

COOCK | OPEN STAD

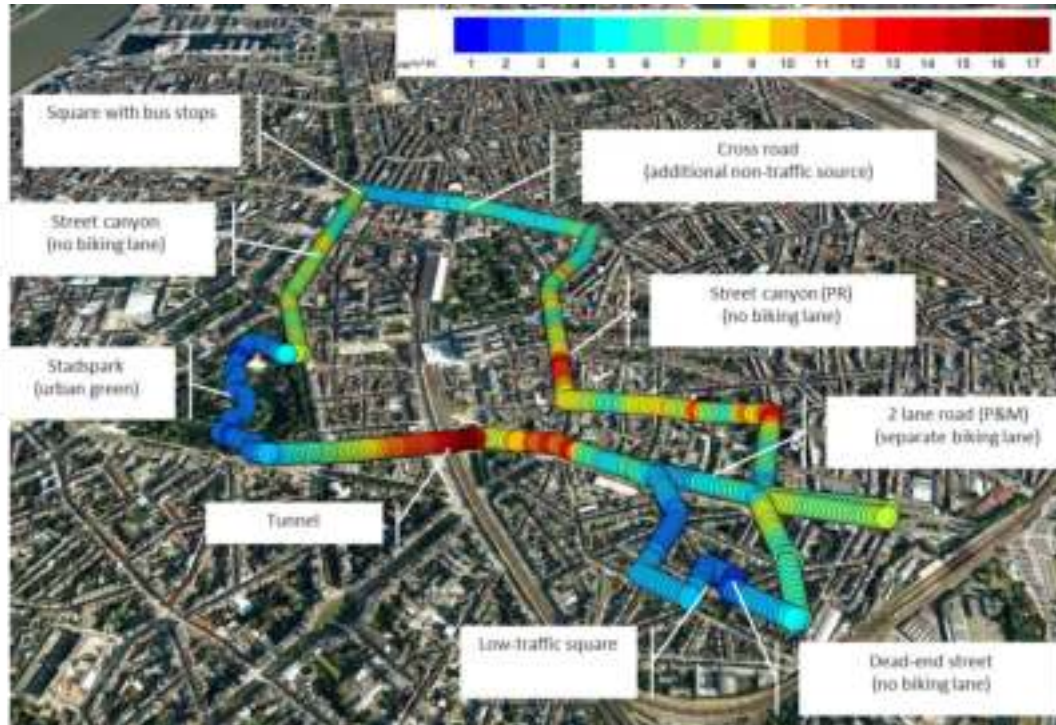


VALUABLE ENVIRONMENTAL INSIGHTS
THROUGH AN OPEN CITY APPROACH

Mobile monitoring using airQmap: data analysis and lessons learned

Jan Peters, Martine Van Poppel - VITO

SPATIAL VARIABILITY OF AQ



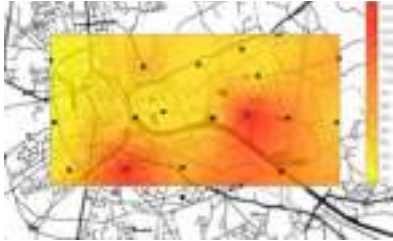
Traditional measurement stations:

- Reference measurements
- Limited locations

Large spatial variability in a city

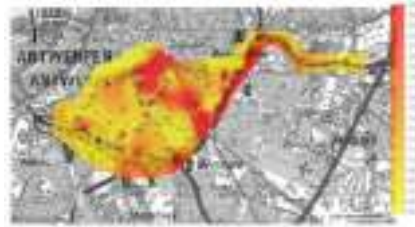
SOME HISTORY OF MOBILE MONITORING AT VITO

2003



5-min measurements + interpolation

2005

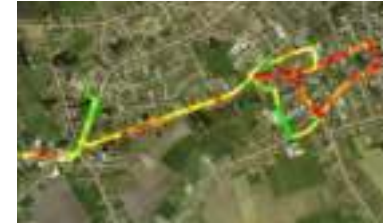


Car as platform + GPS

2010



1s + bike as platform



BC + data processing

today...



airQmap
Aeroflex



useability
data processing
methodology
Visualisation
flexible

THE AIRQMAP APPROACH

AIRQMAP

airqmap is a tool to collect large amounts of mobile BC data and process them into street-level BC exposure maps.



