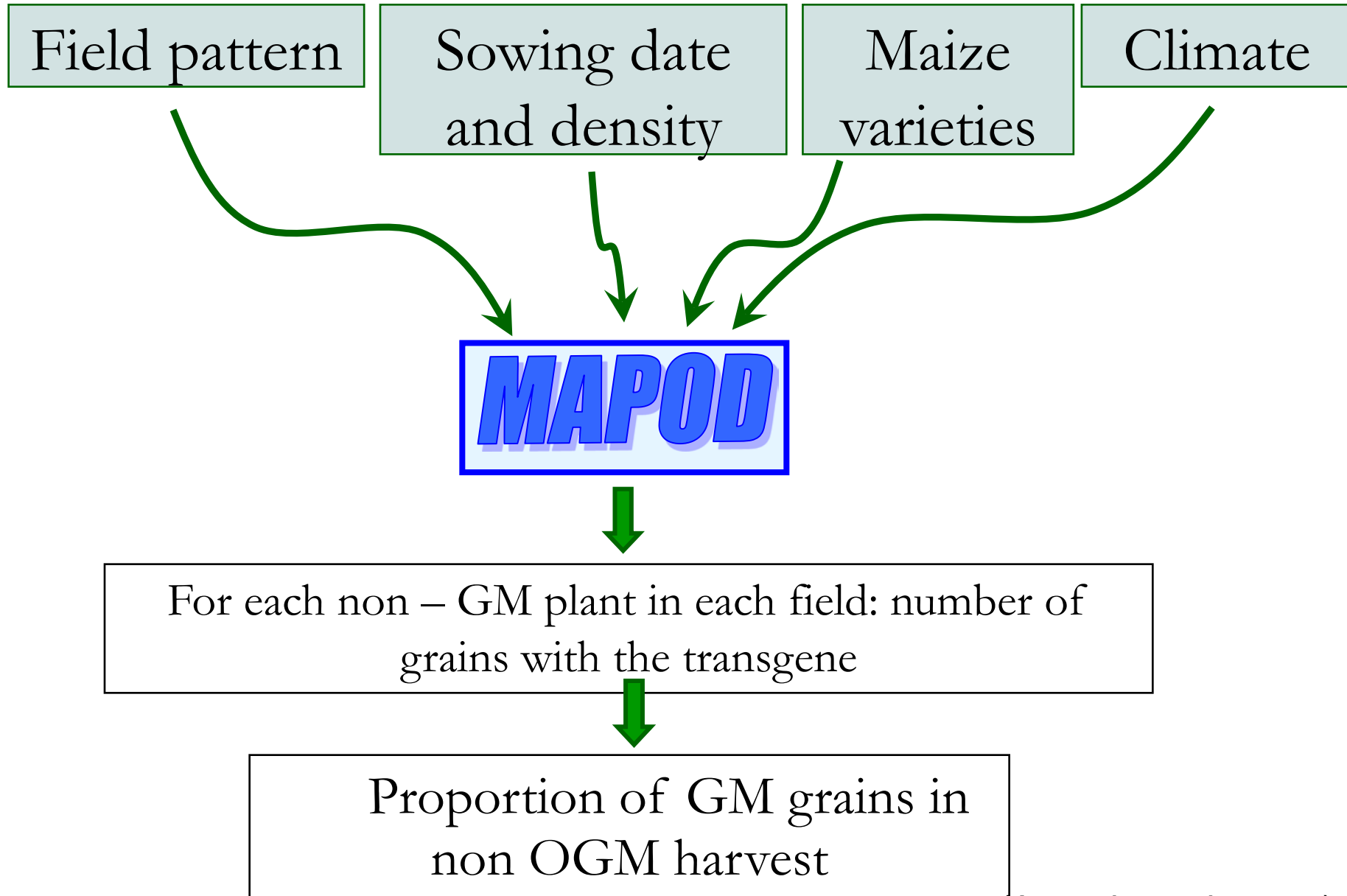
Meta-analysis of maize gene flow datasets  
(Riesgo et al., 2010)



