

Air quality monitoring from global to local scales

Applications of satellites and microsensors – examples of two rapidly evolving technologies

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Outline

1. Satellite-based air quality monitoring

Can satellite observations of atmospheric composition help with air quality monitoring?

2. Low-cost microsensors for air quality mapping

Given the current state of sensor technology, how can we best exploit low-cost microsensors for urban-scale air quality mapping?

PART 1: SATELLITE MONITORING

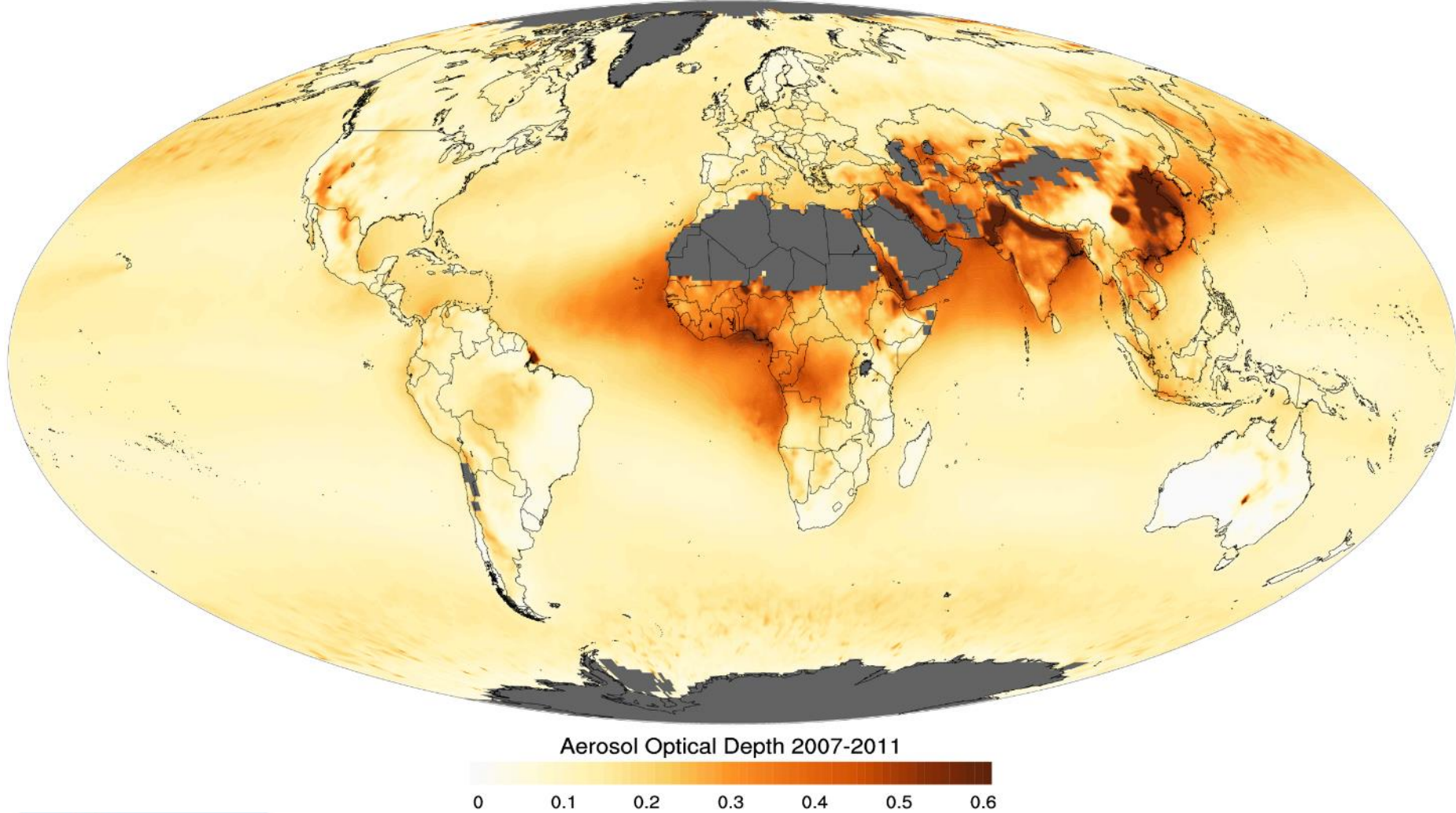
Why the view from space?

- Spatially exhaustive information
- Global data availability
- Consistent and homogeneous time series of data
- Archive of several decades of worldwide data

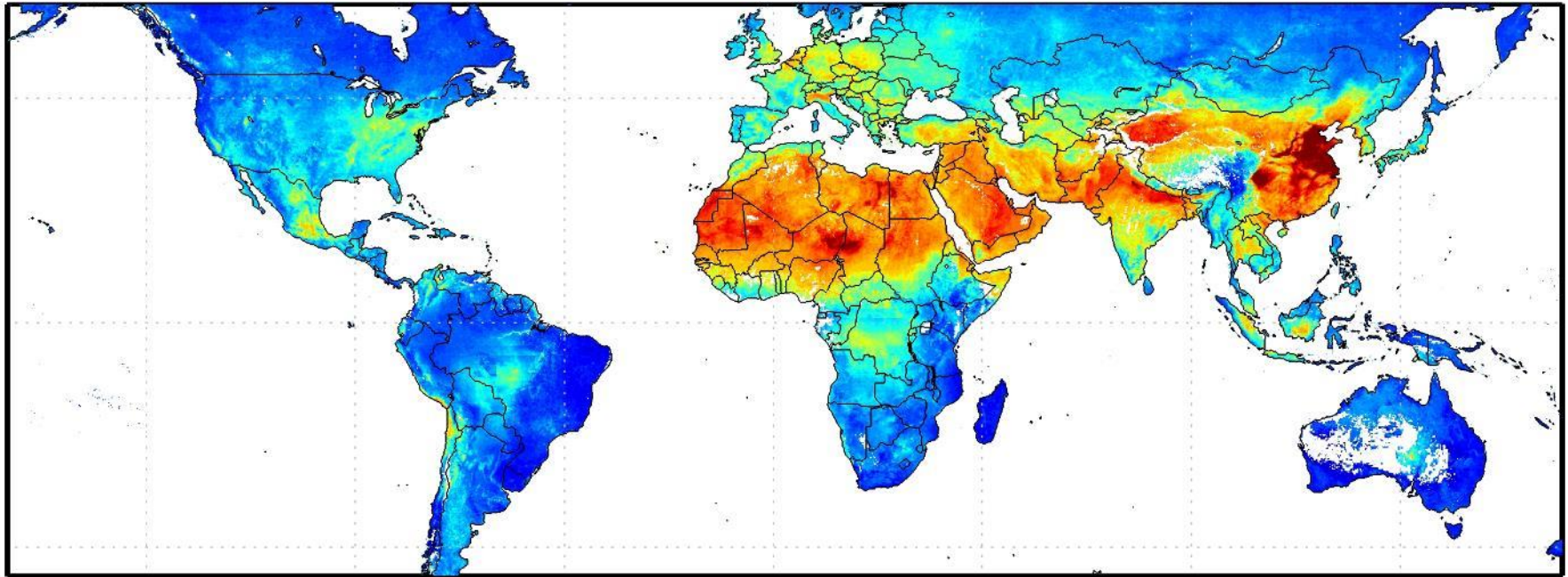
Satellite-based air quality monitoring

- Technically not that “new” ... has been around since mid-1990s
- However: So far mostly used for global and regional applications due to coarse spatial resolution
- Only now are instruments slowly becoming suitable for (slightly) more detailed observations of urban air quality

Aerosol Optical Depth

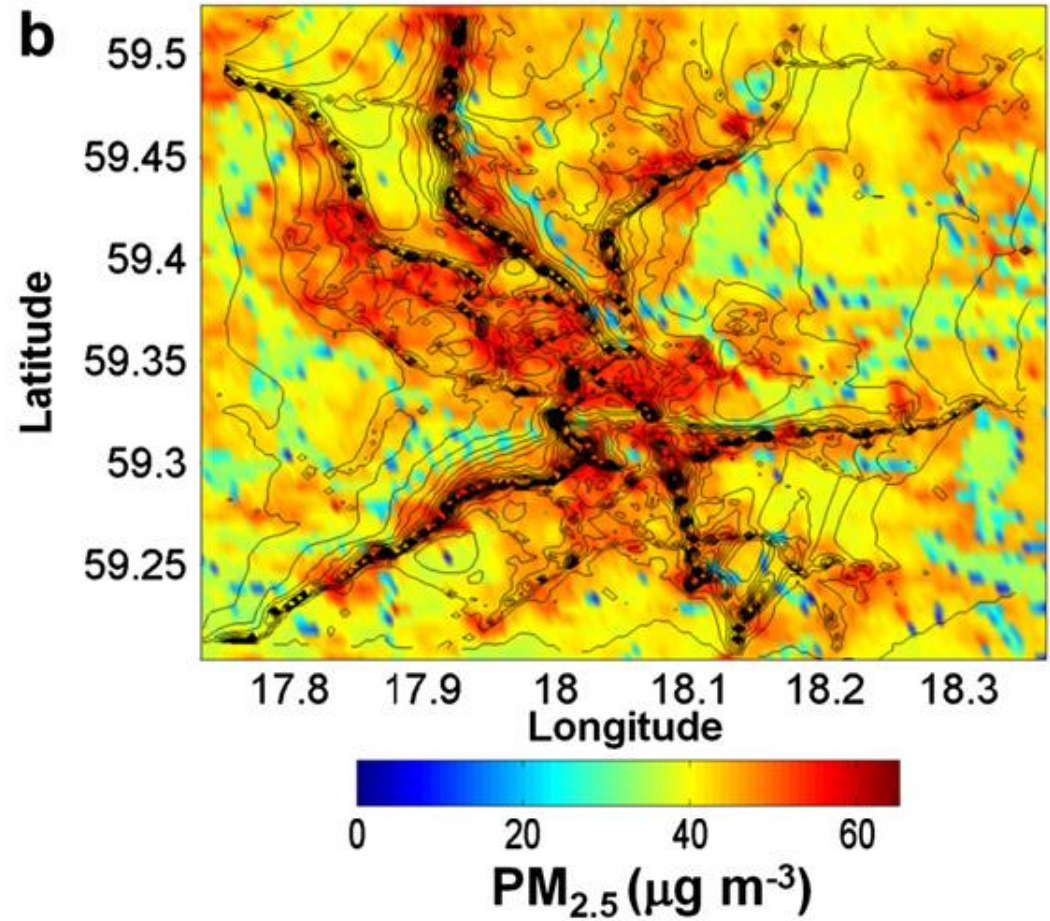
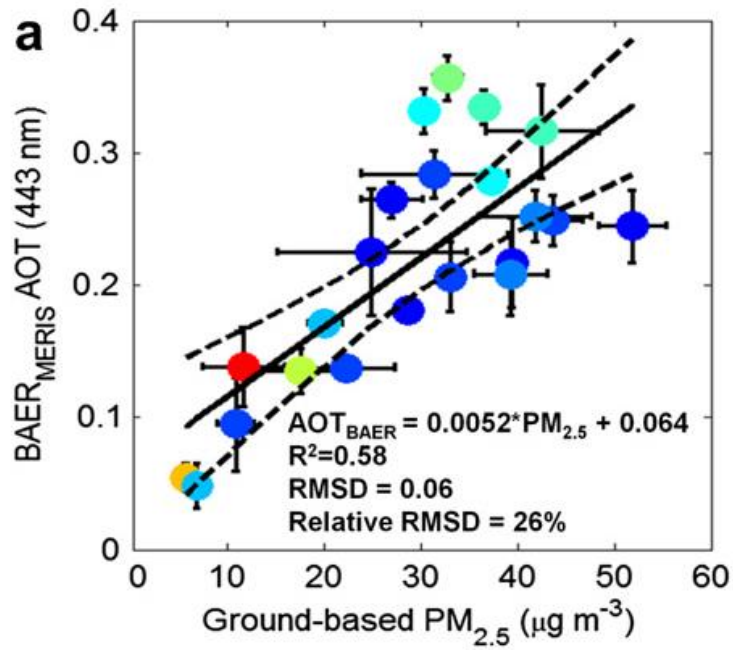


Particulate Matter



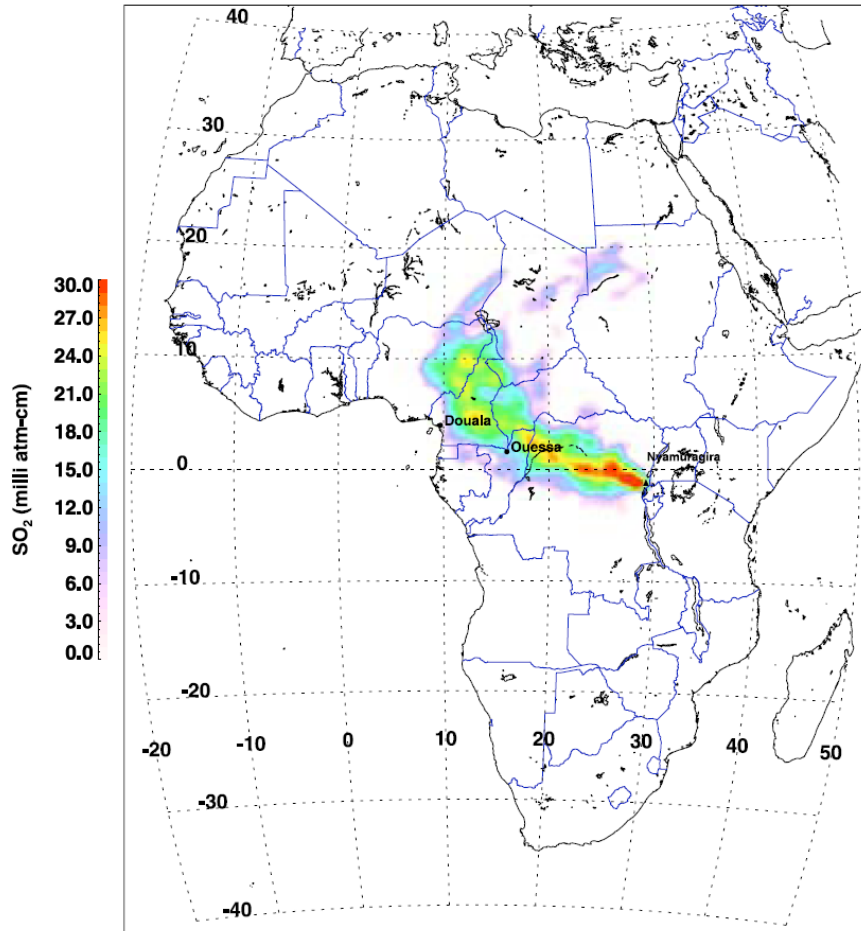
Satellite-Derived PM_{2.5} [$\mu\text{g}/\text{m}^3$]

Global satellite-derived map of PM_{2.5} averaged over 2001-2006. MODIS and MISR AOD was used. From: Van Donkelaar et al. 2010.



