

# What role have changes in climatic conditions and atmospheric pollution played in the spatio-temporal variation of epiphytic bryophyte communities?

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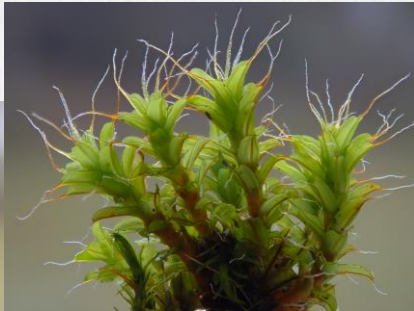


# Introduction

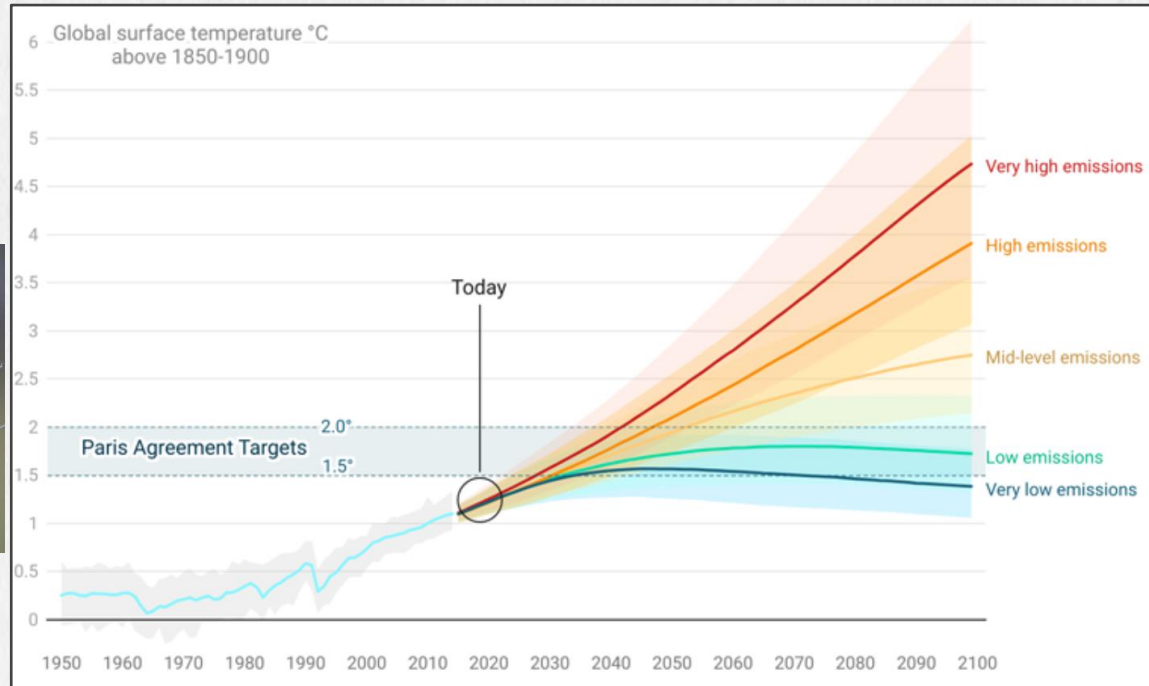


## Climatic change

- Sensitive epiphytic bryophytes
- No stomata
- No roots
- No water content regulator system



Observed and predicted increase in temperatures compared to the pre-industrial period (IPCC 2018)

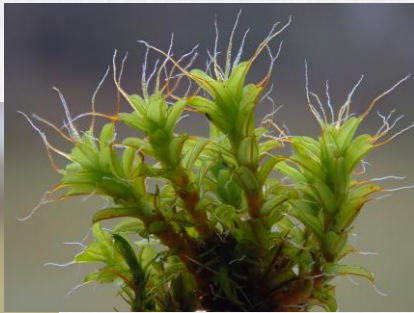


# Introduction

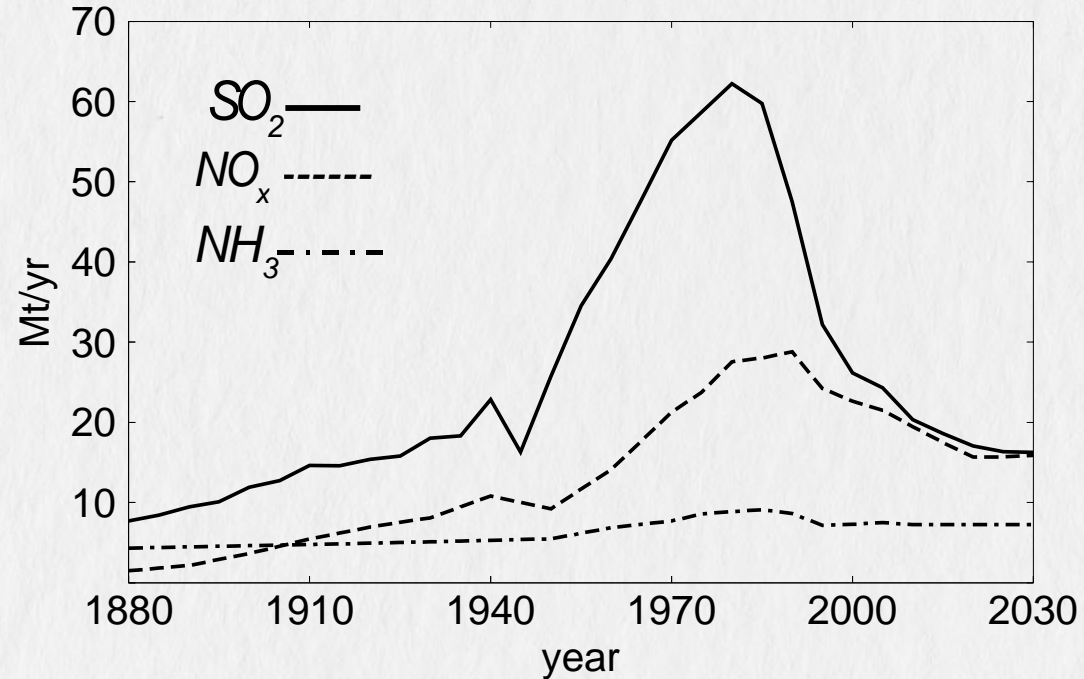


## Atmospheric pollution

- Sensitive epiphytic bryophytes
- Industrial period
- 1950-1980 pollution peak
- Primary sanitary concern in Europe