Measurement devices:
GPS and microAeth



Easy-to-use software



Web application
BC maps



Scientific Methodology



Automated dataprocessing

www.airqmap.com

AIRQMAP MEASUREMENT DEVICE AND POLLUTANT

- Black Carbon (BC)

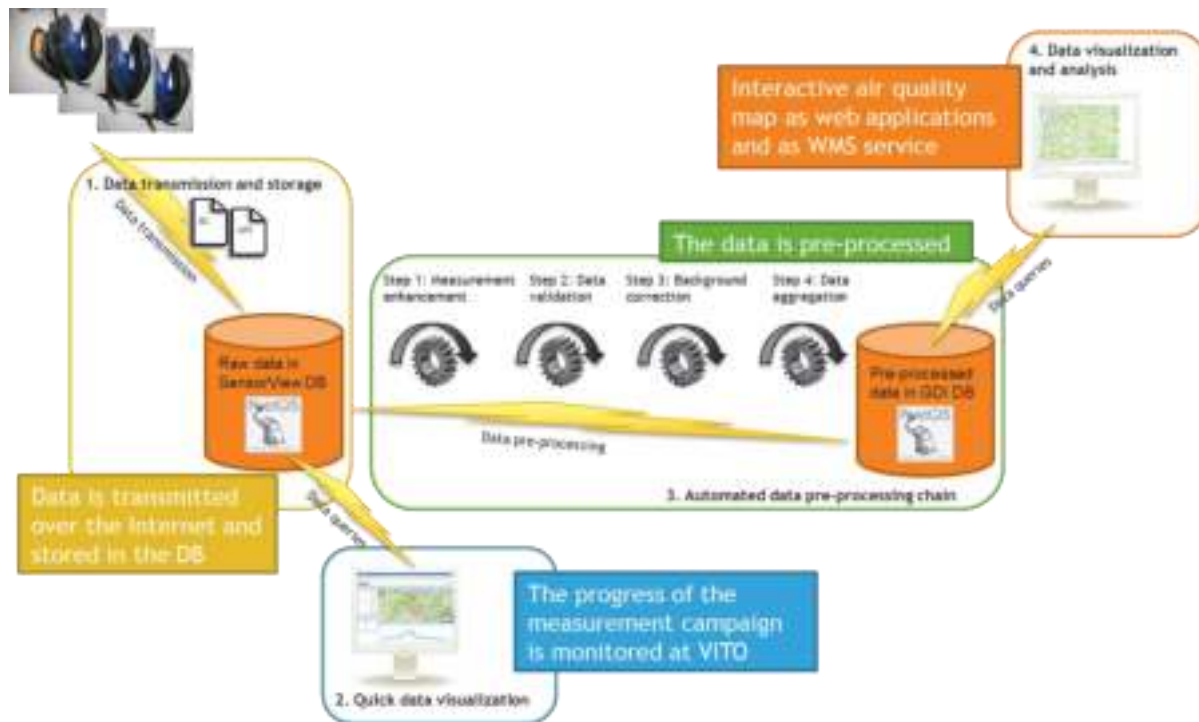
- Indicator for combustion-related aerosol
- Associated with cardiovascular and cardiopulmonary health effects
- WHO claims: “Studies of short-term health effects suggest that BC is a better indicator of harmful particulate substances from combustion sources (especially traffic) than undifferentiated particulate matter (PM) mass.”

- MicroAeth

- Not a sensor but a ‘medium range’ (portable) instrument
- 1 sec time resolution (signal noise – postprocessing)



AIRQMAP DATA PROCESSING

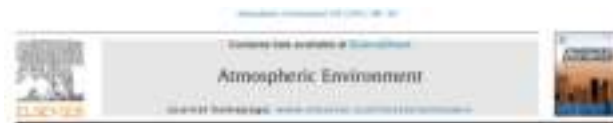


AIRQMAP **STANDARD** METHODOLOGY

- BC measurements
 - Algorithm for noise reduction, outlier removal
- Trajectory
 - Fixed route by bike or foot: representative for exposure of cyclist or pedestrians
- Repeated measurements at fixed times
 - Dependent of the research question e.g. hot spots at peak hours
- Data aggregation over time and space
 - 30m spatial resolution
 - Visualisation of result (no interpolation) : min # of datapoints
- Result
 - Spatial variability, no evaluation of limit values



NUMBER OF REPETITIONS



Mobile monitoring for mapping spatial variation in urban air quality: Development and validation of a methodology based on an extensive dataset

Joels Van den Broecker^{1,2,*}, Jan Peeters³, Jan Vervaeke⁴, Dick Bontadelli⁵, Jan Thoenes⁶, Bernard De Baets⁷

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Table 2

Results of the data experiments: required number of repetitions using a deviation of 25 and 50% at different levels of spatial aggregation: the entire route (route 1), different streets (for abbreviations see Table 1) and 20 and 50 m segments (10th, 50th and 90th percentiles are given), and using different processing methods: arithmetic mean ('standard'), trimmed mean ('trimmed'), with a background normalisation ('background') and the combination of both ('both'). The minimum for each spatial level and for 25 and 50% is shown in *italics*.

