Anonymous reviewer #1

We would like to thank the reviewer for the careful reading and constructive comments.

Please find below our point-by-point replies. For clarity, the reviewer's comments are displayed in black, our replies in blue, and suggestions for revised text in blue italics.

Main points

1.The comparison contrasts sets of CO2 simulations in lowest 20 m near surface made with emissions emitted either at surface or at more realistic heights. It should be mentioned that the observations are often made at higher elevations than 20 m, using either small towers (40-100 m) or tall towers (200-300 m tall). For modeling such observation sites, the conclusions presented in this study can serve more as a warning, rather than ready to use estimate of emission height-related bias.

This is a very valid point. For observations from tall towers the effect will likely be smaller, though not negligible. We added the following sentence at the end of Section 3.3

The impact on observations from tall tower networks measuring CO_2 some 100 m to 300 m above the surface (Bakwin et al., 1995; Andrews et al., 2014) will likely be somewhat smaller than suggested by the numbers above, especially in winter when the atmosphere is less well-mixed.

and the following sentence in the conclusions section:

Since measurements of CO_2 are often taken from towers some 100 m to 300 m above the surface (Bakwin et al., 1995), the impact on actual ground-based observations will likely be somewhat smaller.

2.Lagrangian plume models (eg STILT, FLEXPART) are often used in backward, adjoint mode for inverse modeling, and some are used in studies cited here (Page 3 Line 8). In that setting they have to assume emissions are mixed quickly in surface layer of nonzero thickness. It can be as thick as diurnally varying PBL height (Lin et al., 2003) or assigned a constant value (Ganshin et al., 2012). This is done to minimize sampling errors in estimating adjoint tracer concentration near surface, which is made by counting particles in the surface layer. In case of using relatively thick layer, the assumption may reverse the effect of neglecting CO2 emission height, towards having more errors from surface emissions rather than from elevated stacks.

Thank you for pointing this out. In fact, in FLEXPART the particle residence times can be written out for multiple vertical levels, which would offer the possibility to account for emissions at the surface and at higher altitudes separately. To our knowledge, this possibility has not yet been explored in the context of emissions from power plants and industrial sources, though. We added the following sentences:

In Lagrangian models such as STILT (Lin et al. 2003) or FLEXPART (Stohl et al. 2005), which are often used in backward, adjoint mode for inverse modelling, particles are typically sampled over a fixed vertical depth above the surface or relative to the height of planetary boundary layer to derive source-sensitivities. Similar to the release of emissions at the surface in Eulerian models, this ignores the potentially different sensitivities to emissions from elevated sources.

Technical corrections:

Page 2 Line 11 Add period after CO2 and before "Top-down".

There was already a period.

Page 3 Line 8 Add year to Lauvaux et al.

Year added.

Page 3 Line 15 It is worth noting earlier references to air quality modeling, such as SMOKE-CMAQ modeling system (eg Houyoux et al, 2002), to emphasize that the problem had long been recognized and addressed. For CO2 modelling audience it is also useful to mention that in air quality modeling effort is made to account for plume rise height of biomass burning emissions (eg Achtemeier et al, 2010).

Thank you. We have added these references and modified the paragraph to the following:

In the air quality modeling community, the importance of vertically distributing emissions has been recognized much earlier (e.g., Houyoux et al. 2002) and is now well established, especially for species such as SO_2 that are primarily emitted from power plants and industrial sources (Bieser et al., 2011; Mailler et al., 15 2013; Karamchandani et al., 2014; Guevara et al., 2014). Accounting for plume rise has also been demonstrated to be critical for biomass burning emissions (Achtemeier et al., 2011).

Page 4 line 6 Written as "COSMO is the first NWP model worldwide" - it appears that similar effort with ASUCA model (Shimokawabe et al., 2010) was done in about same time, suggest checking, rephrasing.

Right, efforts for GPU acceleration of NWP/climate models such as for ASUCA or for the CAM-SE model have indeed been acknowledged in the publication of Fuhrer et al. (2014). Nevertheless, we maintain that COSMO is the first NWP model run on GPUs in an operational weather forecasting context. We slightly changed the wording to " *COSMO is the first operational NWP model worldwide*".

Page 7 line 16 Suggest revising "In order to prevent re-heating," as "In order to avoid re-heating,"

Done

Page 12 Line 10 Need to add year to Bagley et al.

Done

Anonymous reviewer #2

We would like to thank the reviewer for the careful reading and constructive comments.

Please find below our point-by-point replies. For clarity, the reviewer's comments are displayed in black, our replies in blue, and suggestions for revised text in blue italics.

Specific comments:

Page 4, Line 23: It might be good to include, after the sentence stating the height of the lowest vertical layer, a sentence stating that the vertical allocations will be described in more detail in section 2.4. Otherwise, an interested/over-eager reader may go searching for that information prematurely (as I just did).

We added a corresponding reference to Section 2.4.

Page 5, Line 16: Is there a citation yet for version 3 of the TNO/MACC inventory?

No, Kuenen et al. (2014) referring to version 2 is still the most appropriate reference.

Page 5, Line 19: It would be nice to have a reminder of the SNAP categories here.

We added a reference to Table2 describing the 10 SNAP categories.

Page 5, Line 21: What are the 7 source categories used by the city of Berlin?

We added a reference to the publicly available final report of the SMARTCARB project, where more details on the Berlin inventory (including the mapping of its 7 categories to SNAP) can be found, see Table 2 in the report (https://www.empa.ch/documents/56101/617885/FR_Smartcarb_final_Jan2019.pdf)

Page 5, Line 24: Please go into more detail about how this "mapping" was performed. There may (necessarily) be inherent assumptions here that the reader should know about.

The sentence was not quite correct, since we did not perform a mapping to SNAP but rather a mapping to three broad source groups represented as separate tracers in our simulations. We changed the sentence to the following:

In our simulations, separate tracers were included for three broad source groups: traffic, industry, heating. The attribution of the 10 SNAP categories in TNO/MACC-3 and the 7 source categories in the Berlin inventory to these groups are described in detail in the final report of the SMARTCARB project (Kuhlmann et al., 2019).

Page 5, Line 28: How was this merging performed? An average of the two at each pixel? Figure 1: In the left image, what do the colors represent? Different source categories or emission fluxes? What are the units? If these really are both in units of flux, then I am wondering why there appear to be more strong point sources in the southern portion of the right image (post-projection) than the left (pre-projection).

We added the following sentence to clarify the merging:

The merged emission e per grid cell was estimated as e = (1-f) e_{TNO} + e_{Berlin}, where f is the fraction of the grid cell area covered by Berlin.

Regarding Figure 1: The lefthand figure is only an illustration of the GIS-type of inventory, showing at the same time area, point and line sources, each of them in different units (in kg s^{-1} for point, kg s^{-1} m^{-1} for line and kg s^{-1} m^{-2} for area sources) each being colored separately between minimum and maximum values. In contrast to the right-hand figure, the left-hand figure can thus only be interpreted in a qualitative way. We will add this information.