# How to set up coexistence measures

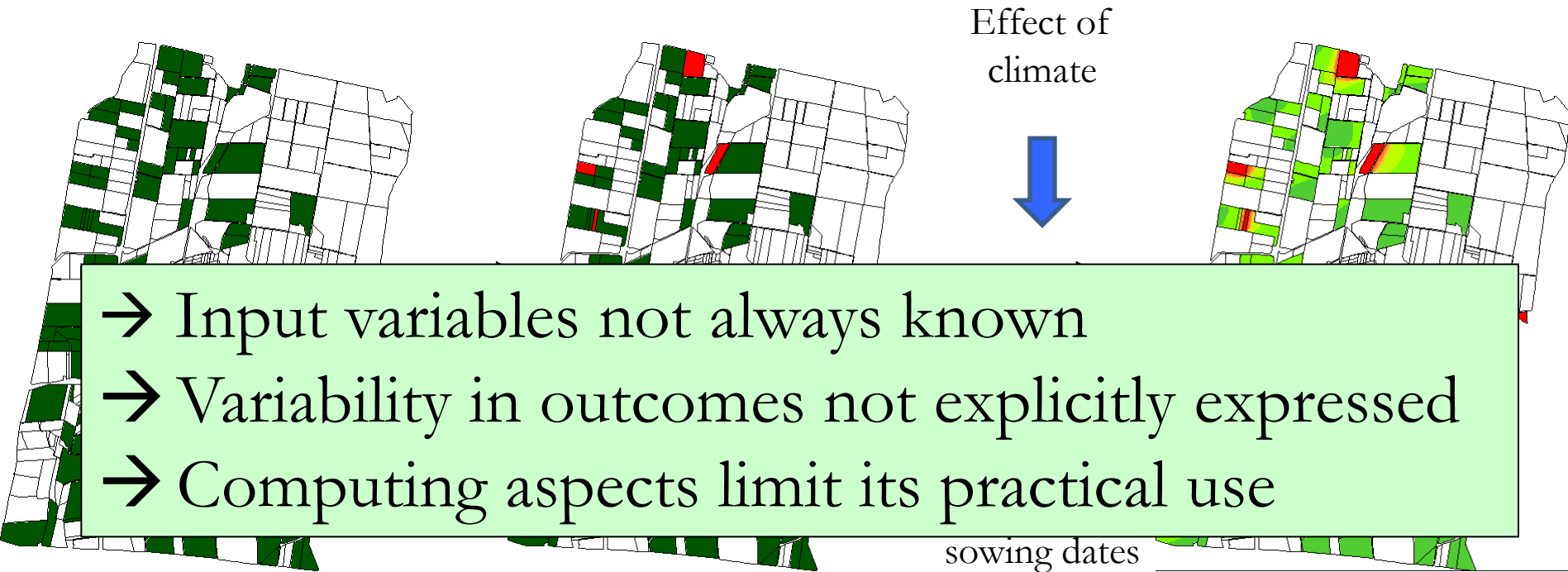
1. A statistical approach which takes advantage of gene flow studies and which determines adequate isolation distances
  - Easy to implement from a regulatory point of view;
  - Not proportional to the risk;
2. A mechanistic approach which aims at considering the effects of landscapes, climate and agricultural practices on gene flow at the landscape level
  - Adaptation to regional contexts
  - Flexible coexistence measures (may change from field to field);
  - More difficult to implement under a regulatory framework;

Adaptation to local conditions are necessary to meet the proportionality principle → gene flow models can help



(Angevin et al., 2008)

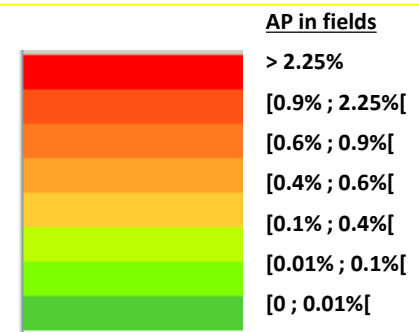
# Assessment of Agronomic Scenarios with MAPOD



Typical agricultural landscape where maize fields are represented in 'green'

Users decide where they would like to grow GM maize (in 'red')

MAPOD estimates the subsequent AP in non-GM fields



# → Three major steps forward

- Develop a dynamic and operational decision-support tool to help secure compliance with given thresholds
  - Development of a web-based prototype
- Associate confidence intervals to AP predictions by taking into consideration uncertainty/variability
  - Design of a Bayesian framework
- Develop cost-effective sampling strategies
  - Use of predictive models to optimize sampling



# Scenarios for the DST

## *Ex ante*

- GM maize fields location (if registered)
- Hypotheses on allocation of non GM fields
- Historical climatic data

## *Ex post 1*

- Actual spatial location of GM and non GM fields
- Sowing dates and estimation of flowering
- More precise climate