| | 25% | | | | 50% | | | |
|---------------|----------|---------|------------|------|----------|---------|------------|------|
| | Standard | Trimmed | Background | Both | Standard | Trimmed | Background | Both |
| Route | 17 | 18 | 11 | 12 | 4 | 5 | 3 | 4 |
| Street level | | | | | | | | |
| PM | 18 | 20 | 13 | 14 | 5 | 6 | 3 | 4 |
| OP | 18 | 19 | 14 | 14 | 5 | 5 | 4 | 4 |
| VO | 63 | 61 | 60 | 57 | 16 | 11 | 15 | 9 |
| KR | 82 | 68 | 75 | 61 | 21 | 10 | 20 | 9 |
| GR | 66 | 50 | 62 | 42 | 18 | 8 | 18 | 7 |
| KA | 44 | 34 | 40 | 25 | 12 | 7 | 12 | 5 |
| 20 m segments | | | | | | | | |
| 10% | 31 | 30 | 26 | 24 | 8 | 6 | 7 | 5 |
| 50% | 59 | 54 | 52 | 48 | 15 | 10 | 13 | 8 |
| 90% | 164 | 112 | 156 | 108 | 44 | 13 | 43 | 12 |
| 50 m segments | | | | | | | | |
| 10% | 33 | 31 | 29 | 24 | 8 | 7 | 6 | 5 |
| 50% | 57 | 52 | 50 | 42 | 14 | 10 | 13 | 8 |
| 90% | 143 | 102 | 141 | 94 | 39 | 13 | 40 | 11 |

25% and 50% Standard deviation

=>Data processing

=>Spatial resolution

AIRQMAP

- The airQmap approach can be used to analyze mobile data in smart city context
- Other pollutants
 - Some tests performed with PM sensors, UFP
- Other platform
 - Some tests with a car/mobile van
- Opportunistic measurements
 - Some tests with city guards
- ...

CHALLENGES OF MOBILE MEASUREMENTS

IMPACT OF DAY

25/10 (weekday)



7/11 (weekday)

