

European Network on New Sensing Technologies for Air Pollution Control and Environmental Sustainability - *EuNetAir*

COST Action TD1105

WGs & MC Meeting at SOFIA (BG), 16-18 December 2015

New Sensing Technologies for Indoor Air Quality Monitoring: Trends and Challenges

Action Start date: 01/07/2012 - Action End date: 30/04/2016 - Year 4: 1 July 2015 - 30 April 2016

High-resolution mapping of urban air quality using low-cost sensors

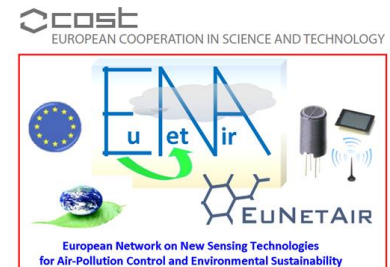


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Making sense of crowdsourced observations: High-resolution mapping of urban air quality using low-cost sensors

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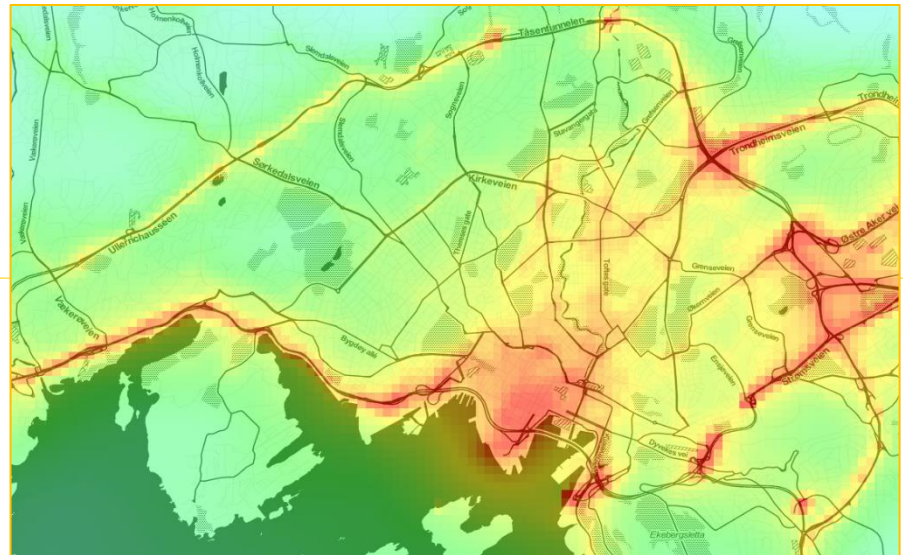
Joris van den Bossche²

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¹NILU – Norwegian Institute for Air
Research

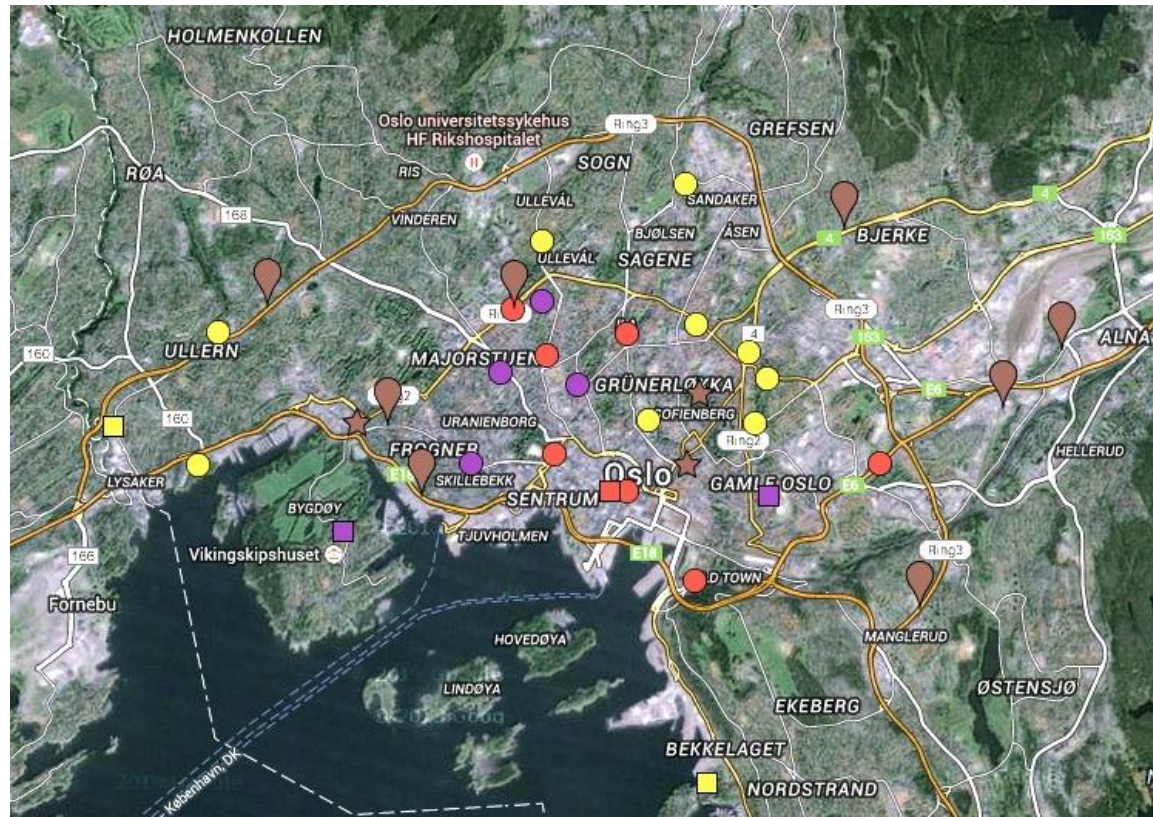
²VITO – Flemish Institute for Technological
Research



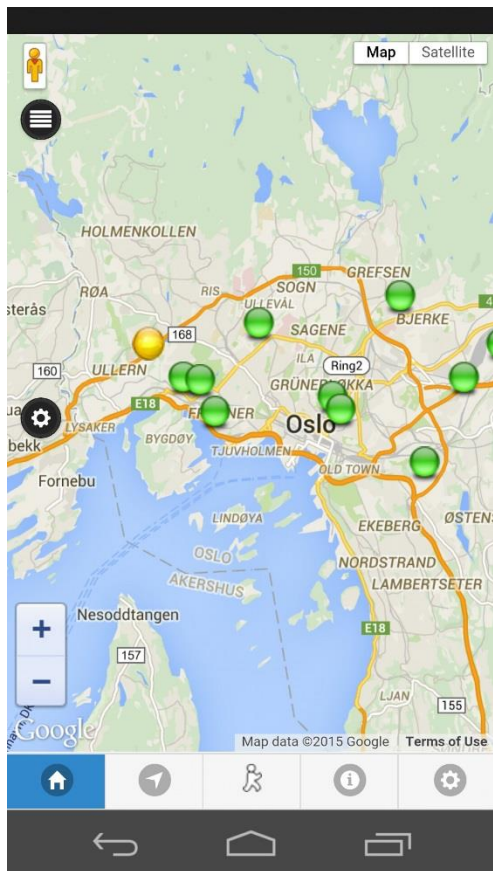
Static observation network

- AQMesh by Geotech
- Wireless air quality monitor
- Measures a variety of pollutants: NO, NO₂, O₃, CO, SO₂, PM₁₀, PM_{2.5}, as well as temperature, humidity, and pressure
- Compares reasonably well with reference equipment (but dependent on species)

Geotech
AQMesh

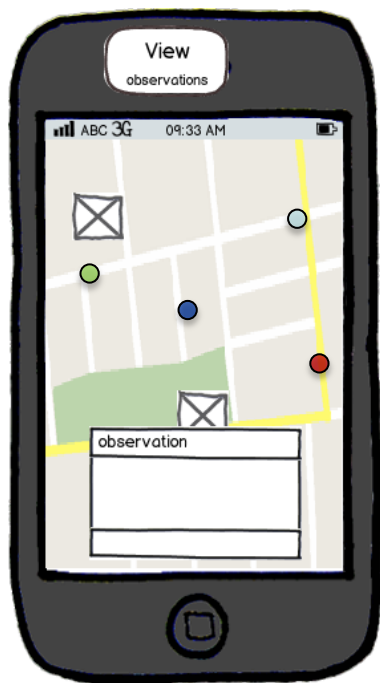


Point-based observations: The problem

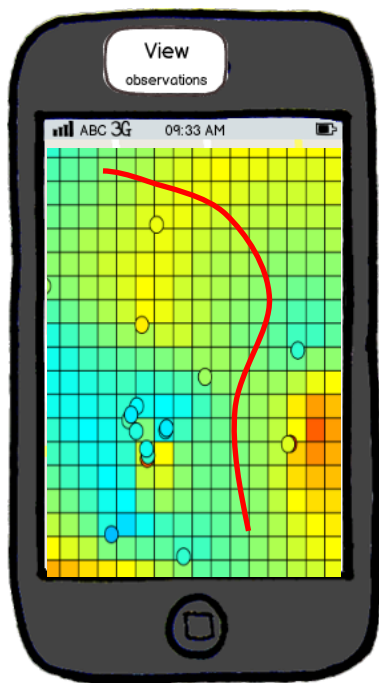


CITI-SENSE app: Android-based mobile app for real-time AQ monitoring

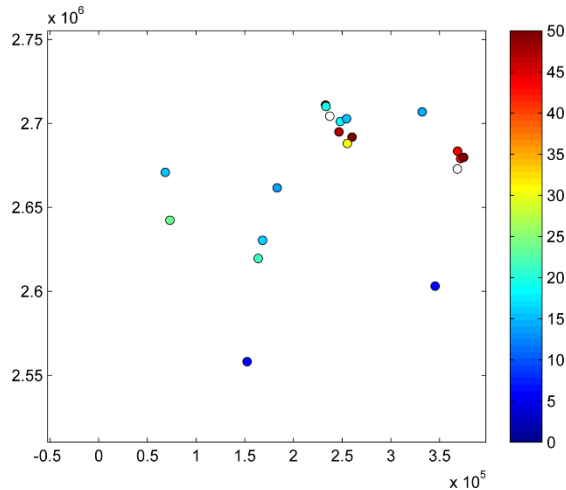
How do we get from this:



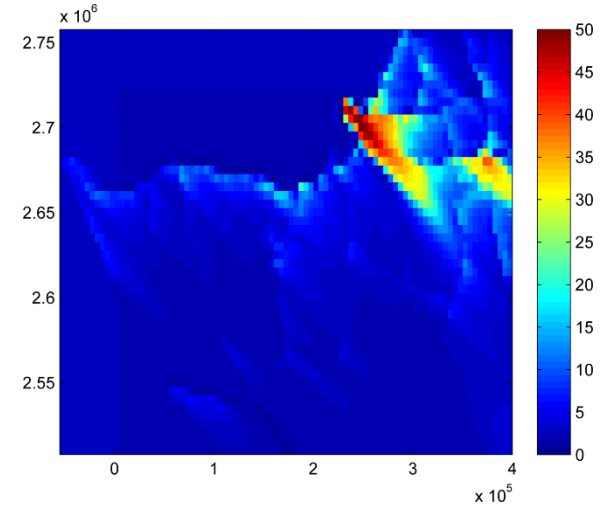
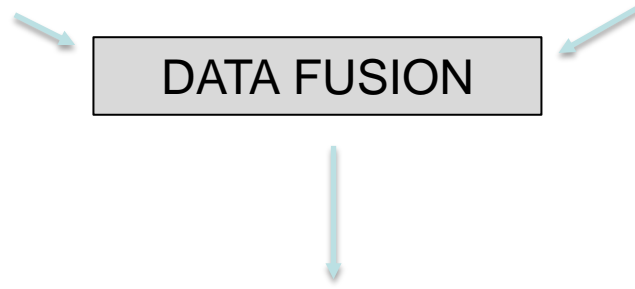
To this:



Data fusion: Basic Premise



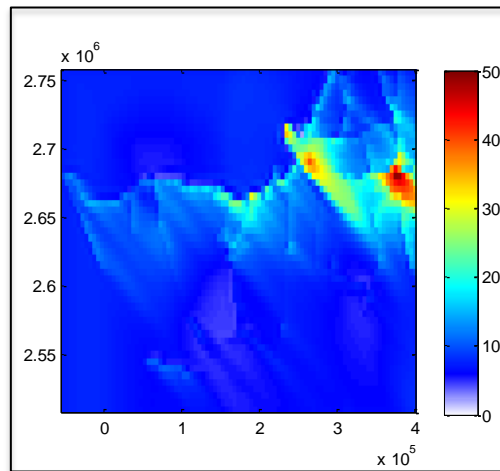
Observations



Modelling results or other auxiliary data

Data fusion (as a subset of data assimilation) creates a value-added product by

- Interpolating the observations in an objective way
- “correcting” the model estimates with true observations

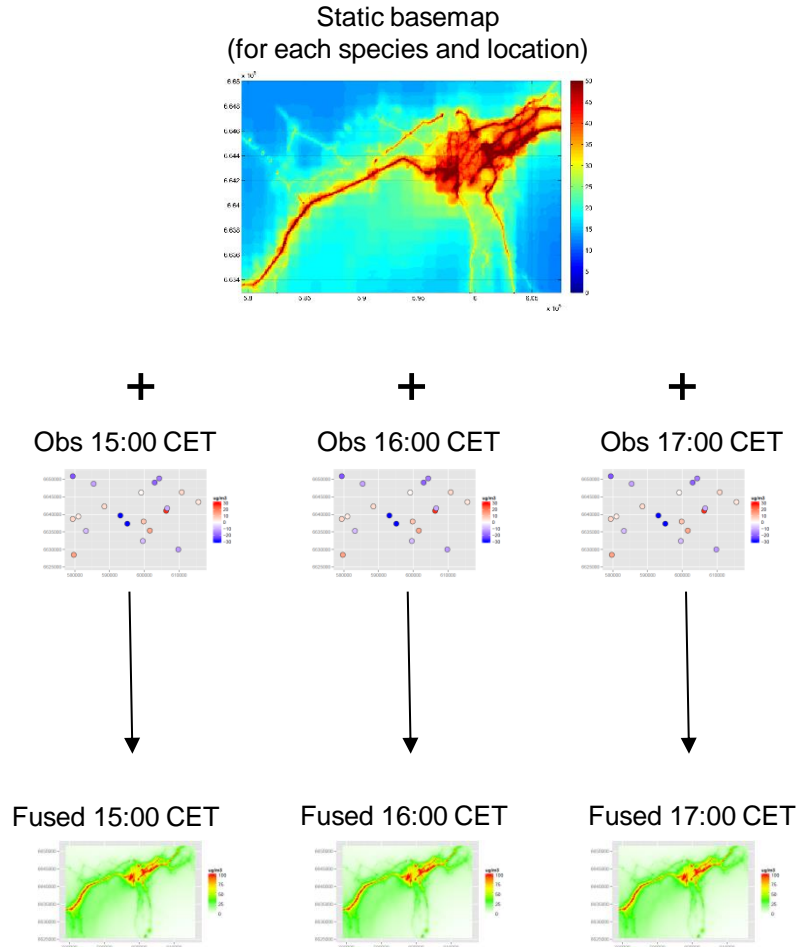


Combined map

Data fusion method used here provides a combined concentration field by regressing the observations against model data and spatially interpolating the residuals

Data fusion for CITI-SENSE

- A static basemap is created for each location and each species of interest to show the long-term spatial patterns
- This basemap is then modified according to the observations made by the static Geotech sensors
- This is essentially a location-dependent level-shift of the basemap
- The final result are hourly maps with the current best guess for the $\text{NO}_2/\text{PM}_{10}/\text{PM}_{2.5}$ concentration field at all CITI-SENSE locations



Basemap:
Provides information about general spatial patterns

Geotech observations:
Provide information about current state of atmosphere at a few sampling locations

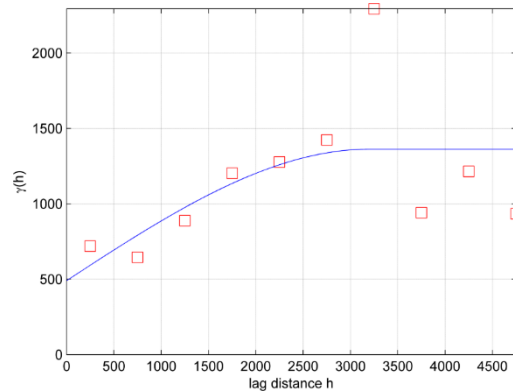
Fused map:
Value-added product providing a best guess of current state of atmosphere for the entire domain

Data fusion methodology

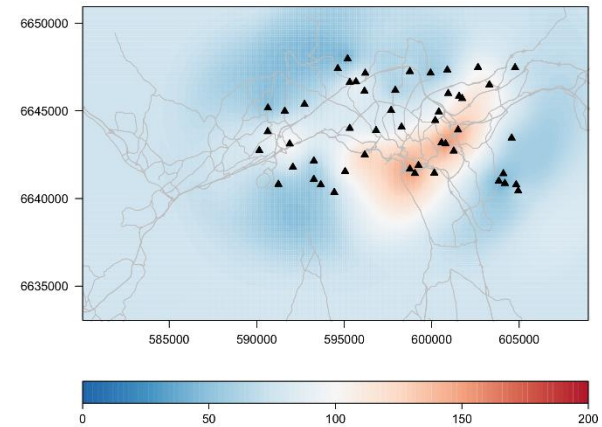
- Data fusion is a subset of data assimilation techniques (Lahoz and Schneider, 2014)
- Uses geostatistical framework
- Analysis performed entirely in log-space
- Universal kriging approach
- Spatial interpolation guided by proxy
- Explicit automated modelling of spatial autocorrelation

Lahoz, W. A., and P. Schneider (2014), Data assimilation: making sense of Earth Observation, *Front. Environ. Sci.*, 2(16), 1–28, doi:10.3389/fenvs.2014.00016.

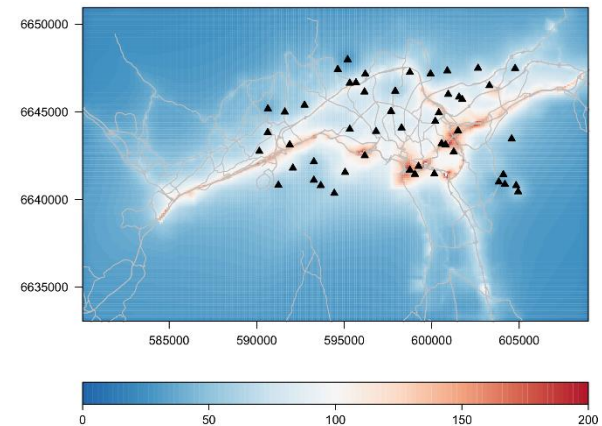
Theoretical model of spatial autocorrelation



Using simple spatial interpolation

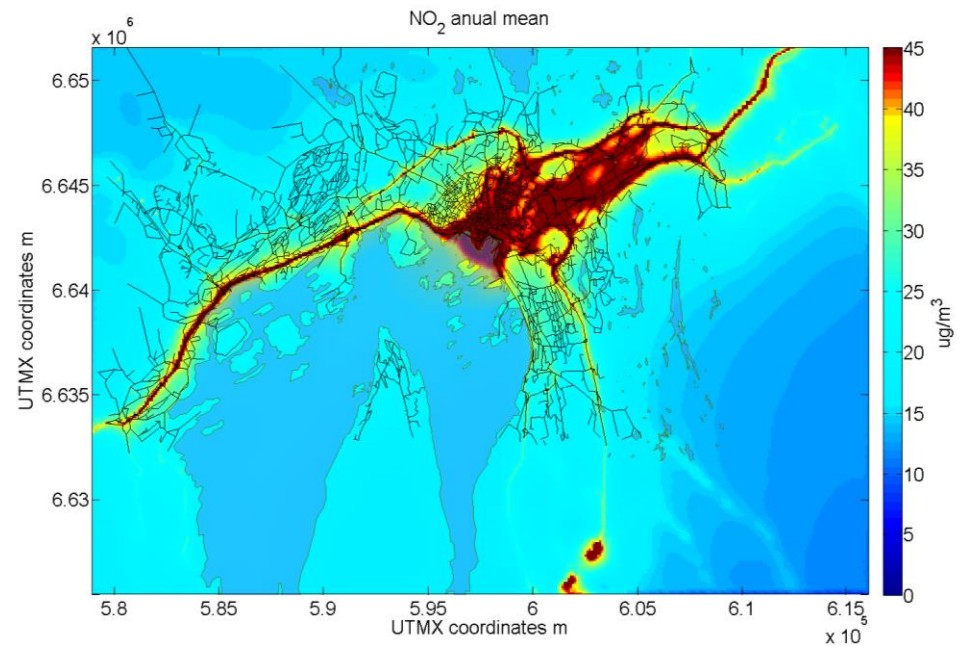


Using data fusion with spatial proxy



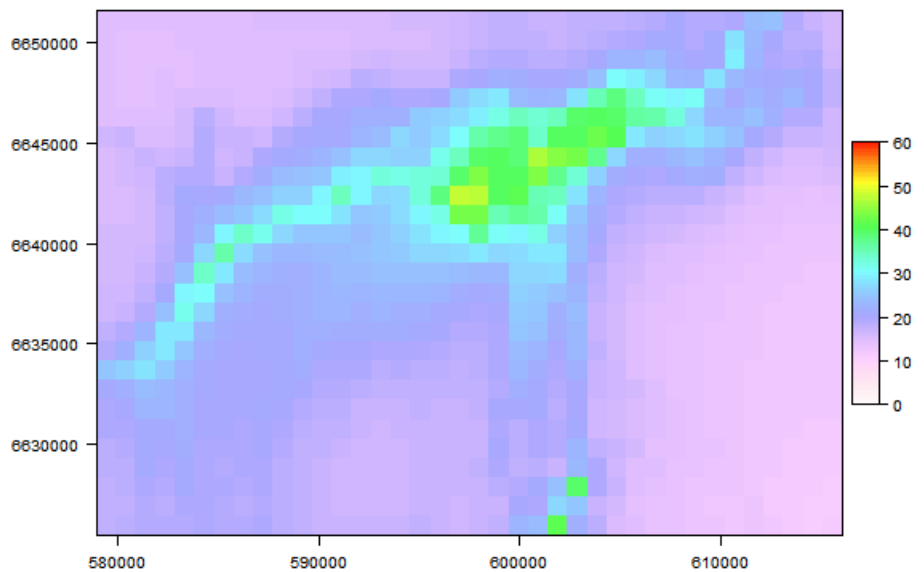
Modelling of the basemaps

- Can be nearly any spatially exhaustive dataset that is related to the observation
- Best to use are urban-scale dispersion models
- Alternatively concentration map created through LUR modelling
- We use the EPISODE model
 - Three-dimensional, combined Eulerian/Lagrangian air pollution dispersion model, developed at NILU
 - Combined modelling and postprocessing approach to obtain basemaps at 10-100 m spatial resolution