Page 6, Lines 8-9: In what context were these flux tower fits performed to tune the remaining VPRM parameters? Was this further part of the SMARTCARB simulations? If so, could we get a little more information about which flux towers were used (or a citation)?

We will add the following sentence explaining the origin of the VPRM parameters used here:

Model parameter values were taken from a previous study (Kountouris et al., 2018), in which they were optimized using data from 47 European eddy covariance measurement sites for the year 2007.

Page 6, Lines 14-15: When the vertical layers were divided up to add new layers, how were the emissions distributed along those layers? For example, if the 4-90m EMEP level was divided into 3 COSMO-GHG levels–4-30, 30-60, and 60-90–were the emissions in 4-30 assigned to be identical, more, or less than those assigned to 60-90?

We added the following sentence to further clarify the procedure:
The attribution of emissions to these sub-layers was done in a rough way, for example by placing a larger proportion of emissions from "combustion in manufacturing industry" into the upper parts but distributing emissions from "Non-industrial (residential) combustion'" rather evenly over the three layers between 4-90m, considering that industrial sources are likely to emit at higher altitudes.

All information on the finally applied vertical attribution of emissions to the layers is given in Tables 2 and Figure 2 of the manuscript.

Page 6, Line 18: Sentence starts with "This modification"; is this referring to the second of the 2 modifications? Or is this intended to read "These modifications"? If it is only referring to the second modification, then where did the 10% number come from in the first modification?

We changed the wording to "*This latter modification*". The 10% was a very rough assumption to account for CO_2 emissions from the waste sector not associated with waste incinerators.

Page 9, Line 14: What is the July difference in ppm, to compare against the 1.6 ppm in the previous (winter) paragraph?

The difference has a strong diurnal cycle in summer. We changed the corresponding sentence as follows:

In summer, the differences are generally smaller and exhibit a pronounced diurnal cycle. Differences are about 1 ppm at night and almost vanish (about 0.1 ppm) in the afternoon.

Page 9, Line 32: Is it stated somewhere how the column mean dry air mole fraction is calculated? I assume it is, as the name suggests, the mean concentration at each pixel along the full vertical column for each pixel in the domain, but it is worth stating this explicitly somewhere. Apologies if I just missed it.

The column mean dry air mole fraction is the column integrated amount of CO_2 (moles per m^2) divided by the column integrated amount of dry air (moles per m^2). We will add this information. We changed all references to "column mean" to "column-averaged".

Page 10, Lines 19-20: This last sentence is very confusing for me. What does the "95th percentile" refer to?

We will add "(see last column in Table 4)", since we are referring to the entries in this table here.

Page 10, Lines 21-22: Maybe not in this sentence, but somewhere in the next paragraph or two, it would be nice to support the "very large" differences assertion with a number (from Figure 12, presumably).

Since Figure 12 is only an illustration of the problem, we would like to refrain from providing a number here. The difference is as large as the amplitude of the plume, which, depending on weather situtation, can be up to a few ppm in total column XCO_2 and even higher in individual vertical levels (as suggested by Fig. 12).

Page 12, Line 18: Similar to a comment above, I am still not clear on what the "highest percentiles" refers to.

As explained on page 9, we did not only look at mean values but also at different percentiles of the frequency distributions . The statistics for different percentiles are presented in Table 4. We think this should be sufficiently clear.

Technical corrections:

Page 1, Line 12: Should read "The results suggest that…"

Done

Page 1, Line 18: Comma should be moved, I think: "…, the contributions from anthropogenic and natural fluxes and their sensitivity to climate change, and political and societal drivers."

Thank you for spotting. Done.

Page 2, Lines 13-16: I suggest putting commas around "such as the WMO's Integrated Global Greenhouse Gas Information System (IG3IS)" to make it an appositive, otherwise the sentence is a run-on.

Commas added.

Page 3, Line 4: "Hogue et al." missing a year, Line 8: "Lauvaux et al." also missing a year

Done

Page 4, Line 10: I am not able to make sense of the phrasing "to offload compute intensive part". Is it trying to say "to offload computationally-intensive parts"? If not, is it possible to reword/clarify?

Yes, that's what it means. "Compute-intensive" is a technically term commonly used in high-performance computing. We added the hyphen and changed "part" to "parts".

Page 4, Lines 11-12: The phrasing "which allows to increase…" feels awkward (and maybe incorrect?). I would consider rephrasing the sentence as "which allows for an increase in spatial and temporal resolution, and which greatly…"

This wording was awkward, indeed. Changed as suggested.

Page 4, Line 16: I think it's supposed to be "an NO_2" not "a NO_2", but I could be wrong.

You are probably right, changed.

Page 4, Lines 19-20: "250-km-wide" should be hyphenated to act as one adjective

Done

Page 12, Lines 3, 5, 8, 10, and 12: Might want to say "near" or "around" or some other phrase instead of "of/on the order of" so that readers do not think this is referring to an order of magnitude

Thank you. Mostly changed to "around".

Page 12, Line 11: "Bagley et al." does not have a year.

Year added.

Page 12, Line 25: Change "However, also these methods…" to "However, these methods also…" to correct awkward phrasing

Changed as suggested.

Accounting for the vertical distribution of emissions in atmospheric CO$_2$ simulations

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Abstract. Inverse modeling of anthropogenic and biospheric CO$_2$ fluxes from ground-based and satellite observations critically depends on the accuracy of atmospheric transport simulations. Previous studies emphasized the impact of errors in simulated winds and vertical mixing in the planetary boundary layer, whereas the potential importance of releasing emissions not only at the surface but distributing them in the vertical was largely neglected. Accounting for elevated emissions may be critical, since more than 50% of CO$_2$ in Europe is emitted by large point sources such as power plants and industrial facilities. In this study we conduct high-resolution atmospheric simulations of CO$_2$ with the mesoscale model COSMO-GHG over a domain covering the city of Berlin and several coal-fired power plants in eastern Germany, Poland and the Czech Republic. By including separate tracers for anthropogenic CO$_2$ emitted only at the surface or according to realistic, source-dependent profiles, we find that releasing CO$_2$ only at the surface overestimates near-surface CO$_2$ concentrations in the afternoon on average by 14% in summer and 43% in winter over the selected model domain. Differences in ~~column mean~~ column-averaged dry air mole fractions XCO$_2$ are smaller, between 5% in winter and 8% in summer, suggesting smaller yet non-negligible sensitivities for inversion modeling studies assimilating satellite rather than surface observations. The results ~~suggests~~ suggest that the traditional approach of emitting CO$_2$ only at the surface is problematic and that a proper allocation of emissions in the vertical deserves as much attention as an accurate simulation of atmospheric transport.

