

Multi-scale measurements combined with inverse modeling for assessing methane emissions of Hamburg

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Quantification of methane emissions in Hamburg using a network of FTIR spectrometers and an inverse modelling approach

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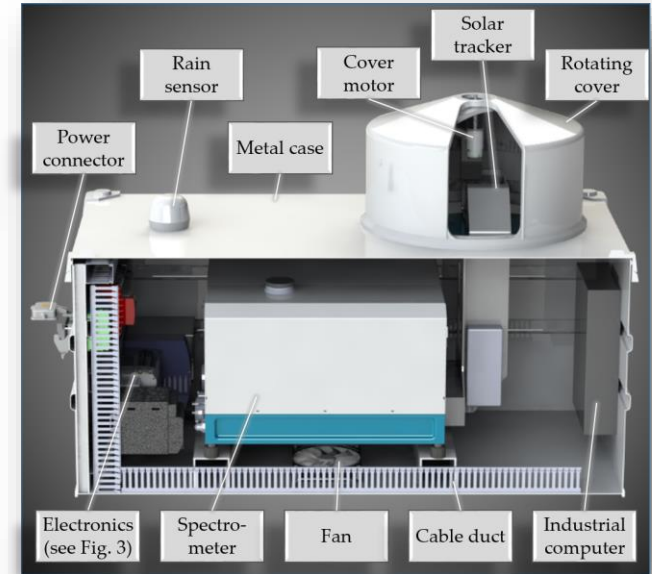
Image by A. Forstmaier, map data are from ©2021 Google, Landsat/Copernicus

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Major components: ground-based remote sensing

(Gisi et al. 2012, Heinle and Chen 2018, Dietrich et al. 2021)

- 4 FTIR sensor systems:
 - EM27/SUN: CH₄ column concentrations
 - Automated enclosure system: Sunny → measure, Rainy → cover close



→ Our system reduces the personnel costs to a minimum and increases the amount of measurement data to a maximum