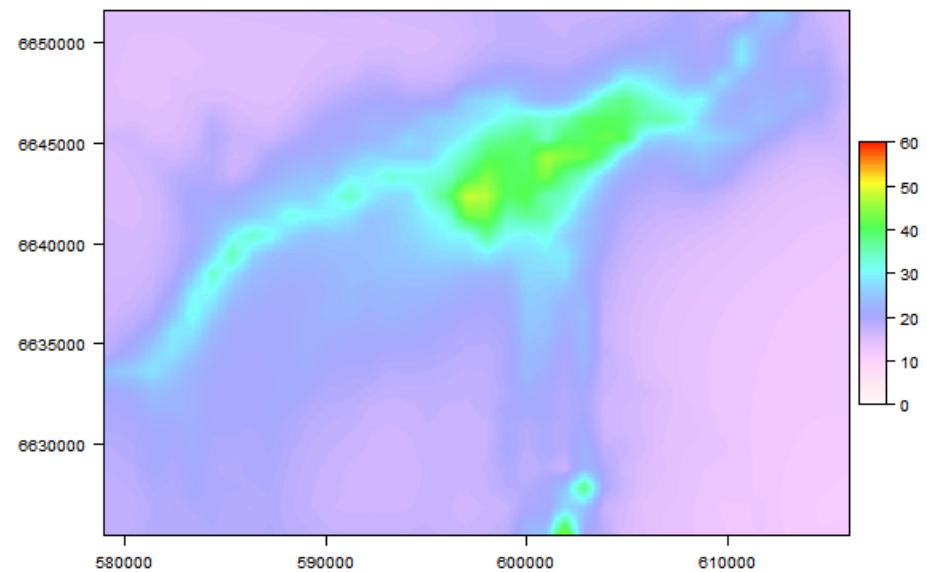


High-resolution map of NO₂ in Oslo from the EPISODE dispersion model. These kind of maps are ideally suited as a spatially distributed auxiliary dataset.

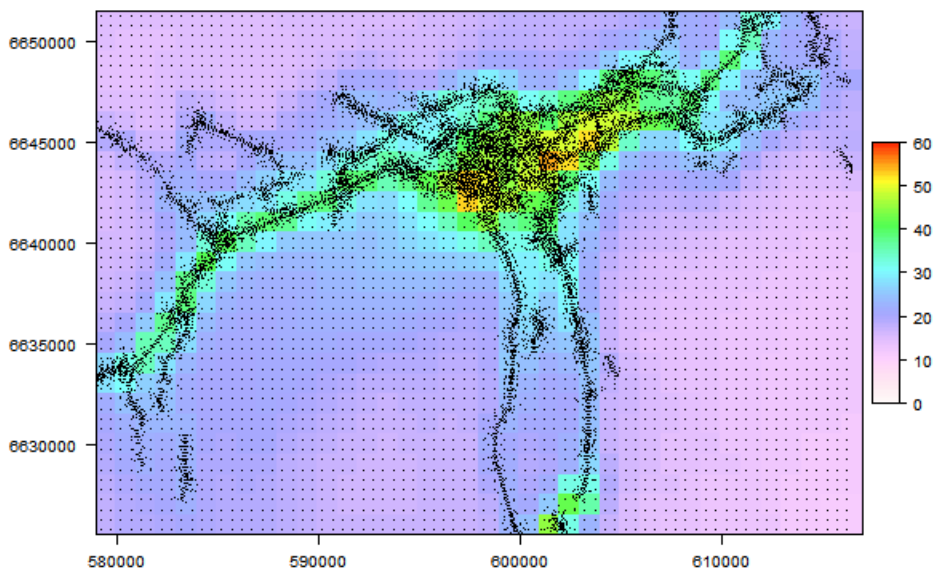
Original EPISODE gridded output



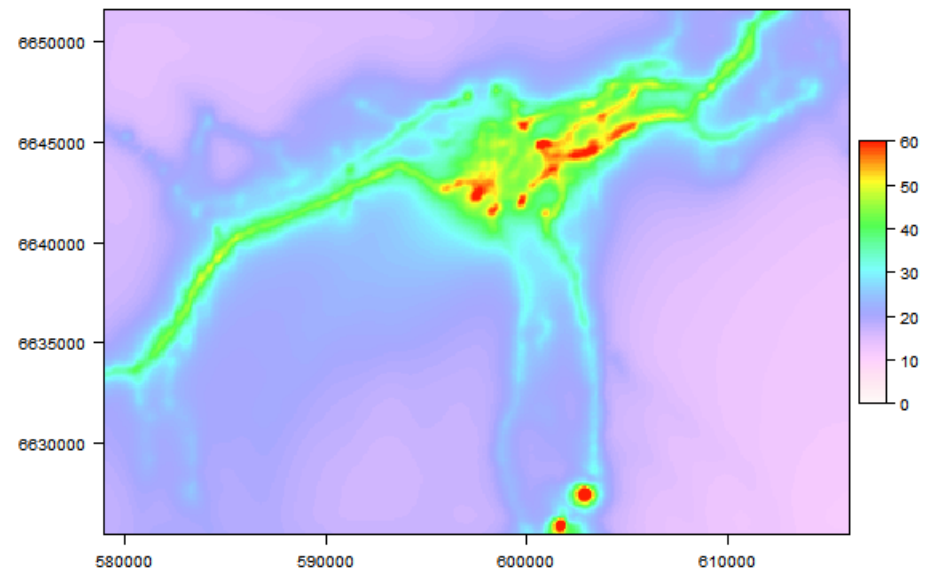
Simple linear interpolation



Distribution of receptor points



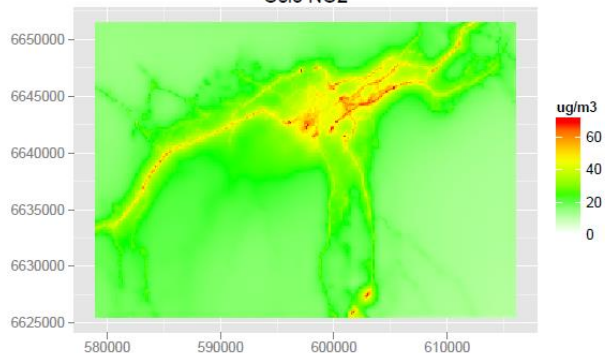
Downscaled using concentrations at receptor points



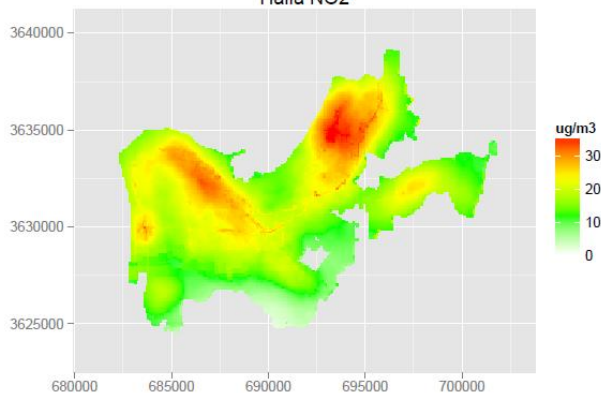
Receptor-point based downscaling of the gridded EPISODE output