Relationship between AOD and station PM_{2.5} (left panel) and satellite-mapped PM_{2.5} in the Stockholm area (right panel). From Glantz et al. (2009)

Sulphur dioxide (SO₂)



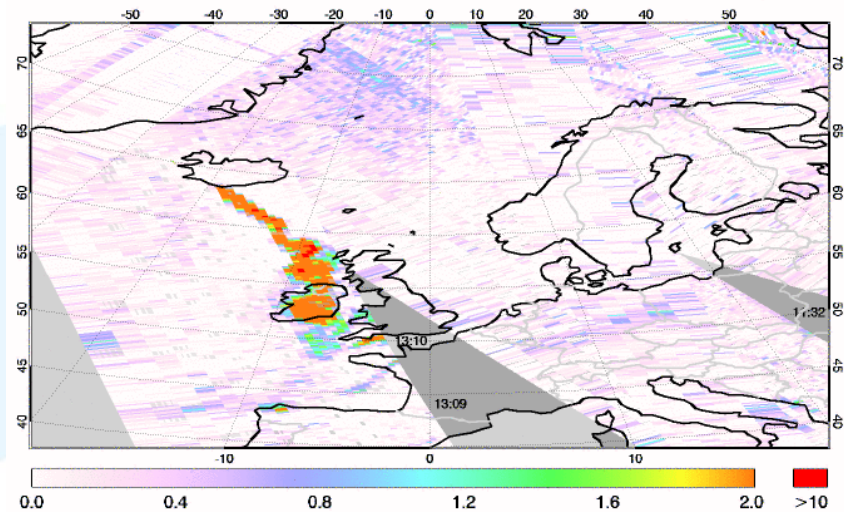
AIRS-derived regional scale SO₂ plume from Nyamuragira volcano, 28 Nov to 3 Dec 2006. From Prata and Bernardo (2007)

SO₂ vertical column [DU]

OMI - KNMI/BIRA-IASB/NASA

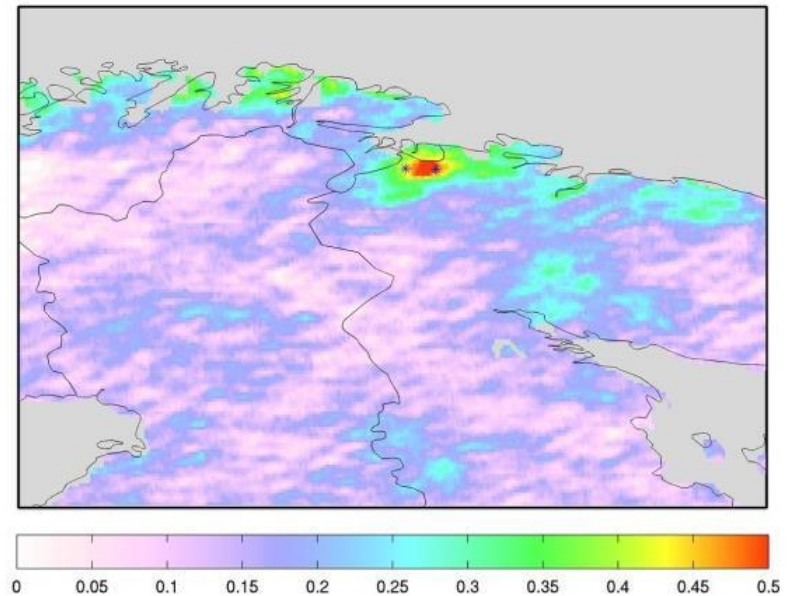
Iceland / Eyjafjöll

5 May 2010



SO₂ plume of the eruption of the Eyjafjallajökull volcano, Iceland, May 2010.

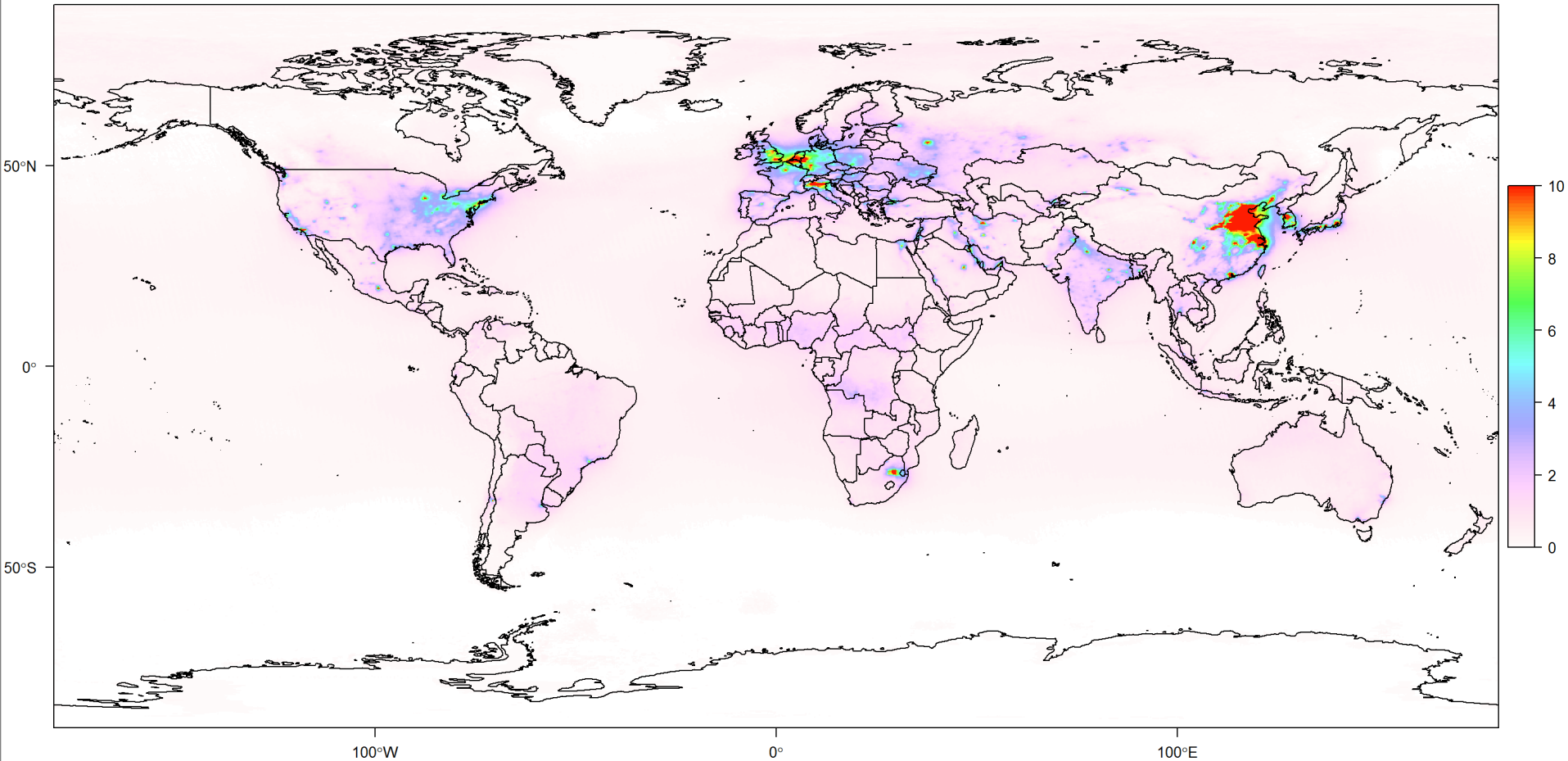
OMI SO₂ VCD [DU] 2005-2013



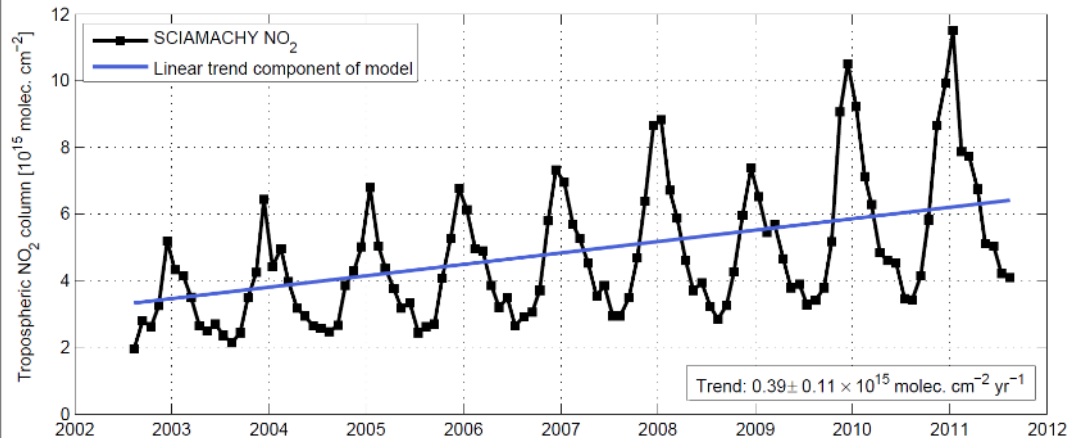
SO₂ hotspot over the Svanvik/Nikel area on the Norway/Russia border, based on OMI data (K. Stebel/N. Theys).

Nitrogen Dioxide (NO₂)

Tropospheric NO₂ column average 2004-2016 [in 10¹⁵ molec. cm⁻²]