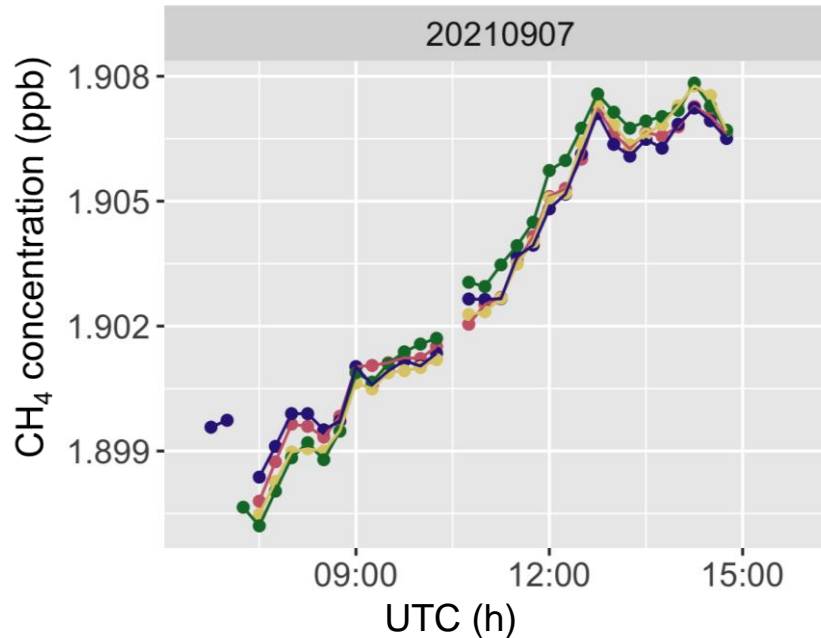
Multiscale sensor network



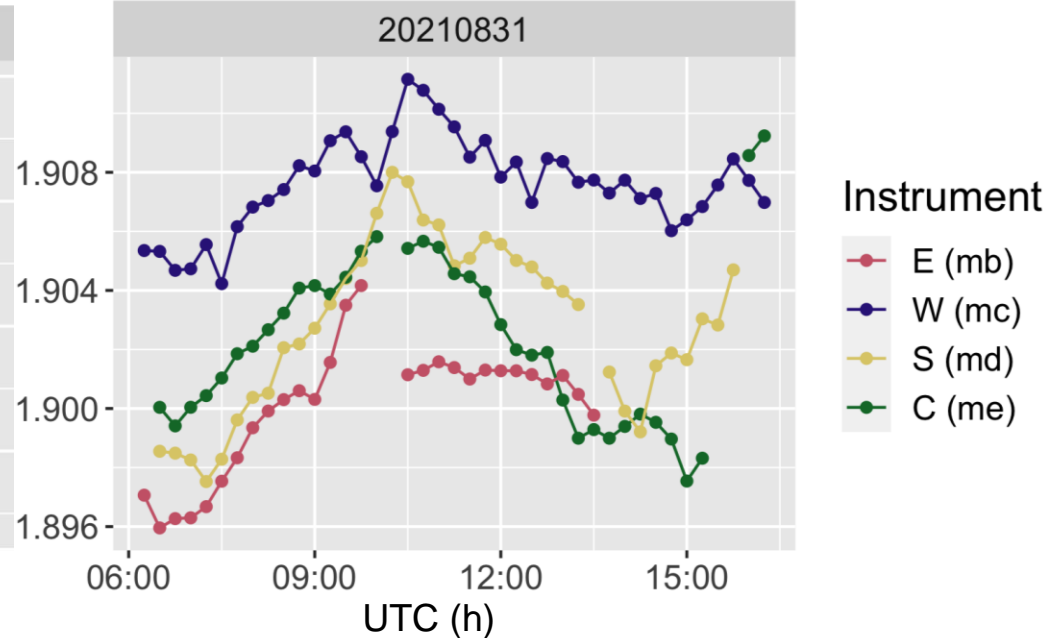
- 4 FTIR systems around high emission zones in Hamburg: upwind/downwind station for arbitrary wind conditions
- 1 LiDAR system in city center captures 3D wind information → improving transport modeling
- Isotope measurement → source attribution
- Mobile Measurements → refining the prior emission inventory

FTIR column measurements

side by side



network configuration



Framework for Estimating Emission (Bayesian Inversion)

(Jones et al. 2021)

$$\min_{x, b} [(Hx + Bb - y)^T S_\epsilon^{-1} (Hx + Bb - y) + (x_a - x)^T S_a^{-1} (x_a - x)]$$

y : observations

H : footprint matrix

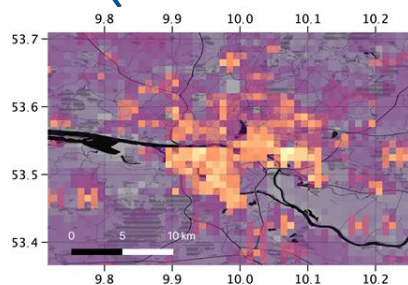
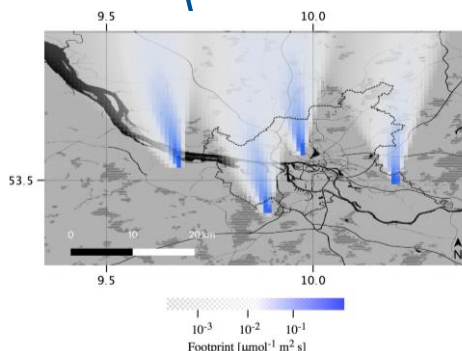
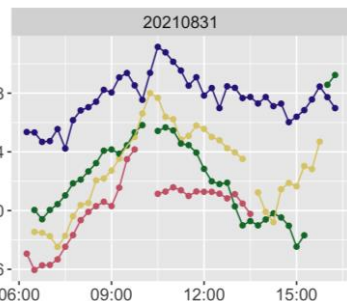
x : emissions