Copyright statement. TEXT

1 Introduction

Reliably predicting future atmospheric concentrations of CO$_2$, the most important long-lived greenhouse gas, requires a profound understanding of the global carbon cycle, the contributions from anthropogenic and natural fluxes ~~,~~ and their sensitivity

to climate change, and political and societal drivers. An important tool for advancing our knowledge of the carbon cycle is the integration of CO_2 observations with atmospheric transport simulations in an inverse modeling framework. Global inverse modeling systems helped to better constrain the terrestrial carbon budget, to allocate the global land sink to different continents and ecosystems, and to assess inter-annual variability and the sensitivity to climate variations (e.g. Bousquet et al., 2000; Chevallier et al., 2010; Peylin et al., 2013; van der Laan-Luijkx et al., 2017; Rödenbeck et al., 2018). Mesoscale inverse modeling systems assimilating observations from dense, regional in situ networks are increasingly being used to study biospheric fluxes at the regional scale (e.g. Sarrat et al., 2009; Goeckede et al., 2010; Broquet et al., 2011; Meesters et al., 2012).

The Paris Climate Agreement adopted in 2015 (United Nations Framework Convention on Climate Change, 2016), which requires each signing partner nation to accurately report its GHG emissions and to reduce emissions in the future following its Nationally Determined Contribution, has boosted the interest of the atmospheric science community in quantifying not only natural fluxes but also anthropogenic emissions of CO_2. "Top-down" inverse estimation of anthropogenic emissions from atmospheric observations has in fact been proposed as an independent method to complement the traditional "bottom-up" collection of national emission inventories (Nisbet and Weiss, 2010). Several initiatives at the national and international level, such as the WMO's Integrated Global Greenhouse Gas Information System (IG3IS), have been launched to advance and harmonize inverse methods with the long-term vision to establish these methods as a policy-relevant verification and support tool. A summary of the state of the science in inverse modeling for verification of greenhouse gas inventories has recently been presented by Bergamaschi et al. (2018a).

Because of the great challenge of accurately estimating CO_2 fluxes and distinguishing between the biospheric and anthropogenic contributions, the scientific community is calling for a globally integrated carbon observation and analysis system. This system should build on a substantially expanded ground-based and satellite observation capacity and should integrate observations and bottom-up data with atmospheric transport modeling in an inverse framework (Ciais et al., 2014). The individual components and necessary steps towards an operational European system for quantifying anthropogenic CO_2 emissions were outlined in two recent reports to the European Commission (Ciais et al., 2015; Pinty et al., 2017).

A central component of this system is a constellation of CO_2 satellites with imaging capability similar to the CarbonSat concept proposed by Bovensmann et al. (2010). The satellites are currently planned to be implemented in the framework of the European Earth observation program Copernicus, and are commonly referred to as Sentinel-CO_2. These satellites should be able to quantify the CO_2 emissions from large sources such as power plants and cities during single overpasses. Corresponding observing system simulation experiments (OSSEs) were presented by Pillai et al. (2016) and Broquet et al. (2018).

In order to establish the requirements for ~~such a constellation of~~ the constellation of Sentinel-CO_2 satellites, the European Space Agency (ESA) has launched several scientific support studies including SMARTCARB, a project that focused on the potential of complementary satellite measurements of NO_2 or CO to improve the quantification of CO_2 emissions from cities and power plants. The present study makes use of the high-resolution atmospheric transport simulations conducted in SMARTCARB to address a specific topic beyond of the main focus of the project. The core results of the SMARTCARB study will be presented elsewhere.

Estimating CO_2 fluxes by inverse modeling requires accurate simulation of observed atmospheric CO_2 concentrations. Systematic model biases are particularly critical as they tend to directly translate into biased flux estimates. Several studies investigated the impact of uncertainties in atmospheric transport on simulated CO_2 concentrations and how they could be accounted for in an inverse modeling framework (e.g. Lin and Gerbig, 2005; Chan et al., 2008; Lauvaux et al., 2009). A particular focus was placed on potential biases introduced by errors in vertical transport in the PBL (Gerbig et al., 2008; Kretschmer et al., 2012, 2014). Uncertainties associated with bottom-up CO_2 emissions were related to uncertainties in the horizontal gridding (Hogue et al., 2016) or the representation of the temporal variability (Liu et al., 2017).

Uncertainties associated with the vertical placement of emissions, conversely, have been largely ignored so far. In fact, in the vast majority of CO_2 atmospheric transport and inverse modeling studies, CO_2 emissions were released exclusively at the surface (e.g. Sarrat et al., 2009; Lauvaux et al., 2009; Broquet et al., 2011; Ganshin et al., 2012; Meesters et al., 2012; Pillai et al., 2016; Lauvaux et al., 2016; Liu et al., 2017; Graven et al., 2018; Fischer et al., 2018). ~~This approach is~~ In Lagrangian models such as STILT (Lin et al., 2003) or FLEXPART (Stohl et al., 2005), which are often used in backward, adjoint mode for inverse modelling, particles are typically sampled over a fixed vertical depth above the surface or relative to the height of planetary boundary layer to derive source-sensitivities. Similar to the release of emissions at the surface in Eulerian models, this ignores the potentially different sensitivities to emissions from elevated sources.

These approaches in Eulerian and Lagrangian models are questionable given the fact that a large proportion of CO_2 is released from point sources such as power plants well above the surface. Smoke stacks are in fact designed to minimize the impact on concentrations at the ground. This is particularly important during situations with stable inversions in winter or on clear nights. However, even in well-mixed conditions, an elevated release may lead to a faster dilution and different propagation of the signal due to wind speeds and directions changing with altitude.

In the air quality modeling community, the importance of vertically distributing emissions ~~is~~ has been recognized much earlier (e.g. Houyoux et al., 2002) and is now well established, especially for species such as SO_2 that are primarily emitted from power plants and industrial sources (Bieser et al., 2011; Mailler et al., 2013; Karamchandani et al., 2014; Guevara et al., 2014). Accounting for plume rise has also been demonstrated to be critical for biomass burning emissions (Achtemeier et al., 2011). The main goal of this study is to demonstrate that this is also critical for the simulation of atmospheric CO_2 concentrations.

We employ very high resolution (1.1 km $\times$ 1.1 km) simulations for the year 2015 conducted over a 750 km x 650 km wide domain centered on the city of Berlin, Germany. The simulations included separate tracers representing CO_2 emitted only at the surface and CO_2 emitted according to source-dependent vertical profiles. As the domain covered several large coal-fired power plants, we also investigate the impact of dynamically accounting for plume rise versus applying static vertical emission profiles. We focus on domain-averaged statistics rather than on individual plumes and on near-surface concentrations in the afternoon and total columns at satellite overpass time as would typically be used in inverse modeling. Conditions with a well-mixed and fully-developed boundary layer during daytime are expected to minimize the mismatch between model and observations.

2 Data and methods

2.1 COSMO-GHG model

COSMO is a limited-area, non-hydrostatic numerical weather prediction (NWP) model developed by the German weather service together with a consortium of seven European weather services (Baldauf et al., 2011). In addition to operational weather prediction, the model is applied widely for climate and air pollution research in various modified and extended versions (e.g Rockel et al., 2008; Davin et al., 2011; Vogel et al., 2009; Zubler et al., 2011), making COSMO a versatile community model.