## *Ex post 2*

- Observed flowering dates;
- Climatic data for the whole growing season

Sowing

Flowering

Harvesting

Time

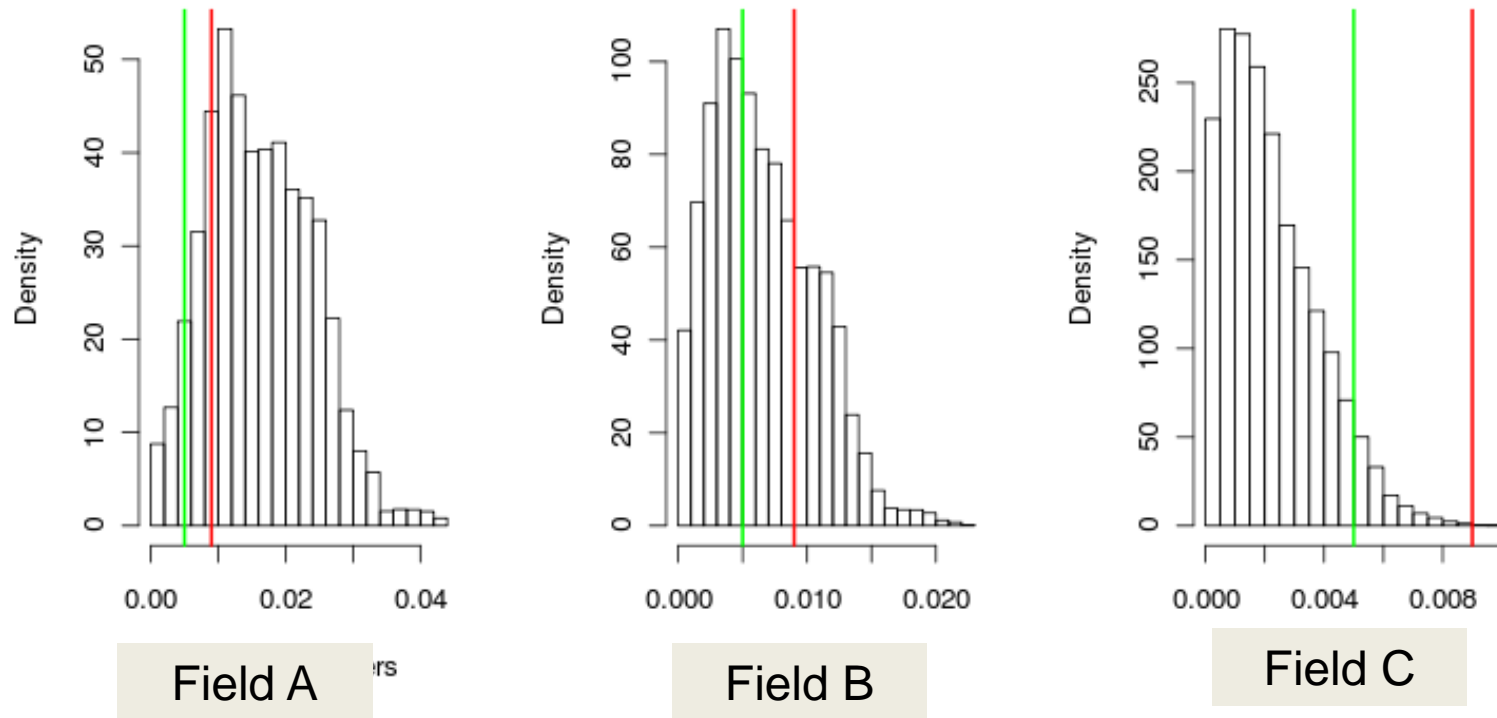
Discussion between neighbouring farmers, could result in changes in GM allocation

Estimation of AP risk, can trigger the need for additional observations and/or field sampling

Harvest allotment to meet specific thresholds

→ Can useful information on AP be delivered all along the growing season and whatever is the level of information?

# Distribution of GM AP in every non-GM field of the landscape rather than one single value



→ End-users have to make decisions knowing the uncertainty → more difficult but more accurate

# Bayesian models

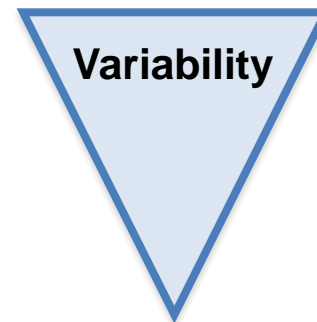
- Observation model  $\rightarrow$  Zero-inflated Poisson