Example basemaps for NO₂

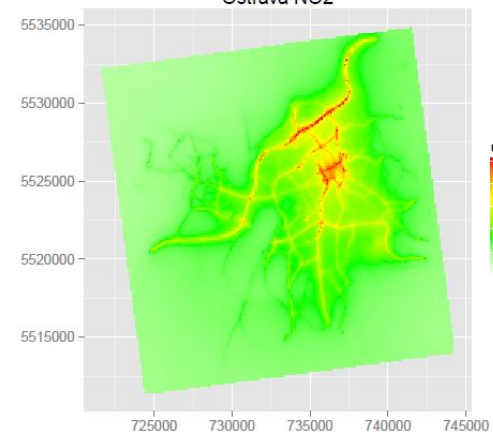
Oslo NO₂



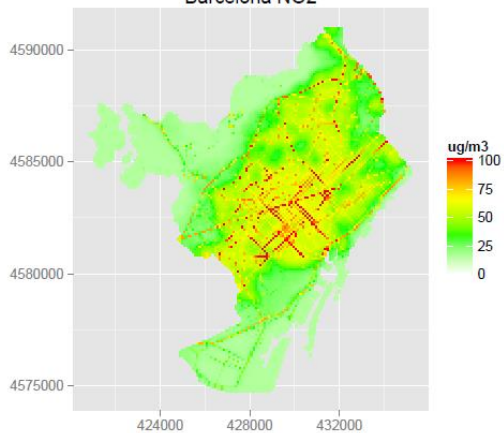
Haifa NO₂



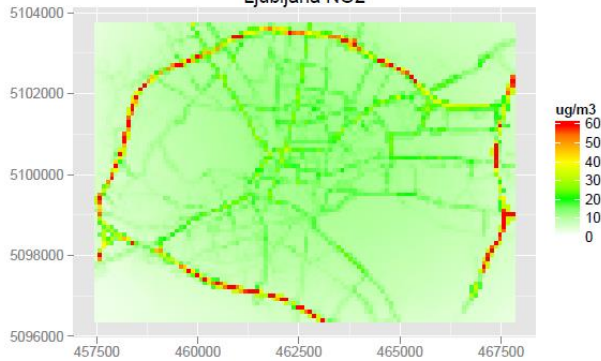
Ostrava NO₂



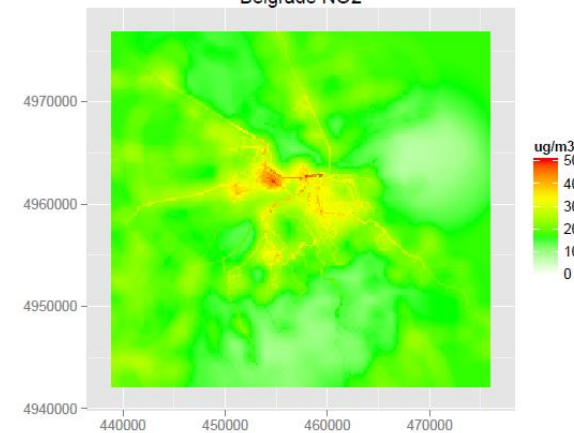
Barcelona NO₂



Ljubljana NO₂

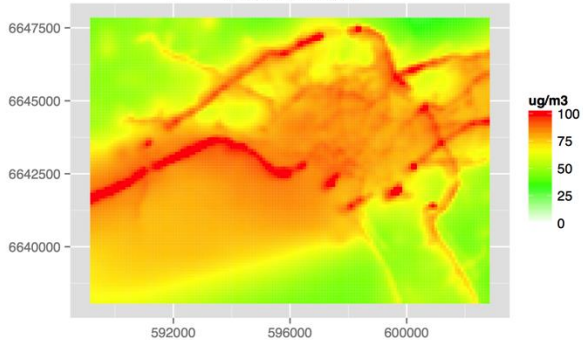


Belgrade NO₂

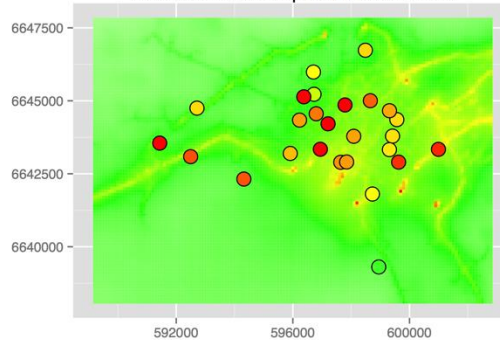


A fusion example for Oslo

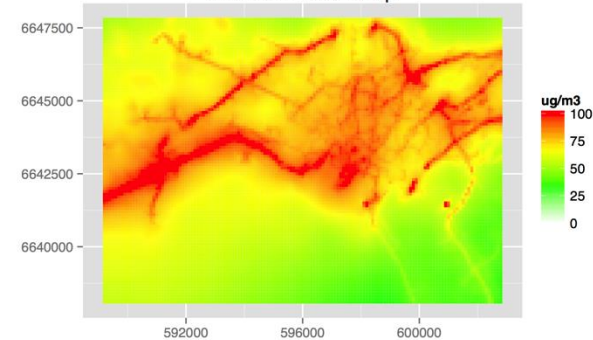
Oslo NO2: Truth



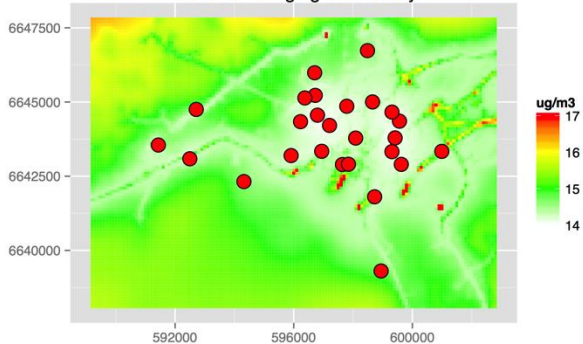
Oslo NO2: Basemap and Observations



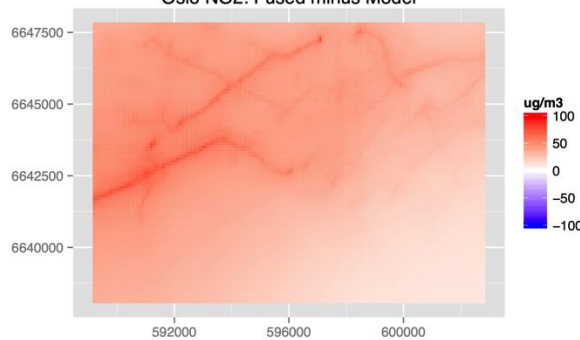
Oslo NO2: Fused map



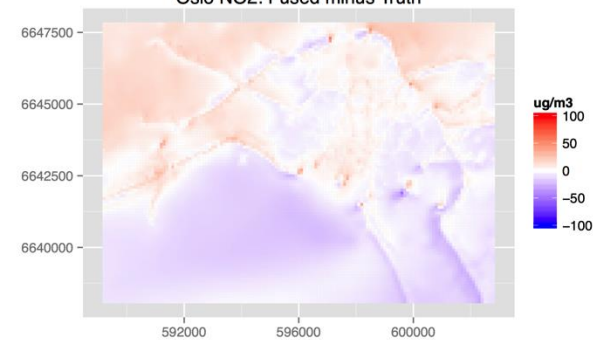
Oslo NO2: Kriging uncertainty



Oslo NO2: Fused minus Model



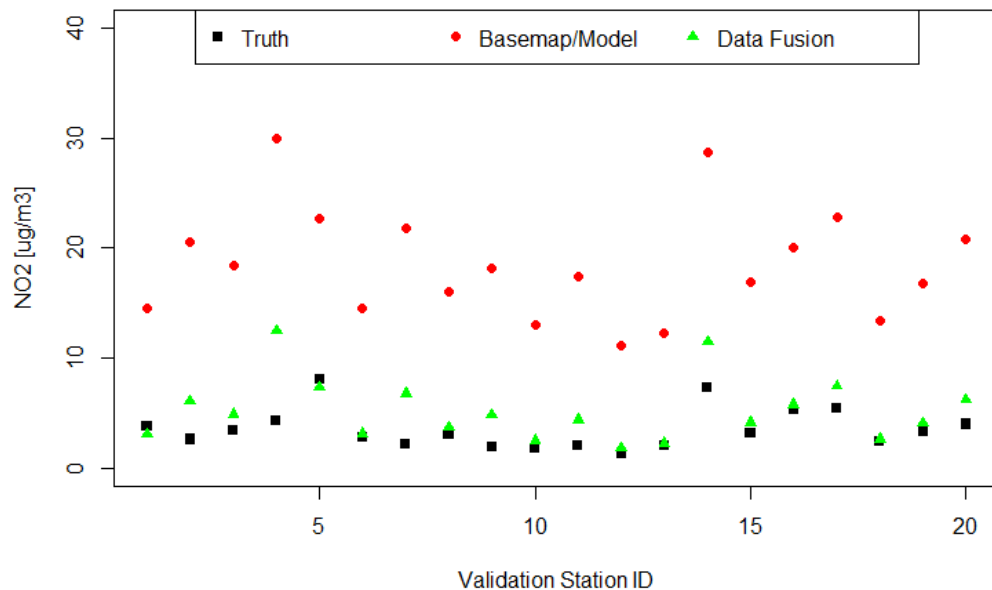
Oslo NO2: Fused minus Truth



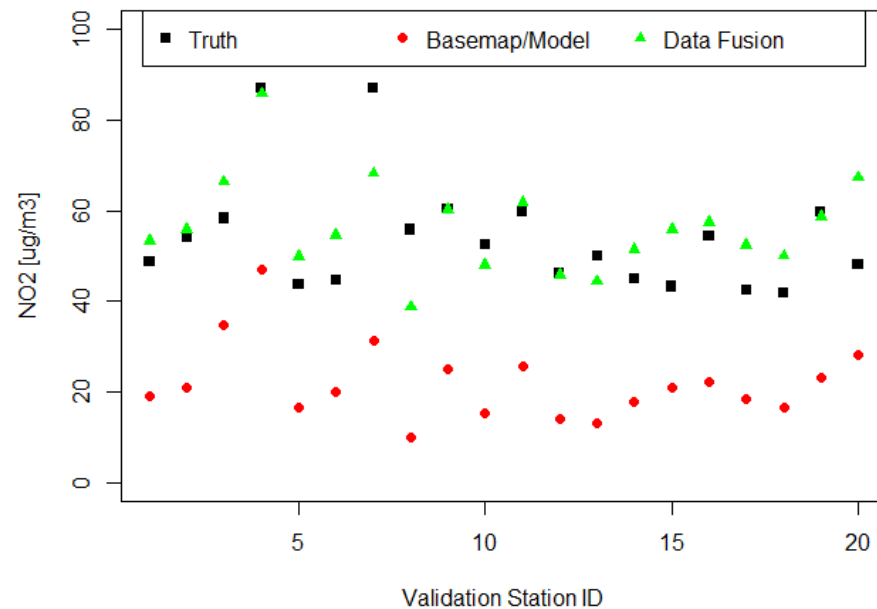
Validation against “Truth”

Validation sites are randomly selected throughout the image. Concentration values at these sites can be extracted from the truth, the basemap, and the fused result and compared.

Example 1: Basemap overestimates “truth”



Example 2: Basemap underestimates “truth”



This shows that the method can predict the true concentration field quite well even in areas where no observations are available.



Typical example of a data fusion-based surface concentration field of NO₂ for Oslo, Norway, at 100 m spatial resolution.

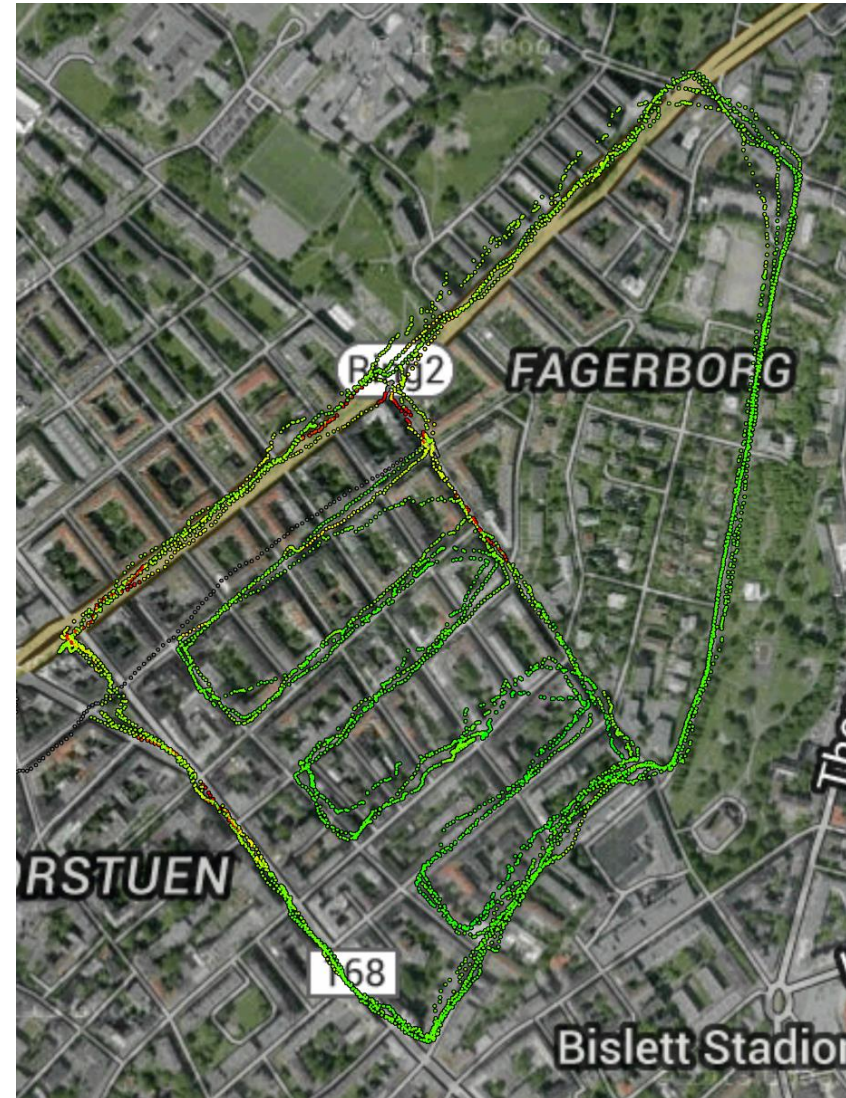
Urban AQ mapping using mobile measurements of Black Carbon