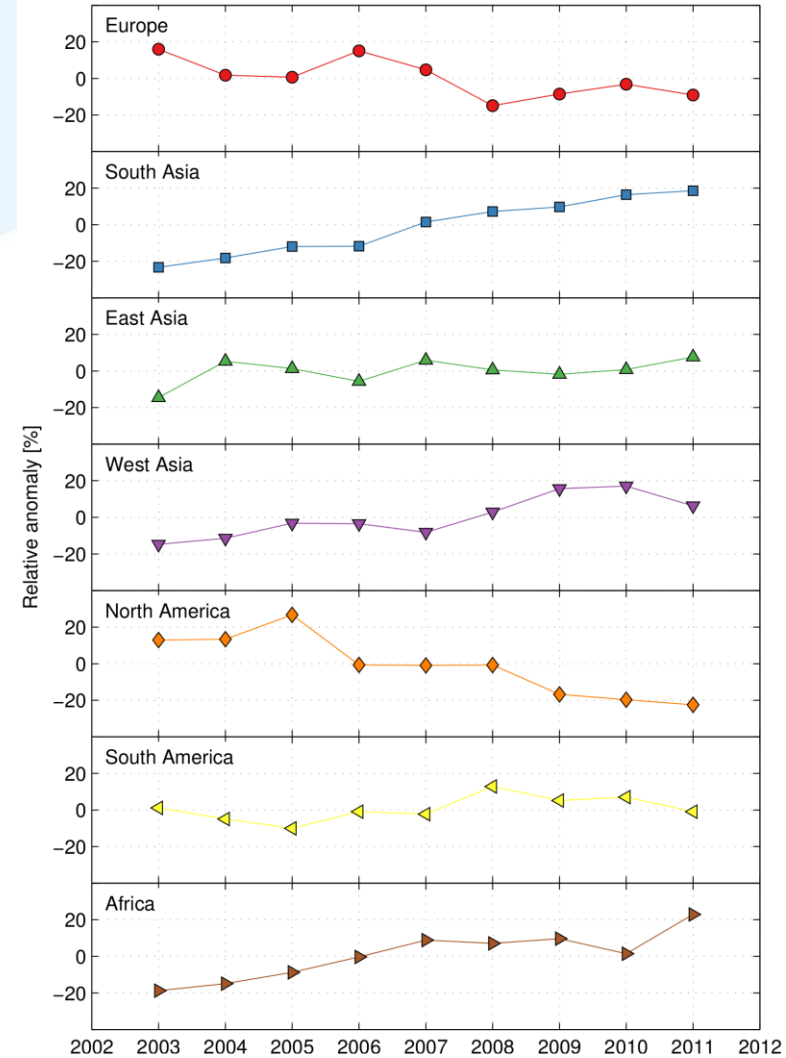
Temporal evolution



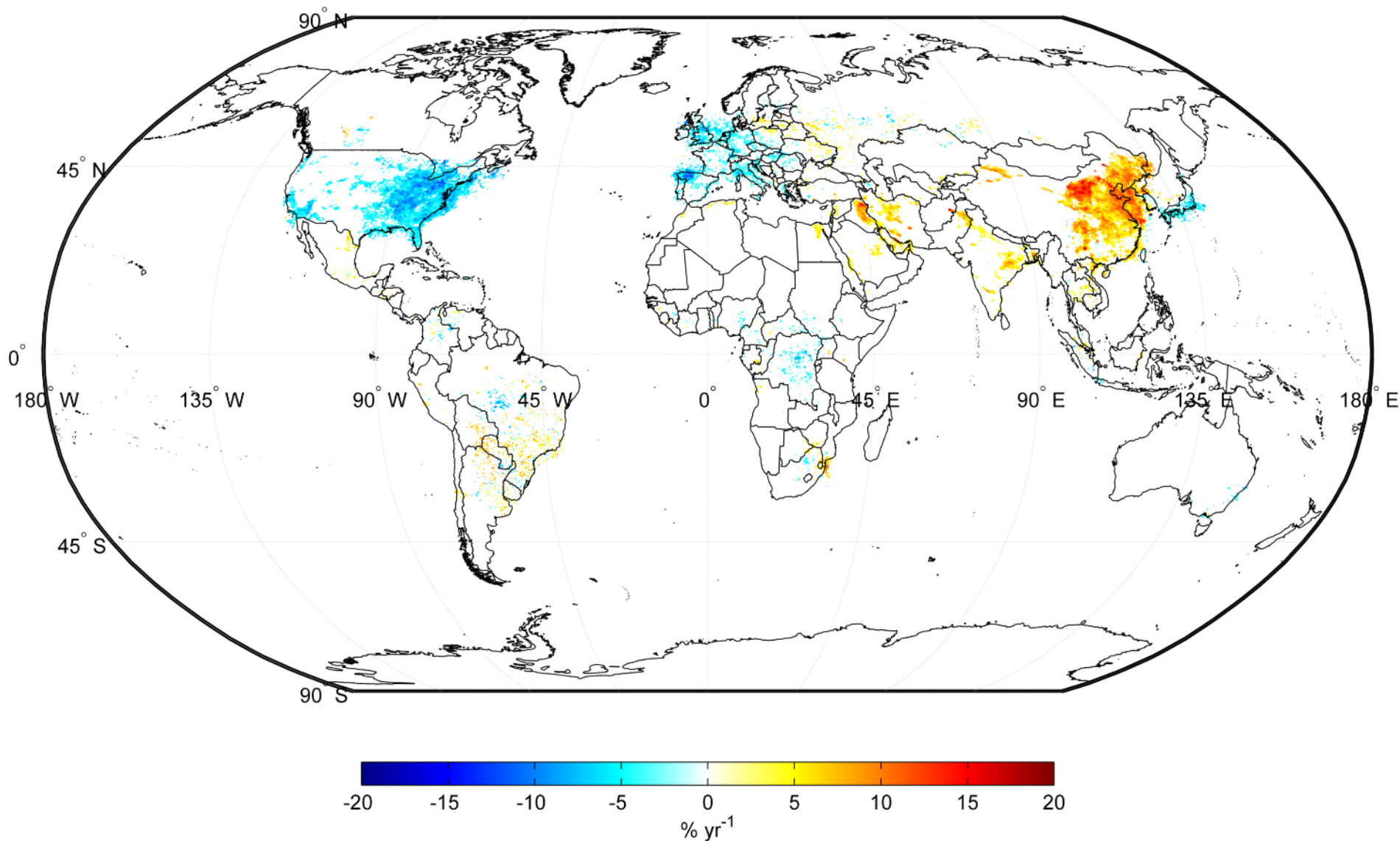
Time series of tropospheric NO₂ over China

Note on measurement uncertainty

The average relative uncertainty of individual (daily) satellite observations of NO₂ is approximately 30%. This random error is reduced significantly when multiple measurements are averaged over time (e.g. monthly averages)

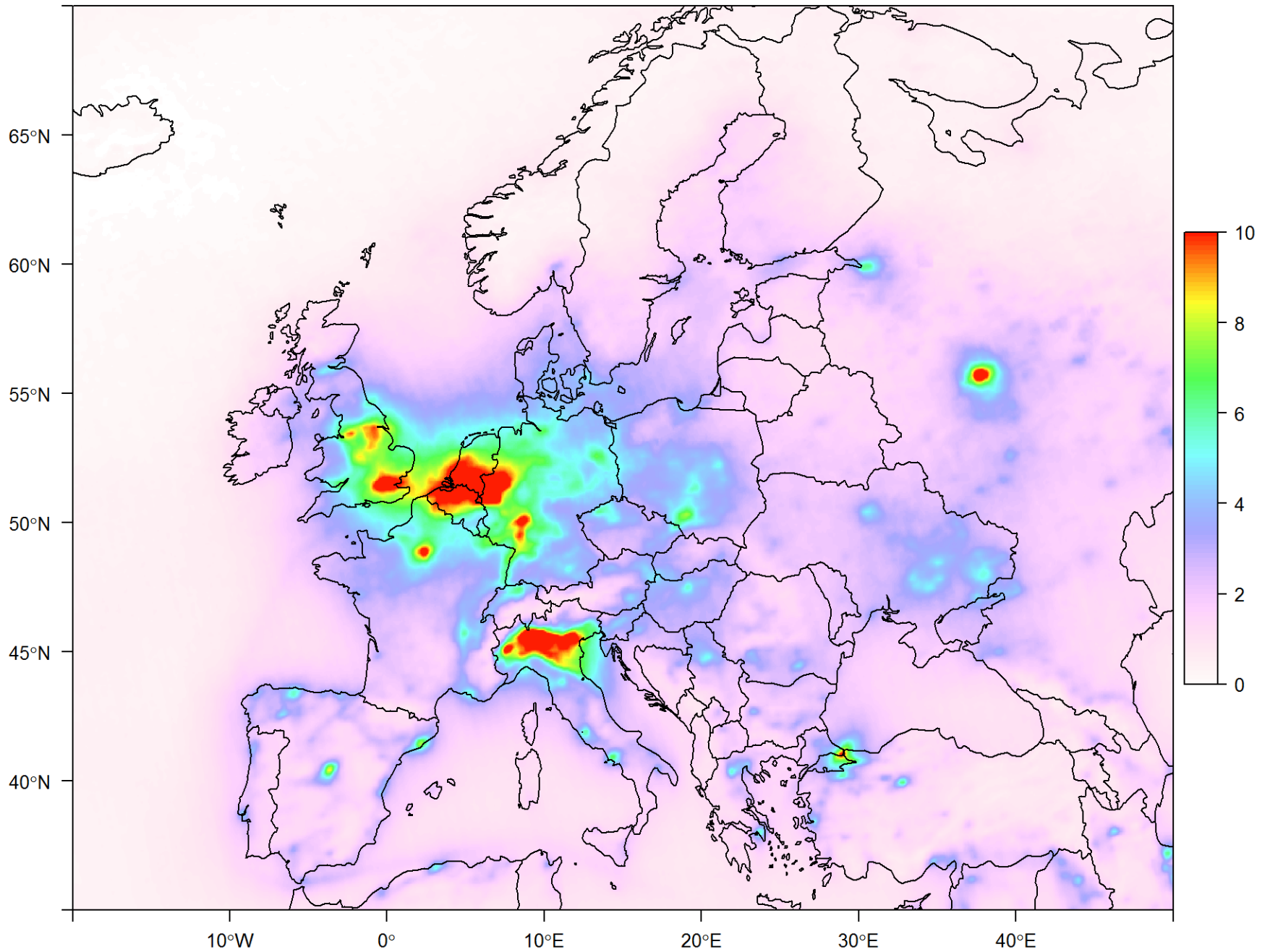


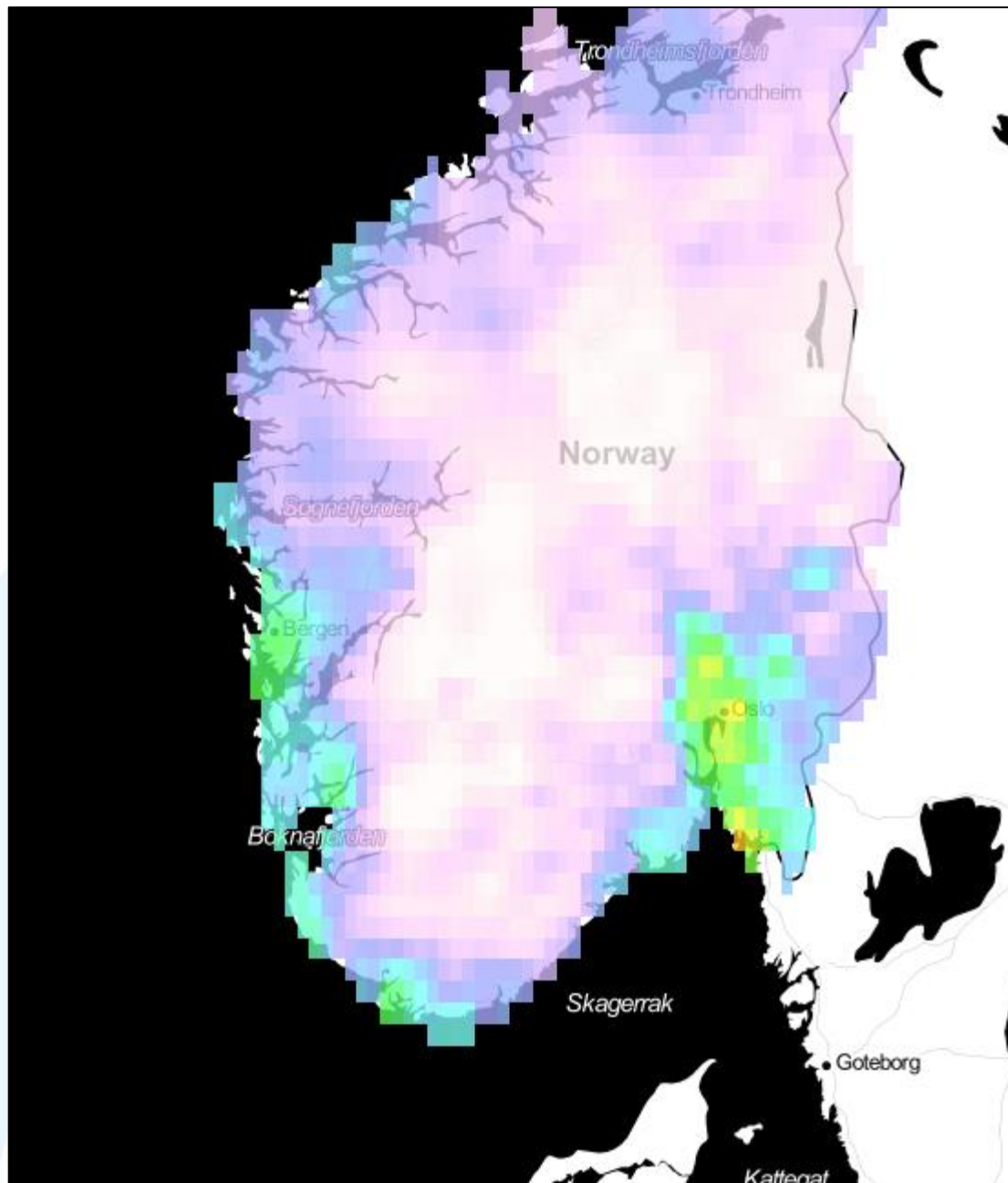
Satellite based annual average NO₂ by region



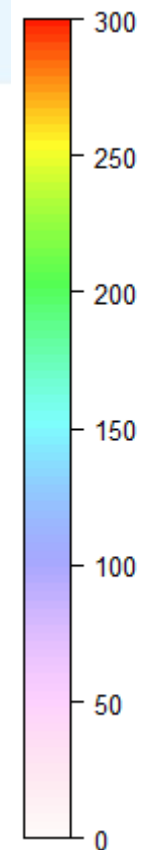
OMI-based estimate of relative change (in percent of long-term average) of nitrogen dioxide (NO₂) between 2004 and 2016.
(Schneider et al., 2015, ACP)