In the project CarboCount-CH (Oney et al., 2015), COSMO was extended with a module for the simulation of greenhouse gases, hereafter referred to as COSMO-GHG. The extension was built on a newly developed tracer module (Roches and Fuhrer, 2012), which replaced the previous nonuniform treatment of meteorological tracers in the model. A similar, independently developed COSMO-based CO_2 model system was presented by Uebel and Bott (2018).

COSMO-GHG was first applied by Liu et al. (2017) to study the spatiotemporal patterns of fossil fuel CO_2 in Europe. They concluded that fossil-fuel CO_2 accounts for more than half of the total (anthropogenic plus biospheric) temporal variations in atmospheric CO_2 over large parts of Europe. Furthermore, they evaluated the model against observations demonstrating that the simulated fossil fuel variations favorably agreed with observations of fossil fuel CO_2 derived from concurrent measurements of CO and $^{14}CO_2$.

COSMO is the first operational NWP model worldwide that is running ~~operationally~~ on hardware accelerated using graphical processing units (GPU) (Fuhrer et al., 2014). This highly efficient code is operationally used by the Swiss weather service MeteoSwiss and has been applied for decadal convection-resolving climate simulations (Leutwyler et al., 2017). In the framework of SMARTCARB, the modules of the GHG extension were ported to the GPU version following the concept of OpenACC compiler directives, which is a high level approach to offload ~~compute intensive part~~ compute-intensive parts to a GPU accelerator (Lapillonne and Fuhrer, 2014). Benchmark tests showed that the GPU version achieved a speedup by a factor of six, which ~~allows to increase the~~ allowed for an increase in spatial and time resolution~~and~~, and which greatly reduced the computational (and energy) costs of the simulations conducted in this study.

2.2 Model domain and setup

The model simulations were conducted in the framework of the project SMARTCARB, which aimed at studying the potential benefit of adding ~~a~~ an NO_2 or CO channel to the instrument package of a future CO_2 satellite mission with respect to quantifying CO_2 emissions from strong localized sources such as cities and power plants. For this purpose, a model domain was selected covering the city of Berlin as well as several large power plants in Germany, Poland and the Czech Republic. The domain extended approximately 750 km in the east-west and 650 km in the south-north direction to cover at least a complete ~~250 km wide~~ 250-km-wide satellite swath on either side of the city. Berlin was selected not only because it is one of the largest cities in Europe, but also because it is rather isolated, and because it has already been investigated in previous CO_2 modeling and observation studies (Pillai et al., 2016; Hase et al., 2015). Simulations were conducted for the complete year 2015 at 1.1 km x 1.1 km horizontal resolution and 60 vertical levels with a top at 24 km. The lowest model layer had a thickness of 20 m. The

~~high~~ vertical allocation of the emissions relative to this vertical resolution of the model will be described in Sect. 2.4. The high horizontal resolution was selected to comply with the small pixel size of 2 km x 2 km of the planned CO_2 imaging satellite, keeping in mind that the effective resolution of Eulerian transport models is much lower than the spacing of the model grid (Kent et al., 2014).

The simulations included a total of 50 different passively transported tracers of CO_2, CO and NO_x. The tracers represented different sources, release times or release altitudes and included background tracers constrained at the lateral boundaries by global-scale models. Two additional tracers for biospheric fluxes (respiration and photosynthesis) were included for CO_2, and four additional tracers with varying e-folding lifetimes of 2, 4, 12 and 24 hours for NO_x. In this study, however, we focus on only a few of these tracers, notably on the following four anthropogenic CO_2 emission tracers:

- CO2_VERT: All anthropogenic emissions in the model domain released according to source-specific vertical profiles.

- CO2_SURF: Same as CO2_VERT but all emissions released at the surface, i.e. into the lowest model layer.

- CO2_PP-PR: Emissions from the six largest power plants with explicit plume-rise calculations.

- CO2_PP-EMEP: Emissions from the six largest power plants released according to a fixed vertical emission profile.

The simulations were nested into the operational European COSMO-7 analyses of MeteoSwiss, which provided the lateral boundary conditions for meteorological variables (temperature, pressure, wind, humidity, clouds) at 7 km horizontal and hourly temporal resolution. Because of the relatively small domain, these boundary conditions provided a strong constraint for the meteorology within the domain. For CO_2, lateral boundary conditions were obtained from a global free-running high-resolution CO_2 simulation (T1279, 137 levels, ~15 km horizontal resolution) provided by the European Center for Medium Range Weather Forecast (ECMWF) through the European Earth observation program Copernicus (Agustí-Panareda et al., 2014). The simulation was indirectly constrained by in-situ data by using a global climatology of optimized biospheric fluxes computed with the CO_2 assimilation system of Chevallier et al. (2010).

2.3 Emissions and biospheric fluxes

Anthropogenic emissions were obtained by merging the European TNO/MACC-3 inventory with a detailed inventory available for the city of Berlin. Dedicated municipal inventories can accurately account for the specific conditions and activities in a city and may, therefore, significantly deviate from inventories obtained by simple downscaling from larger scale inventories (Timmermans et al., 2013; Gately and Hutyra, 2017).

Version 3 of the TNO/MACC inventory was used, which has an improved representation of point sources as compared to version 2 described in Kuenen et al. (2014). It has a nominal resolution of 1/16° x 1/8° (approximately 7 km x 7 km) but additionally reports the emissions from strong point sources at their exact location. Emissions are provided separately for different source categories following the Standardized Nomenclature for Air Pollutants (SNAP) classification (Agency, 2002). The emissions of the year 2011 were taken, which was the most recent year available from the inventory.

The city of Berlin has developed a very detailed inventory for about 30 air pollutants and greenhouse gases for ~~seven~~ 7 major source categories. The inventory was provided in a Geographic Information System (GIS) format as a collection of shape files representing individual area, point and line sources. The latest available year was 2012. ~~Since the source categories did not follow the SNAP nomenclature, a mapping onto SNAP categories was applied to ensure consistency with the~~ In our simulations, separate tracers were included for three broad source groups: traffic, industry, heating. The attribution of the 10 SNAP categories in TNO/MACC-3 ~~inventory~~ and the 7 source categories in the Berlin inventory to these groups are described in detail in the final report of the SMARTCARB project (Kuhlmann et al., 2019).

Both inventories were projected onto the COSMO model grid (rotated latitude/longitude grid with north pole at $43°N$ and $10°W$, $0.01° \times 0.01°$ resolution) using mass-conservative methods. Point sources were placed into the proper COSMO grid cell. As a last step, the two inventories were merged using a mask for the city of Berlin. The merged emission e per grid cell was estimated as $e = (1 - f)e_{TNO} + e_{Berlin}$, where f is the fraction of the grid cell area covered by Berlin.