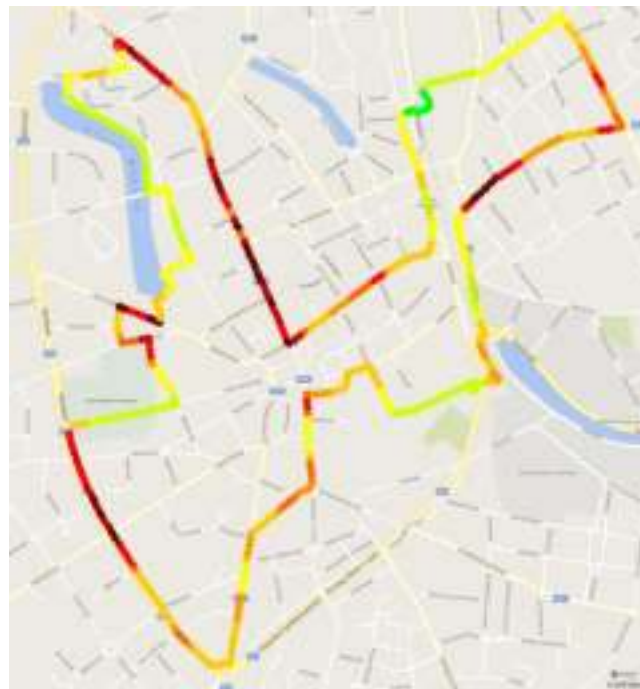


IMPACT OF TIME OF DAY

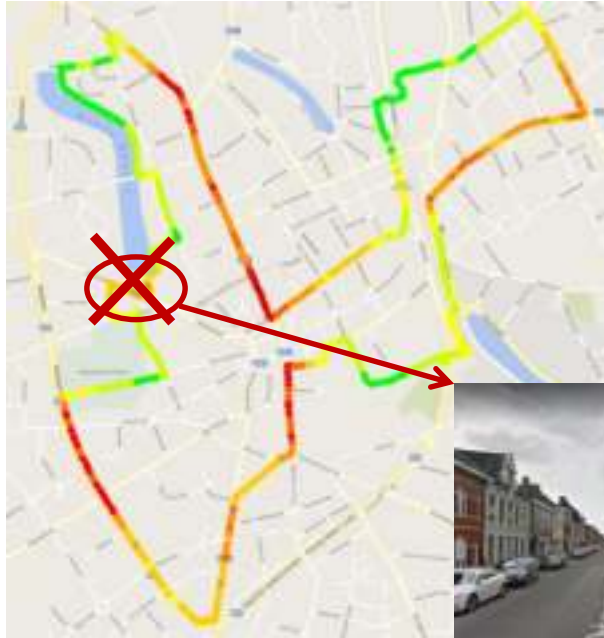
before noon



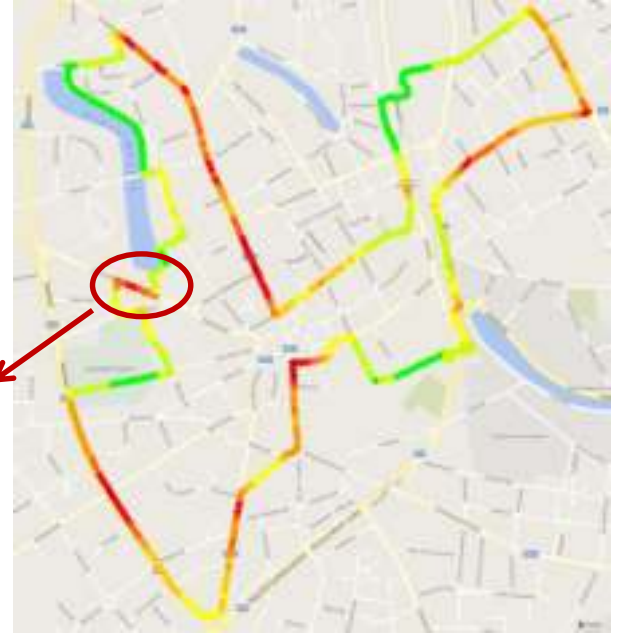
afternoon



TIME-SPACE SYNCHRONIZATION, RESPONSE OF SENSORS

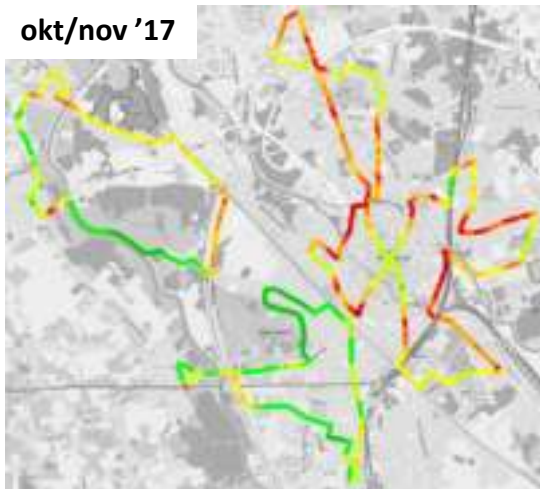


Aggregated map (#: 25 runs)