$$Y_s \sim ZI\mathcal{P}(1 - q_s, \mu'_s)$$

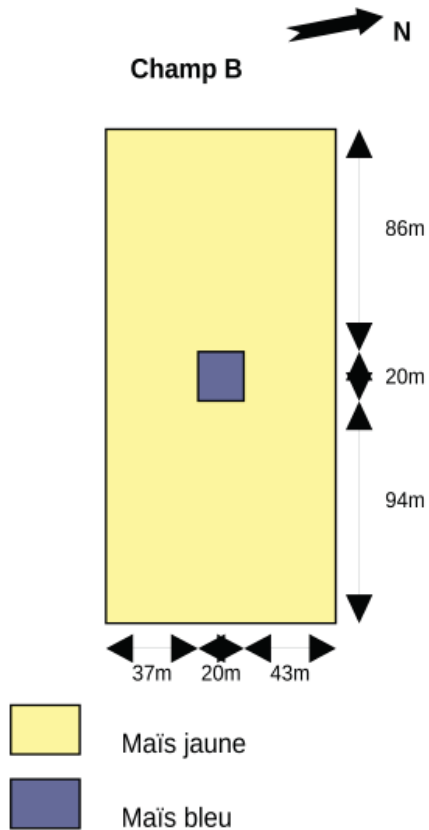
$$\mu'_s = \mu_s \times \text{FD}$$

or  $\ln(\mu'_s) \sim \mathcal{N}(\ln(\mu_s \times \text{FD}), \sigma^2)$

- 3 pollen dispersal models: exponential, 2Dt, NIG
- 3 « co-variate » models
  - Distance
  - Distance + Wind
  - Distance + Wind + Flowering