Fig.1 presents a map of the CO_2 inventory of Berlin in the original format and after projection onto the COSMO grid. The rasterized map reveals 22 strong CO_2 point sources, which together account for 41% of the total emissions in the city. The two horizontal stripes correspond to the paths of airplanes taking off and landing at the two main airports.

In order to calculate hourly emissions as input for the model simulations, temporal scaling factors were applied describing diurnal, day-of-week and seasonal variations. The same temporal profiles were used as in Liu et al. (2017), which are based on factors originally developed for air pollution modeling (Builtjes et al., 2003). These profiles depend on SNAP category and, except for diurnal profiles, on country. Diurnal profiles were matched to the local time of Germany.

Biospheric CO_2 fluxes due to gross photosynthetic production and respiration were computed at the resolution of the COSMO model using the Vegetation Photosynthesis and Respiration (VPRM) model (Mahadevan et al., 2008). This diagnostic model is based on meteorological input (2-m temperature and downward shortwave radiation at the surface) along with the Enhanced Vegetation Index (EVI) and Land Surface Water Index (LSWI) calculated from MODIS satellite reflectances. These indices were available as an 8-day product (MOD09A1, V006) at 500 m spatial resolution. Vegetation classes were determined from the 1-km SYNMAP land cover map (Jung et al., 2006). ~~Further parameters required for VPRM were based on fits to flux tower measurements representative of these vegetation classes.~~ Model parameter values describing the fluxes from different vegetation types were taken from a previous study (Kountouris et al., 2018), in which they were optimized using data from 47 European eddy covariance measurement sites for the year 2007. The model was run off-line for the whole year 2015, driven by the highest resolution ECMWF meteorological data available.

2.4 Vertical allocation of emissions and plume rise

Emissions were distributed in the vertical using source-specific profiles developed for the European Monitoring and Evaluation Program (EMEP) (see e.g. Bieser et al. (2011)). For the purpose of the present study, the number of vertical layers was increased from seven to ten to enable a finer allocation to the model layers of COSMO-GHG. Specifically, the layers 4-90 m and 90-170 m of the original EMEP profiles were divided into three and two sub-layers, respectively. The attribution of emissions to these sub-layers was done in a rough way, for example by placing a larger proportion of emissions from "combustion in

manufacturing industry" into the upper parts but distributing emissions from "Non-industrial (residential) combustion" rather evenly over the three layers between 4-90 m, considering that industrial sources are likely to emit at higher altitudes. The profile for SNAP 9 (waste) was modified in two ways: (i) by moving 10% from higher layers to the lowest layer to account for CO_2 emissions from landfills and waste water treatment plants, and (ii) by moving the large fractions originally placed into the layers 170-310 m (40%) and 310-470 m (35%) into lower layers between 90 m and 310 m. This latter modification was made following the study of Pregger and Friedrich (2009), which showed that the emission-weighted height of waste incinerator stacks in Germany is only about 100 m and that 90% of the corresponding emissions including plume rise are expected to occur below 300 m.

The modified EMEP profiles are presented in Tab. 2 and illustrated in Fig. 2a. They were only applied to point sources such as power plants and large industrial facilities. A different set of profiles with lower average emission heights was applied to area sources, as these represent much weaker and more dispersed sources. The corresponding profiles are presented in Tab. 3 and Fig. 2b. The motivation for this distinction is best illustrated for SNAP category 2, residential and other non-industrial heating. In the case of point sources, these are large heating facilities such as combined heat and power plants with tall stacks. In the case of area sources, in contrast, these are mostly private heating systems releasing CO_2 through chimneys at roof level. This is reflected in our profiles for SNAP 2 area sources being limited to the layers between 4 m and 60 m above surface.

As noted by Bieser et al. (2011), the EMEP profiles are based on very limited information originally collected for the city of Zagreb, Croatia. They therefore proposed a different set of profiles based on plume rise calculations for a large number of point sources across Europe. Their study indicated that the vertical placement of emissions in the EMEP profiles is too high for combustion processes (SNAP 1, 2, 3, and 9) but too low for production processes (SNAP 4 and 5). For SNAP 1 (power plants), for example, they proposed a median release height of about 300 m, whereas the median in the EMEP profiles is about 400 m.

Although we did not apply the profiles of Bieser et al. (2011), our modification of SNAP 9 and the distinction between point and area sources effectively reduces the emission heights of CO_2 compared to the standard EMEP profiles. Furthermore, for the 22 largest point sources in Berlin as well as the 6 largest power plants in the model domain in Germany (Jänschwalde, Lippendorf, Boxberg, Schwarze Pumpe) and Poland (Turów, Pątnów), the static EMEP SNAP 1 profiles were replaced by dynamic plume rise simulations for each hour of the year.

The effective emission height can be much higher than the geometric height of a stack because of the momentum and buoyancy of the flue gas. In general, plume rise depends on stack geometry (height and diameter), flue gas properties (temperature, humidity, exit velocity) and meteorological conditions (wind speed, atmospheric stability). Plume rise and the vertical extent of the plumes were calculated using the empirical equations recommended by the Association of German Engineers (VDI - Fachbereich Umweltmeteorologie, 1985), which are based on the original work of Briggs (1984). Hourly profiles of wind and temperature for the year 2015 were extracted from the COSMO-7 analyses of MeteoSwiss at the locations of the individual stacks. For power plants outside Berlin, typical stack and flue gas parameters were mainly taken from published statistics for Germany (Pregger and Friedrich, 2009), since these parameters are not publicly available (see Table 1). For stacks in Berlin, parameters were included in the emission inventory provided by the city of Berlin.

Two complicating factors were not considered in the present study. The European standards for large combustion plants ($http : //data.europa.eu/eli/dec\_impl/2017/1442/oj$) require the flue gas to be cleaned for sulfur- and nitrogen-oxides. As a consequence of the chemical washing process, the temperature of the flue gas is reduced to a level where it can no longer be released via a classical smoke stack (Busch et al., 2002). In order to avoid re-heating, the flue gas is therefore often directed to the cooling tower and mixed into its moist buoyant air stream. This is true for the major German power plants in the domain, whereas the power plants Turów and Pątnów in Poland were, to the best of our knowledge, still equipped with smoke stacks in 2015. Plume rise computations are much more complex in the case of cooling towers due to latent heat release and the interaction with moisture in the ambient air (Schatzmann and Policastro, 1984). The additional release of latent heat may enhance plume rise by 20% to 100% compared to a dry plume (Hanna, 1972). Second, large power plants such as Jänschwalde are equipped with multiple cooling towers a short distance from each other. Their plumes tend to interact in a way that enhances plume rise, an effect that additionally depends on the alignment of the towers relative to the flow direction (Bornoff and Mokhtarzadeh-Dehghan, 2001). Neglecting these effects may thus underestimate true plume rise in our calculations.