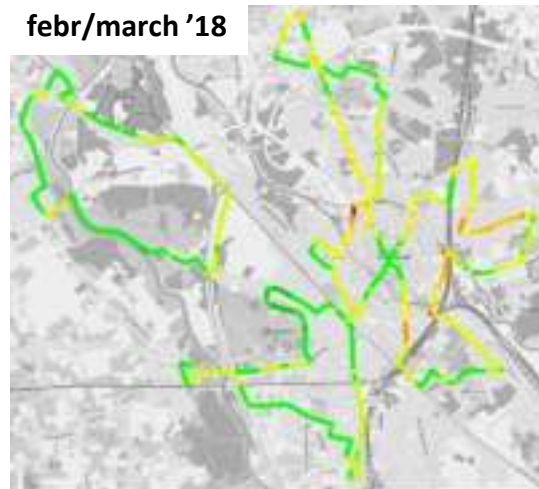


SEASONAL DIFFERENCES

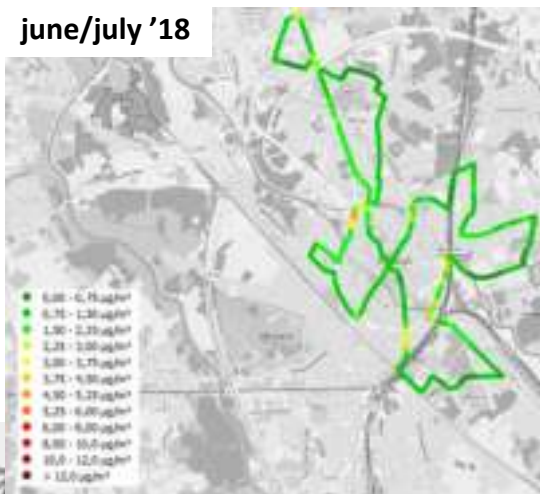
okt/nov '17



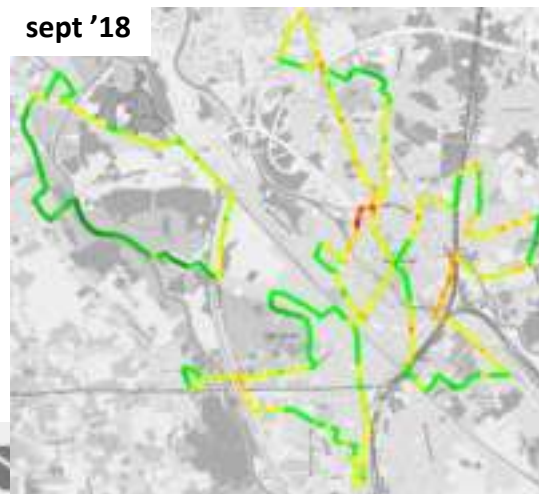
febr/march '18



june/july '18



sept '18



SEASONAL DIFFERENCES

- Summary statistics per campaign and BC concentrations at fixed monitoring stations

| | Min. | 1 st Qu. | Median | Mean | 3 rd Qu. | Max. | VMM virtual Traffic station |
|--------------------------|------|---------------------|--------|------|---------------------|------|--------------------------------|
| Campaign 1 | 1.0 | 2.4 | 3.3 | 3.7 | 4.7 | 12.4 | 3.3 |
| Campaign 2 | 1.3 | 2.0 | 2.3 | 2.5 | 2.9 | 10.5 | 2.4 |
| Campaign 3 | 0.5 | 0.9 | 1.2 | 1.4 | 1.7 | 6.2 | 1.4 |
| Campaign 4 | 0.5 | 1.8 | 2.5 | 2.5 | 3.1 | 8.3 | 2.5 |
| Yearly average 2017: 1.9 | | | | | | | |

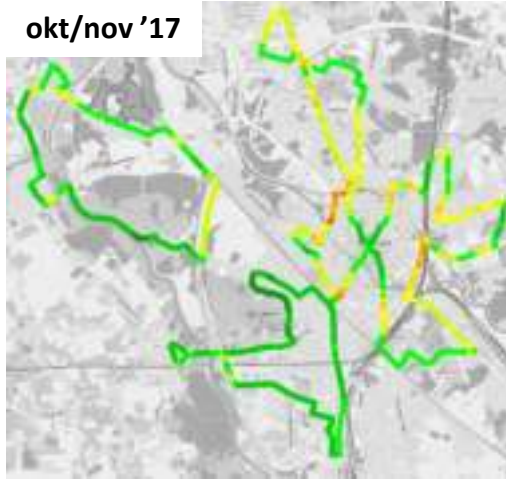
→ Max concentration

→ Min concentration

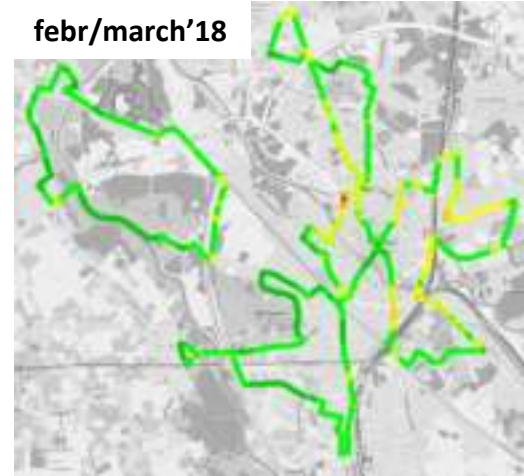
- Rescaling** based on data on ‘virtual’ traffic station
ratio = yearly average/campaign average @VMM traffic station
- Aggregation** of different campaigns based on weighed average
(weighing factor ~ frequency of occurrence of the average BC concentration)

SEASONAL DIFFERENCES

okt/nov '17



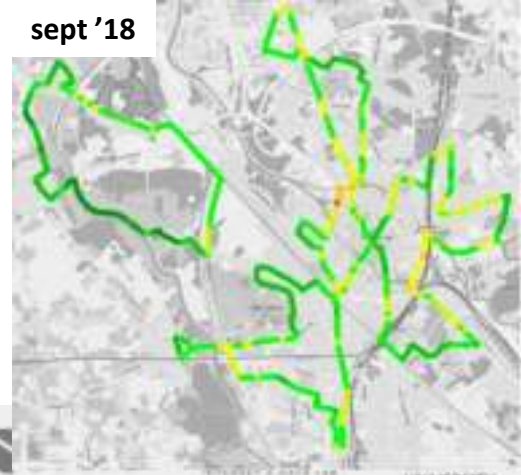
febr/march '18



june/july '18



sept '18



MEASUREMENT SET-UP

OPPORTUNISTIC VERSUS DEDICATED MEASUREMENTS

DEDICATED MEASUREMENTS VERSUS OPPORTUNISTIC APPROACH

- Dedicated measurement set-up
 - Measurement route is defined (in function of question)
 - At specific time(s) of day
 - Gives a representative picture of these moments in time
 - Easy to compare different locations
 - Additional effort to collect data
- Opportunistic measurements
 - Measurements are performed using a 'platform'
 - Collection of spatio-temporal data
 - Additional effort/attention to data processing
 - Risk of sampling bias



