



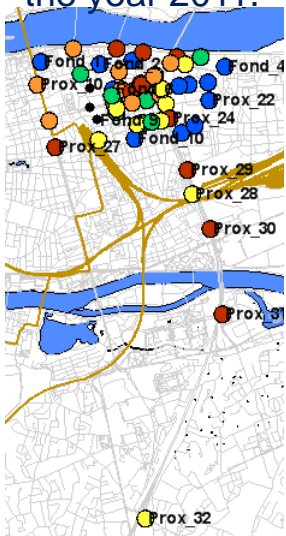
# Spatial representativeness and station classification

- Local assessment of station representativeness based on sampling surveys and (where possible) geostatistical data analysis
  
- European/national scale: on-going studies on station classification and data quality for model evaluation and air quality mapping
  - ✓ Classification according to Joly and Peuch methodology (2012), comparison with AirBase classification
  
  - ✓ Detection of outliers

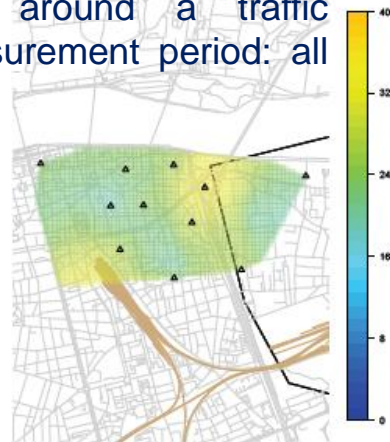
## Local assessment of spatial representativeness

- Implementation of a geostatistical approach based on passive sampling surveys (Bobbia et al., 2008; LCSQA, 2007, 2010-2012)

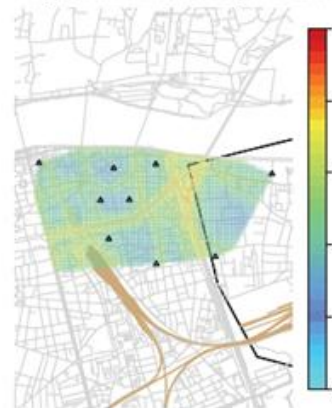
City of Tours. NO<sub>2</sub>. Passive sampling survey conducted by Lig'Air around a traffic monitoring station. Measurement period: all the year 2011.



*Background pollution: kriging with NO<sub>x</sub> emissions and population density as external drift.*



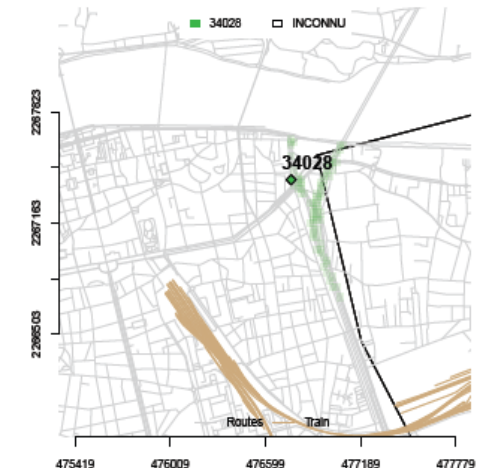
Estimation du NO<sub>2</sub> sur Tours en 2011  
Application du modèle de surplus de trafic



*Background + traffic-related pollution (statistical adjustment along the roads using sampling data at traffic points)*



Carte de Tours Année : 2011  
Seuil de représentativité : 10µg.m<sup>-3</sup>  
Seuil de probabilité : 10%



**Estimation of the corresponding representativeness area**

**Estimation of NO<sub>2</sub> annual mean concentration**

- Main criterion: concentration difference with respect to the station measurement
- For a station  $S_0$  located in  $x_0$ , a given pollutant (ex:  $\text{NO}_2$ ), a given concentration variable  $Z$  (ex: annual mean) and a given period (ex: one year),
  - $x$  is considered as part of the representativeness area of  $S_0$  if:

$$|Z(x) - Z(x_0)| < \delta$$

$\delta$  : threshold in  $\mu\text{g}/\text{m}^3$

- Method:
  - $Z(x)$  is estimated from sampling data and auxiliary variables: external drift kriging + statistical correction along roads.
  - The estimation uncertainty is taken into account by considering the probability  $\eta$  of wrongly including a point  $x$  in the representativeness area of  $S_0$ :