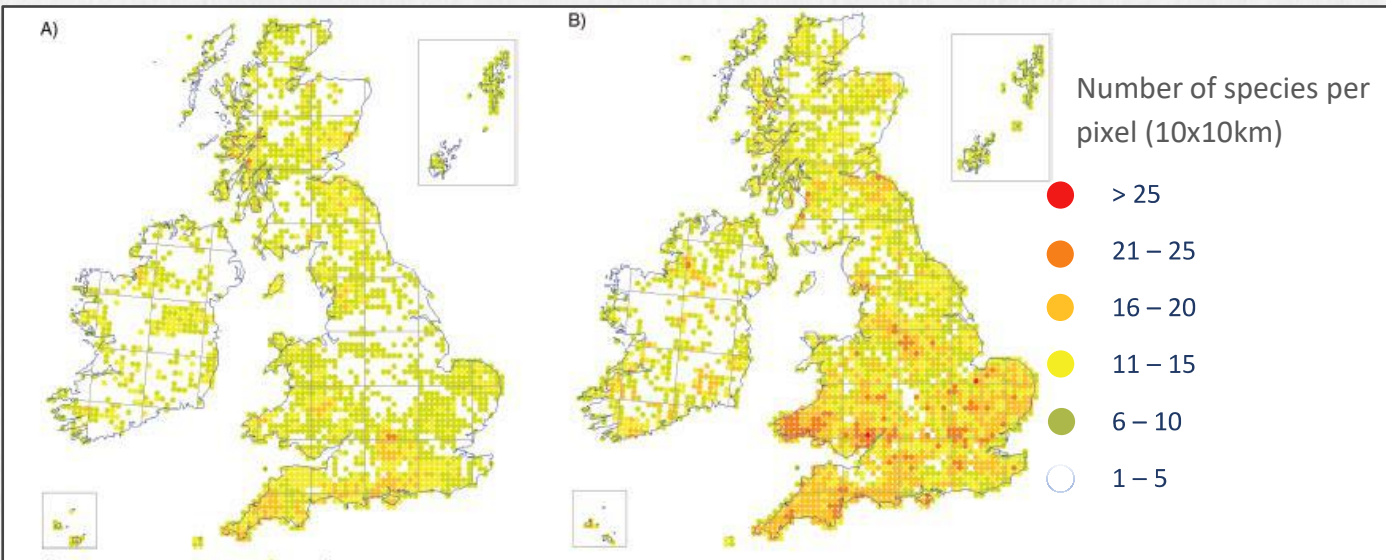
Variation in emission of major atmospheric pollutants 1880-2030 in Europe (Schöpp et al. 2003, Hydrol. Earth Syst. Sci.)



# Introduction



The spectacular expansion of epiphytic flora



1960-1980

1990-2010

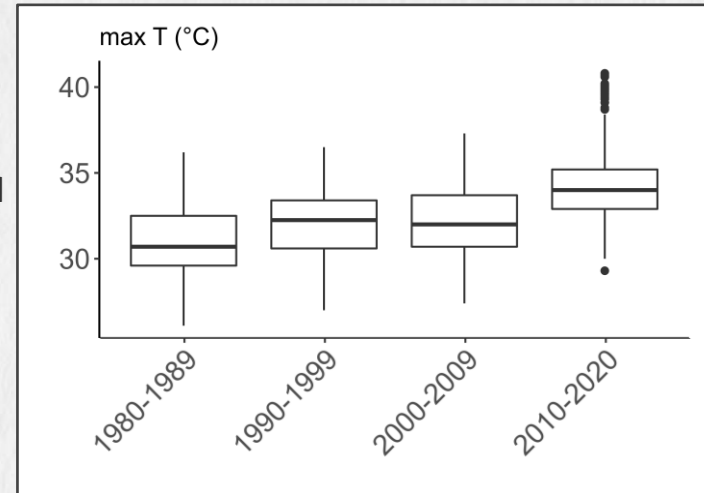
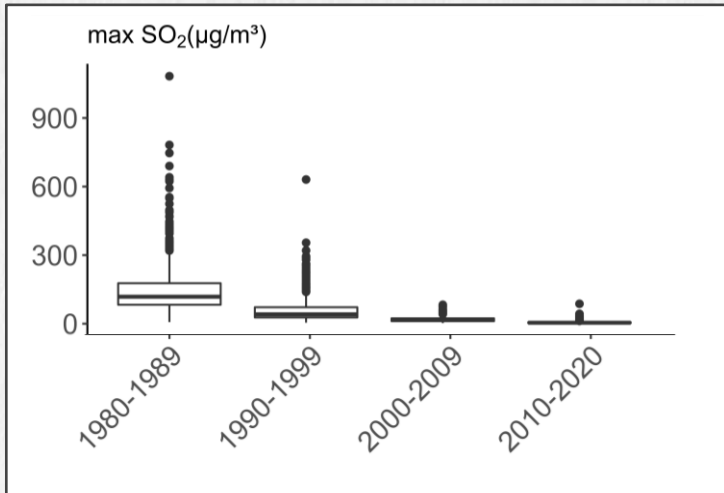
Frequency of 28 species of epiphytic bryophytes in 1960-1980 (A) and 1990-2010 (B) in Great Britain (Hill et Preston 2014).

# Concomitance of climate change and variation in air quality



## Objective

To disentangle the relative impact of climate change and variation in air pollution on the temporal shifts of epiphytic bryophyte species composition