AethLabs
Microaethalometer

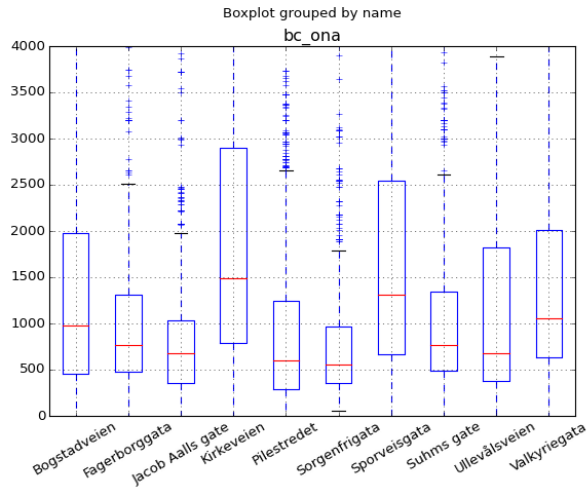


Black Carbon
measurements
campaign in
Majorstuen/Oslo,
May 2015

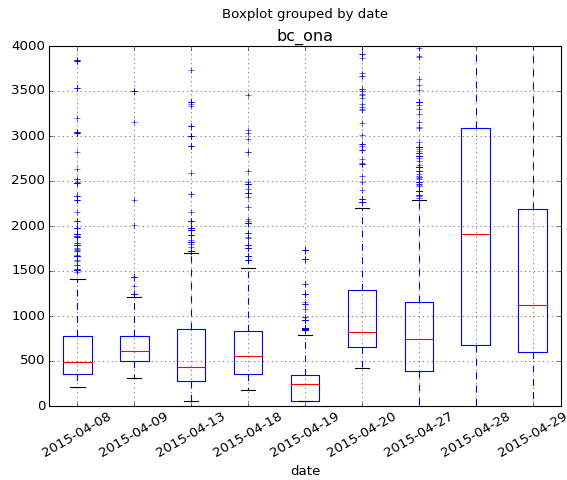


Spatial and temporal averages

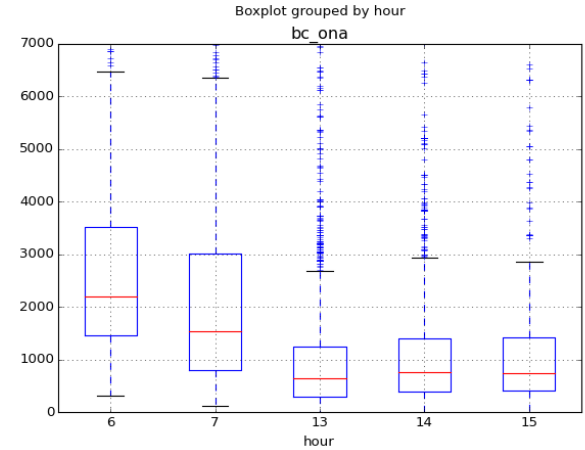
Streetwise averages

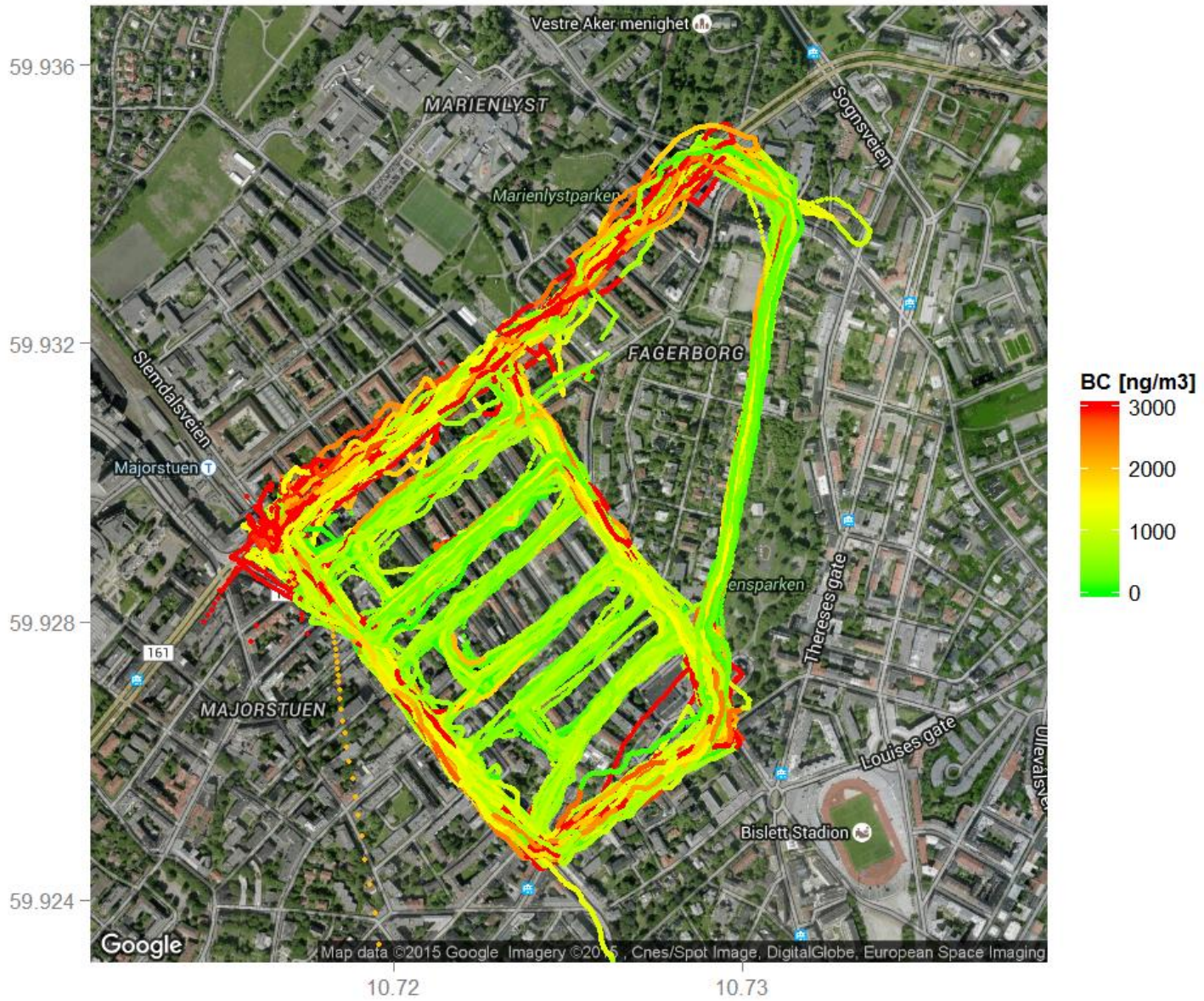


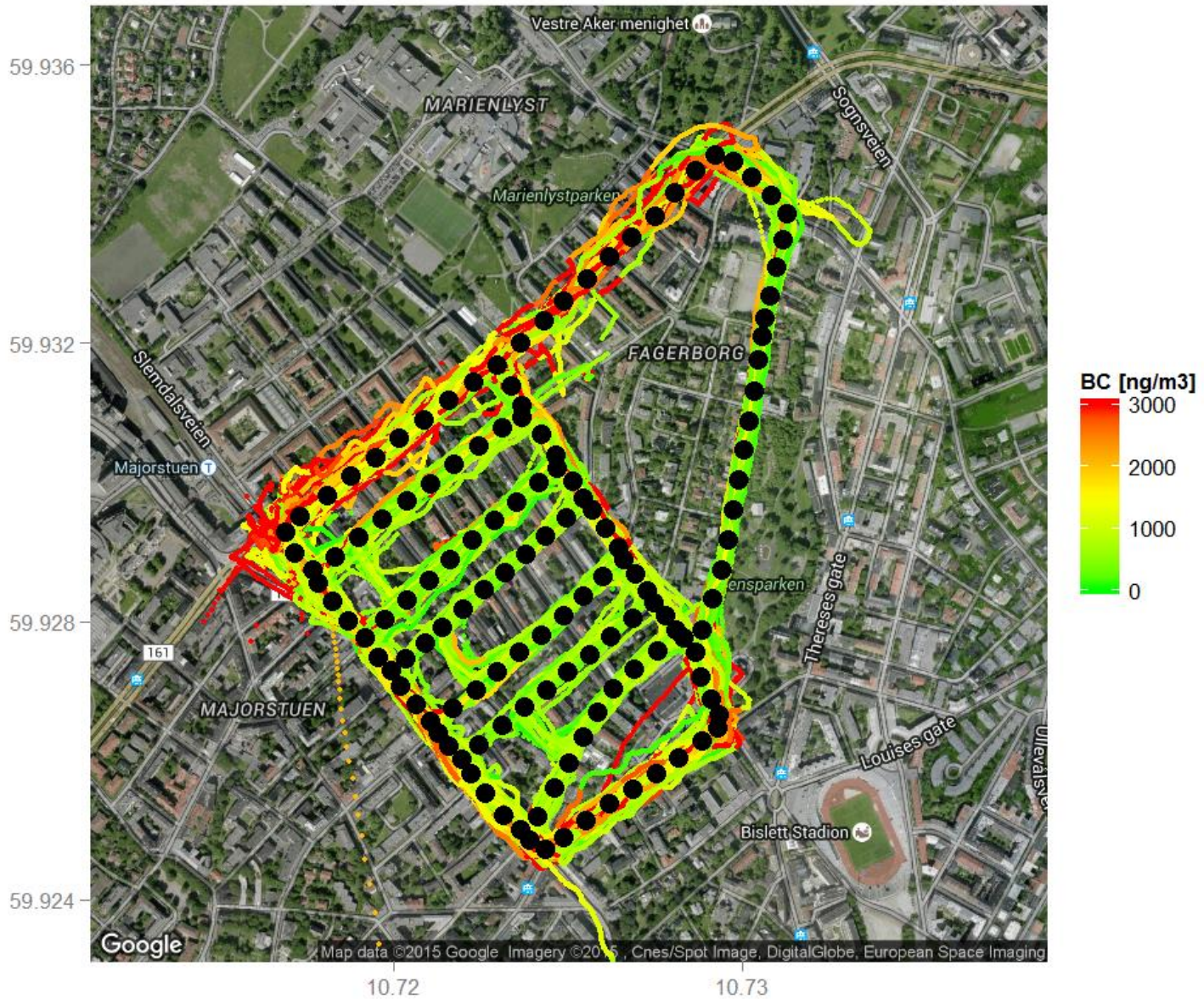
Daily averages



Hourly averages





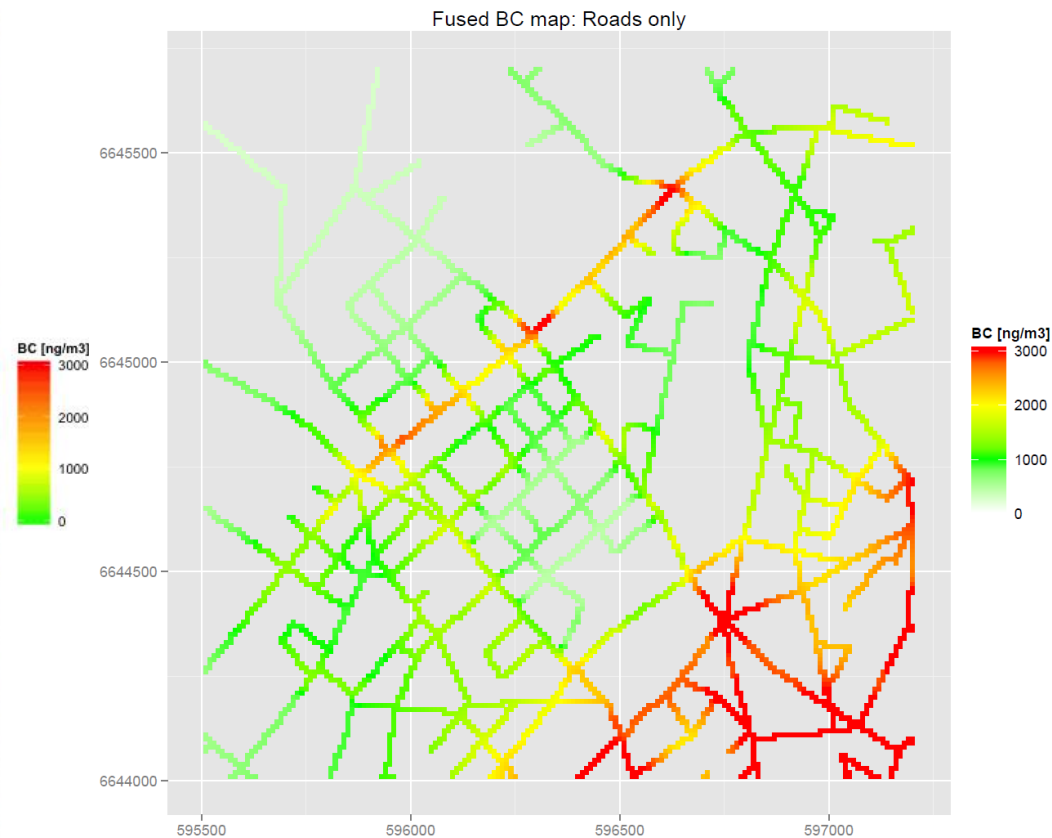




Data fusion of BC observations

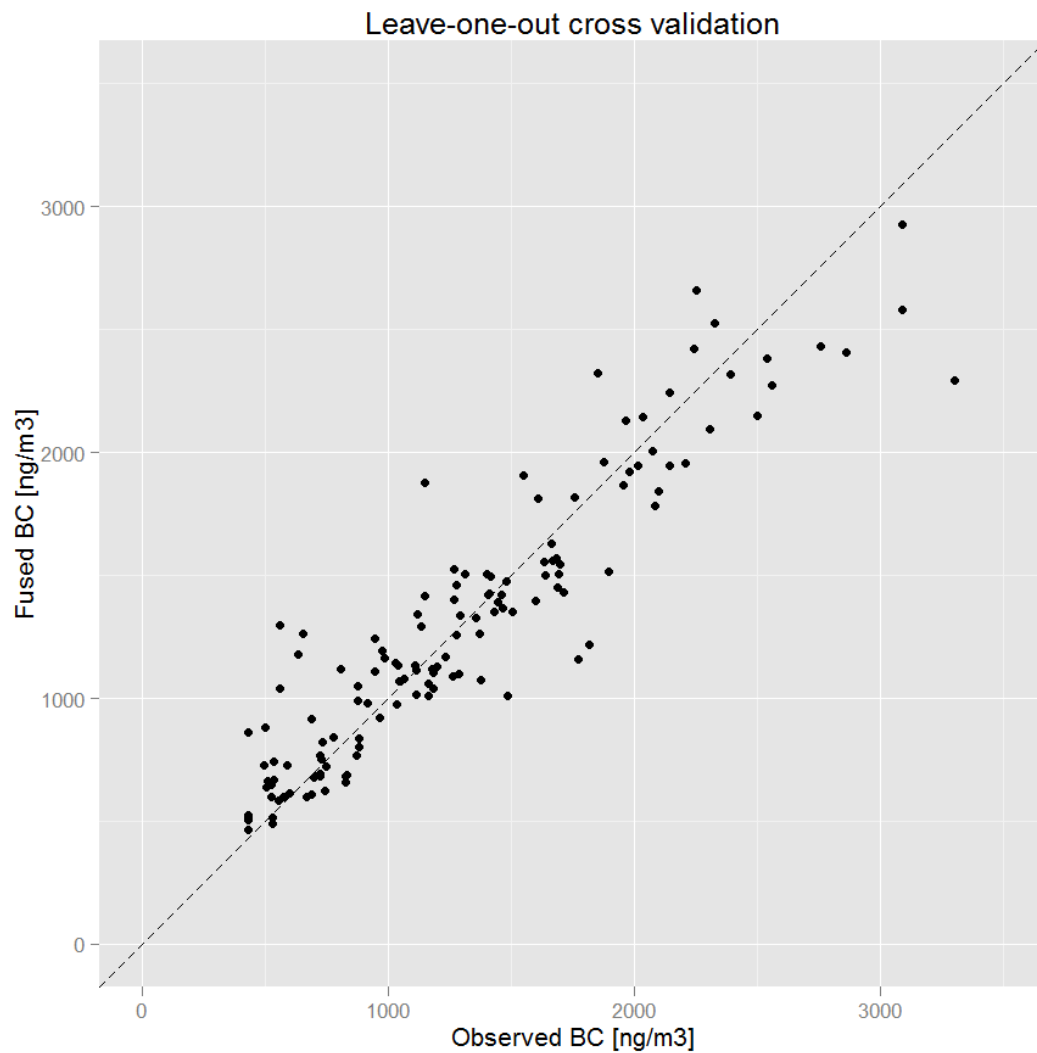


Spatially and temporally averaged BC observations



BC observations fused with EPISODE model output -> Extension to the entire road network

Validation



Leave-one-out cross validation of the fused road map against averaged BC concentrations

$$R^2 = 0.87$$

$$\text{RMSE} = 240.8 \text{ ng/m}^3$$

Conclusions

- We developed a technique for merging point-based crowdsourced observations of air quality with model information
 - **Geostatistics**-based: Builds upon decades of experience; best linear unbiased estimator; provides uncertainty estimates
 - **Fully automated** implementation: Can be run operationally in real time with large datasets
- Provides a much more **realistic estimate of true concentration field** than observations or model data alone