Modified condition for representativeness:

$$|Z^*(x) - Z(x_0)| < \delta - \sigma_k(x) * q_{1-\frac{\eta}{2}}$$

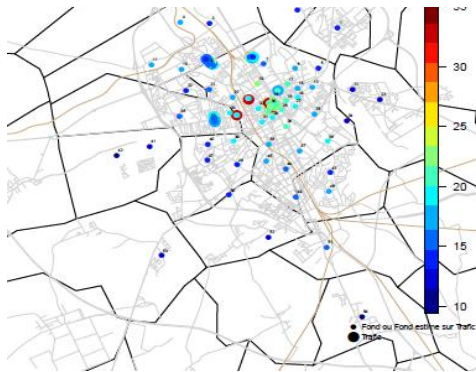
Kriging  
standard  
deviation

Quantile of  
the normal  
distribution

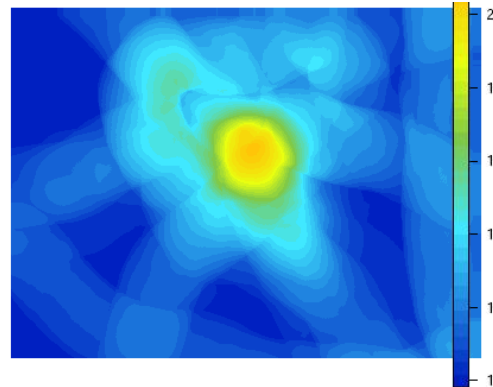
# Spatial representativeness

- Methodology applicable on the urban scale

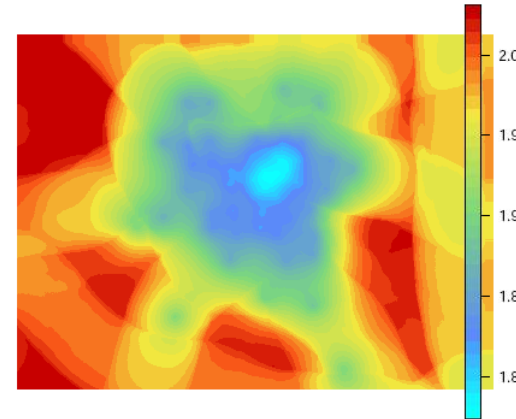
Sampling points: several periods during the year 2009



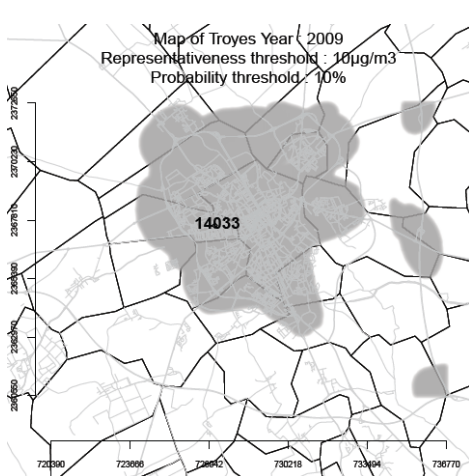
Estimation map of NO2 annual mean concentrations: kriging with NOx emissions as external drift



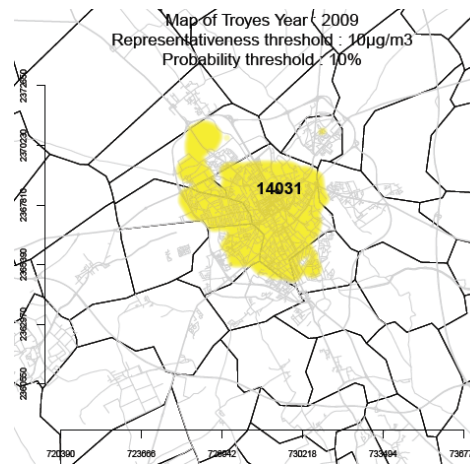
Kriging standard deviation



City of Troyes (campaign conducted by ATMO Champagne-Ardenne) Annual mean concentrations of background NO<sub>2</sub>. 2009.