x_a : prior emissions

B : background influence matrix

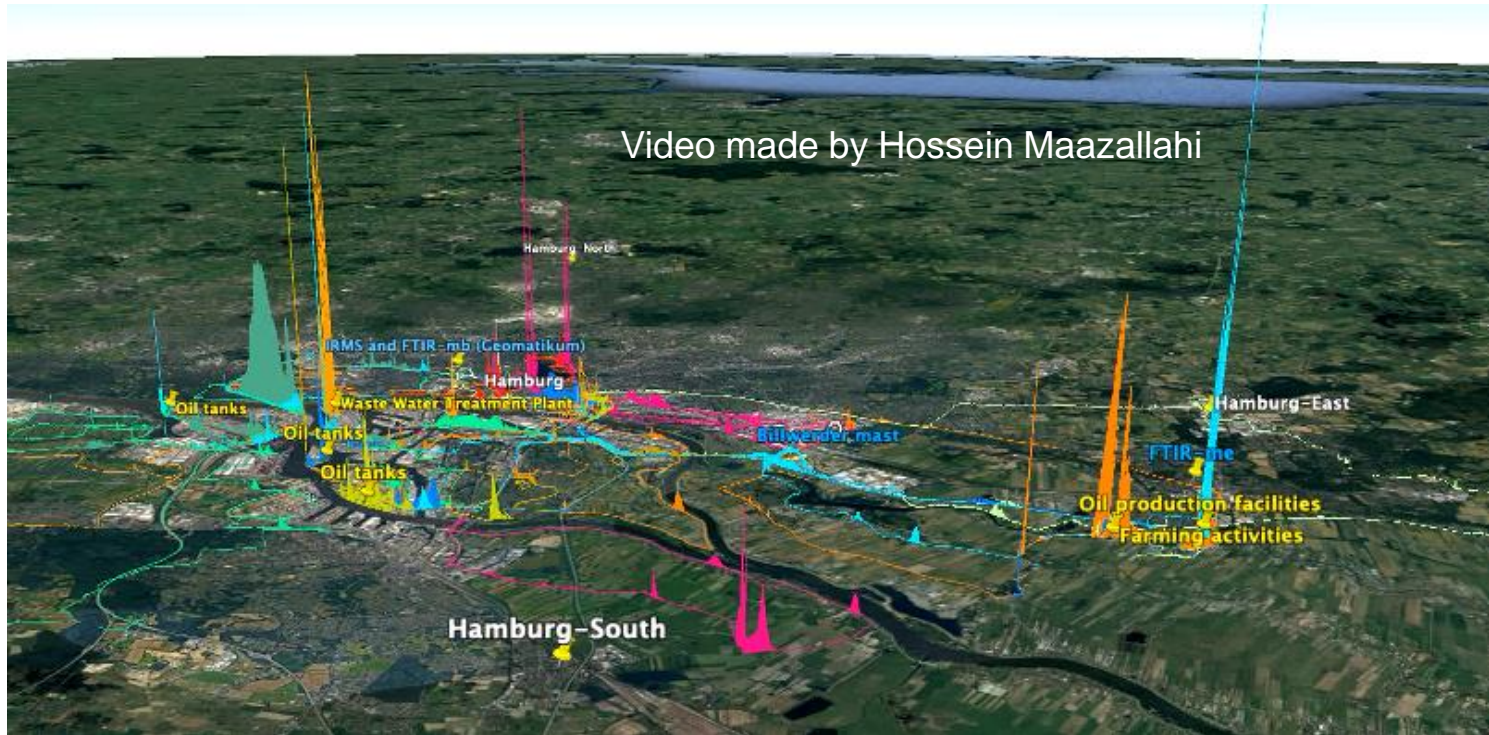
b : background concentration

S : error covariance matrix