# Study area

- South of Belgium = Wallonia





# Study area

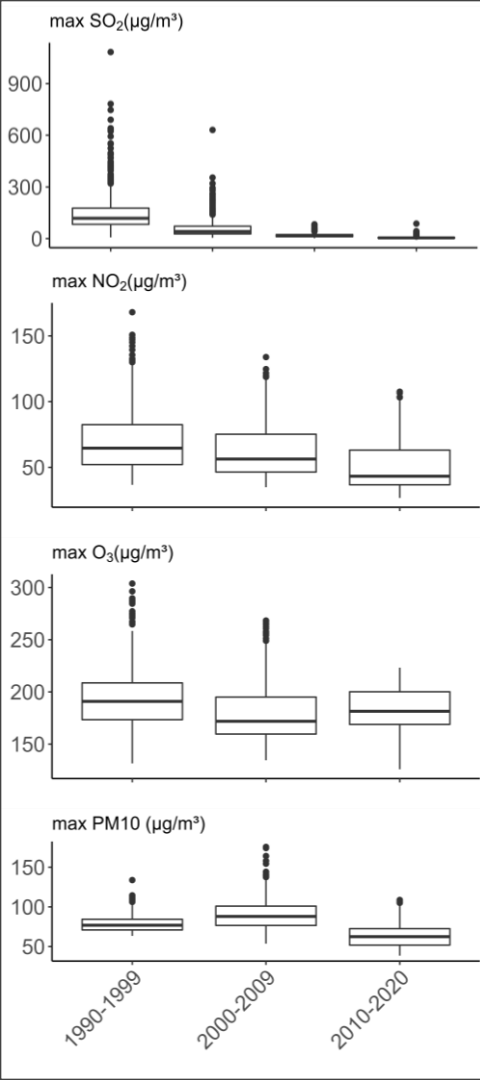
- South of Belgium = Wallonia



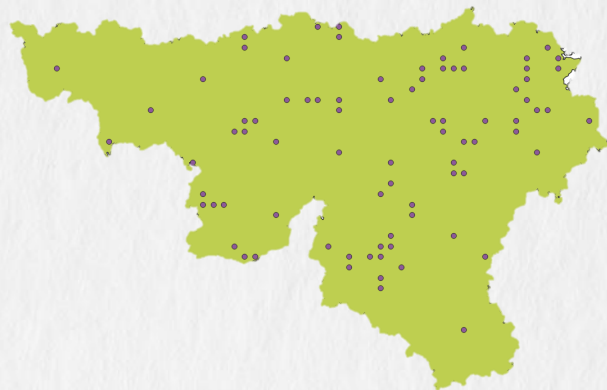


# Method

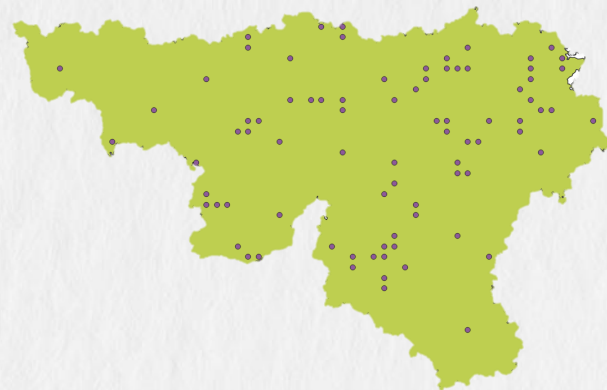
- South of Belgium = Wallonia
- Bryophytes records from 1980-1990-2000-2010 and 2015-2020.
- Atmospheric pollutants : from 1980 for SO<sub>2</sub>, 1990 for NO<sub>2</sub>, O<sub>3</sub>, PM10, interpolated (stored by the Belgian Interregional Environment Agency, IRCEL—CELINE).
- Climatic data were provided daily since 1980 by the Royal Meteorological Institute of Belgium (RMI).



1980-2010



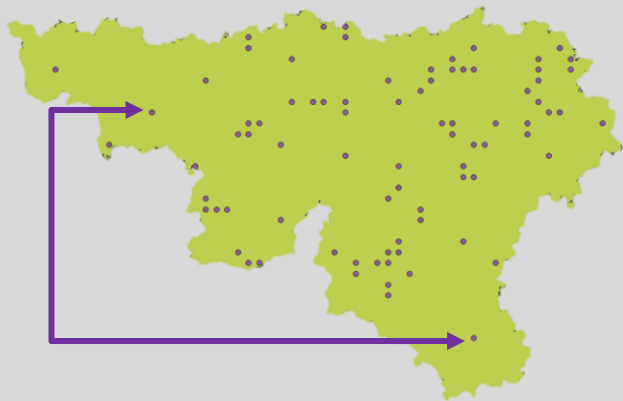
2015-2020






# Beta diversity

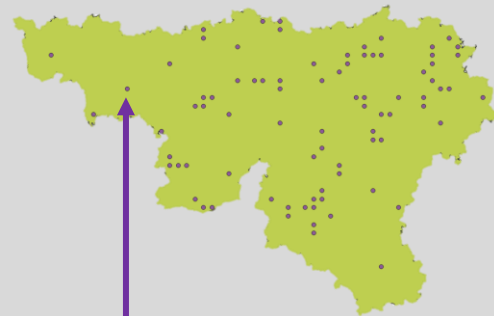
 spatial  $\beta$



 temporal  $\beta$

1980-1990-2000-2010

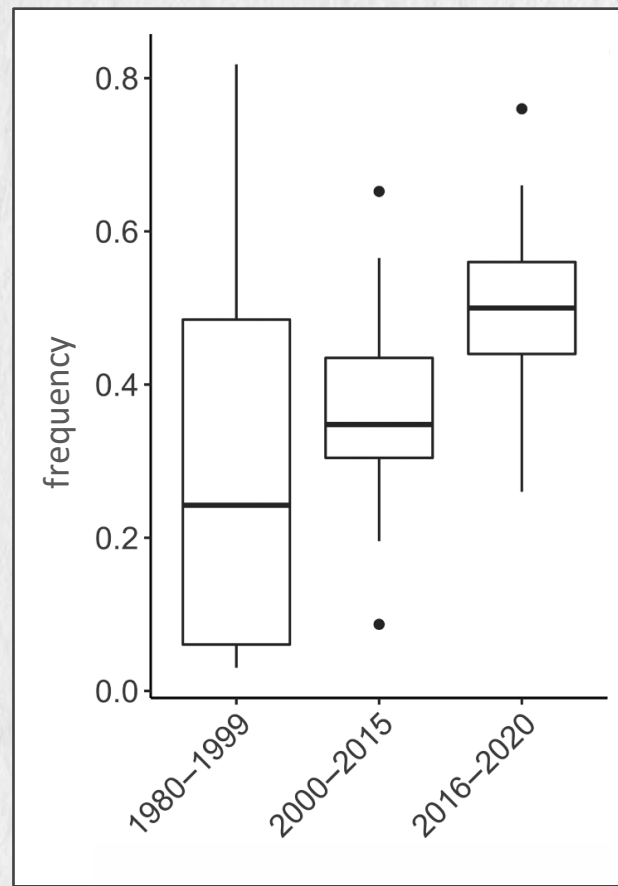
2015-2020



# Results



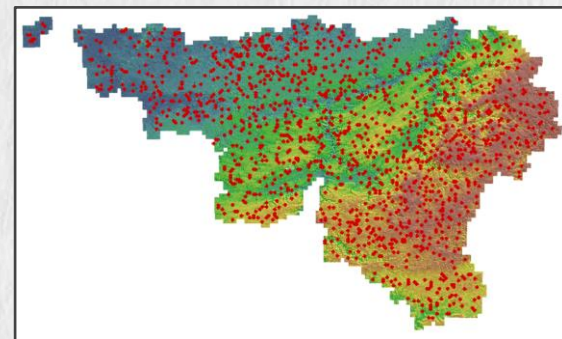
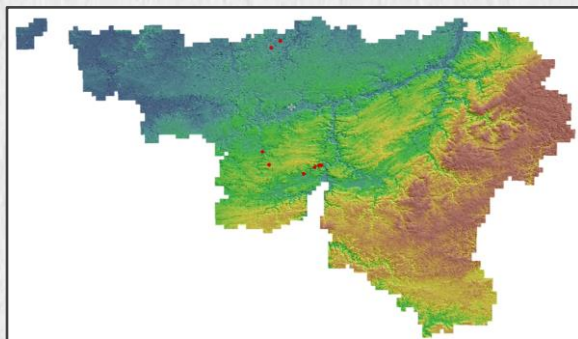
Temporal variation of species frequency



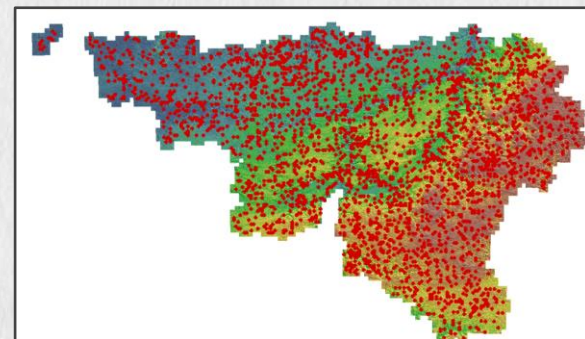
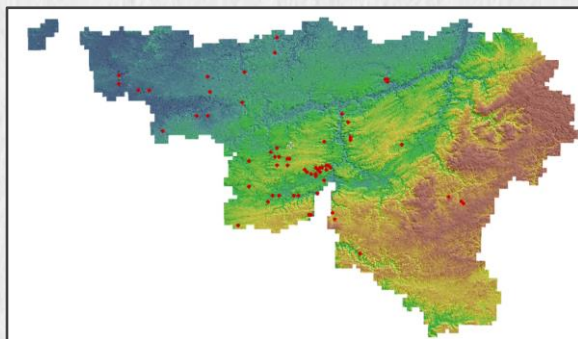
1980-1985