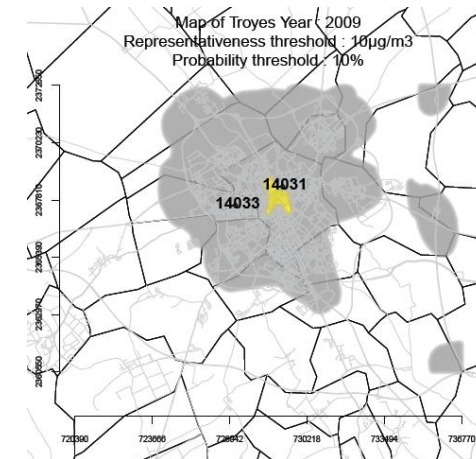


Representativeness area for site 14033



Representativeness area for site 14031

Suppression of the overlap. Different criteria tested. Retained criterion: minimum concentration difference



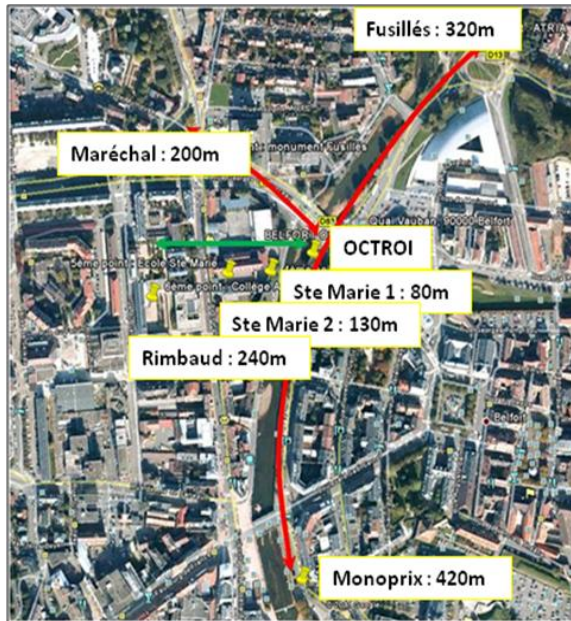
⇒ Partly redundant information. 14033: the most suitable for comparison with large scale modelling results.

- Remarks

- Application limited by the possibility of conducting dense sampling campaigns.
- Methodology mostly adapted to NO<sub>2</sub> or benzene annual, seasonal or monthly average concentrations.
- Requires information on the uncertainty of the concentration map.
- To investigate: how could the methodology be extended to other types of spatial estimates and wider spatial scales?

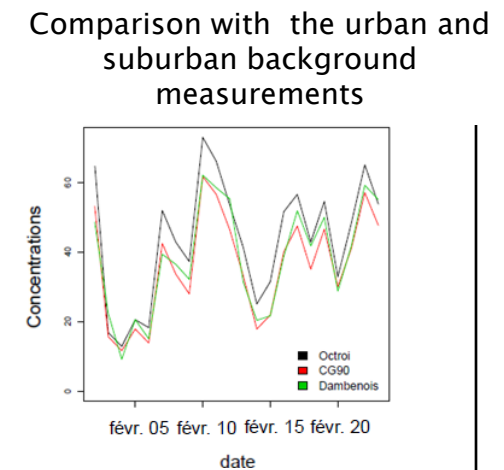
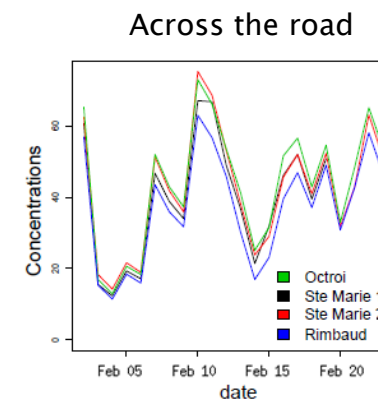
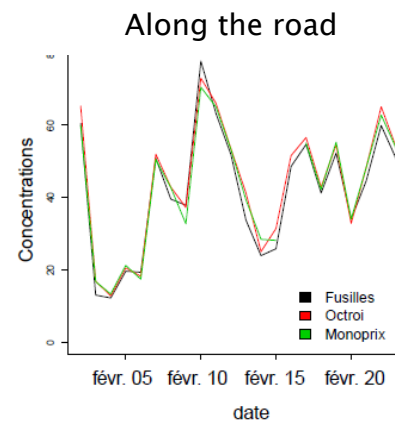
## Spatial representativeness