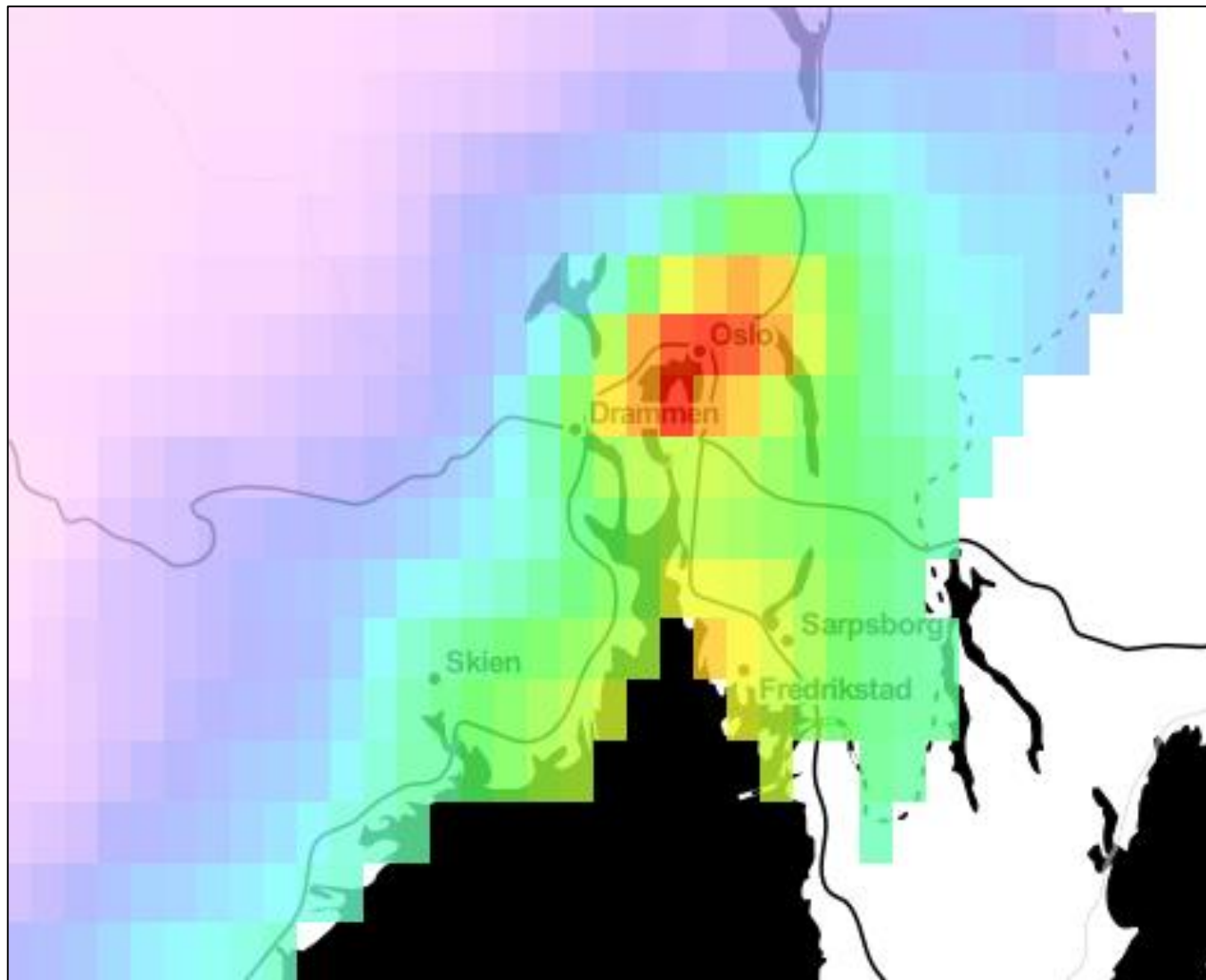
Tropospheric NO₂ column average 2004-2016 [in 10e15 molec. cm²]



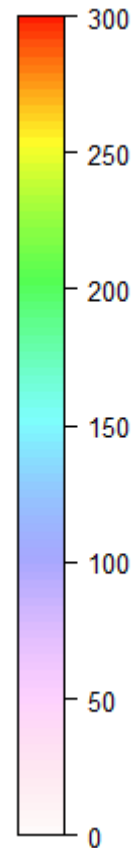


10^{13}
molec./cm²



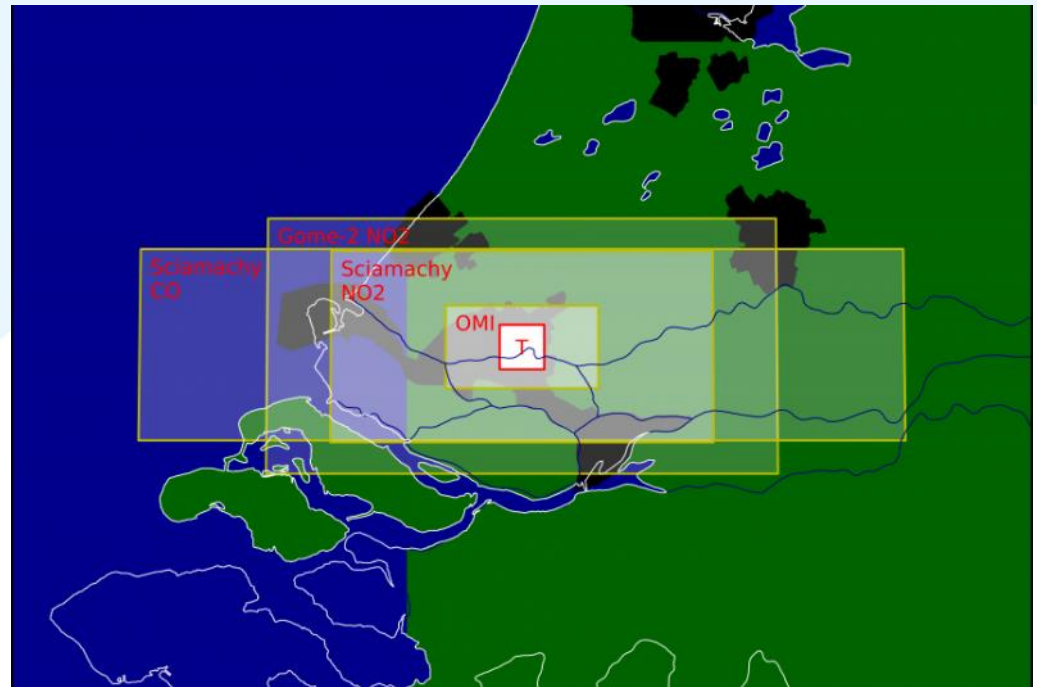


10^{13}
molec./cm²

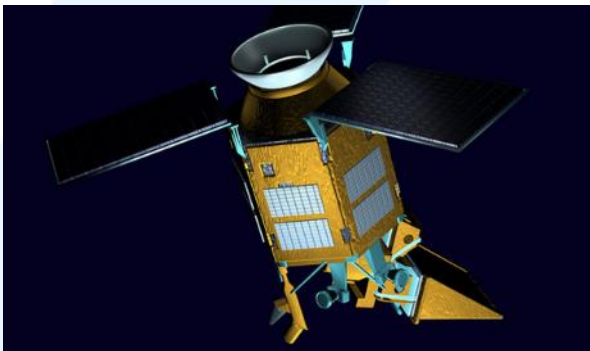


New opportunities: TROPOMI

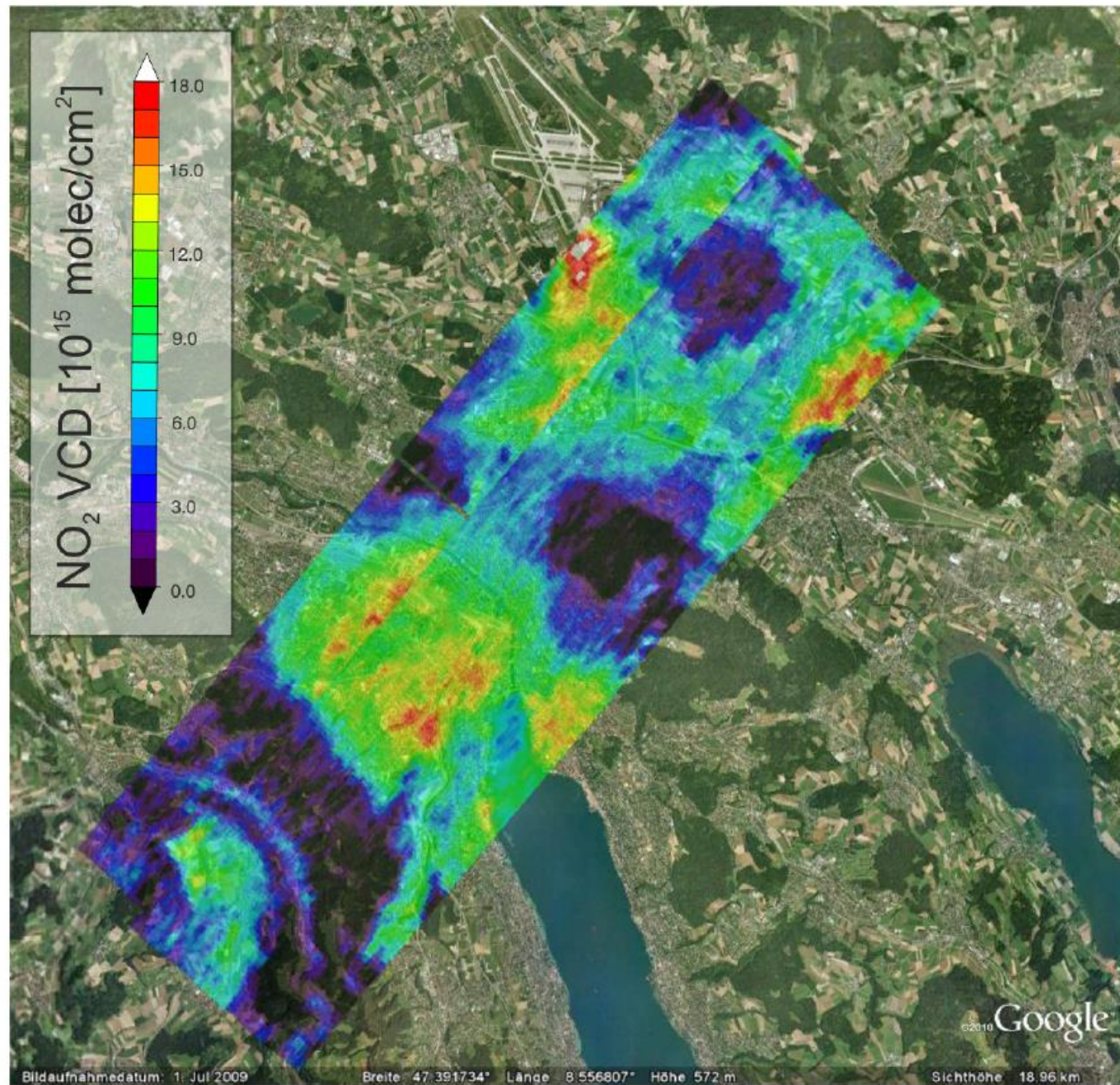
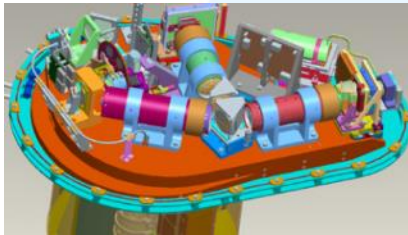
- TROPOMI is an instrument on the upcoming Sentinel-5P mission
- Launch in June 2017
- Spatial resolution of 7x7 km²
- Products for NO₂, O₃, SO₂, CO, HCHO, CH₄, and Aerosols



A comparison of the footprint of TROPOMI (nadir pixel) to past and current air quality satellite instruments



Airborne NO₂ mapping with the APEX instrument



PART 2: LOW-COST MICROSENSORS

Traditional Air Quality Monitoring

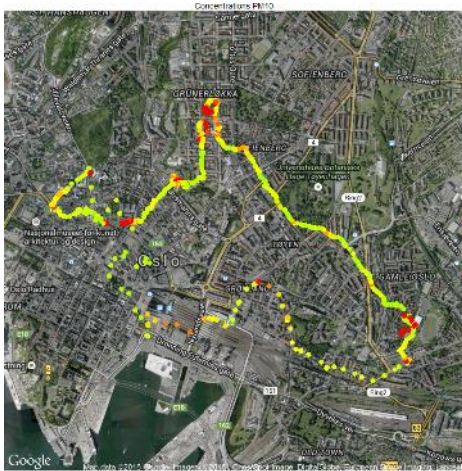
- Large
- Complex
- High-maintenance
- Expensive

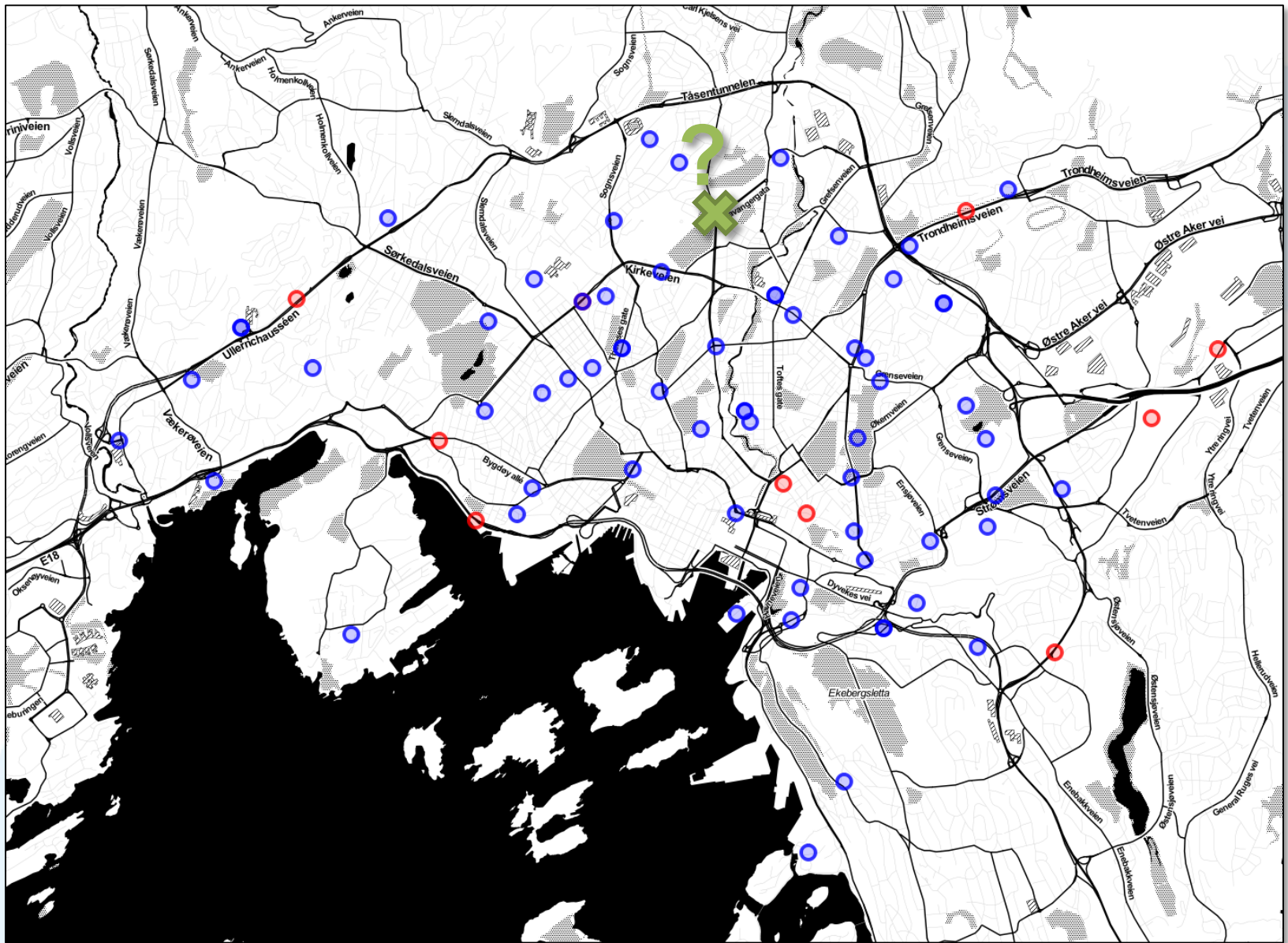
→ Very sparse

Is there another way?