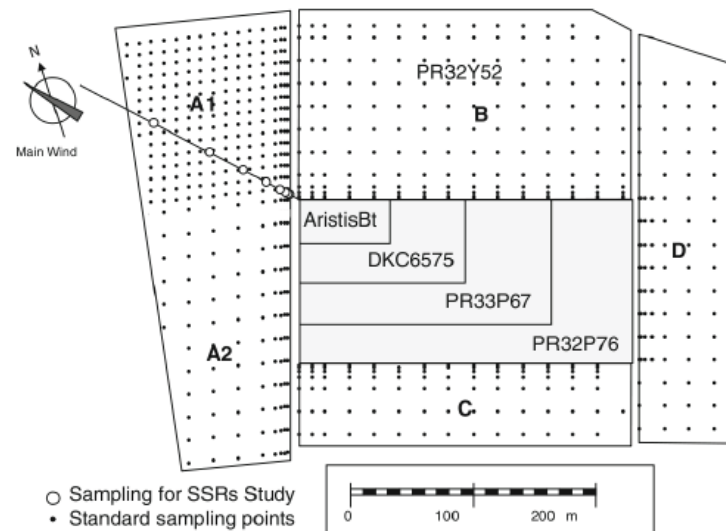


# Datasets used

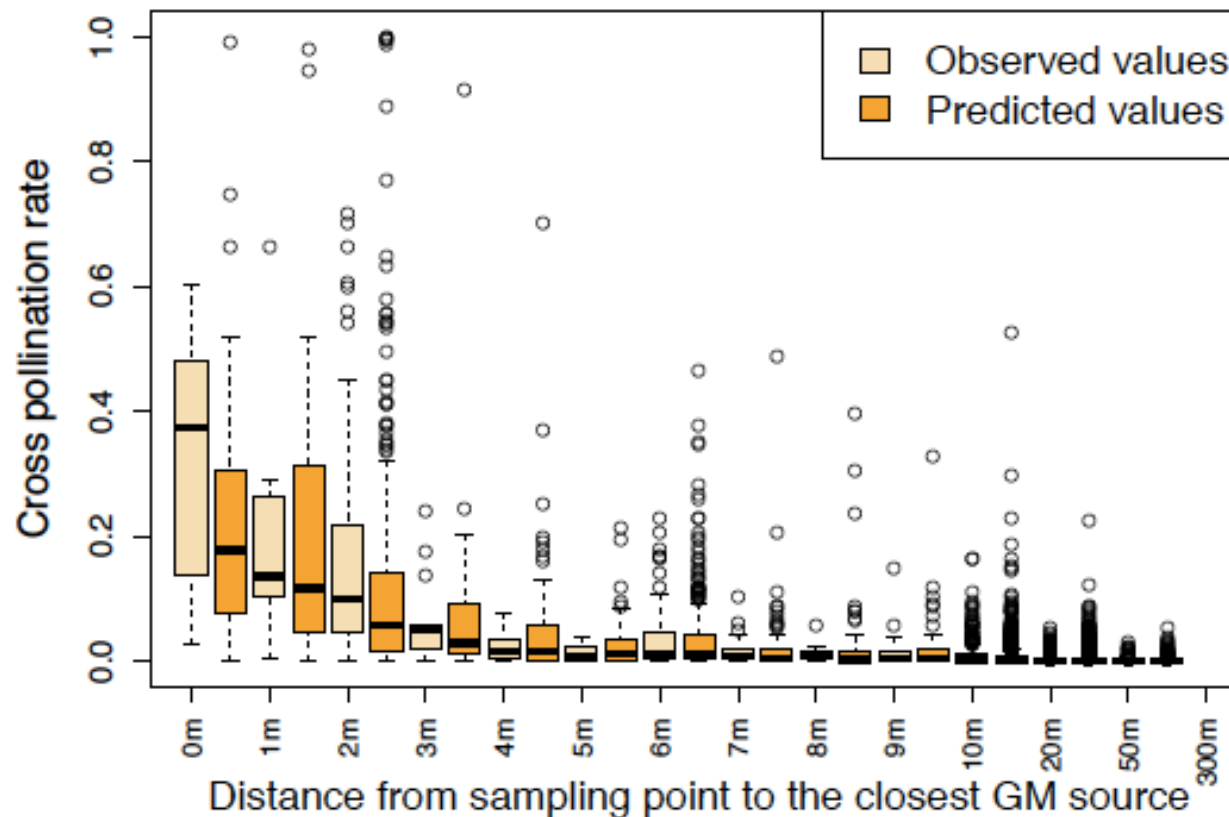


Montargis, France, 98/99  
(Klein et al., 2003)

Mas Cebria, 2006  
(Palaudemas, 2012)



# Correct prediction of AP and of its within-field variability



(Montargis, 1998/1999)

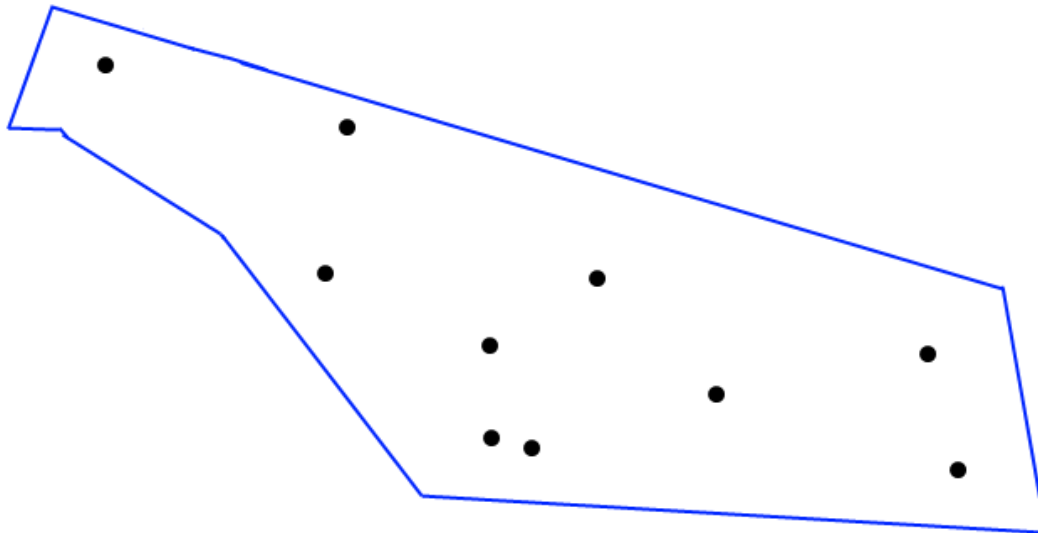
# Model-based sampling strategies

- Simulation of the **within-field spatial distribution** of GMO AP using our Bayesian gene flow model predictions
- Calculation of mean and variance of the simulated AP values
- Possible stratification of fields (mean and variance)
- Comparison of various estimation strategies with random sampling