2015-2020

*Cryphaea heteromalla*



x200



x50

*Lewinskya affine*

# Results

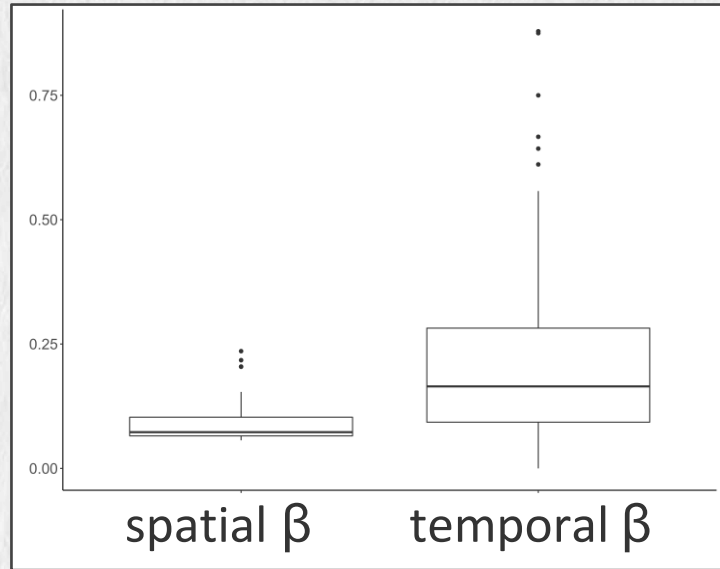


temporal  $\beta$

## Temporal vs spatial variation of the bryophyte community composition

1980-1985

2015-2020



# Results

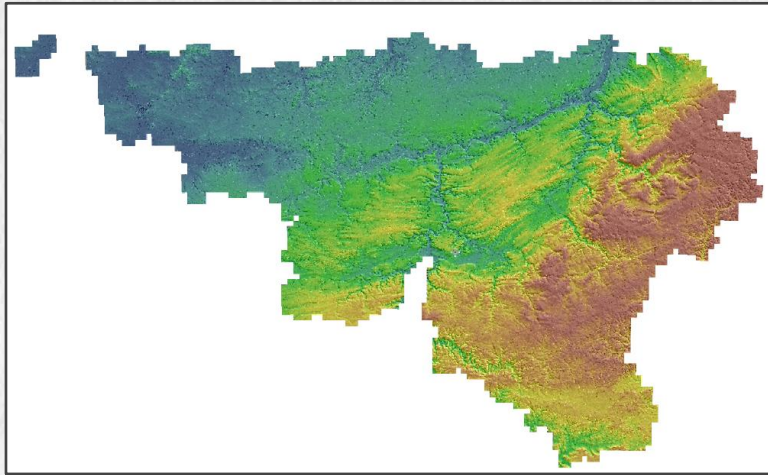


spatial  $\beta$

## Temporal vs spatial variation of the bryophyte community composition

Low land

High land

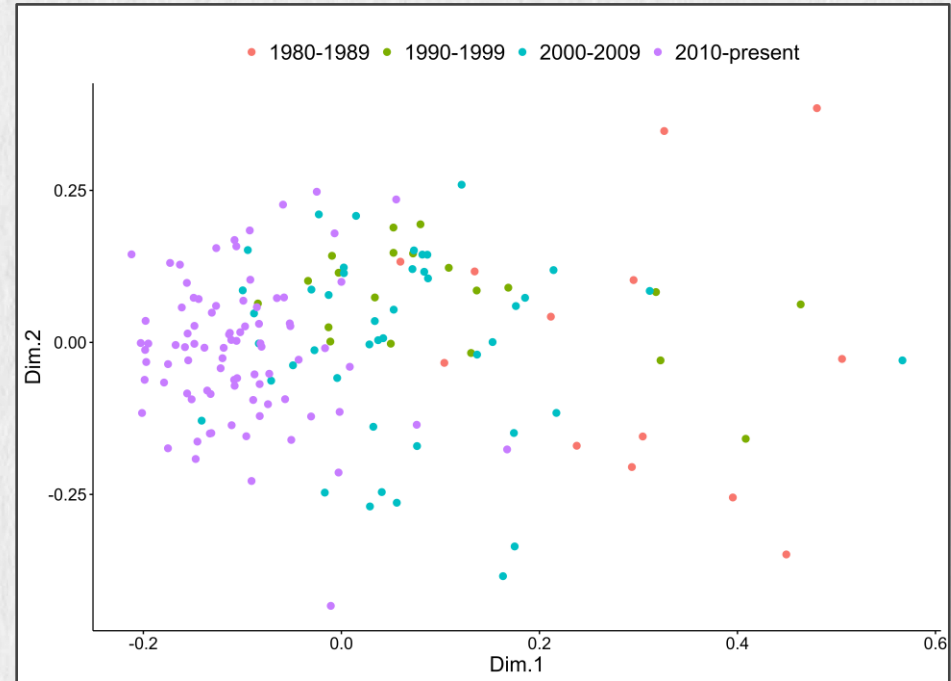
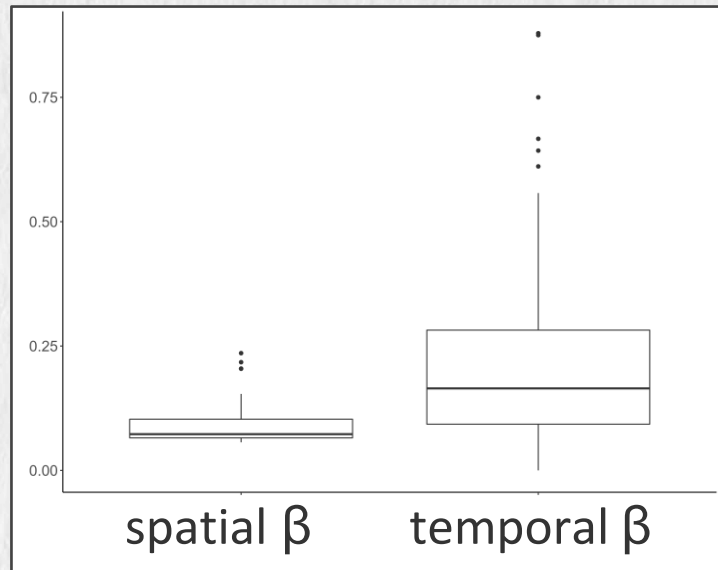


# Results



MDS

## Temporal vs spatial variation of the bryophyte community composition



MDS ordination of 81 of 16 km<sup>2</sup> pixels in southern Belgium based on a Sorensen distance matrix of their composition in epiphytic bryophytes