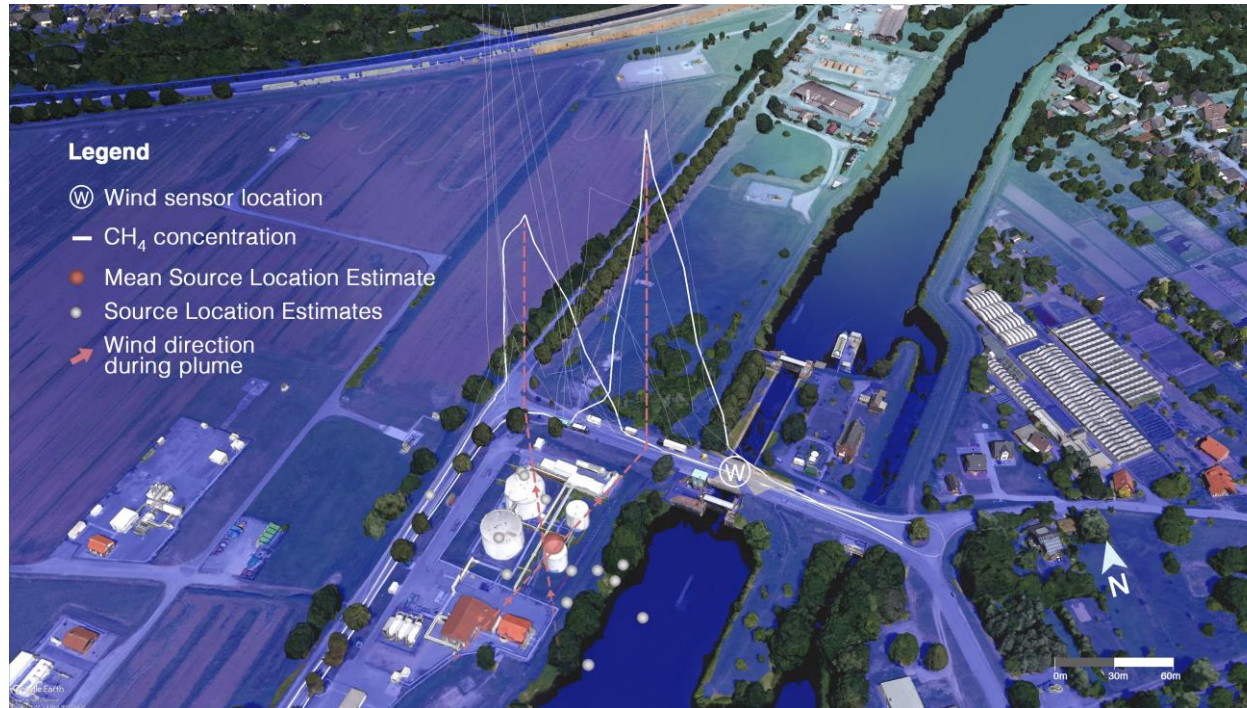
→ Approach: Minimizing a cost function to determine the emissions and the background influence