- Representativeness of PM<sub>10</sub> monitoring sites: feasibility study of an experimental approach



Ex: City of Belfort, PM<sub>10</sub> measurement campaign around a traffic site (Octroi). Campaign conducted in collaboration with ATMO Franche-Comté, February 2011

Gravimetric measurements with DA-80 samplers along the main roads and at increasing distances from the station



Comparison of time series → qualitative assessment of spatial representativeness (in terms of concentration and daily exceedances)

### □ Station classification

To qualify monitoring sites on a wider scale

Possible application for model evaluation and air quality mapping

#### ➤ Study on national scale (LCSQA, 2012)

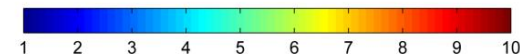
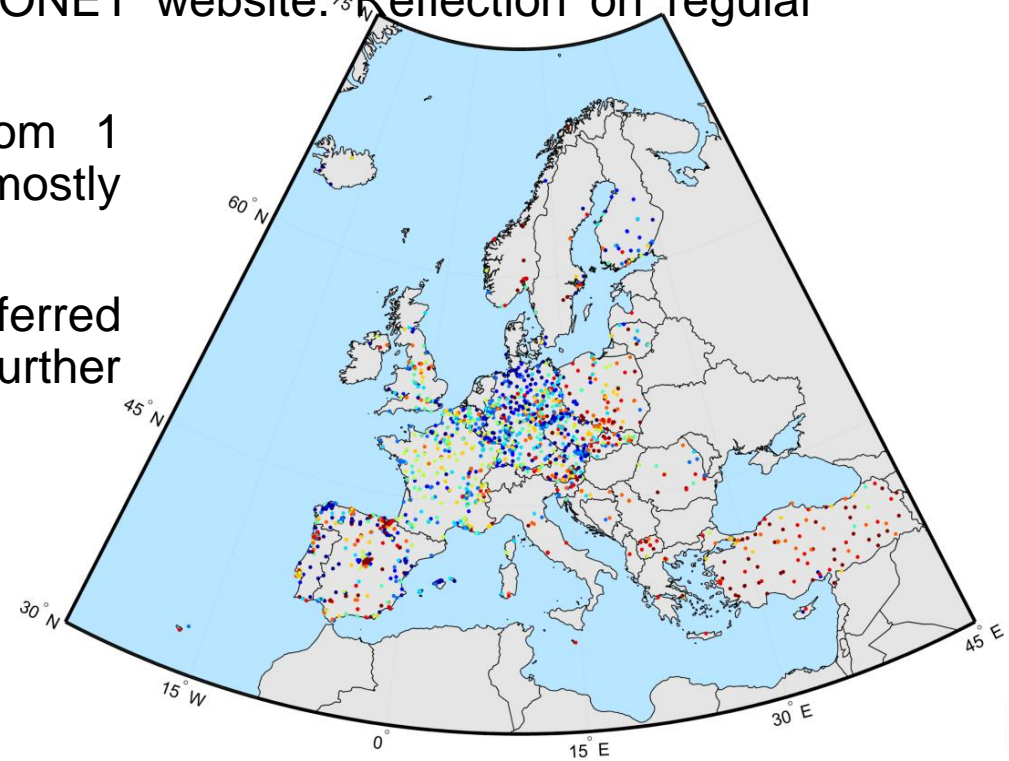
- ✓ Classification through principal component analysis based on environmental parameters (terrain height, population density, land cover,  $\text{NO}_x$  emissions from traffic) and average concentration data (ratio  $\text{NO}/\text{NO}_2$ ,  $\text{PM}_{10}/\text{NO}_2$ )
- ✓ The stations split into five groups which can be interpreted in relation to the environment (urban, agricultural, forest...) and emission sources.