3 Results

3.1 Plume rise at power plants

An example for the results of the plume rise simulations are presented for Jänschwalde, the largest power plant in the domain. Figure 3a shows the hourly evolution of the plume center heights during the year 2015. Plumes typically rise 100 m to 400 m above the top of the cooling tower, but occasionally much further when both winds and atmospheric stability are low. Plume rise varies strongly from day to day and shows a pronounced seasonal cycle. Plumes rise on average to 360 m above ground in summer but only to 250 m in winter. Due to the diurnal evolution of the planetary boundary layer (PBL), plume rise also varies with the hour of the day, except during winter. In summer, the amplitude of this diurnal variability is about 150 m, with a broad minimum at night from 21 to 08 LT (19 to 06 UTC) and a peak in the early afternoon.

Figure 4 compares the histograms of plume rise at Jänschwalde in summer and winter to the standard EMEP SNAP 1 profile of Fig. 2. In agreement with Bieser et al. (2011), the computed profiles tend to place emissions significantly lower compared to EMEP, even in summer. Median and mean effective emission heights in 2015 were 266 m and 310 m above ground, respectively. For the smaller power plant Lippendorf, these numbers were 187 m and 210 m. Simulated plume rise for these two power plants was thus at least 100 m lower compared to the EMEP profile. The large fraction of emissions placed above 500 m in the EMEP profile seems particularly unrealistic. Our plume rise calculations are more consistent with the profiles recommended by Bieser et al. (2011).

3.2 Emission profiles over Berlin

The main emission sources of CO_2 in Berlin are "traffic", "private households and public buildings", "industry", "trade and others", and the "transformation sector" which includes large facilities for heat and electricity production. In 2015, traffic only accounted for 29% of all emissions, households and industry for 27% (Amt für Statistik Berlin-Brandenburg, 2018). By far the largest single source was the transformation sector with a share of 42% of the total. As a result, a significant fraction of emissions is released through stacks.

Figure 5 shows the monthly mean vertical profiles of CO_2 emissions from Berlin in a winter month (February) and a summer month (August) as used in our simulations. The profiles reflect the superposition of emissions from different sectors with different vertical profiles (see Fig. 2) and different seasonal contributions. In February, for example, residential heating is an important source between 4 m and 60 m above ground, whereas emissions in these layers are almost completely absent in August. Although in both months the profiles have a pronounced peak at the surface, 36% of CO_2 is released above 90 m in February, with that share rising to 58% in August when residential heating is small. These numbers suggest that a proper vertical allocation of emissions is important even in cities.

The vertical profiles of the emissions of CO and NO_x are overlaid in Fig. 5 for comparison, since coincident measurements of NO_2 or CO may be used by a future CO_2 satellite for emission quantification. Both CO and NO_x have a more pronounced peak at the surface due to the larger proportion of traffic emissions. Similar to CO_2, a large (albeit smaller) fraction of NO_x is emitted well above ground by large point sources. Interestingly, these sources emit very little CO, suggesting efficient combustion or cleaning of the exhaust. The fraction of CO released above 90 m is below 5%, in sharp contrast to CO_2.

3.3 CO_2 at the surface

Maps of the monthly mean afternoon CO_2 dry air mole fractions (hereafter referred to as "concentrations") in the lowest model layer (0–20 m) from all anthropogenic emissions within the model domain are presented in Fig. 6 for January and July, respectively. Afternoon values were selected since current inversion systems usually assimilate only observations in the afternoon when vertical concentration gradients are smallest (e.g. Peters et al., 2010). The left hand panels show the results for the tracer CO2_VERT, with vertically distributed emissions, while the right hand panels show the tracer CO2_SURF, with all emissions concentrated at the surface. For all power plants labeled in the figure, plume rise was explicitly simulated for the tracer CO2_VERT.

The concentrations are generally higher in January than in July due to higher emissions and reduced vertical mixing in winter. The concentrations are also significantly higher when all CO_2 is released at the surface. The main reason for these differences are power plants and other point sources, which stand out prominently in the maps for the tracer CO2_SURF but not for CO2_VERT.

As shown in Fig.7, the differences are much larger in winter than in summer. In summer, power plant emissions are mixed efficiently over the depth of the afternoon PBL. Since this mixing is not instantaneous, differences are noticeable close to the sources but fade out rapidly with increasing distance. In winter, conversely, when vertical mixing is weak, the differences

between the two tracers remain well above 1 ppm over distances of a few tens of kilometers downstream, occasionally over 100 km or more.

Domain-averaged monthly mean diurnal cycles of the two tracers are presented in Fig. 8 for the months of January and July. Consistent with the maps, the concentrations of CO2_SURF are significantly higher than those of CO2_VERT. This difference is ~~on the order of~~ around 1.6 ppm in January and almost constant over the day. The variations over the day appear to be dominated by the diurnal cycle of the emissions rather than by the dynamics of the PBL.

In summer, the differences are generally smaller and ~~tend to be very small~~ exhibit a pronounced diurnal cycle. Differences are about 1 ppm at night and almost vanish (about 0.1 ppm) in the afternoon~~, consistent with Fig. 7b.~~. Due to the low PBL at night, the concentrations increase over the night despite relatively low emissions. This increase is much more pronounced for tracer CO2_SURF, which is susceptible to surface emissions from point sources that do not stop at night. The tracer CO2_VERT only shows a marked increase during the early morning hours when traffic increases and the PBL is still low. Emissions from point sources, on the other hand, are likely released above the nocturnal PBL leading to marked differences between CO2_VERT and CO2_SURF at night.

Statistics of the afternoon concentrations of the two CO_2 tracers are summarized in Tab. 4 in terms of mean values and different percentiles of the frequency distribution. Domain-averaged CO_2 concentrations are 43% higher in January when all CO_2 is released at the surface compared to when emissions are distributed vertically. The differences are larger for the high percentiles, suggesting that background values are less affected than peak values. This is understandable as vertical mixing tends to reduce the differences with increasing distance from the sources.

In summer, the differences are generally much smaller, but as suggested by Fig.8, this is only true for afternoon concentrations. Mean differences in the afternoon are of the order of 14%. Again, higher percentiles tend to show larger differences. The impact on observations from tall tower networks measuring CO_2 some 100 m to 300 m above the surface (Bakwin et al., 1995; Andrews et al., 2014) will likely be somewhat smaller than suggested by the numbers above, especially in winter when the atmosphere is less well-mixed.

3.4 ~~Column mean~~ Column-averaged dry air mole fractions XCO$_2$

As shown in the previous section, the choice of vertical allocation of the emissions has a large impact on ground-level concentrations. Since the tracers CO2_VERT and CO2_SURF are based on the same mass of CO_2 emitted into the atmosphere and only differ by the vertical placement of this mass, differences in ~~column mean~~ column-averaged dry air mole fractions (ratio of moles of CO_2 to moles of dry air in the vertical column, XCO$_2$) are expected to be small. However, since wind speeds tend to increase with altitude, CO_2 emitted at higher levels is more likely to be transported away from the sources and leave the model domain more rapidly.