# Random sampling

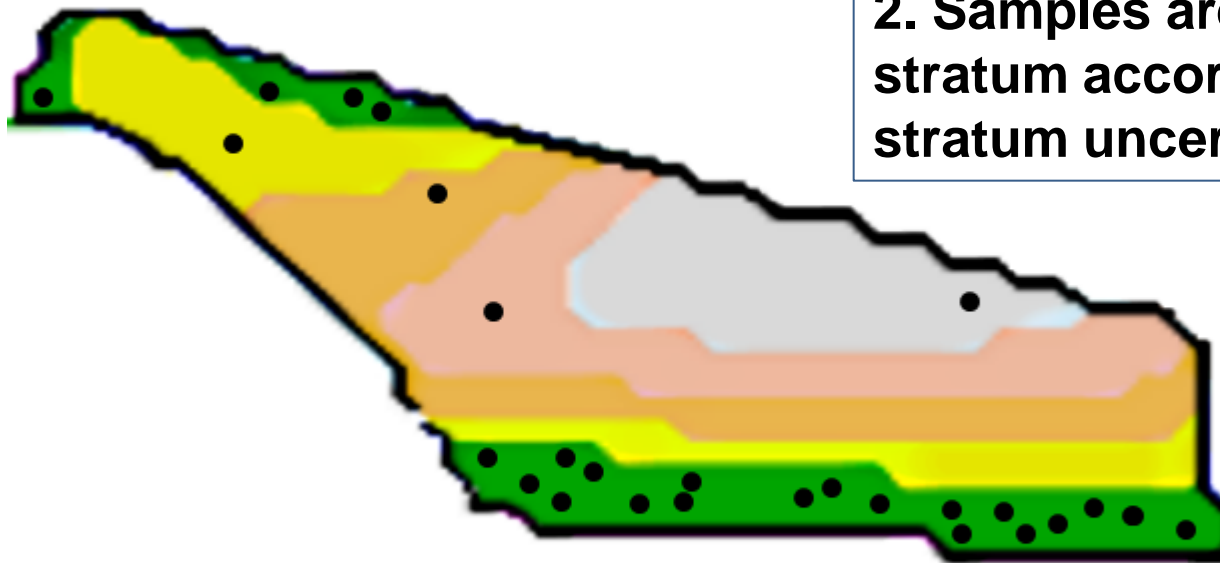
- The random sampling without replacement was used as a reference method, since it does not need any prior information. The estimator of the field AP is the sample mean.



1. The field is divided in strata of same