## Station classification

- Study on European scale (ETC/ACM, 2012 & 2013)
  - ✓ Classification based on the temporal variability of concentrations: diurnal cycle, weekend effect, high frequency variability. AirBase type of area and type of station are used as a priori information in the classification process. Methodology developed by Joly and Peuch (2012).
  - ✓ Underlying idea: spatial representativeness and temporal variability are linked.
  - ✓ Application of the methodology to AirBase v6 and update with AirBase v7. Report and results available on EIONET website. <sup>75</sup> Reflection on regular update within MACC project
  - ✓ Pollutant specific classification, from 1 (rural behaviour) to 10 (behaviour mostly influenced by urban traffic)
  - ✓ Identification of specific situations referred to as « outliers » that require further investigation

*Classification of PM<sub>10</sub> monitoring stations according to Joly & Peuch (2012) methodology*



## ➤ Use of station classification in model evaluation and air quality mapping

- ✓ Currently : selection of stations based on AirBase classification (type of area and type of station) and local expertise
- ✓ On-going investigations on the use of Joly & Peuch methodology for air quality mapping :

Comparison of different selections of stations for air quality mapping (observations + CHIMERE combined in an external drift kriging)

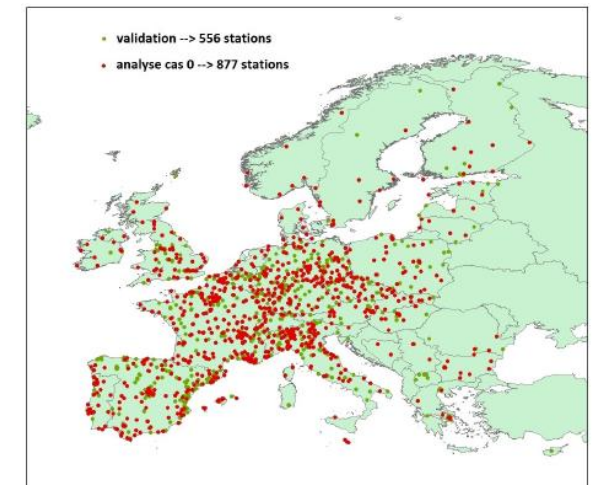
Study carried out on the European scale, O<sub>3</sub> and PM<sub>10</sub>

Stations split into two sets:

1/3 of stations randomly taken out from the different Joly & Peuch classes: used as independent validation stations in all the tests

Different selections of stations taken from the remaining 2/3: used as input in the kriging

- background stations
- stations classified as 1 to 3
- stations classified as 1 to 4
- (...)
- stations classified as 1 to 10



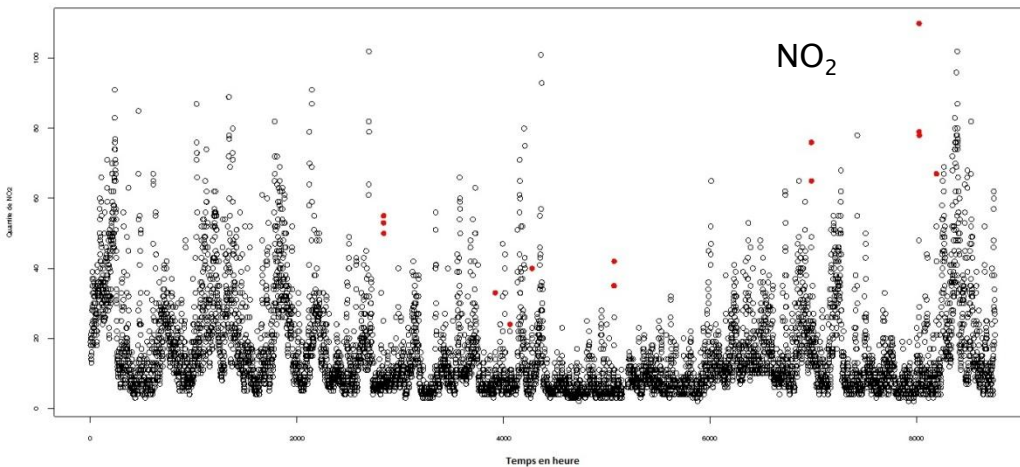
Computation of performance indicators by validation station and on average by class