- Methodology can be used for a **static** sensor network as well as observations from **portable** instruments
- Realistic high-resolution near real-time concentration fields in urban areas for the first time allow for **personalized air quality information**
 - “How much particulate matter will I breathe in if I ride my bike **from home to work right now?**”
 - “What **route** to work is the **least polluted/healthiest?**”
- A first step towards **making sense of highly distributed observations** in the age of *crowdsourcing, Citizen Science, ubiquitous sensing, and Big Data*



EuNetAir 5th Scientific Meeting
16-18 December 2015, Sofia, Bulgaria



Thank you for your attention!

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www.citi-sense.eu

Extra Slides

Outline

- The general problem
- High-resolution mapping using a network of static AQ sensors
- High-resolution mapping of black carbon concentrations using observations with portable sensors



Air pollution
is the top
environmental
risk factor for
premature death

In 2011, 458000 premature deaths in Europe were attributed to particulate matter in the air



Traditional Air Quality Monitoring

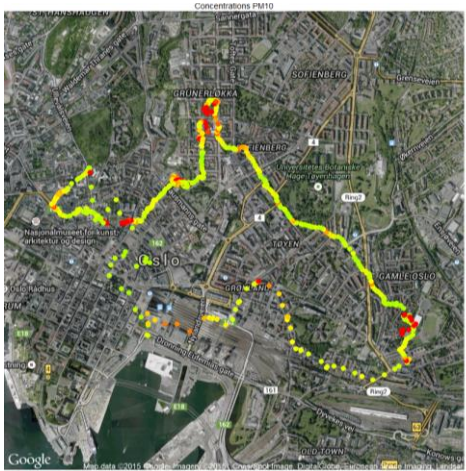
- Large
- Complex
- High-maintenance
- Expensive