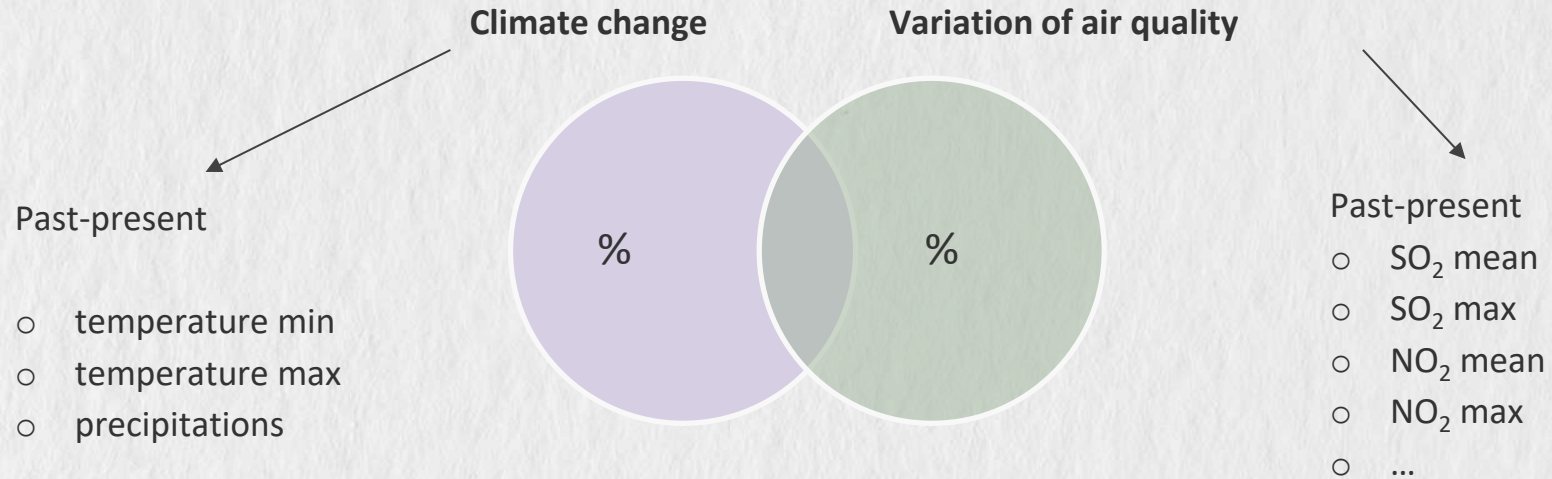
# Results



Variance partitioning



## Drivers of temporal variation of the epiphytic bryophytes communities



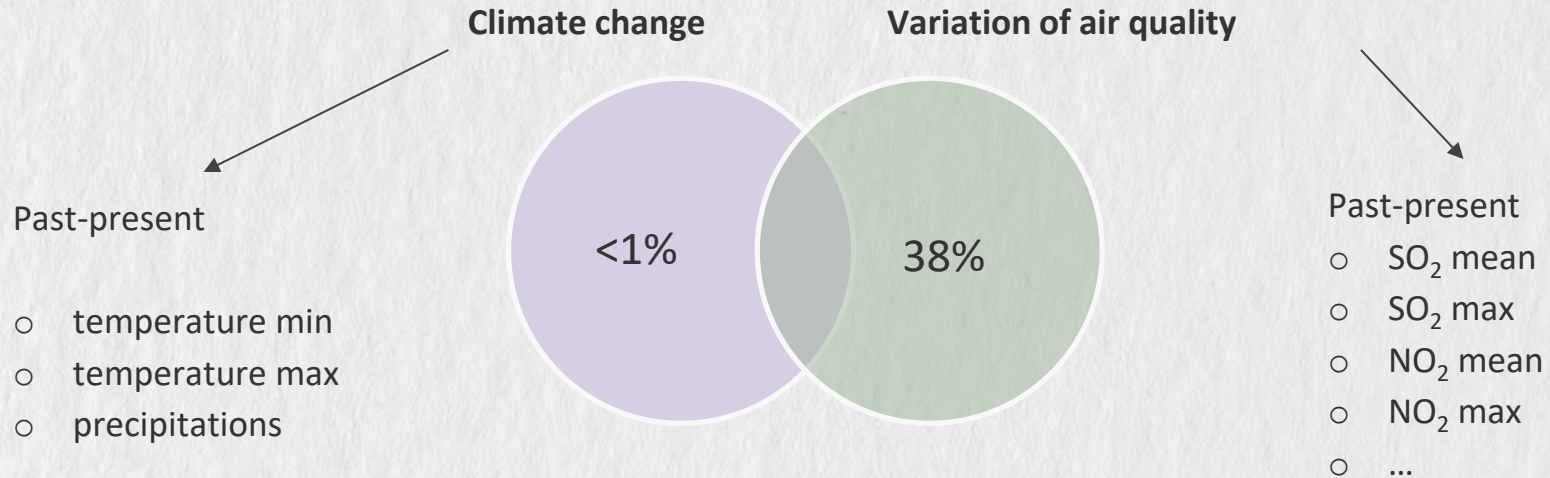
# Results



Variance partitioning



## Drivers of temporal variation of the epiphytic bryophytes communities

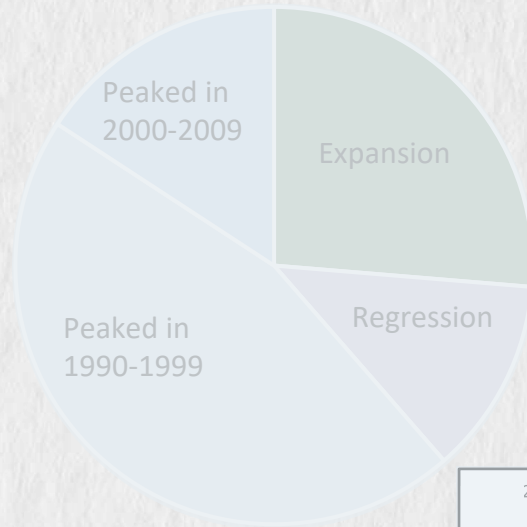
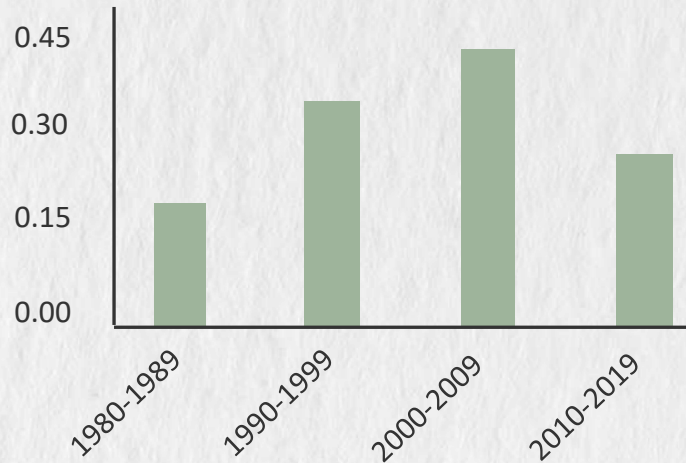


→ Negligible impact of variations of temperature and precipitation.