DATA PROCESSING OF OPPORTUNISTIC MEASUREMENTS

- Data collection and aggregation
 - Based on opportunistic measurements (city wardens)
 - 50 m road segments, 1h time
- Land Use Regression model (annual average)
 - Spatial predictor variables: traffic, land use, street geometry,...
 - Temporally adjusted BC values (annual average) additive and multiplicative method
- Spatio-temporal Land Use Regression model
 - Time-dependent variables as predictor variables
 - Real time dynamic pollution map: model update based on recent collected data



Development and evaluation of land use regression models for black carbon based on bicycle and pedestrian measurements in the urban environment

Jero Van den Broeke^{1,2,*}, Bram De Baets¹, Jan Vermeiren¹, Dick Beeldens^{1,3}, Jan Thieme¹

¹ Ghent University, Ghent, Belgium; ² Ghent University, Ghent, Belgium; ³ Ghent University, Ghent, Belgium

*Corresponding author: jero.vandenbroeke@ugent.be



A spatio-temporal land use regression model to assess street-level exposure to black carbon

Jero Van den Broeke^{1,2,*}, Bram De Baets¹, Dick Beeldens^{1,3}, Jan Thieme¹

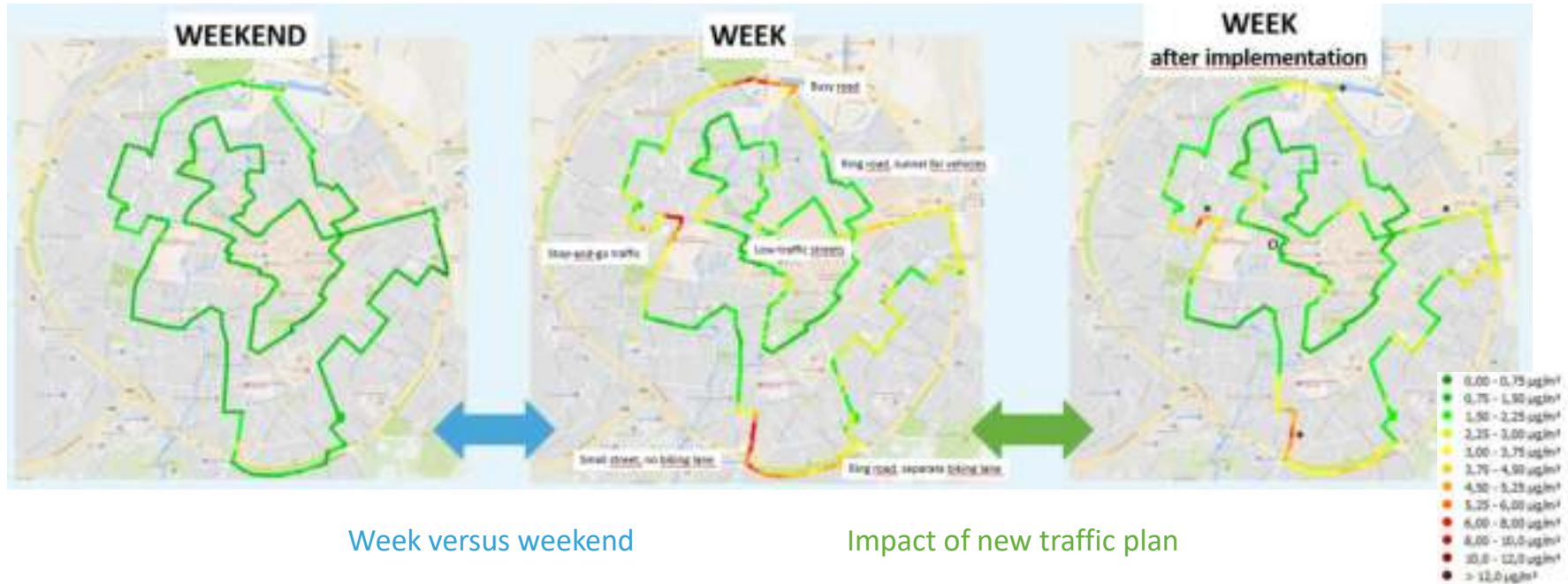
¹ Ghent University, Ghent, Belgium; ² Ghent University, Ghent, Belgium; ³ Ghent University, Ghent, Belgium