There might be...

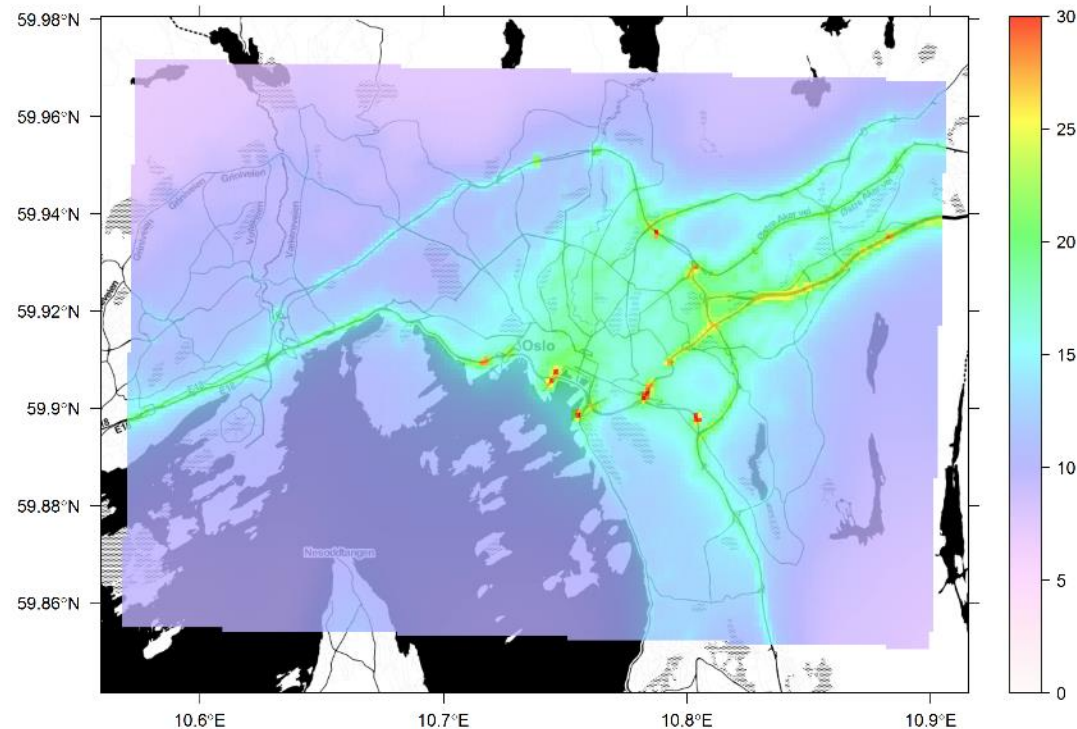




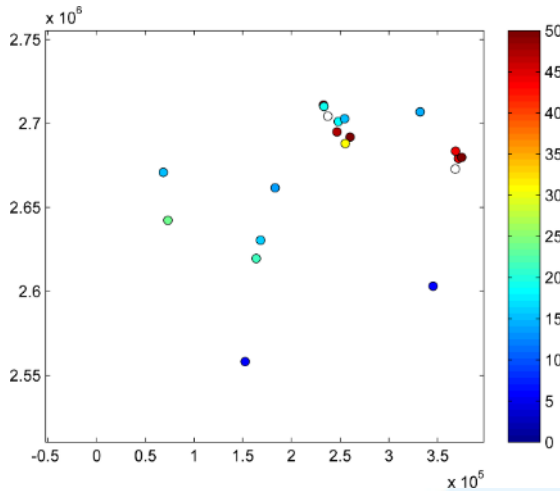
Red markers: Locations of Air Quality Monitoring stations for NO₂
Blue markers: Deployment sites of low-cost microsensors

Combination with model output

- To map the observations from the low-cost sensors onto a high-resolution grid in a scientifically meaningful way we need to use a spatial auxiliary dataset that guides the interpolation
- We use here the output from the EPISODE air quality model (high-resolution long-term average concentration maps)

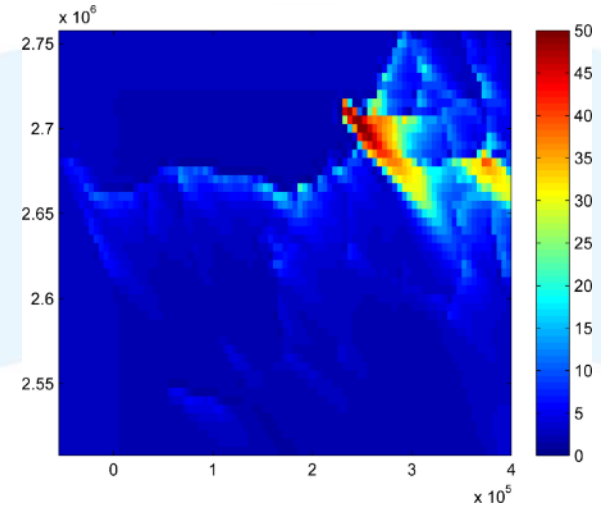


Annual average concentration of NO₂ for 2014 as compute by the EPISODE air quality model.

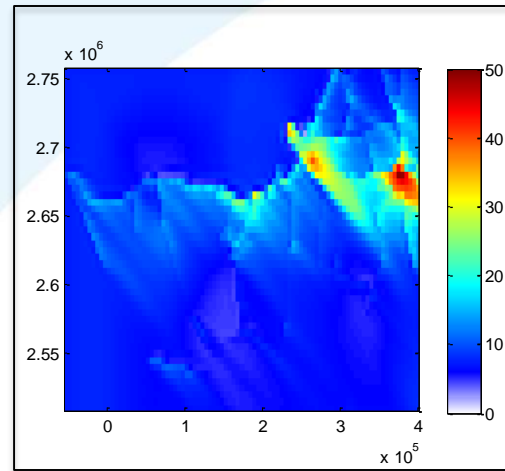


Observations

DATA FUSION



Modelling results or other auxiliary data



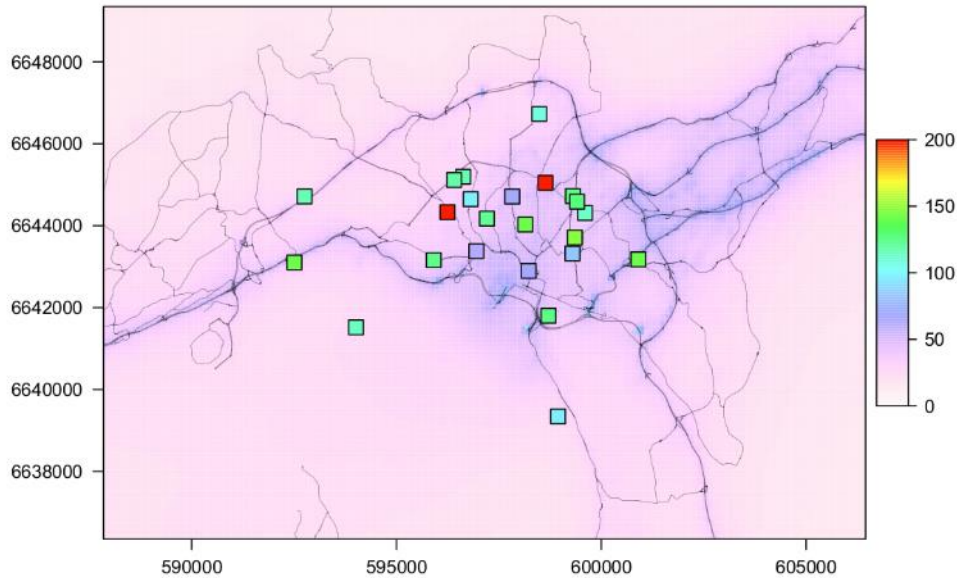
Combined map

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

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

Data fusion method used here provides a combined concentration field by separately interpolating the observational residuals from a regression model and then combining both.

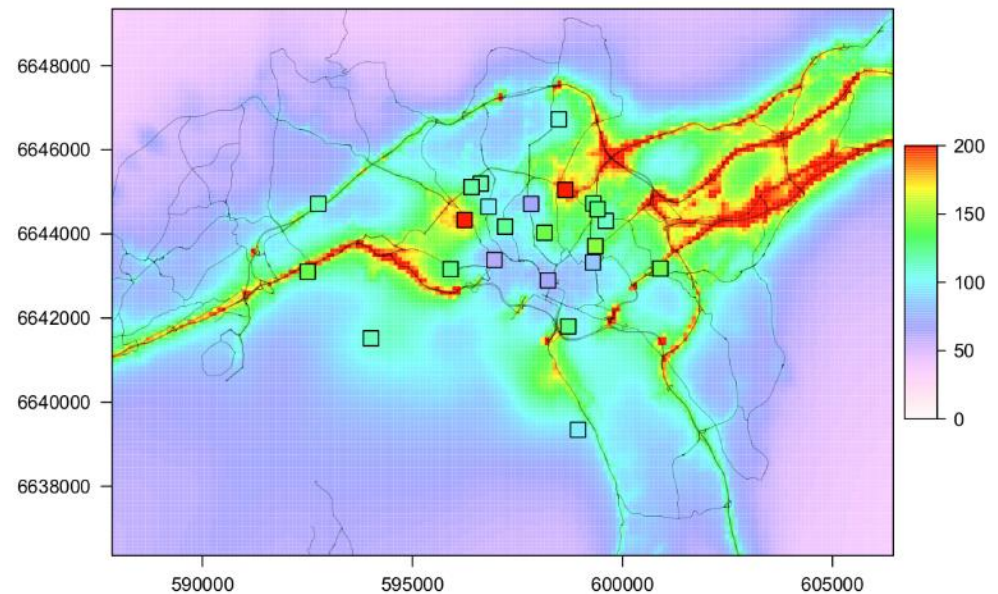
Basemap and observations [$\mu\text{g}/\text{m}^3$]



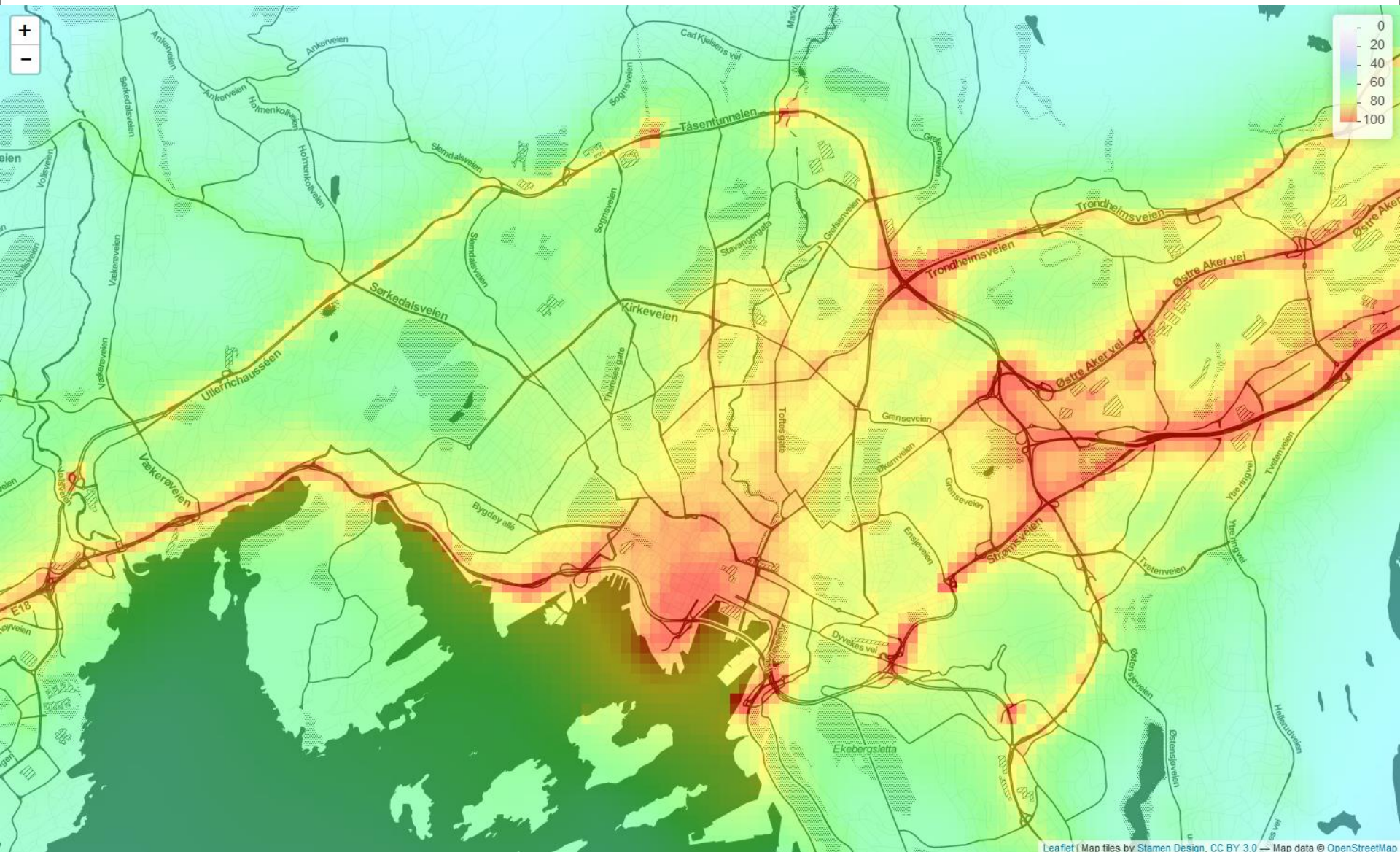
Oslo NO₂ model-derived annual average basemap (background) and observations from AQMesh nodes (markers) on 6 January 2016 at 9:00 UTC. Units in $\mu\text{g}/\text{m}^3$.