→ Very sparse

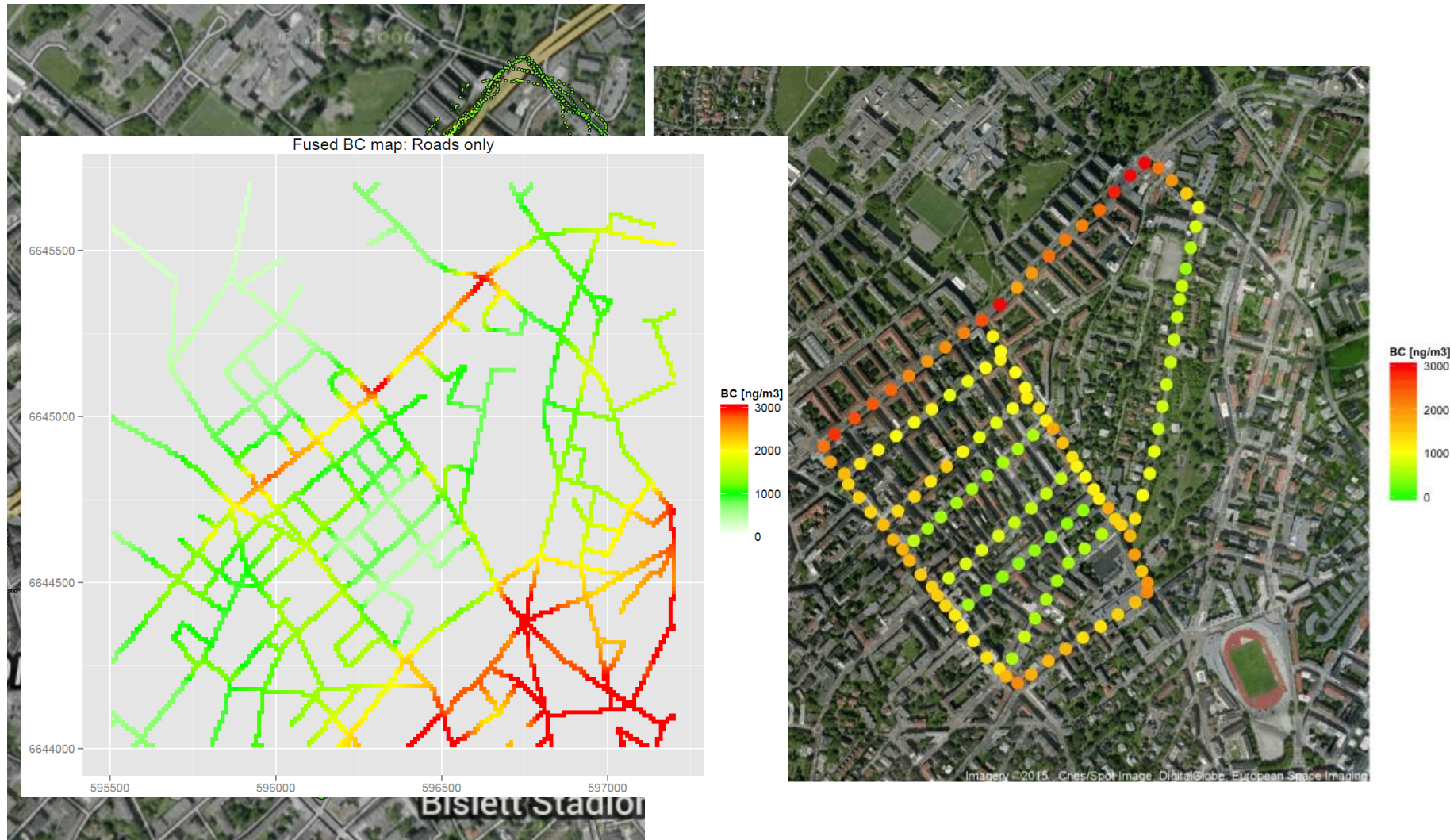
Is there another way?



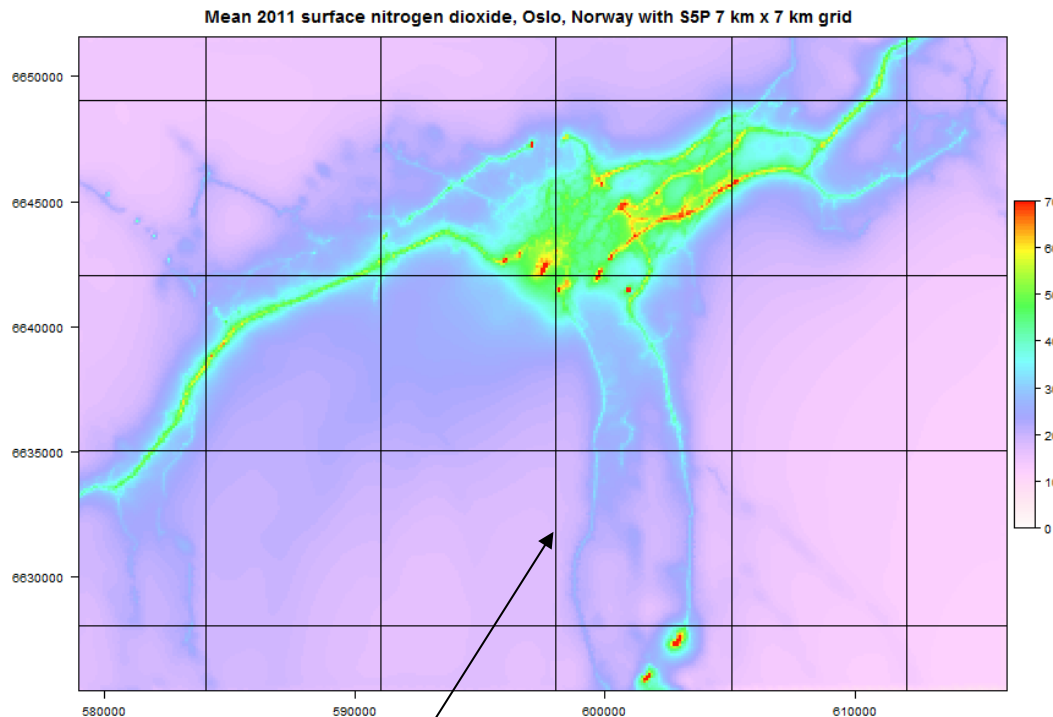
There might be...



Data fusion of mobile measurements of Black Carbon



Applications for Sentinel-5P



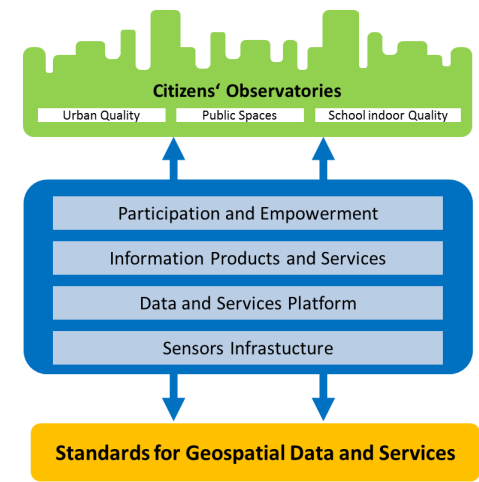
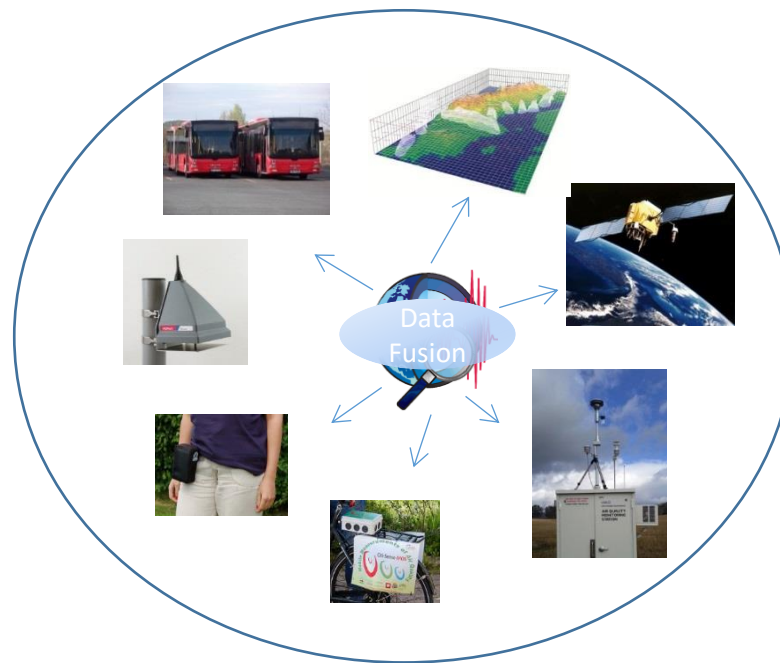
7 km by 7 km grid of Sentinel-5P products for atmospheric composition

High-resolution urban air quality maps based on the combination of crowdsourced observations and model data provides sub-pixel information for Sentinel-5P

Could be used for:

- Validation/verification of S5P data (e.g. NO_2)
- Downscaling of the S5P products to higher spatial resolution using the fused map as proxy for spatial patterns

- Collaborative Project funded by FP7-ENV-2012
- 28 project partners from 12 countries (Europe, South Korea, and Australia)
- Objective: Development of sensor-based Citizen's Observatories for improving urban quality and for empowering citizens to
 - Contribute to and participate in environmental governance
 - Support and influence community and policy priorities and associated decision making
 - Contribute to the Global Earth Observation System of Systems (GEOSS)



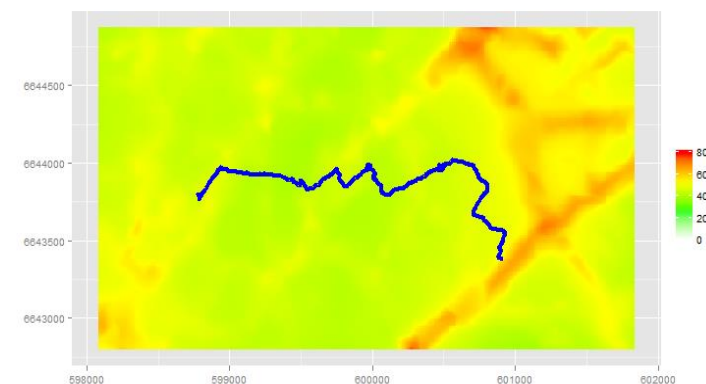
Observations

- There are many aspects to CITI-SENSE
 - Air quality observations using static and mobile sensor nodes
 - Indoor environment in schools
 - Public Spaces
- Here we focus on a network of static sensor nodes for air quality that are being deployed in various cities throughout Europe
 - Measuring the major air pollutants
 - Mounted at stakeholder's premises