*Corresponding author: jero.vandenbroeke@ugent.be

WHAT CAN WE DO WITH MOBILE MEASUREMENTS: SOME CASE STUDIES

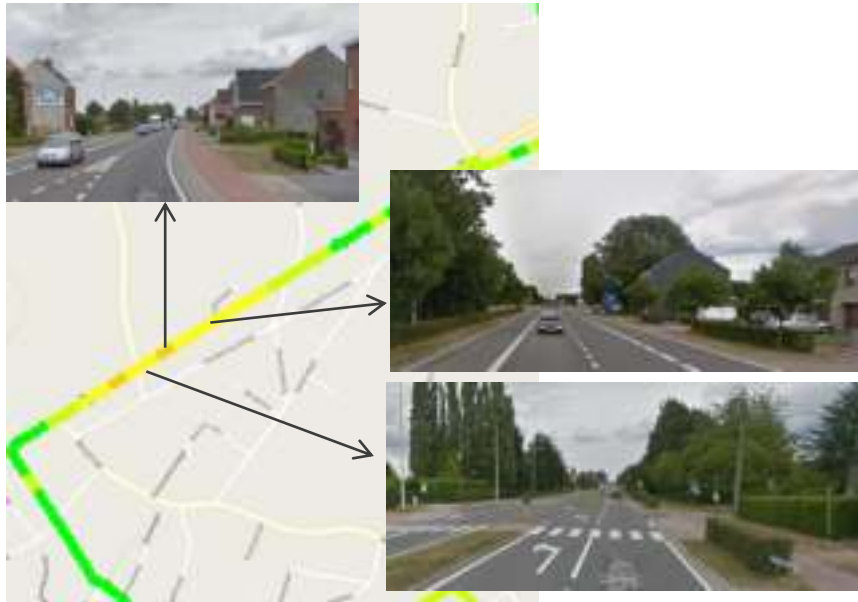
IMPACT OF NEW TRAFFIC PLANS

Impact of traffic plans, evaluated with **airOmap** by citizens in Leuven



IMPACT OF STREET INFRASTRUCTURE ON EXPOSURE

Impact of orientation of biking path



Impact of street characteristics: ventilation and distance

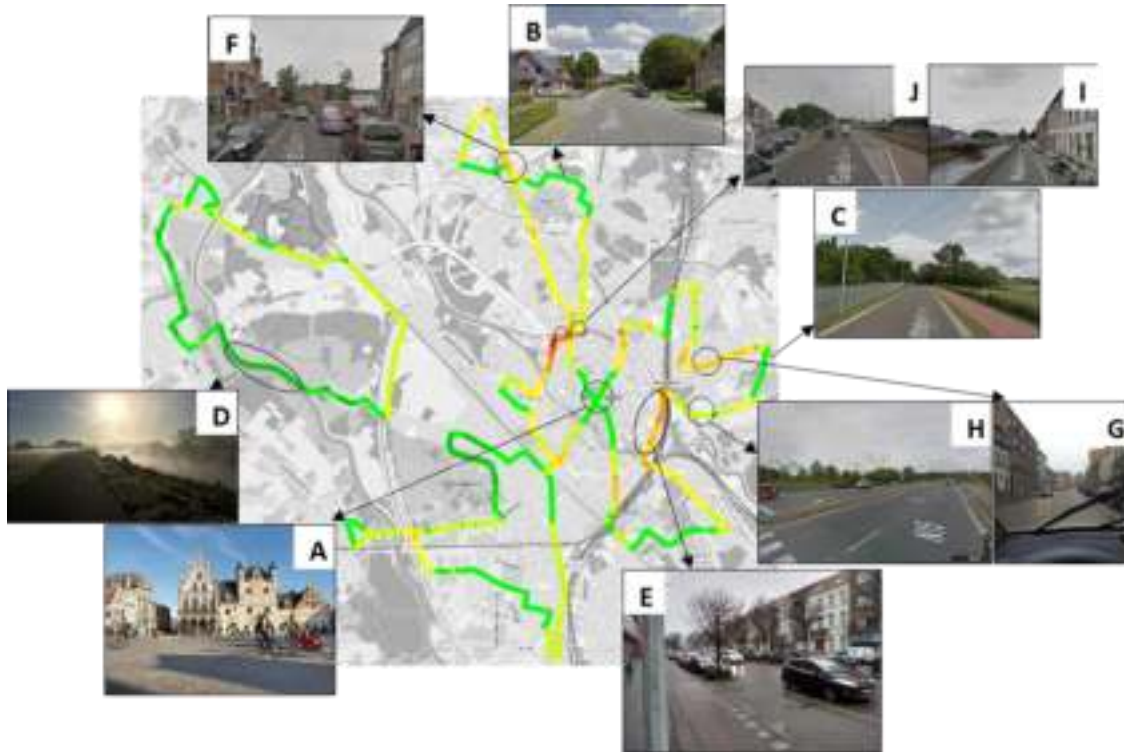


INTERACTIONS WITH CITIZENS

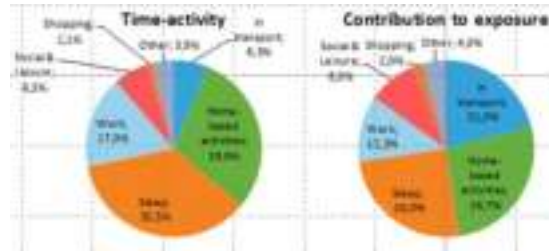
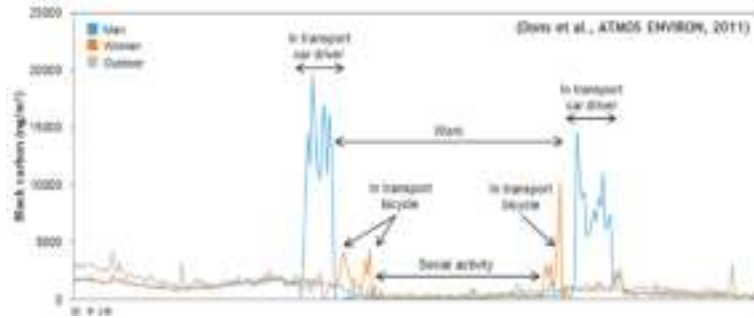
- Citizen Observatory (CO) “Meet Mee Mechelen”
- As part of two Citizen science projects GroundTruth 2.0 and Flamenco
- Involve citizens in AQ measurements and management
- A good data interpretation and feedback is needed
 - Low concentrations are easier to explain than high ones 😊 ?
 - It's all about the scale!



DETECTION OF HOT SPOTS/IMPACT OF TRAFFIC ON AQ



PERSONAL EXPOSURE TO AIR POLLUTION



IMPACT OF WOOD BURNING



Impact of wood burning on AQ by mobile measurements:

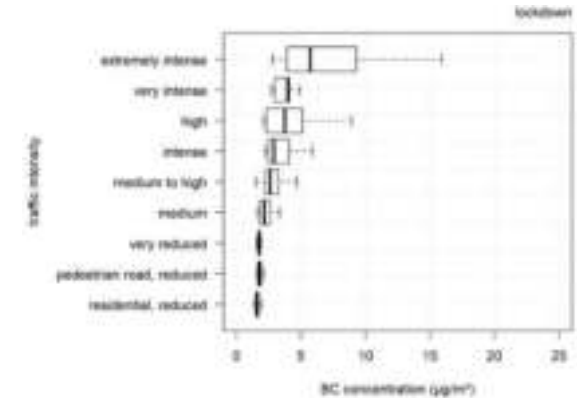
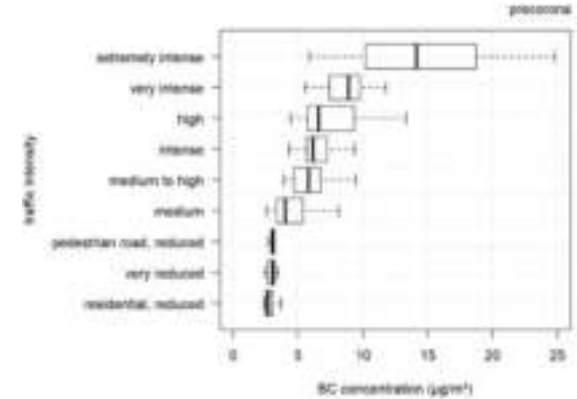
- BC at different wavelengths:
 - BC(WB) : wood burning
 - BC(FF) : fossil fuel

Result show:

- different patterns for wood burning compared to fossil
- Sources identified

IMPACT OF REDUCED TRAFFIC DUE TO COVID 19

- Impact of traffic due to COVID-19
- Case study Romania
- Background correction
 - Concentration at park



CONCLUSIONS

- A mobile measurement approach can be used to assess local differences in air quality
- Repeated measurements are needed to get representative results
- Data processing needs to take into account different background concentrations
 - When collecting data at different time at different locations
 - When comparing scenarios
- Data collection process needs attention
- Setting up ICT infrastructure and Apps is important in data-collection and processing
- Selection of sensors and parameters
 - Different sensors with different performance, calibration
 - What are we looking for? What do we want to measure?

NEXT STEPS

- Expertise of airQmap and methodology can be applied in smart city mobile applications
- Apply knowledge for use cases:
 - Set-up
 - Data processing
 - Data interpretation
 - Calibration
- Contact:
martine.vanpoppel@vito.be – jan.peters@vito.be