Note: The sensors used here were all **co-located for several weeks** at the Kirkeveien AQ monitoring station **before deployment** and are thus **field-calibrated!**

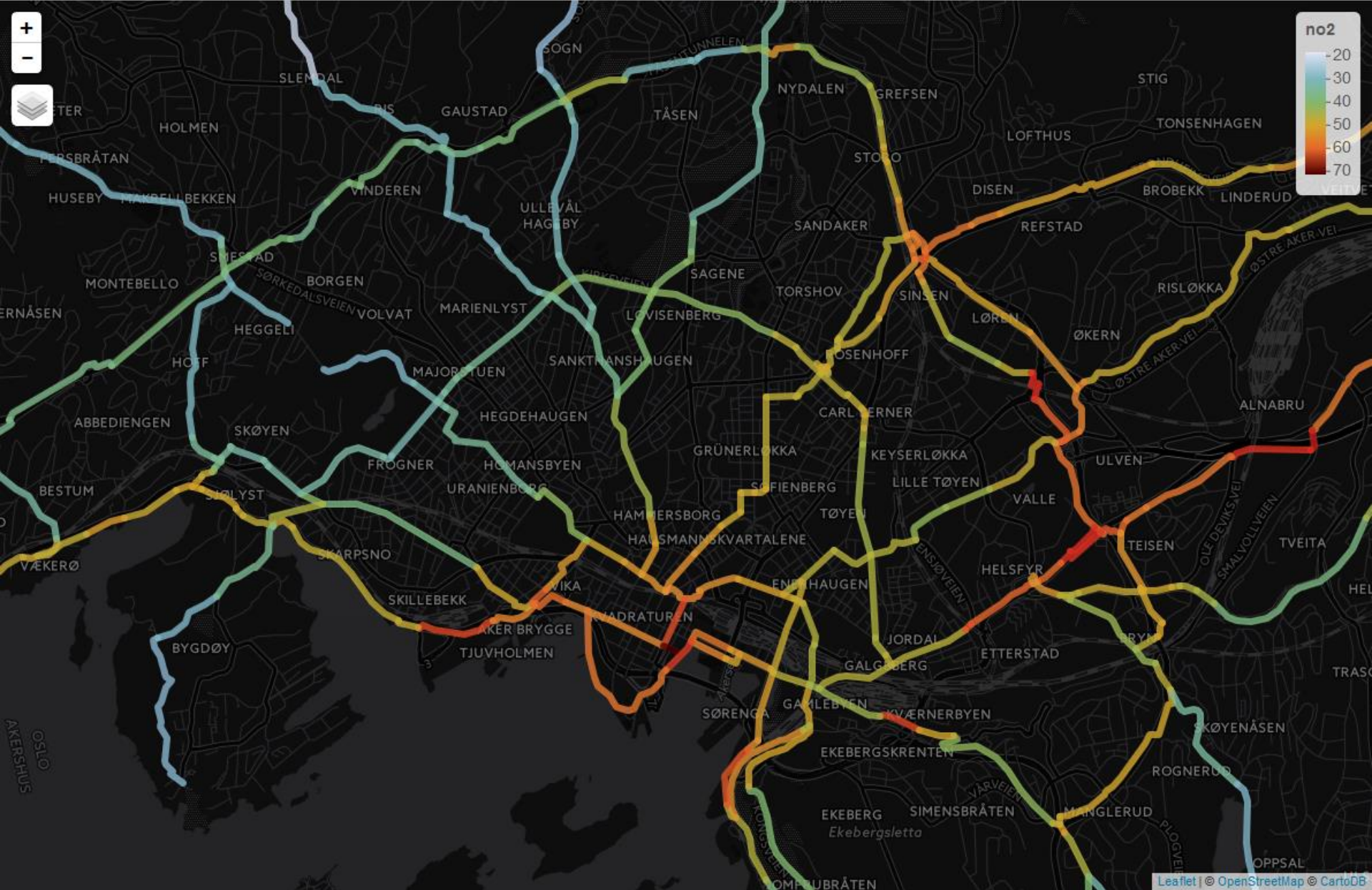
Data fusion result [$\mu\text{g}/\text{m}^3$]



Result of data fusion process (background) and observations from AQMesh nodes (markers). Units in $\mu\text{g}/\text{m}^3$.

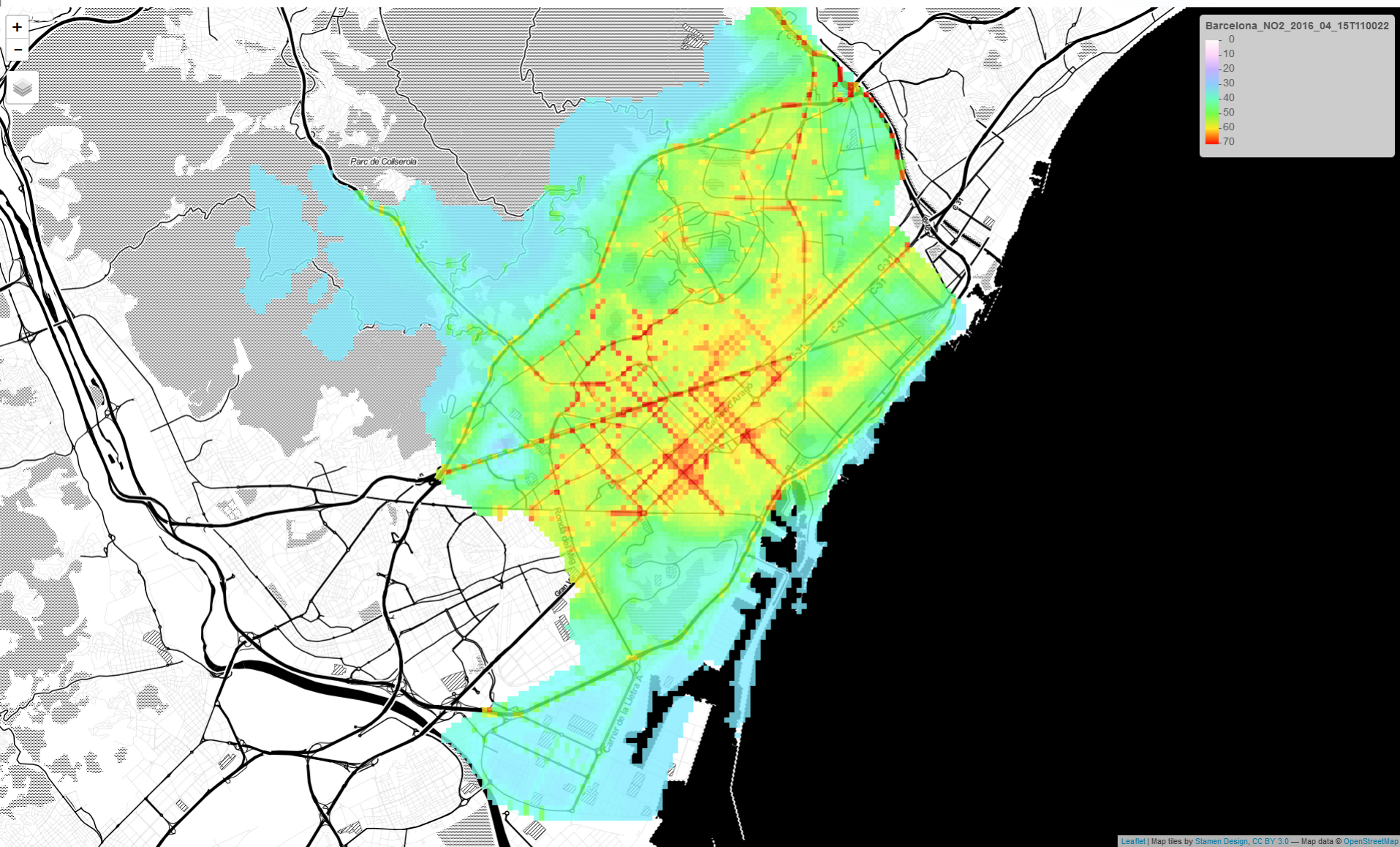


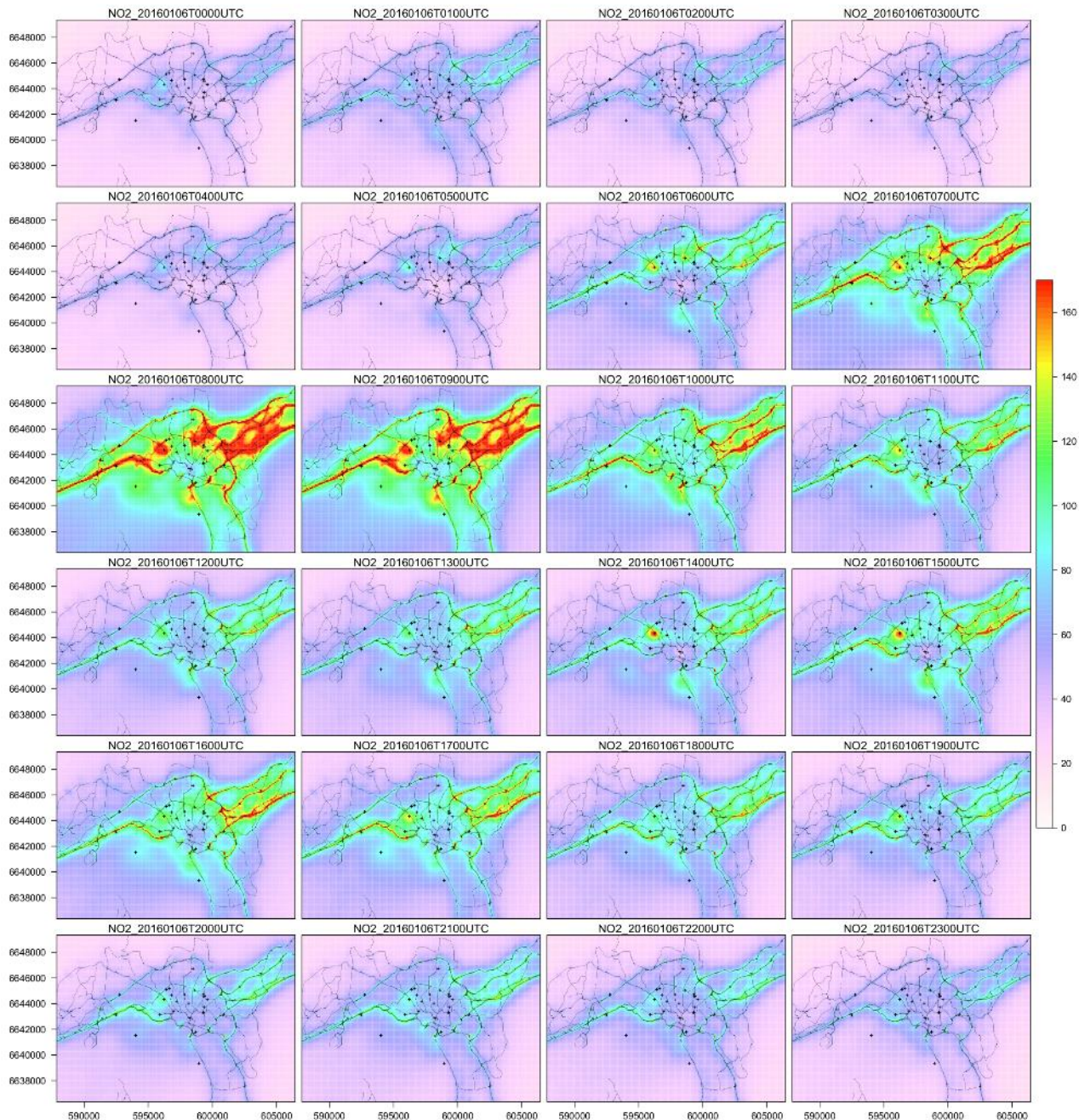
Example of a data fusion-based surface concentration field of NO₂ for Oslo, Norway, at 100 m spatial resolution ([link](#)).



