

Deriving CO₂ emissions of localized sources from OCO-3 XCO₂ and TROPOMI NO₃ data

B. Fuentes Andrade, M. Buchwitz, M. Reuter, H. Bovensmann, J.P. Burrows Institute of Environmental Physics, University of Bremen

Deutscher Wetterdienst Wetter und Klima aus einer Hand



Introduction

Carbon dioxide (CO₂) is the most important anthropogenic greenhouse gas leading to climate change and is mostly emitted by localized sources in the combustion of fossil

Emissions need to be monitored to track reduction efforts to comply with the objectives of the Paris Agreement. Top-down emission estimates can complement and verify bottom-up estimates [1].

The detection of CO, emission plumes and quantification of anthropogenic fluxes is challenging due to the small columnaverage concentrations resulting from anthropogenic emissions from individual point sources, compared to the background concentration and the satellite's instrument noise.

NO, is co-emitted with CO, in the combustion of fossil fuels and its vertical column densities can exceed background values and sensor noise by orders of magnitude in emission plumes. Therefore, it is a suitable tracer for recently emitted CO₂ [2].

Datasets

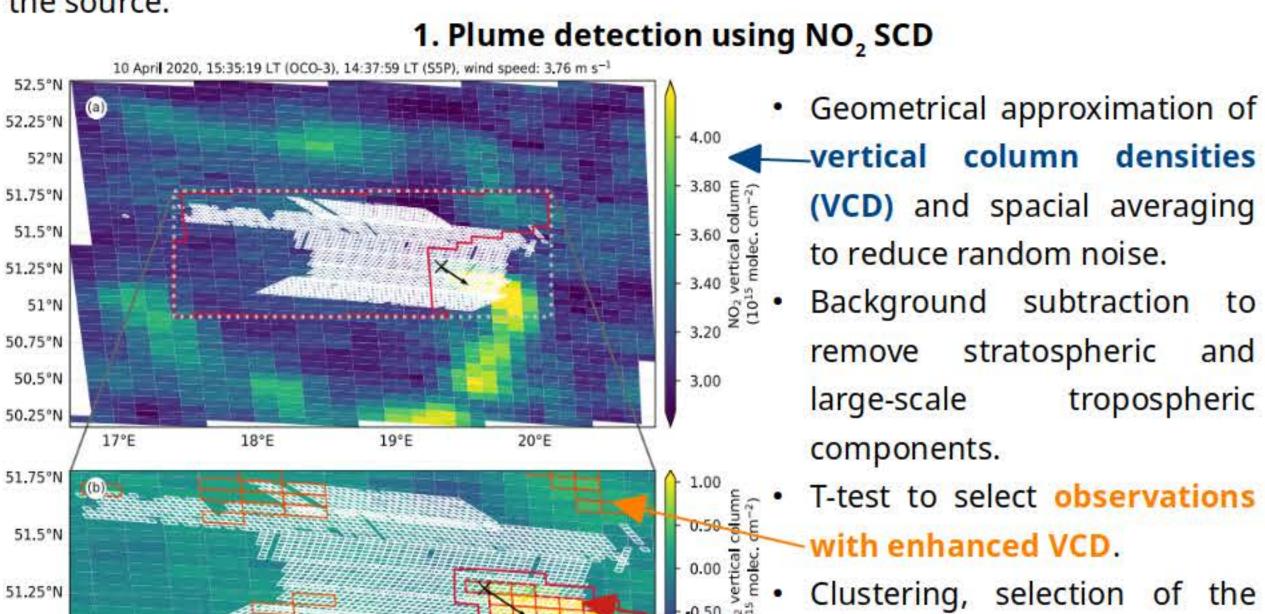
XCO, NASA's v10 level 2 Snapshot Area Maps (SAMs) retrieved by the Orbiting Carbon Observatory 3 (OCO-3) onboard the International Space Station.

NO₂ slant column densities (SCD) from the IUP Bremen retrieved by TROPOMI, onboard the Sentinel 5 Precursor (S5P), using DOAS (Differential Optical Absorption Spectroscopy).

Meteorological data from ECMWF's ERA5. Generated power per unit from the European Network of Transmission System Operators for Electricity (ENTSO-E).

Method

We estimated the CO₂ emissions using a cross-sectional flux method: by mass balance, the source rate is given by the flux through all cross sections downwind of the source.



background subtraction) (b) over the Bełchatów Power Station, whose location is marked as a black cross. The borders of the SAM pixels are marked as white polygons.

vertical column densities (VCD) and spacial averaging

- Background subtraction to remove stratospheric and tropospheric
- T-test to select observations with enhanced VCD.
- Clustering, selection of the cluster closest to the source (Fig. 1) and extension parallel to its borders to obtain the potential plume.