surface, created by ranking each ear with the model output.



2. Samples are distributed in each stratum according to the intra-stratum uncertainty of the model.



Low AP

High AP



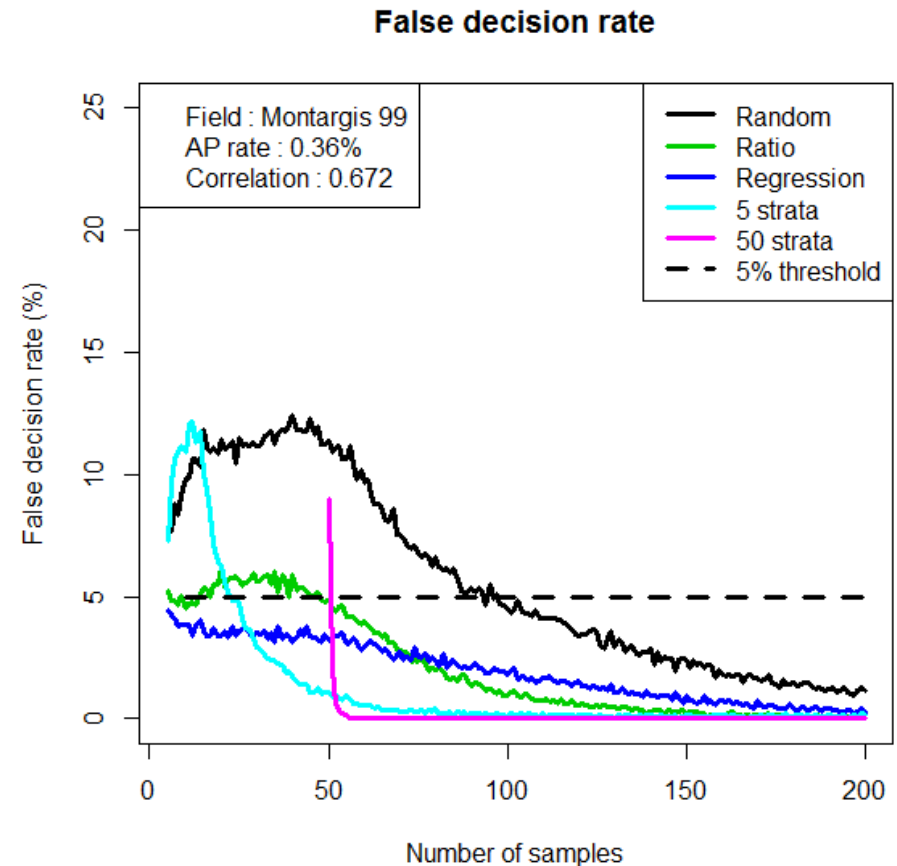
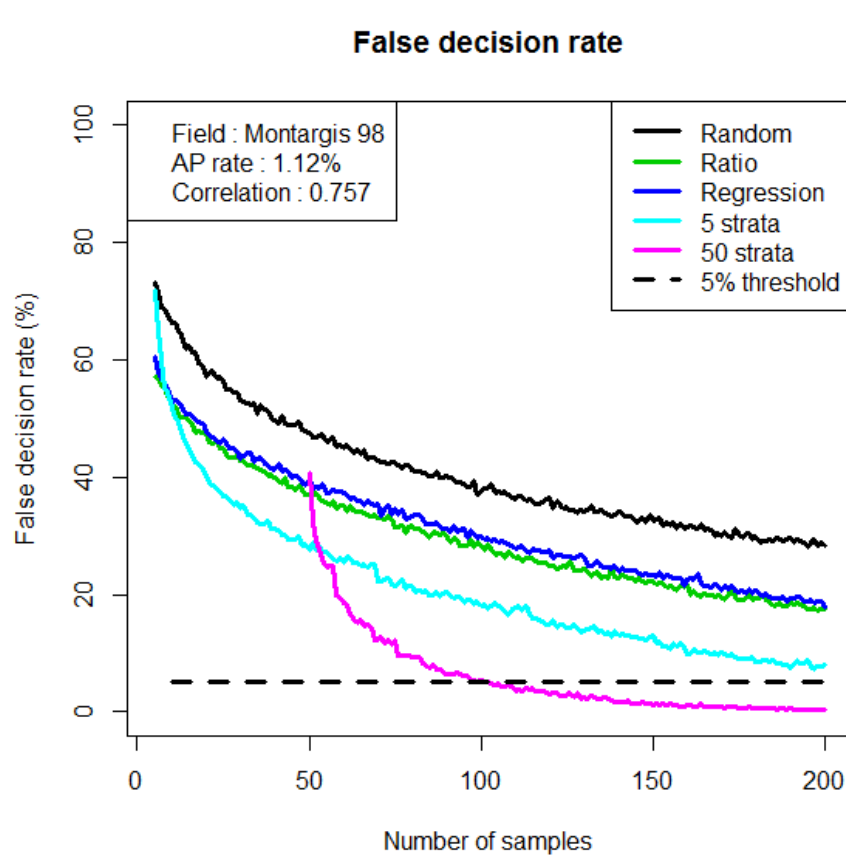
1% of the samples