Maps of XCO$_2$ and of the differences between the two tracers are presented in Figures 9 and 10 in the same way as for the surface concentrations. Instead of presenting mean afternoon values, the figures show the situation at 11:30 LT (the average of output at 10 UTC and 11 UTC) corresponding to the expected overpass time of the planned European satellites. Note that we did not account for daylight savings time in the diurnal cycles of emissions but assumed a constant offset of +1 hour

between local time in the domain and UTC. Differences in XCO_2 between the two tracers are indeed much smaller than the differences in surface concentrations, suggesting that total column observations are much less sensitive to the vertical placement of emissions. However, differences are not negligible, especially in winter (Fig. 10). The largest differences in January are seen in the northern parts of the Czech Republic, which is the main coal-mining region of the country featuring a large number of power plants. Since plume rise was not explicitly calculated, CO2_VERT emissions from these power plants followed the EMEP profiles, which tend to place emissions too high, as mentioned earlier. Differences over the power plants with explicit plume rise are smaller, but not negligible, with values ~~on the order of~~ around 1 ppm close to the sources.

Domain-averaged mean diurnal cycles are presented in Fig. 11 and overall statistics in Table 4. Similar to the results for near-surface concentrations, the differences are larger in winter than in summer. In contrast to the situation at the surface, the differences in XCO_2 remain fairly constant over the day not only in winter but also in summer. Relative differences in mean XCO_2 are only 7.7% in January (compared to 43% at the surface) and 4.8% in July (compared to 14% at the surface). Note that the differences in the columns are related to the synthetic nature of our model experiment, since no anthropogenic CO_2 emissions are advecting into the domain from sources outside. Nevertheless, the analysis reveals significant differences in the contributions from sources within the domain, which is the information used by any regional inverse modelling system. Consistent with the findings for near-surface concentrations, the differences tend to increase for higher percentiles, reaching about 10% at the 95th percentile (see last column in 4).

Although differences in monthly mean XCO_2 values are relatively small, the differences can be very large at a given location and time as illustrated in Fig. 12 for 2 July 2015. The figure shows the differences in XCO_2 between two CO_2 tracers representing emissions from the largest power plants in the domain. The two tracers were released either using explicit plume rise simulations (CO2_PP-PR) or according to EMEP SNAP-1 profiles (CO2_PP-EMEP). No power-plant-only tracer was simulated with emissions at the surface. Plume rise was rather moderate (to about 340 m above ground) at this time due to pronounced easterly winds. The red (positive) parts in the figure correspond to plumes produced by CO2_PP-PR whereas the blue parts (negative) correspond to the tracer CO2_PP-EMEP.

Due to wind directions changing with altitude and due to the different emission heights of the tracers, the plumes are transported in different directions. Spiraling wind directions are typical of the boundary layer where winds near the surface have an ageostrophic, cross-isobaric component due to surface friction, while winds become increasingly geostrophic at higher levels. This is known as the Ekman spiral. The plumes of CO2_PP-EMEP show a stronger lateral dispersion because the tracer is released over a large vertical extent (blue line in Fig. 2). The vertical cross-section transecting the plumes about 15 km to 30 km downwind of the sources suggests that both tracers are partially mixed over the full depth of the boundary layer at this distance, but that the centers of the plumes are clearly higher above the surface for the tracer CO2_PP-EMEP than for CO2_PP-PR.

4 Discussion

The results for near surface concentrations revealed a strong sensitivity to the vertical placement of emissions, especially in winter. Similarly large sensitivities were reported for air pollution simulations. By conducting a set of five one-year European scale model simulations differing only in the vertical allocation of emissions, Mailler et al. (2013) found that ground-based concentrations of SO_2, an air pollutant primarily released by point sources, increased on average by about 70% when reducing all emission heights by a factor four. The changes were less significant (about 15%) for NO_2 due to the larger contribution from traffic emissions. Sensitivities of a similar magnitude were reported by Guevara et al. (2014) for simulations over Spain when replacing the EMEP profiles for power plants by more realistic plume rise calculations.

The results of the present study may be considered as upper limits of the sensitivity for two reasons. First of all, our simulations covered a region in Europe with a particularly high density of coal-fired power plants. Simulations over other regions such as France, where electricity is mostly produced by nuclear power, would likely have yielded lower sensitivities. Nevertheless, averaged over Europe, large point sources are responsible for at least half of all anthropogenic CO_2 emissions (52% in 2011) according to the TNO/MACC-3 inventory, which underscores the general importance of properly allocating their emissions vertically.

Second, for most point sources in our domain the standard EMEP profiles were applied, which tend to place emissions too high in the atmosphere as suggested consistently by Bieser et al. (2011); Guevara et al. (2014); Mailler et al. (2013) and by our own comparison with explicit plume rise simulations. Several improvements were already implemented in the present study, each of them leading to a reduction of the effective emission heights and likely to a more realistic representation as compared to EMEP. This included a modification of SNAP 9 profiles, the distinction between point and area sources, and the explicit computation of plume rise for the largest sources in the domain. Due to a lack of representative studies, these modifications were somewhat arbitrary. More studies like Bieser et al. (2011) and Pregger and Friedrich (2009) are needed, but should not only target large point sources but also emissions from the remaining 48% of emissions, including residential heating, even though their vertical placement will be less critical. Explicit plume rise computations for all large point sources as performed in some air quality models (Karamchandani et al., 2014) would be the best alternative to using static profiles, but this adds significant complexity to the model, and individual stack and flue gas parameters are not publicly available in Europe (Pregger and Friedrich, 2009).

None of the studies mentioned above considered the issue of interaction between multiple plumes and latent heat release in cooling tower plumes, which tend to enhance plume rise. The SNAP 1 profiles recommended by Bieser et al. (2011) and the simulations conducted here are only representative for isolated stacks, which is not consistent with any of the German coal-fired power plants in our domain. Comprehensive models for plume rise from single and multiple interacting cooling tower plumes have been presented by Schatzmann and Policastro (1984) and Policastro et al. (1994), but they seem not to be widely applied, although the code of Schatzmann and Policastro (1984) is available through the Association of German Engineers (VDI, $https://www.vdi.de/index.php?id = 4791$). Most plume rise studies date back 20 to 50 years, and modern computational

fluid dynamics simulations are lacking, with the exception of the study of Bornoff and Mokhtarzadeh-Dehghan (2001) on the interaction of plumes from two adjacent cooling towers.