# Glory of epiphytes in the 2000s, then decline

% occurrence of epiphytes  
among all observations per decade

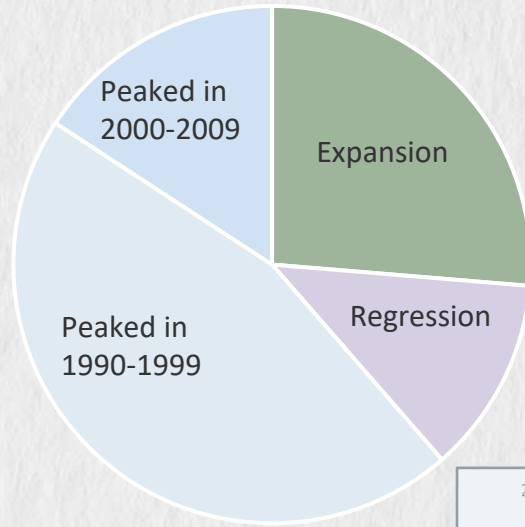
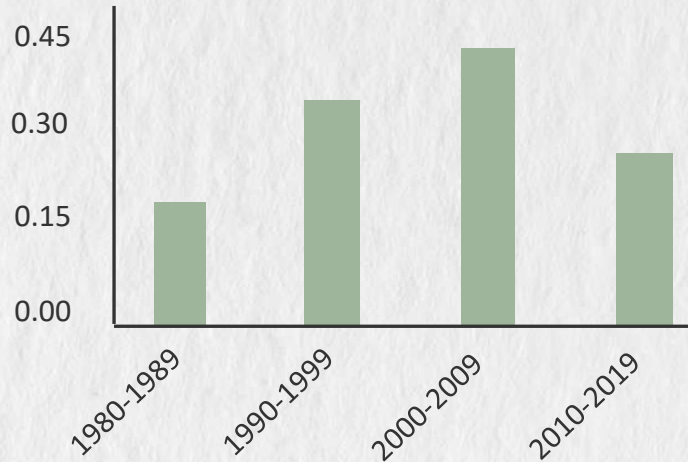


*Platygyrium repens*



# Glory of epiphytes in the 2000s, then decline

% occurrence of epiphytes among all observations per decade

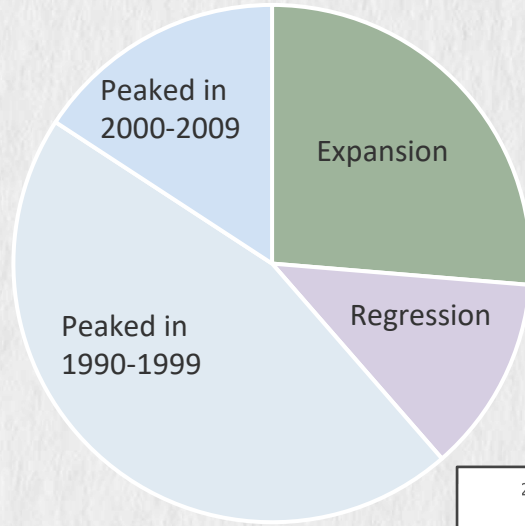
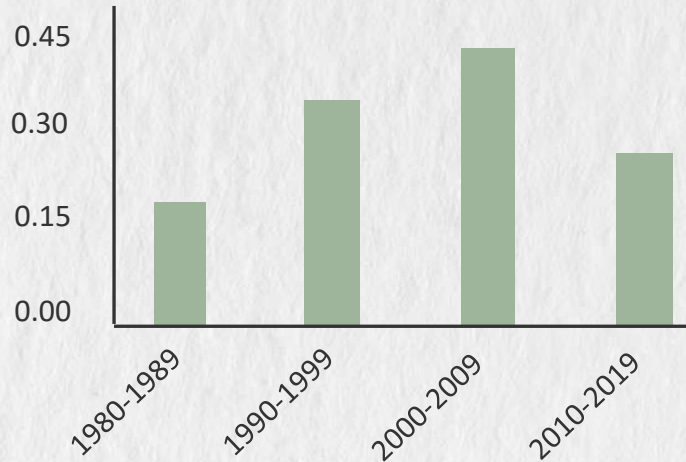


*Platygryium repens*

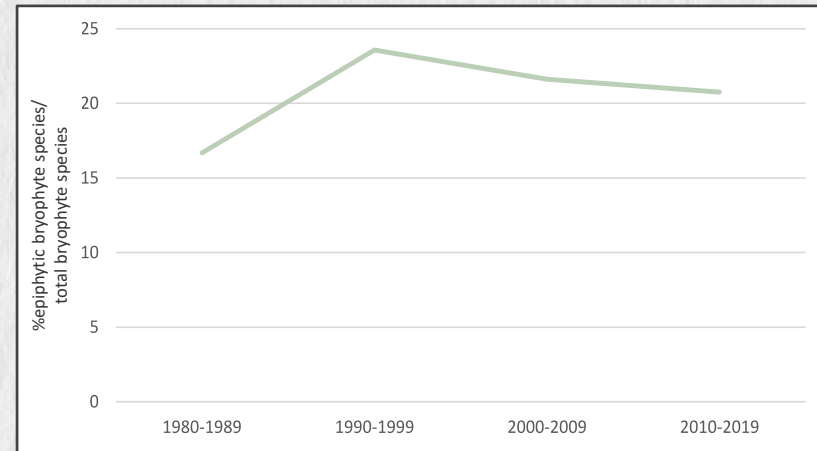


# Glory of epiphytes in the 2000s, then decline

% occurrence of epiphytes among all observations per decade



*Platygyrium repens*



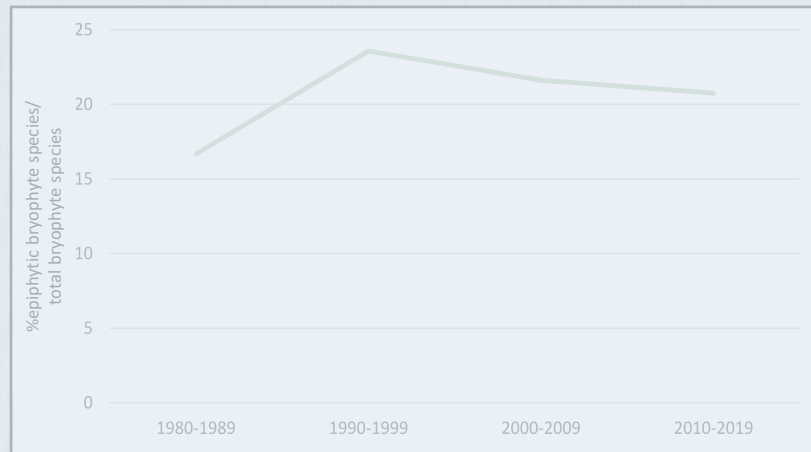
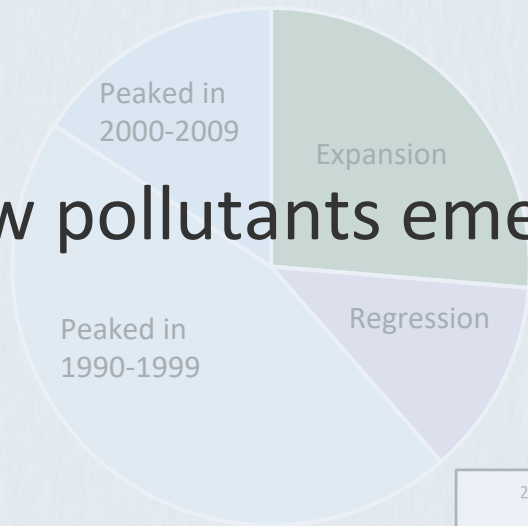
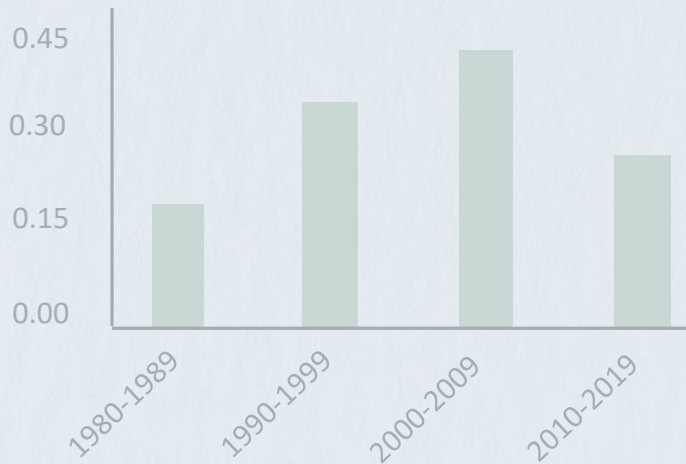
# Glory of epiphytes in the 2000s, then decline