2. Flux estimation from XCO, data

- Background modelling (linear function of longitude and latitude), and subtraction from XCO₂ values to obtain the enhancements, ΔXCO₂ (Fig. 2).
- Average horizontal wind within the boundary layer weighted by the number of dry air particles (n_o), at the centre of each OCO-3 pixel.
- Filling in missing XCO₂ values by Inverse Distance Weighting (IDW) interpolation.
- Flux for each cross section (CS), k, perpendicular to plume track (Fig. 2), as:

$$\Phi_{k} = \frac{M_{CO_{2}}}{N_{A}} \sum_{i} v_{\perp,i} n_{e,i} \Delta l_{i} (\Delta XCO_{2})_{i},$$
 i: each pixel along k-th CS v_{\perp} : horizontal wind speed

 v_{\perp} : horizontal wind speed perpendicular to CS Δl;: length of pixel i Estimated emission

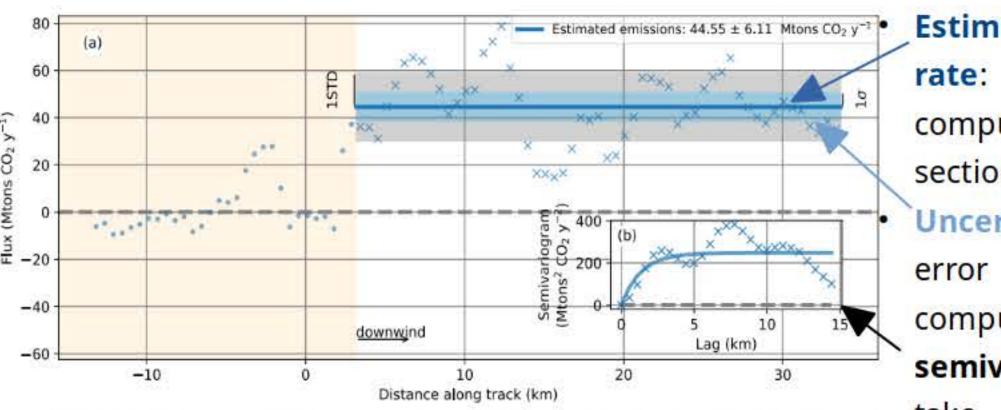


Figure 3. (a) Computed cross-sectional fluxes down-wind (upwind) of the source, represented as crosses (dots). (b) Semivariogram.

computed sectional fluxes. **Uncertainty: standard** error of the mean from computed semivariogram take auto-correlation

into account.