Leaflet | © OpenStreetMap © CartoDB

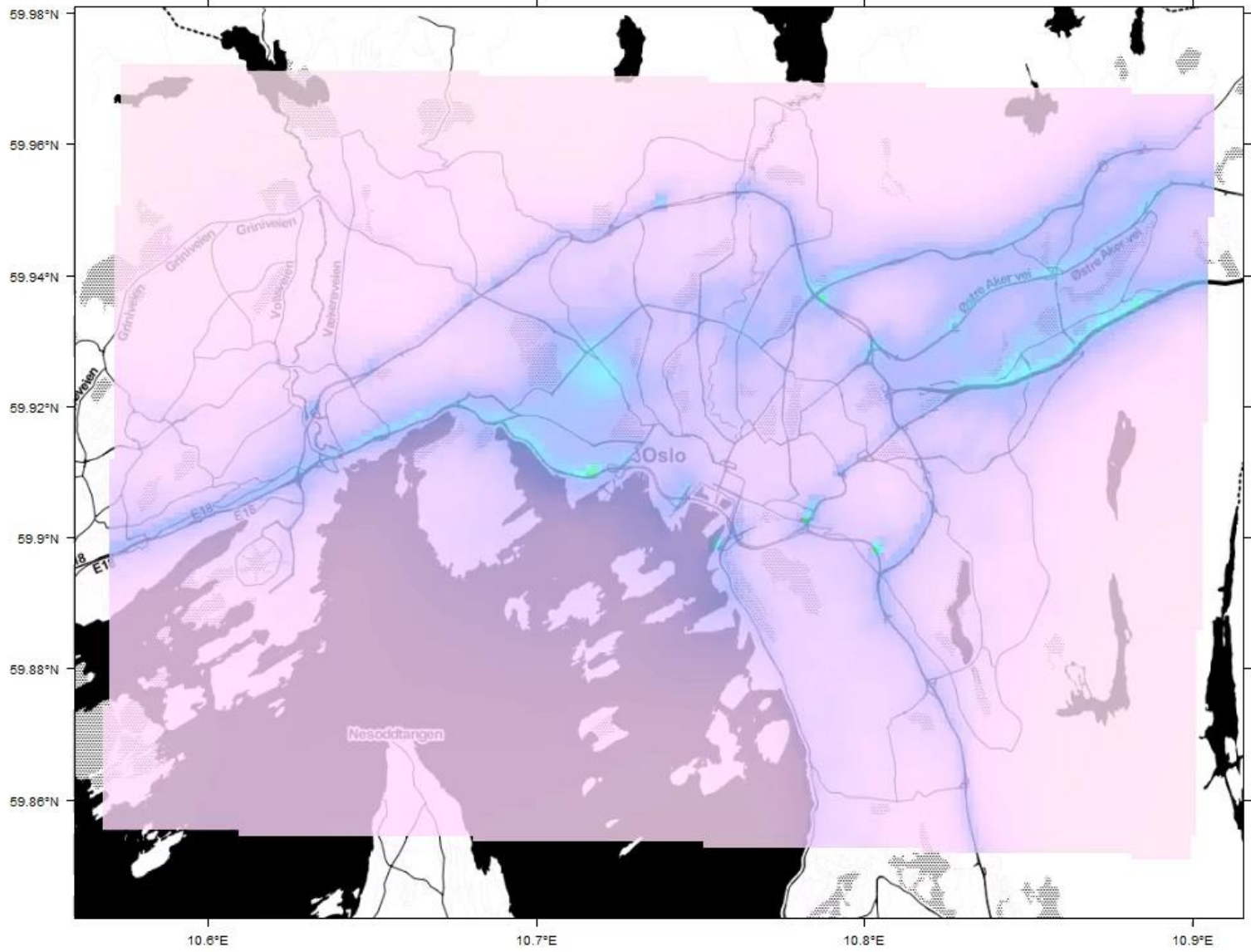


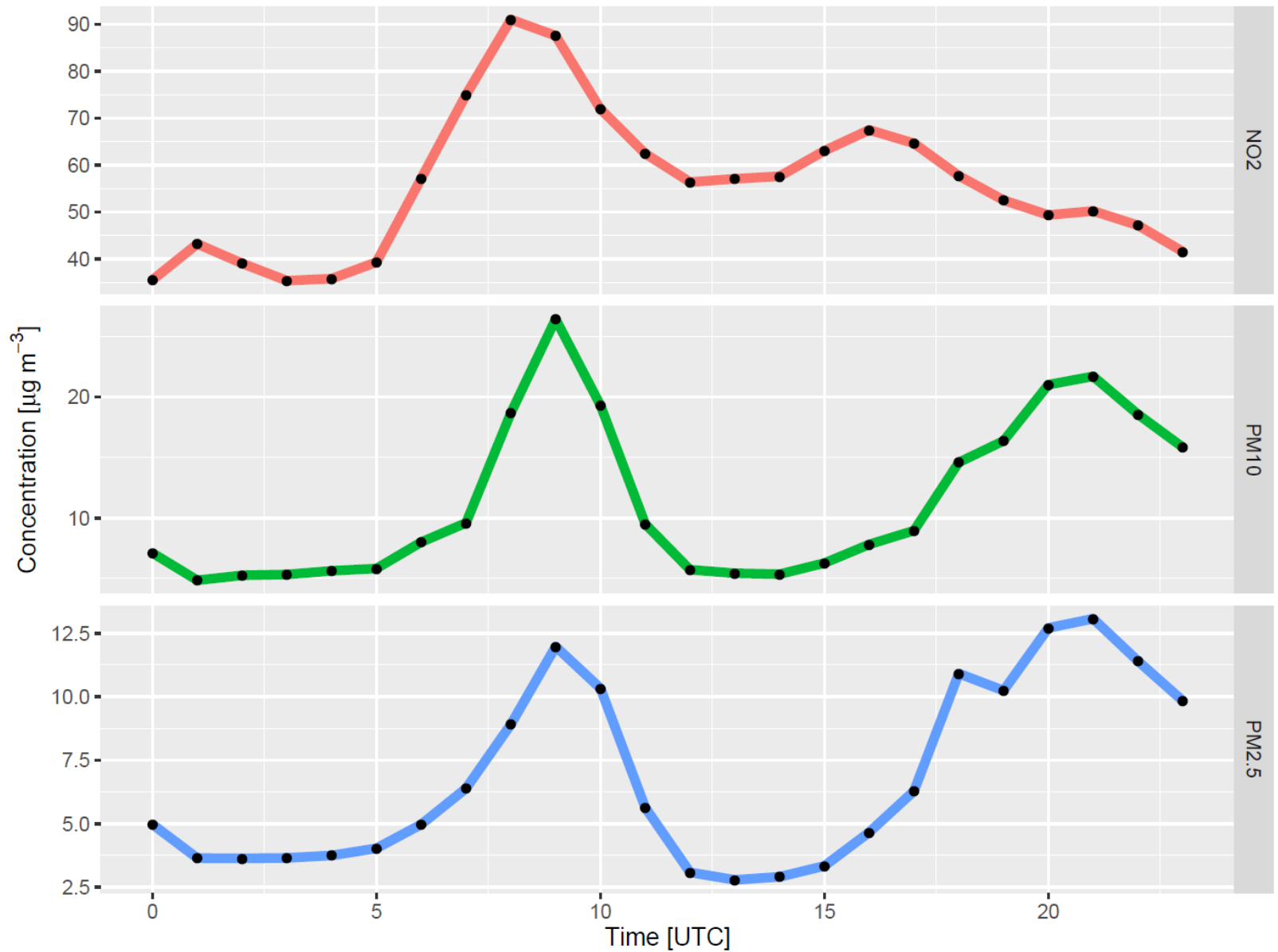




Example of 24 hours of data fusion results in Oslo, combining NO₂ measurements from the AQMesh units with a long-term average basemap derived from the EPISODE model, here shown for 6 January 2016

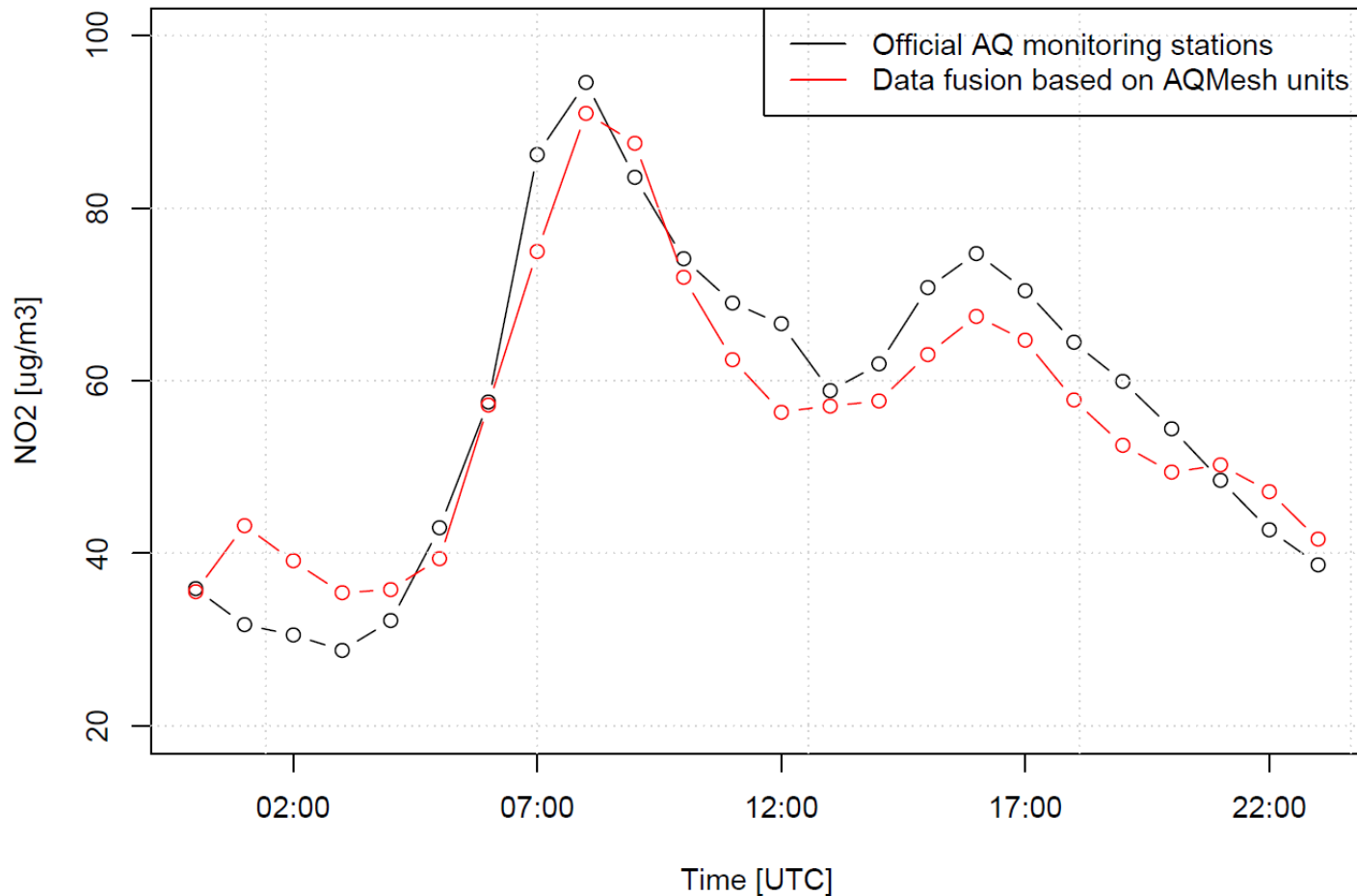
0000UTC



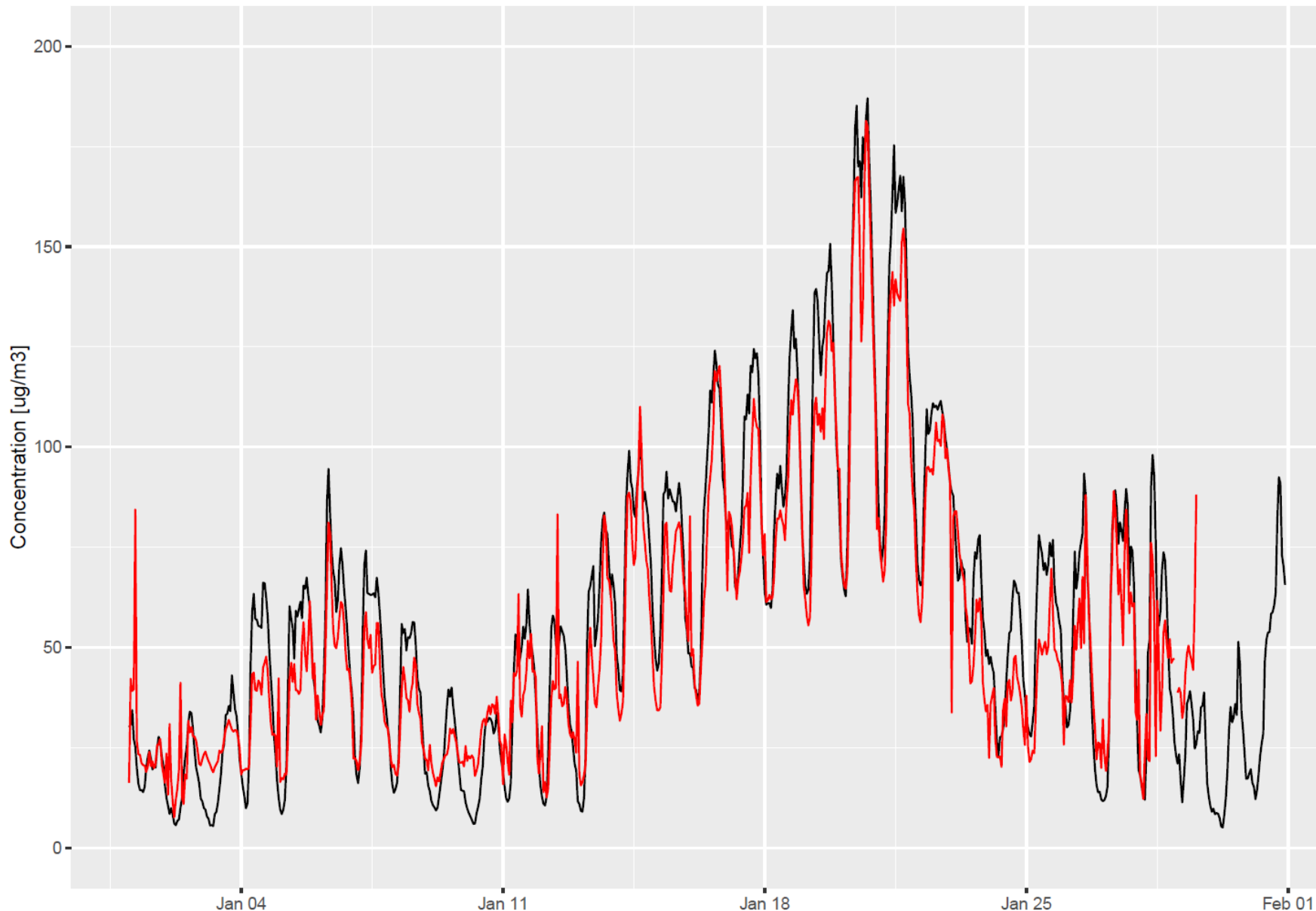


Data fusion maps: Daily cycle of NO₂, PM₁₀, and PM_{2.5} for Oslo on January 6 2016 (NO₂) and 22 March 2016 (PM).