## Did new pollutants emerge?



*Platygyrium repens*

% occurrence of epiphytes among all observations per decade



# Impact of new air pollutants on extant species with SDM

## Climatic variables

Temperature

Precipitation

Relative humidity



## Major pollutants

NO<sub>2</sub>

SO<sub>2</sub>

O<sub>3</sub>

PM10 and PM2,5

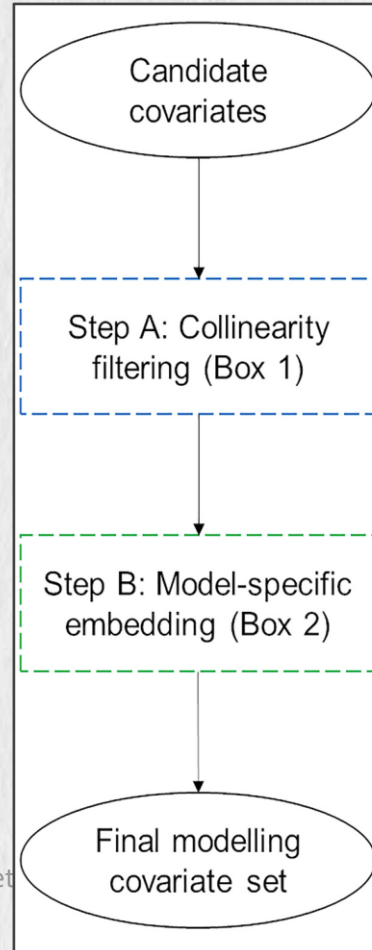


## New pollutants

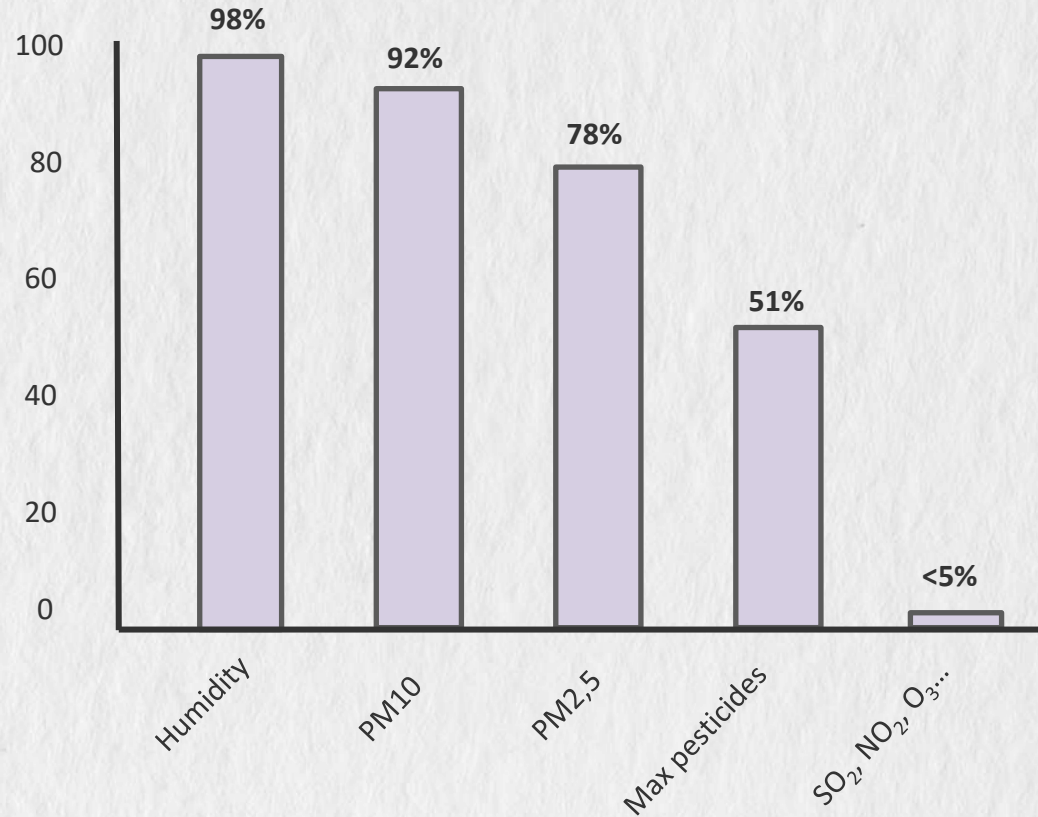
Pesticides

Heavy metals

Black carbon



# Contemporary spatial variation of new pollutants with SDM



- Climatic variation prevail over major pollutants
- Emergence of alternative pollutants: fine particles and pesticides.



# Conclusion : little impact of climate change ?

- No impact of climate change from 1980 to 2020
- Extant species distributions is largely driven by spatial variation in climatic conditions



- A flora on the edge of its climatic tolerance threshold?
- A debt of extinction?
- Emergence of new air pollutants

Any questions ?