Mobile measurements



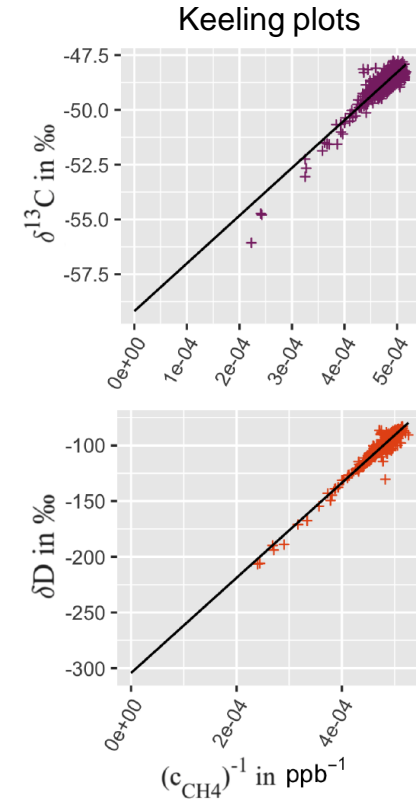
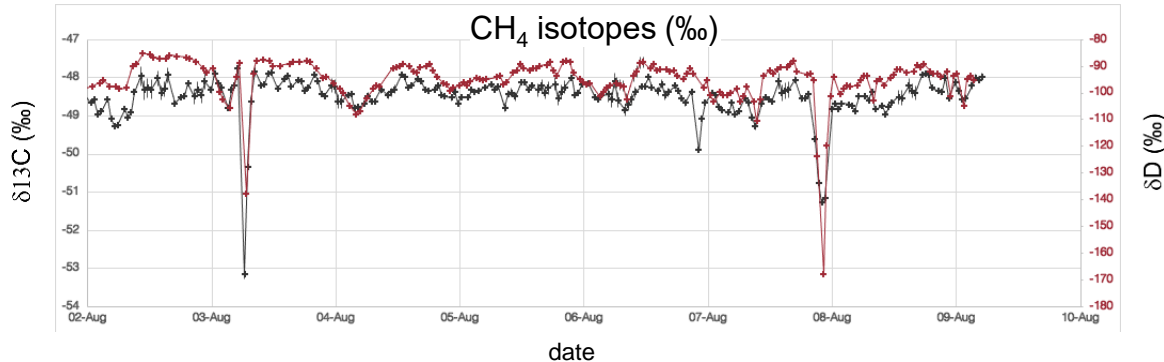
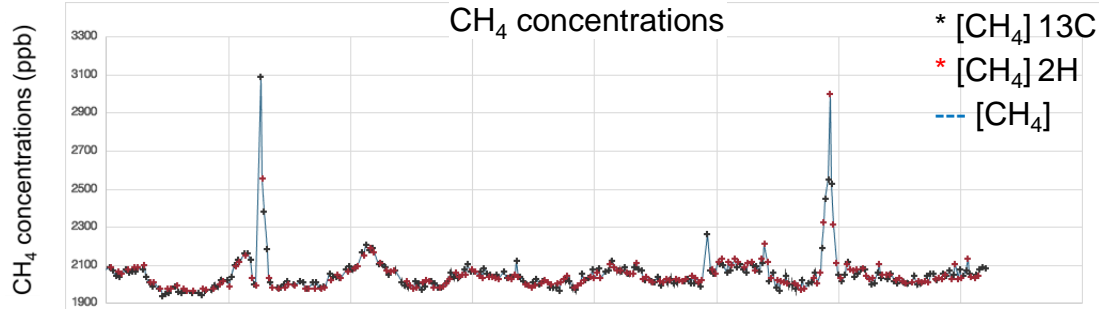
https://youtu.be/LupRJ_h9K5c

Quantification of unknown anthropogenic sources

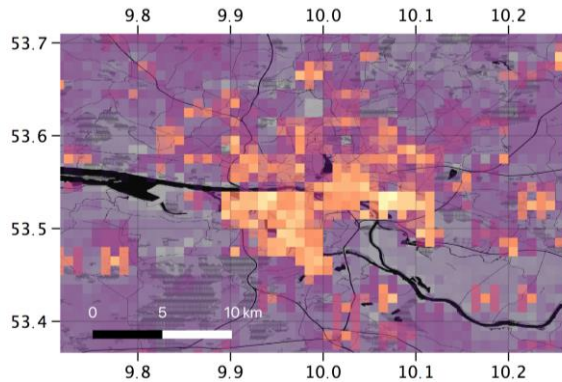


- Unknown point **sources** were detected in **Oil and Gas infrastructure** during mobile survey
- Source strength corrected → up to **100 times** stronger

Indication biogenic sources



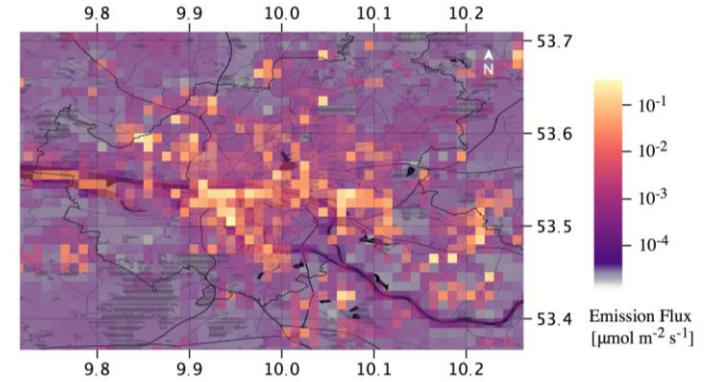
Improving an a-priori emission inventory