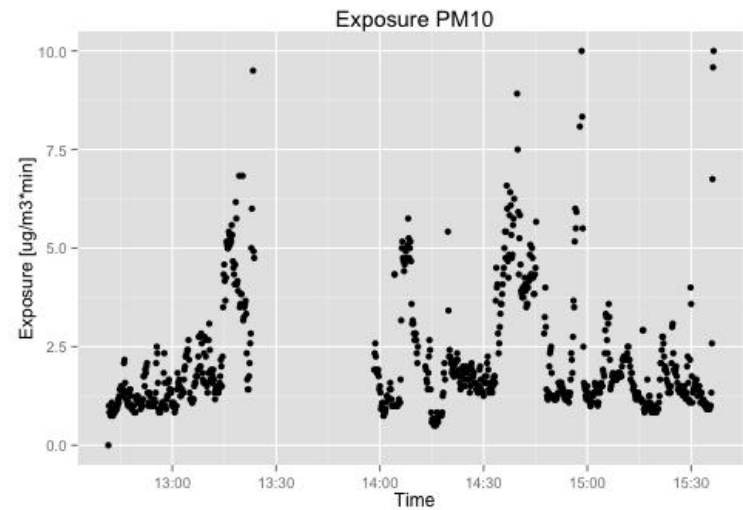
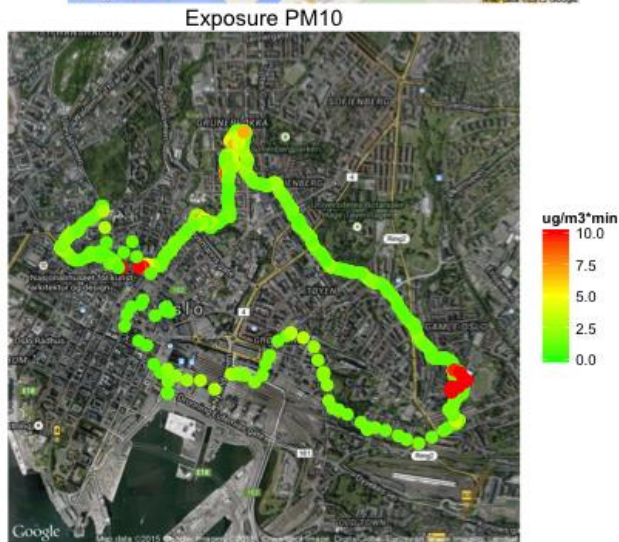
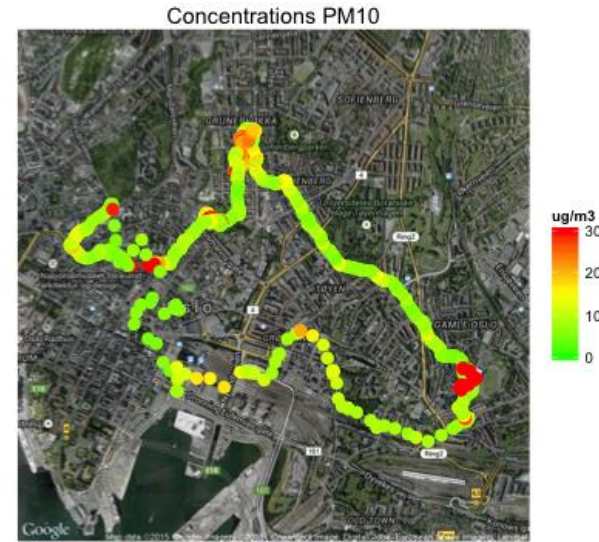
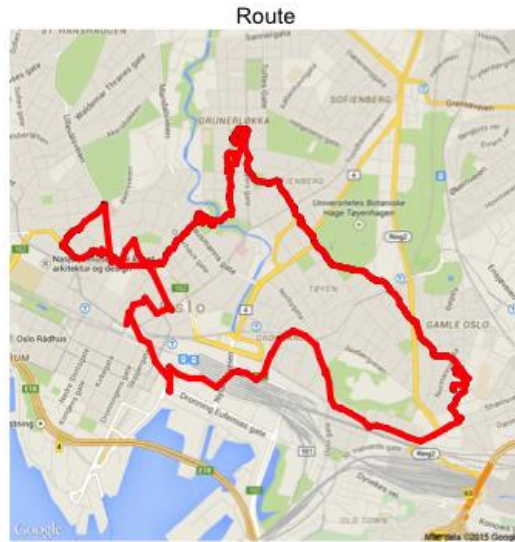


Towards personal exposure estimates

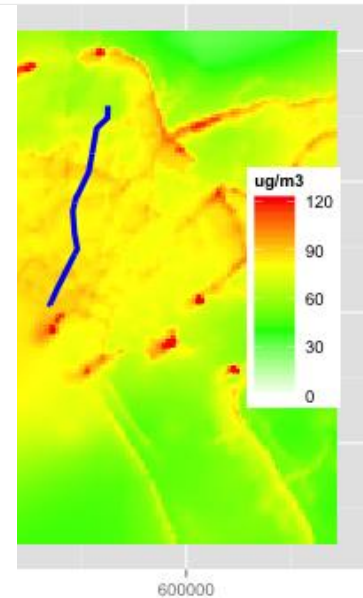
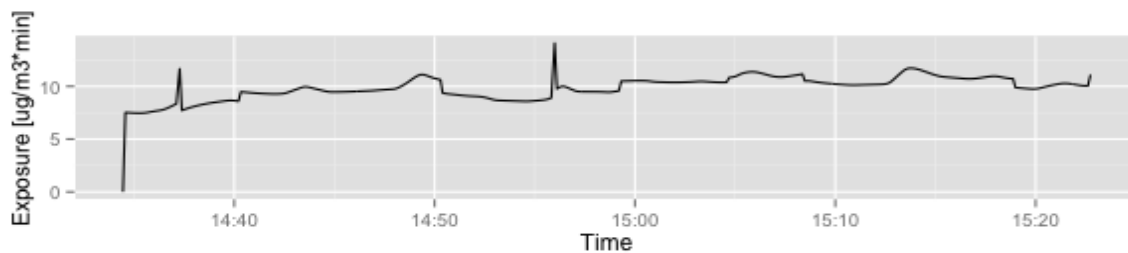
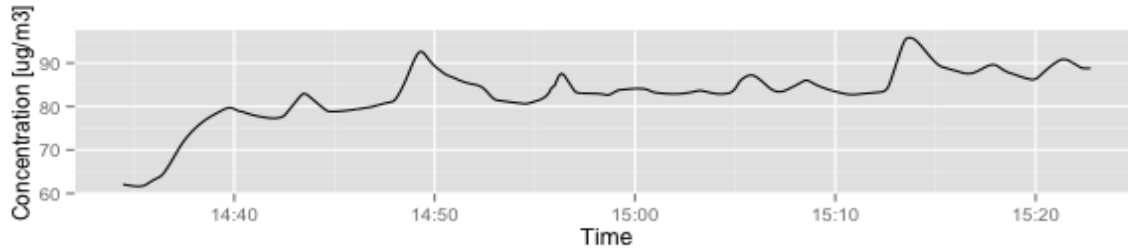
- There are two alternative ways for accomplishing personalized exposure/dose estimates
 - Approach 1: **Direct use of sensors**
People move through the urban environment with portable sensors measuring concentrations
 - Approach 2: **Indirect use of sensors**
Sensor data is used with model info and data fusion techniques to provide up-to-date air quality maps for the city -> these maps are then used to estimate exposure along a given track



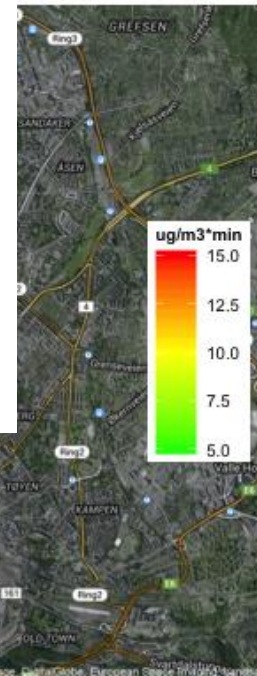
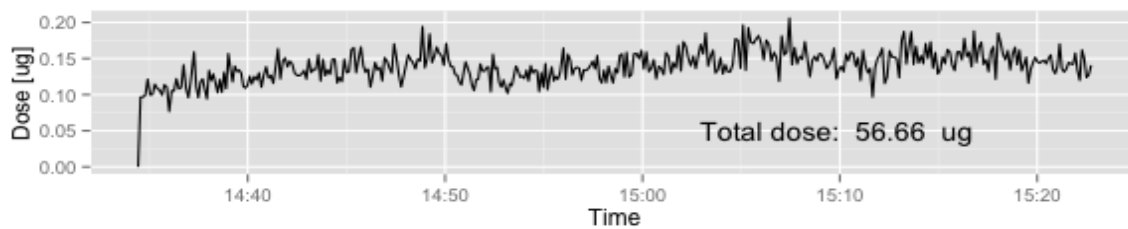
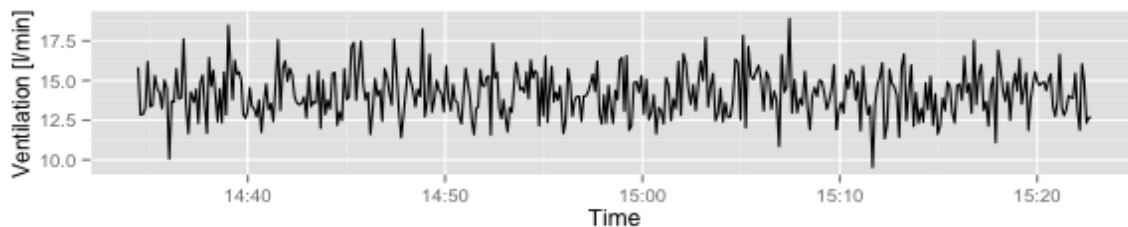
Sensor-based exposure and dose



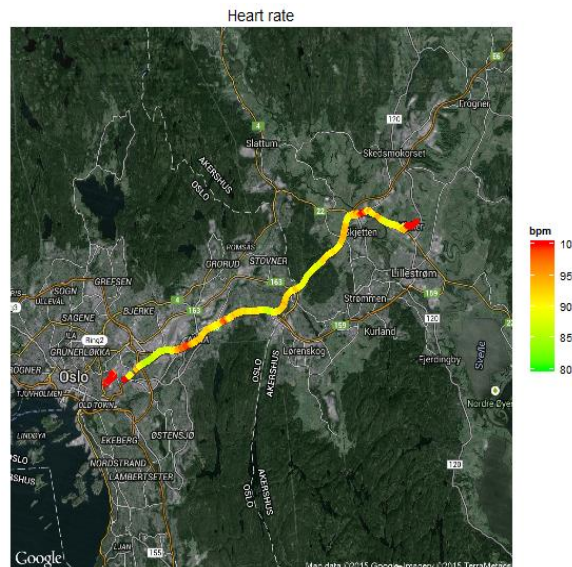
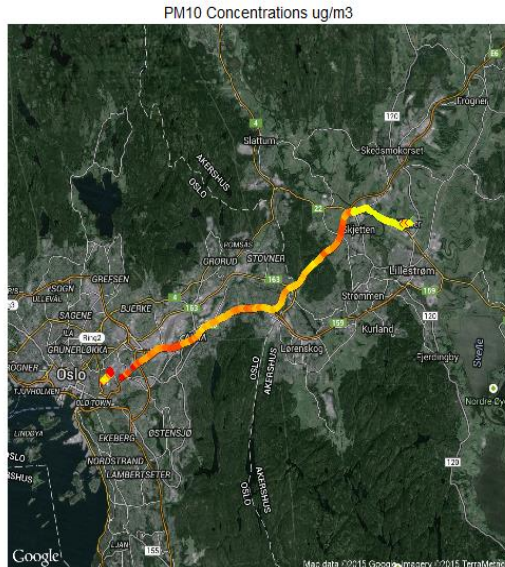
Route



Conce



Heart rate -> Dose



If we know the subject's heart rate we can compute the inhalation rate (ventilation) and the inhaled dose

Heart rate can either be

1. Measured by a heart rate monitor (mostly Approach 1)
2. Derived from accelerometer data (Approach 1+2)
3. Estimated by activity (mostly Approach 2)

$$\text{Inhaled dose} = \text{Concentration} \times \text{Ventilation} \times \text{Duration}$$