Comparison to AQ monitoring stations



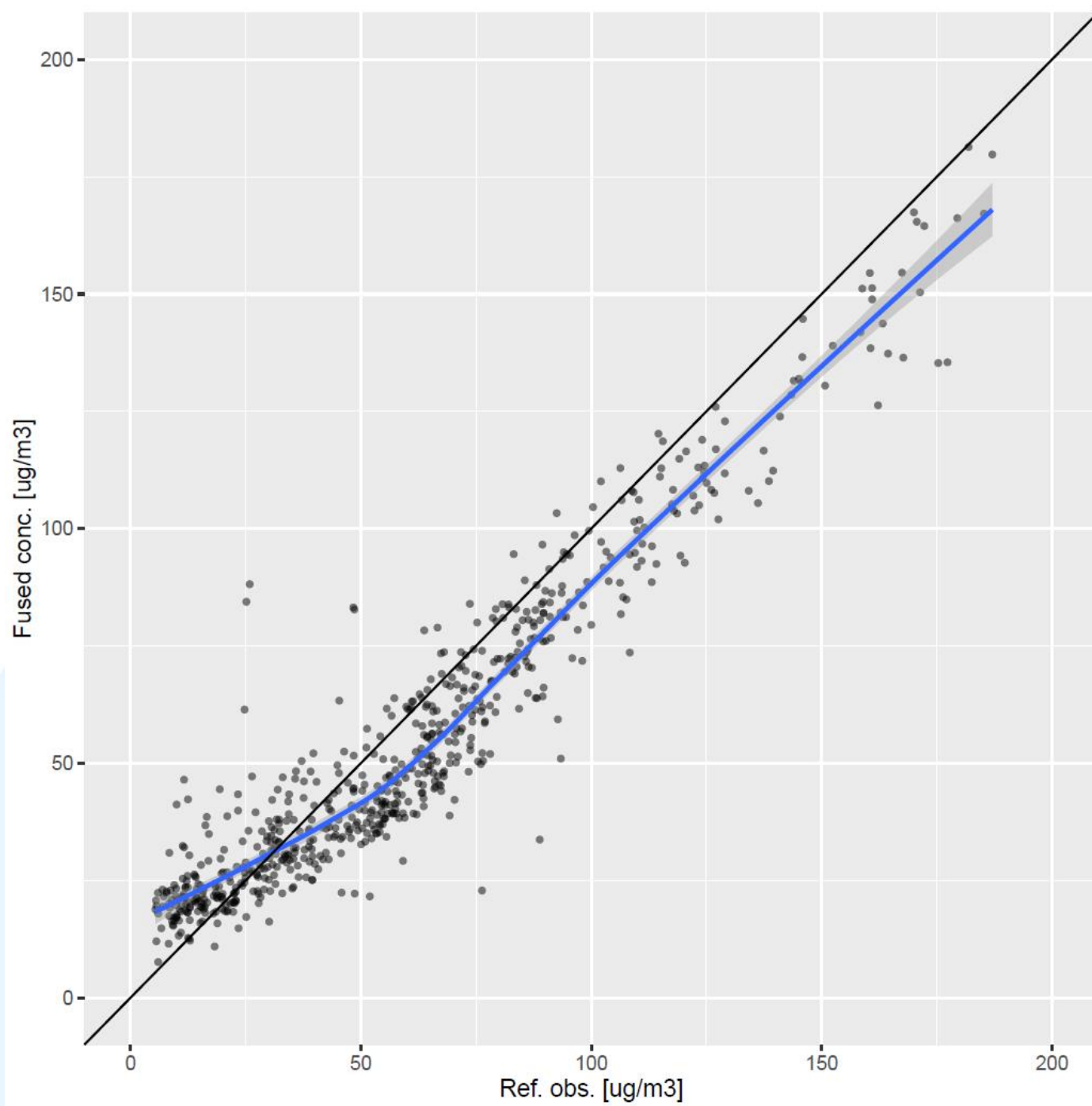
Entire daily cycle of NO₂ as measured by the reference air quality monitoring stations versus the NO₂ concentrations provided by the data fusion map.



Average NO₂ concentration in Oslo in January 2016

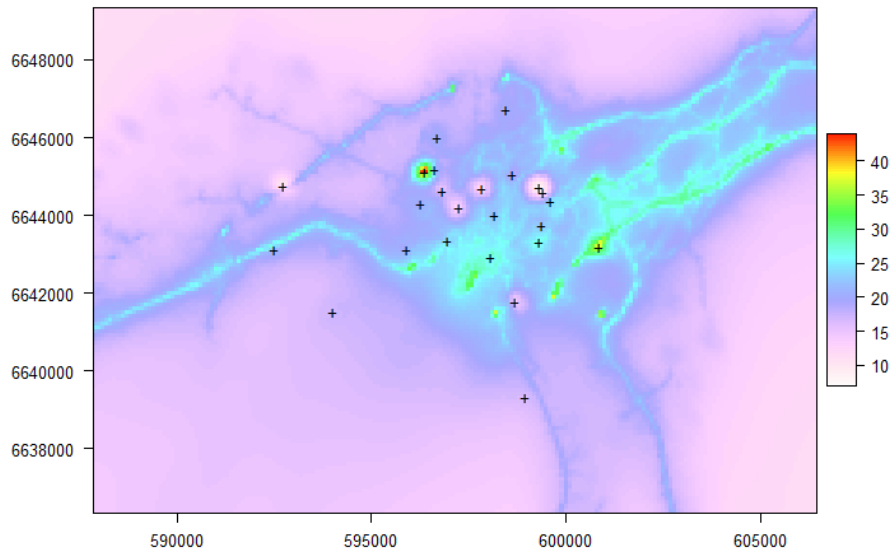
Black line: Reference AQM stations

Red line: Data fusion of AQMesh low-cost sensor network and EPISODE

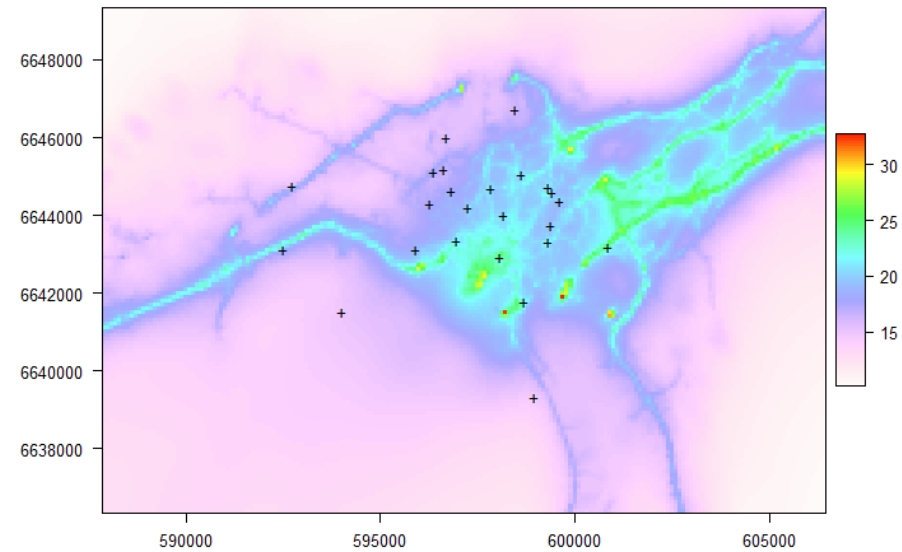


Field calibration of the sensors is crucial!

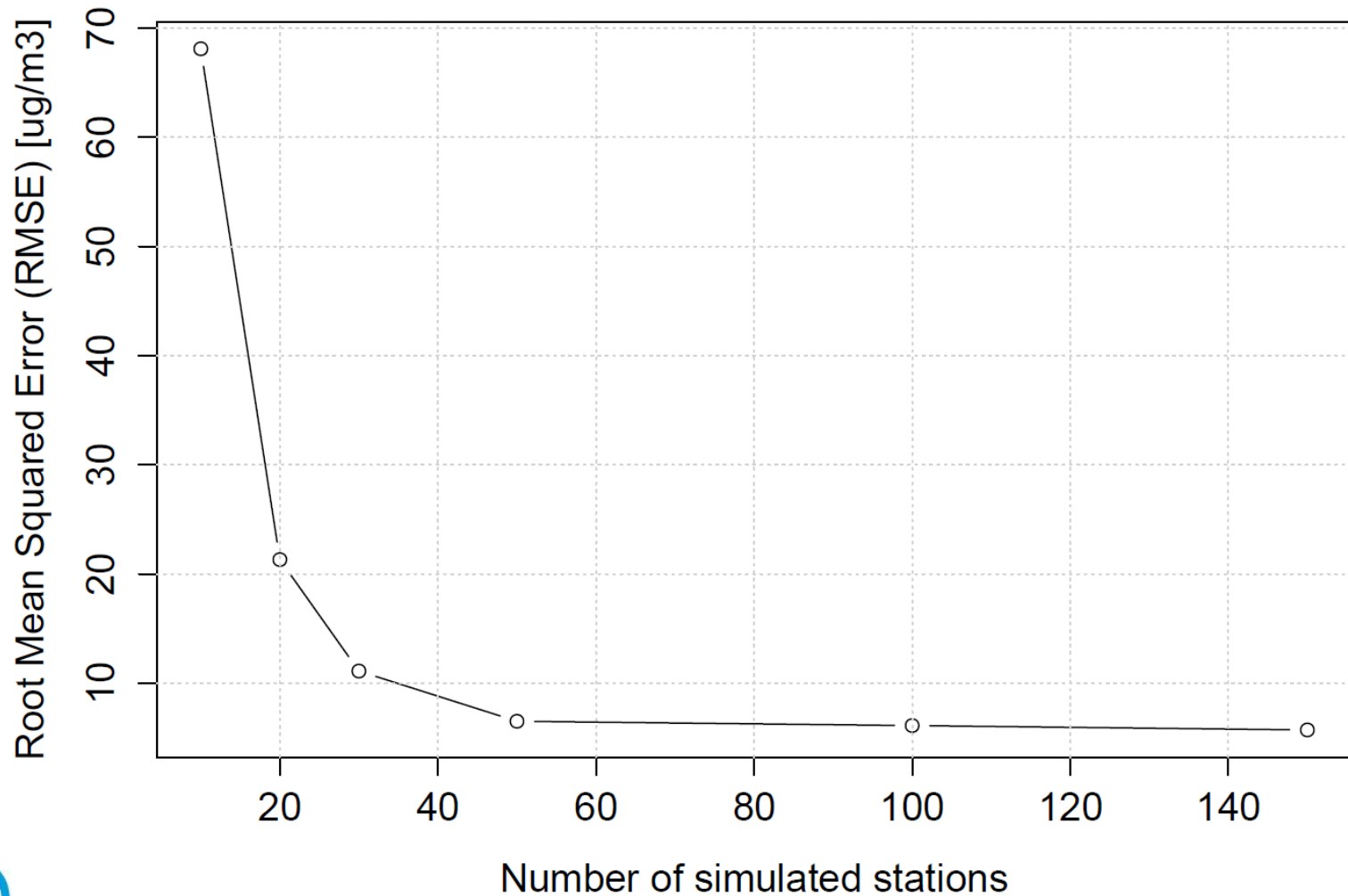
Without field calibration



With field calibration



Dependency of map quality on network size



Take-home messages

There are currently two exciting developments regarding new monitoring techniques:

1. Higher-resolution satellite data for AQ monitoring

Satellite observations of air quality, which so far are mostly useful for global and regional applications are finally reaching a spatial resolution that make them somewhat usable for (limited) urban-scale applications

2. Low-cost microsensors for AQ monitoring

Despite many challenges at the individual sensor level, low-cost microsensors allow for detailed high-resolution urban-scale mapping of air quality if several conditions are met:

- a) the “swarm knowledge” of the entire network is used
- b) the sensor observations are combined with output from an air quality model
- c) the sensors are calibrated in the field



NILU Seminar on new measurement
technologies, 31. 1. 2017, Kjeller, Norway



Thank you for your attention!

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