Original emission inventory
(e.g. TNO GHGco)



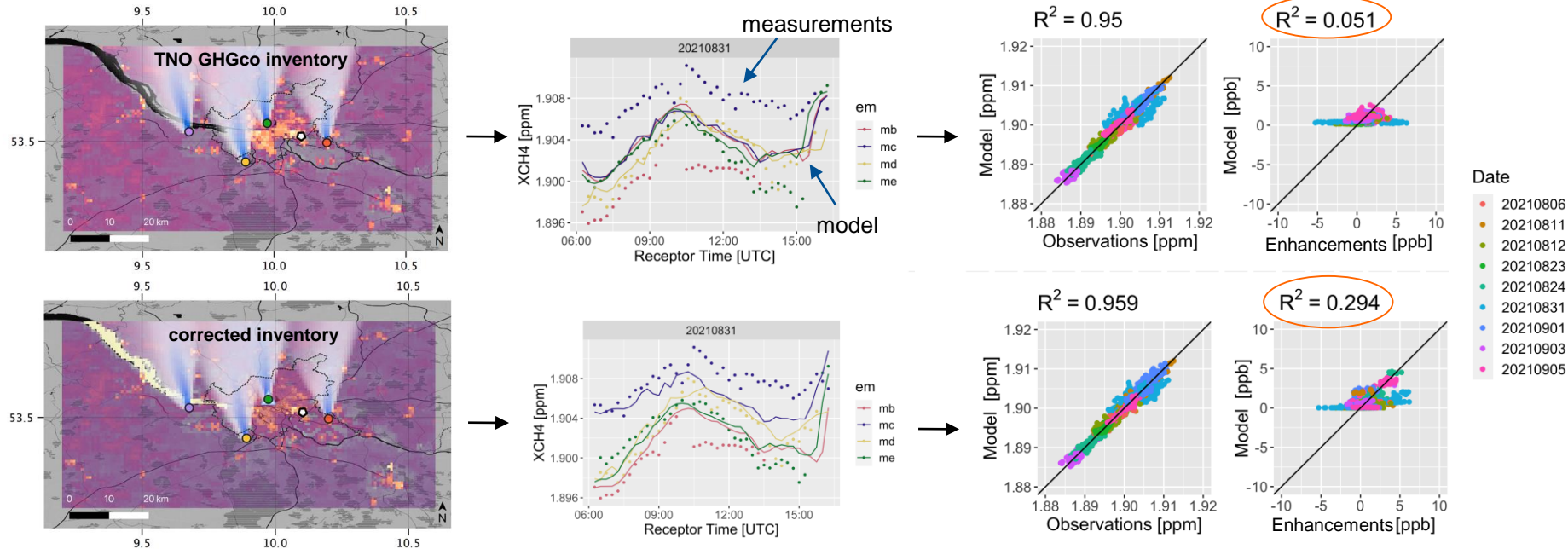
Mobile measurements
(e.g. using Picarro instruments)



Corrected emission inventory

- Correcting spatial distribution of an emission field using mobile measurements
- Adding natural sources (the Elbe river and the associated wetlands)

Inversion with unknown sources \rightarrow modeling improved



\rightarrow The modelled curves fit the measurements better, correlation of the enhancement improved significantly

Conclusion

- **Unknown sources** in the emission inventory uncovered:
 - Natural sources: Elbe river
 - Anthropogenic sources: Refineries, etc.
- **Combination** of ground-based remote sensing and mobile measurements leads to a better inversion
- Hamburg: **Anthropogenic emissions** have largest share; Natural source contribution is significant