## □ Detection of outliers

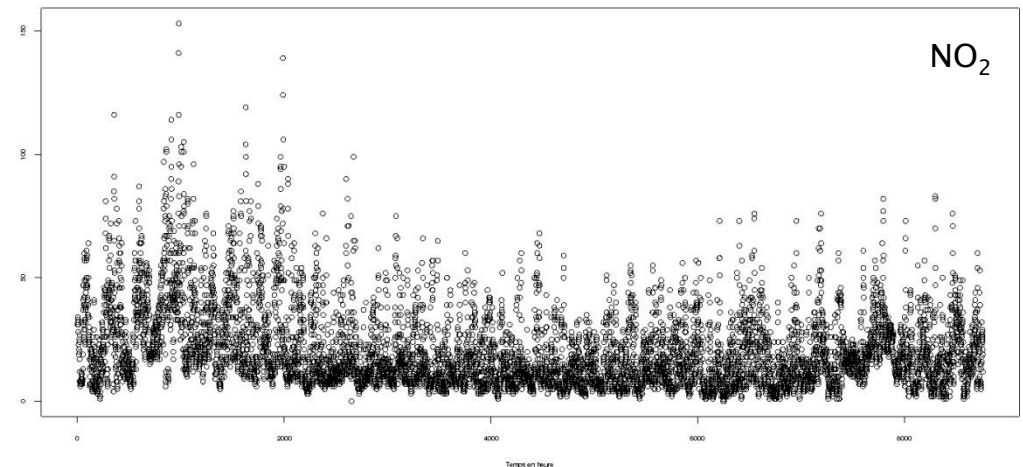
### ➤ Preliminary study

- ✓ Tests performed on AirBase timeseries
- ✓ Adjustment of a method studied by Gherarz et al. (ETC/ACM 2011)
- ✓ Application of a moving window filter (parameters adjusted for each pollutant):

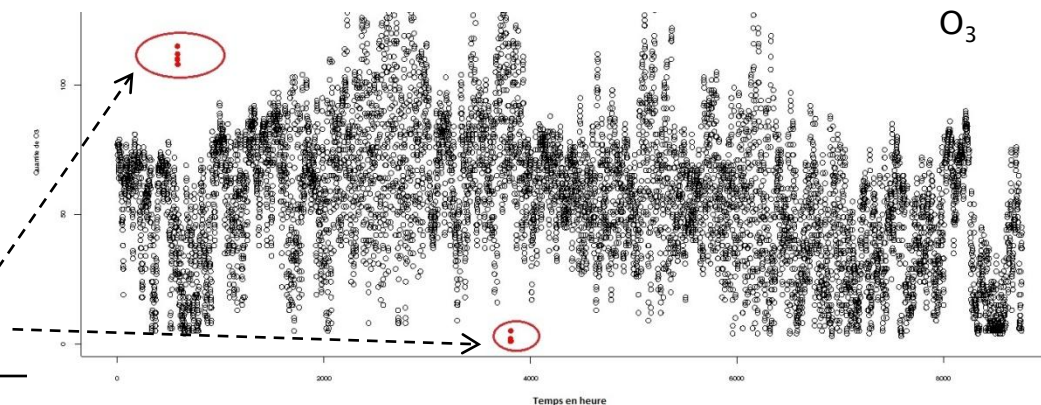
Detection de valeurs aberrantes pour le station FRS2181



Detection de valeurs aberrantes pour le station FRS3005



Artificially modified data



- Support to French local AQ monitoring networks interested in better characterizing station representativeness
- Classification according to Joly and Peuch methodology (2012) :
  - ✓ Get feedback from data providers, e.g. on the stations identified as « outliers » in ETC/ACM 2013 study.
  - ✓ Update of the classification to include more stations.
- Evaluation of CTMs:
  - ✓ Definition of a validation strategy taking the spatial distribution and the classification of stations (AirBase, Joly & Peuch) into account.
  - ✓ Analysis of the model skill scores as a function of the classification. Focus on the model performance for the stations identified as “outliers”.
- Mapping:
  - ✓ Detection of outliers : operational implementation for near-real-time data.
  - ✓ Impact of the selection of stations used in the mapping on the quality of the final maps.