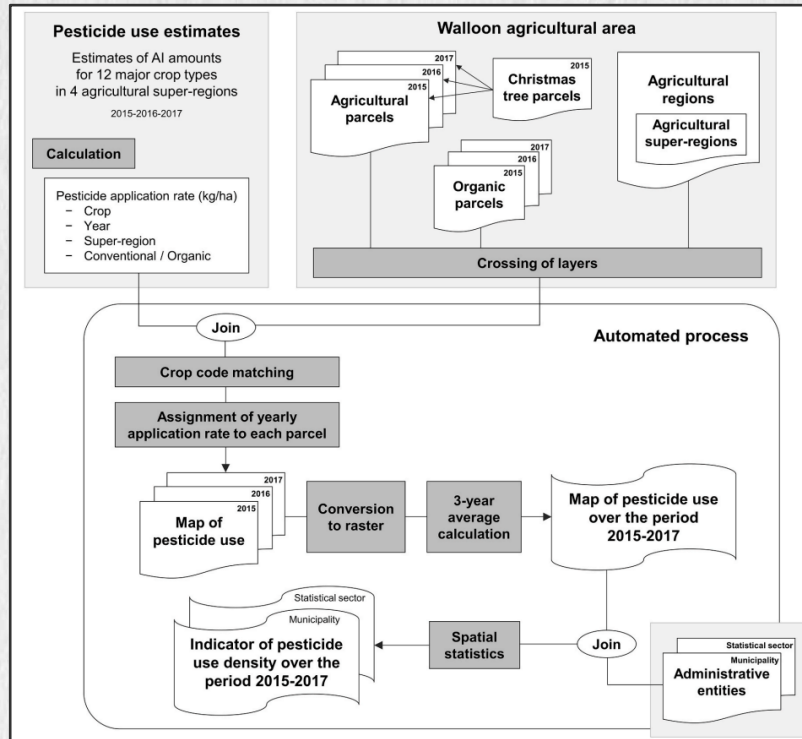


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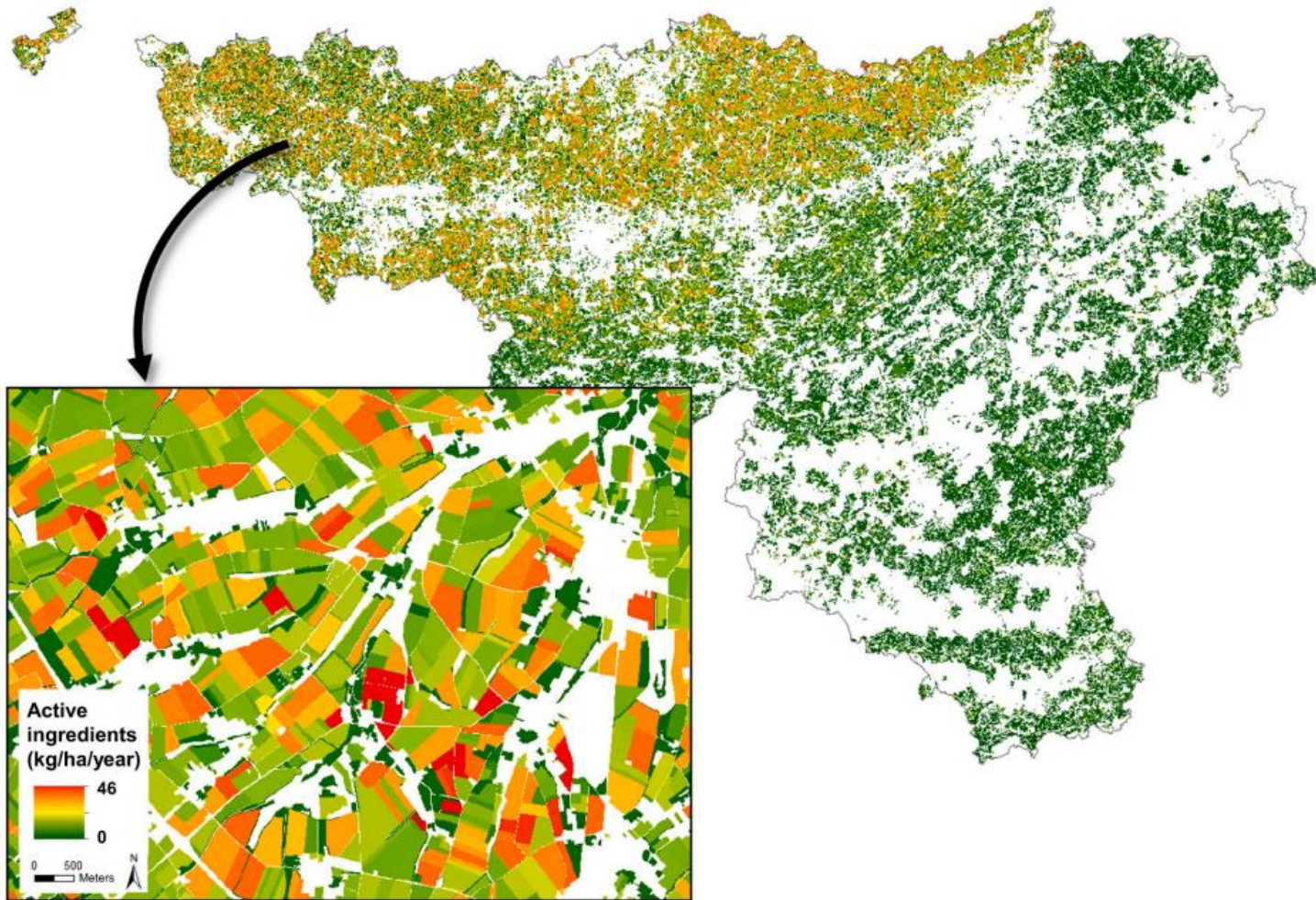




# Mapping agricultural use of pesticides to enable research and environmental health actions in Belgium



- Max amount of pesticides legally authorized
- Crop type (corn, wheat, beet, potato, ...)
- Organic/non-organic
- Region of Wallonia
- Over 3 years to consider crop rotation

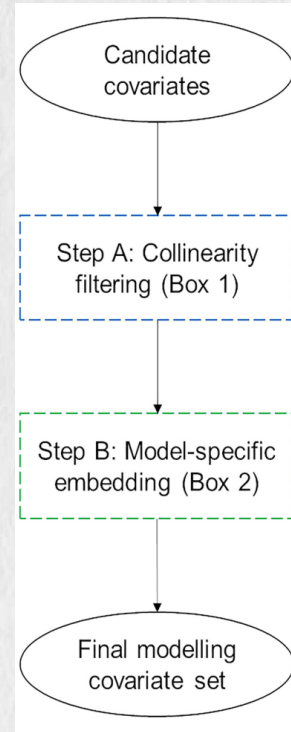


# Embedded covariate selection procedure for species distribution modelling with the covsel R package

Two-step “embedded” covariate selection procedure. The procedure combines a collinearity-filtering algorithm (Step A) with three model-specific embedded regularization techniques (Step B), including generalized linear model with elastic net regularization (GLM-EN), generalized additive model with null-space penalization (GAM-NP), and guided regularized random forest (RF-GR).

In Step A (Collinearity filtering), a panel of many candidate covariates is reduced in number after collinearity analyses. Collinearity analyses are usually based on variance inflation factors (VIFs) or correlation tests. Principal components analysis (PCA) is also applied for reducing the dimensionality of covariate spaces.

Step B (Model-specific embedding) algorithms are doing covariate selection at the same time as model fitting, allowing to account early-on for the specificities of the algorithms and the multivariate context. Furthermore, they have a more reasonable computational cost and limit overfitting.



# Interpolation : RIO model

These data served to calibrate the RIO model. RIO is an interpolation model based on land use, a semivariogram based on the distances to the nearest measuring stations and the levels of air pollution, which was employed to compute, on an hourly basis, the background concentrations at the centroid of all the investigated pixels. Based on the interpolated data, the maximum hourly concentration and annual average concentration of each pollutant were computed every year for each pixel. (Hutsemékers et al., 2023)

- Around 20 measuring stations
- Distance to the nearest measuring station
- Level of air pollution

# The spectacular expansion of epiphytic flora

The significant increase both in frequency and cover of the liverwort *Metzgeria furcata* (Friedel & Müller, 2004).

The bryophytes *Hypnum cupressiforme*, *Brachythecium rutabulum* and *Plagiothecium laetum* var. *laetum/denticulatum* have significantly (Friedel & Müller, 2004).

Increase in cover and number of species of epiphytic bryophytes on forest ecosystem monitoring areas in North Rhine-Westphalia (western Germany) between 1999 and 2001 (Stapper, 2002).

The increase of nitrophytic species suggests that the problem of air pollution due to sulphur inputs may have shifted to a problem of nitrogen fertilization (Frahm, 2001).