mean of the

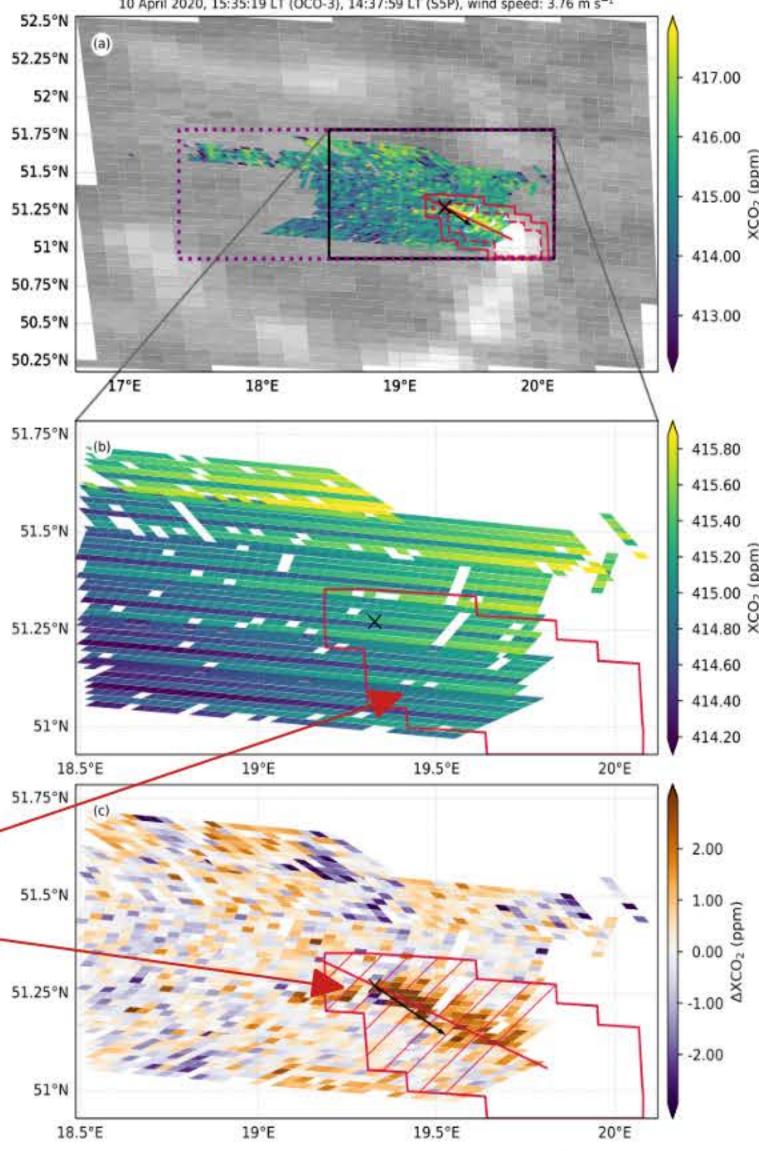
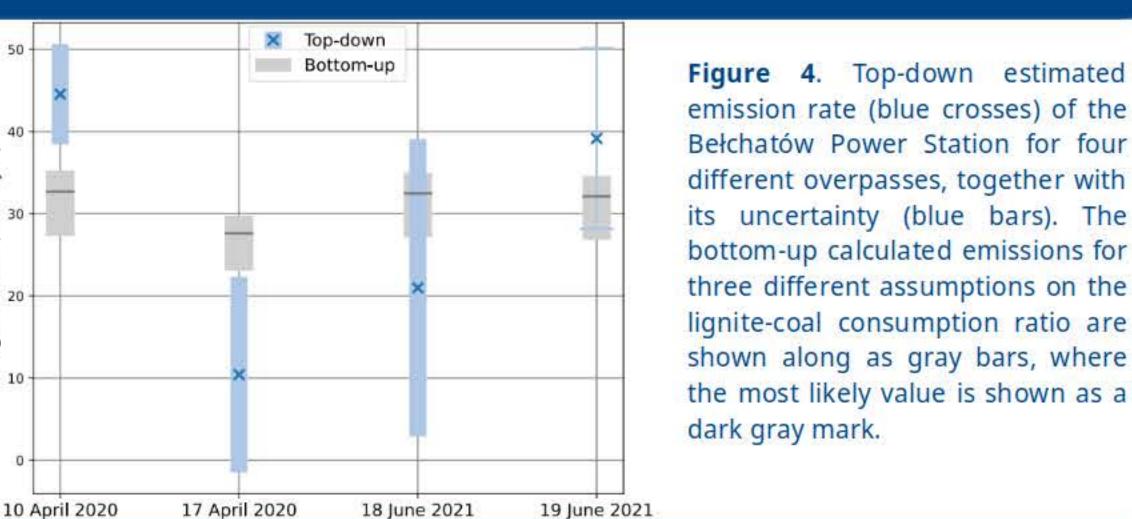


Figure 2. (a) OCO-3 XCO, SAM over the Belchatów Power Station. (b) Background model. (c) XCO, enhancements. The average horizontal wind within the potential plume is depicted as a black arrow at the source location. A subset of CS are shown as red solid lines perpendicual to the plume

3. Bottom-up emissions

Estimation of emitted emissions from the hourly generated power times an emission factor depending on the type of coal used. We compute a higher (lower) limit assuming the use of only lignite (bituminous coal), and a most likely estimate assuming the average lignitebituminous coal ratio used by the power plant. We average the hourly emissions within the estimated time interval that the CO, spends in the scene.





Conclusions and discussion

- Co-located observations of NO, and CO, help us detect the emission plume and characterize its shape.
- We are able to repeatedly monitor power plant CO₂ emissions. Our top-down estimates agree in 2 of 4 cases, within their uncertainty range, with the bottom-up calculated emissions.
- We estimated the emission uncertainty from the variability of the computed cross-sectional fluxes. However, this estimate is still incomplete, as, e.g., wind speed uncertainty is not yet considered. This may explain the differences between our top-down and bottom-up estimates shown in Fig. 4.

Acknowledgments

Financial support was provided by the German Meteorological Service, DWD, grant number 4819EMF06A. The OCO-3 XCO₂ data were produced by the OCO-3 project at the Jet Propulsion Laboratory, California Institute of Technology, and obtained from the OCO-3 data archive maintained at the NASA Goddard Earth Science Data and Information Services Center. The TROPOMI NO, data were produced by the Institute of Environmental Physics, University of Bremen. ERA5 meteorological information was obtained from the Copernicus Climate Change Service (C3S) (2017), operated by the ECMWF.

References

- [1] Janssens-Maenhout, G. et al. Toward an Operational Anthropogenic CO, Emissions Monitoring and Verification Support Capacity. Bulletin of the American Meteorological Society 101, no. 8 (1 August 2020): E1439-51.
- [2] Reuter, M. et al. Towards monitoring localized CO, emissions from space: co-located regional CO, and NO, enhancements observed by the OCO-2 and S5P satellites. Atmospheric Chemistry and Physics 19, 9371-9383 (2019).
- [3] Varon, D. J. et al. Quantifying methane point sources from fine-scale satellite observations of atmospheric methane plumes. Atmos. Meas. Tech. 11, 5673-5686 (2018).