96% of the samples



# Simulation study on real datasets

- Using the gene flow information helps reduce the sampling effort
- Stratification further reduces the effort

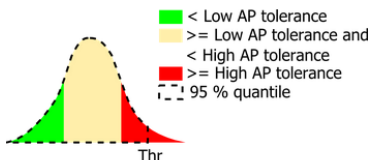


# Decision support tool taking into account local conditions

Bayesian

**Display**

- ☒ 95% probability value Set low and high tolerance value to see fields where the quantile 95% of AP is below or above
- ☐ Probability above threshold Set a tolerance value and quantiles boundaries to see fields where the tolerance is below or above quantiles



## Low and high tolerances

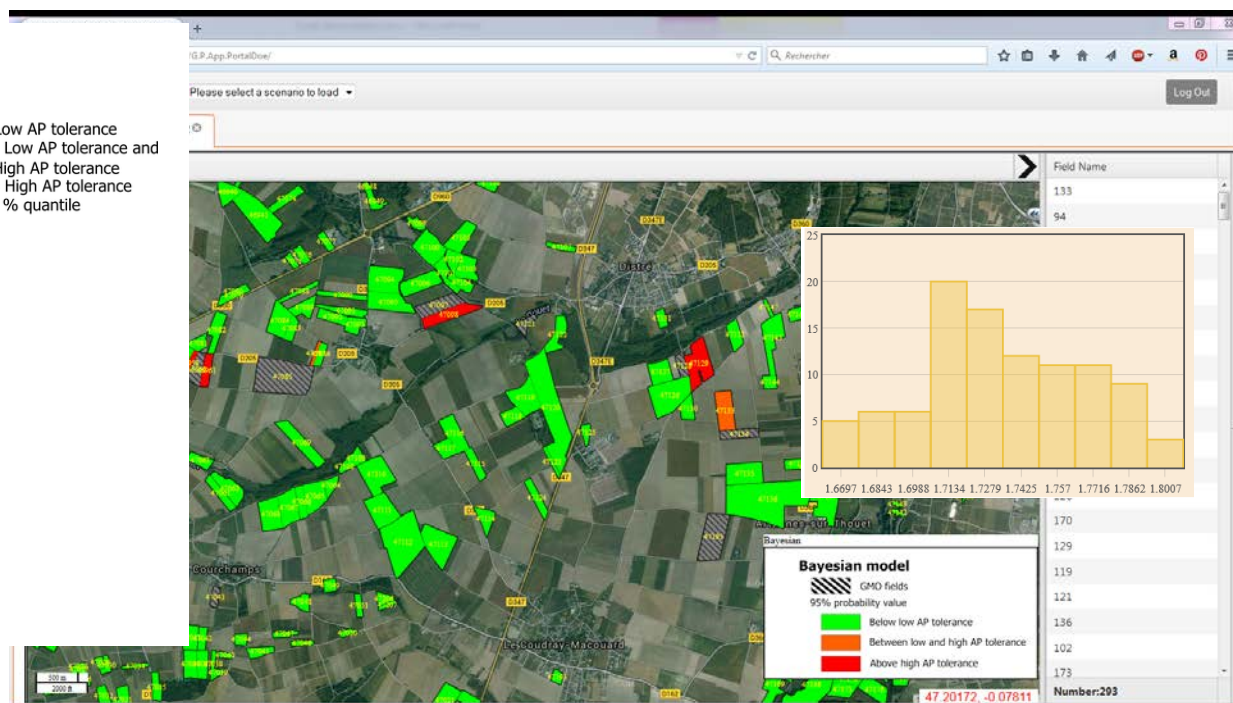
Low AP tolerance  %

High AP tolerance  %

The modeling is based on distance between fields, select an option below to take into account more parameters.

- ☐ Use wind in the pollen dispersal process
- ☐ Use flowering date in the pollen dispersal process

**Simulate**



Set inputs



Run  
model

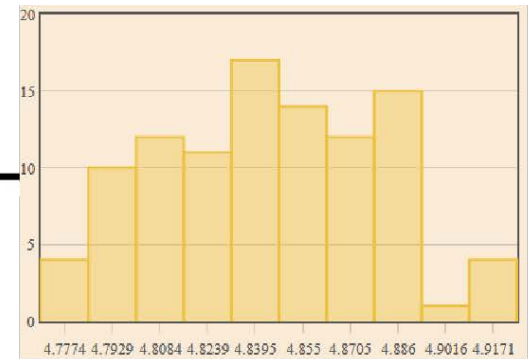


Analyze  
results

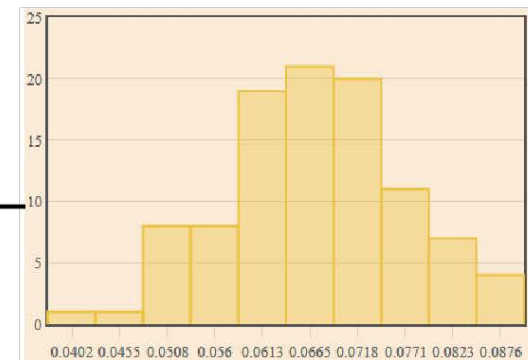


Format  
outputs

# Simulated Distribution of AP (%) at the Landscape Level



No adventitious presence in this field.



# Conclusions

- A comprehensive Bayesian statistical approach has been successfully implemented (for the first time) to the case of gene flow and has made it possible to:
  - Give information on the uncertainty related to adventitious presence in non-GM fields;
  - Inform the decision-maker on adventitious presence whatever is the level of information available on practical situations;
  - Model the variability of adventitious presence within a field;




# Conclusions (2)

- The cost-effectiveness of conventional sampling strategies is poor when the AP is close to targeted thresholds;
- More cost-effective sampling strategies can be proposed by:
  - Using the information given by the above Bayesian gene flow models to optimize sampling within fields;

# Conclusions (3)

- A prototype of a web-based platform implementing decision-tools to support coexistence at the landscape level has been designed;
  - To allocate GM and non-GM fields meeting given **uniform isolation distances**, directly applicable to a wide range of crops (soybean, rapeseed) or cropping systems (seed production);
  - The Bayesian approach featuring the adaptability to available information and reflecting the level of uncertainty has been implemented for maize;
- The Bayesian approach can be extended to cover other compartments of the coexistence supply chain as well as other dispersal issues (pests)

An aerial photograph of a rural landscape. A light-colored dirt road or path runs diagonally from the bottom left towards the center. To the left of the road is a large, dense field of green trees or shrubs. To the right is a field of tall green crops, possibly corn. In the background, there are rolling green hills under a clear blue sky with a few small clouds. The text is overlaid in the center of the image.

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