Mean afternoon differences at the surface in January between the tracers CO2_VERT and CO2_SURF are of the order of 1.5 ppm, which is close to 50% of the total anthropogenic CO_2 signal due to emissions within the model domain. In summer, the relative differences in the afternoon are much smaller due to more efficient vertical mixing, but still ~~on the order of~~ as large as 14%. Inaccurate vertical placement of the emissions may thus lead to biases of a magnitude comparable to other error sources reported in the literature. Random uncertainties ~~on the order of~~ around 30-50% of the regional biospheric CO_2 signal have been estimated, for example, for model errors in PBL mixing (Gerbig et al., 2008) and in wind speed and direction (Lin and Gerbig, 2005). Systematic biases in simulated near-surface wind speeds of mesoscale weather prediction models like COSMO or WRF are typically on the order of 0.5-1 m s^{-1} or about 10-20% of the mean wind speeds (Jiménez and Dudhia, 2012; Brunner et al., 2015; Bagley et al., 2018), which may translate into similar biases in near-surface concentrations. Systematic differences in emission estimates using different regional transport and inverse modeling systems were reported to be ~~on the order of~~ around 20%-40% (Hu et al., 2015; Bergamaschi et al., 2018b). Our analysis focused on the situation in the lowest model layer at 0-20 m. However, the recommended strategy for CO_2 monitoring is to sample from tall towers well above the surface. At higher elevations the model sensitivity to the vertical placement of emissions is likely smaller, which is another benefit of tall tower measurements in addition to their greater spatial representativeness.

Mean differences in total column XCO$_2$ between the tracers CO2_VERT and CO2_SURF are much smaller, around 8% in winter and 5% in summer. However, differences are larger at the highest percentiles (about 10% at the 95% percentile), which are more relevant for satellite missions like CarbonSat designed to image individual plumes. Furthermore, the vertical placement of emissions has a significant effect on the speed and direction of individual plumes, suggesting that an accurate vertical placement is a critical requirement for inverse modeling of power plant emissions from satellite observations. Irrespective of a correct vertical placement of emissions, an appropriate simulation of power plant plumes will remain a great challenge for any mesoscale atmospheric transport model. Current approaches for estimating power plant emissions are circumventing this problem by directly matching the "observed" wind direction defined by the location of the plume, e.g., by fitting a Gaussian plume model (Krings et al., 2018; Nassar et al., 2017). However, ~~also these methods~~ these methods also need to make a realistic assumption about the height of the plume, since the mean advection speed of the plume needs to be estimated from a simulated or observed vertical wind profile.

5 Conclusions

We investigated the sensitivity of model-simulated near-surface and total column CO_2 concentrations to a realistic vertical allocation of anthropogenic emissions as opposed to the traditional approach of emitting CO_2 only at the surface. The study was conducted using kilometer-scale atmospheric transport simulations for the year 2015 for a domain covering the city of Berlin and numerous power plants in Germany, Poland and the Czech Republic. More than 50% of CO_2 in Europe is emitted from large point sources, mostly through stacks and cooling towers, suggesting that a proper representation of these sources in

the vertical dimension may be critical. Our results indeed confirm a strong sensitivity of near-surface afternoon concentrations: A regional CO_2 tracer released in the model domain only at the surface was on average 43% higher in winter and 14% higher in summer than a tracer released according to realistic vertical profiles. Since measurements of CO_2 are often taken from towers some 100 m to 300 m above the surface (Bakwin et al., 1995), the impact on actual ground-based observations will likely be somewhat smaller.

Differences were smaller but not negligible (5%-8%) for total column XCO_2, suggesting that the assimilation of satellite observations is less sensitive to the vertical placement of emissions than the assimilation of ground-based observations. Individual plumes as imaged by future CO_2 satellites, however, may propagate more rapidly and in different directions when using realistic vertical profiles instead of releasing CO_2 only at the surface.

Plume rise was explicitly simulated for the six largest power plants in the domain and for the 22 largest point sources in Berlin. Power plant plumes typically rose between 100 m and 400 m above the top of the stacks. Plume rise showed significant seasonal and diurnal variability that would be missed when applying static vertical profiles. Simulated plume rise was on average more than 100 m lower than suggested by the frequently used EMEP profile for power plant emissions. An accurate vertical placement of emissions is not only critical for power plants but may also be relevant for the simulation of city plumes. In the case of Berlin, for example, more than 35% of CO_2 is released through stacks presumably more than 90 m above surface.

We strongly recommend the representation of CO_2 emissions in all three dimensions in regional atmospheric transport and inverse modeling studies. The specific impact on the model results will depend on factors such as vertical model resolution and boundary layer scheme, but in general cannot be expected to be negligible. Current gridded emission inventories only provide information in two dimensions. Information on the source-specific vertical allocation of emissions as used in this study, conversely, is still sparse and should receive more attention in the future.

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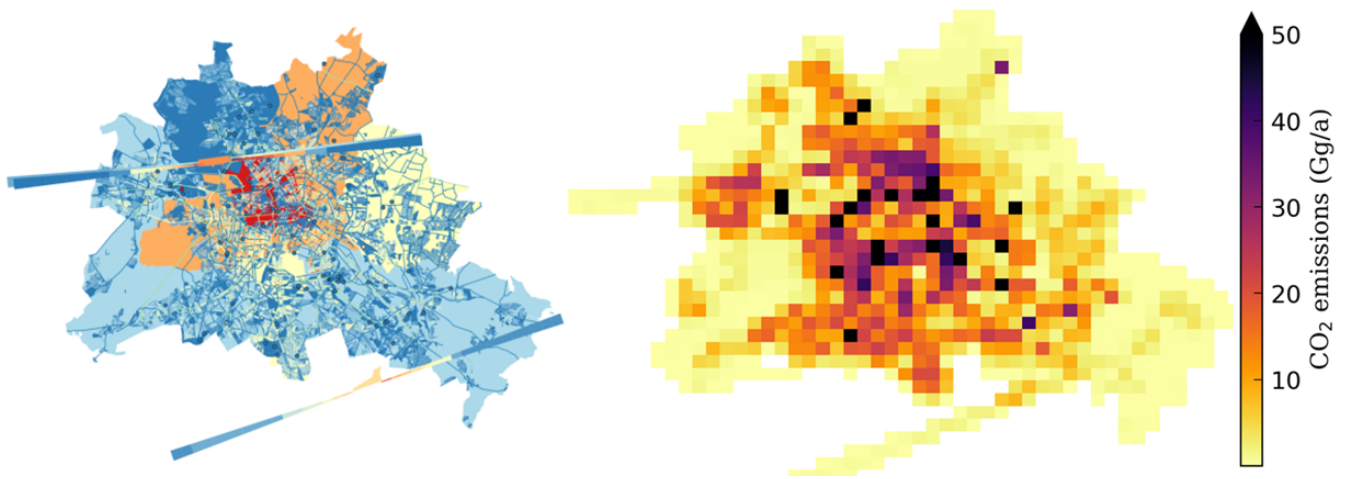
Figure 1. (Left) ~~Sketch~~ Schematic of CO_2 emission in the Berlin inventory with point, line and area sources from different emission categories. <u>Point, line and area sources are in different units (in $kg \cdot s^{-1}$ for point, $kg \cdot s^{-1} \cdot m^{-1}$ for line and $kg \cdot s^{-1} \cdot m^{-2}$ for area sources) each being colored separately between minimum and maximum values.</u> (Right) Total CO_2 emissions re-projected and rasterized onto the COSMO model grid.

(a)   (b)
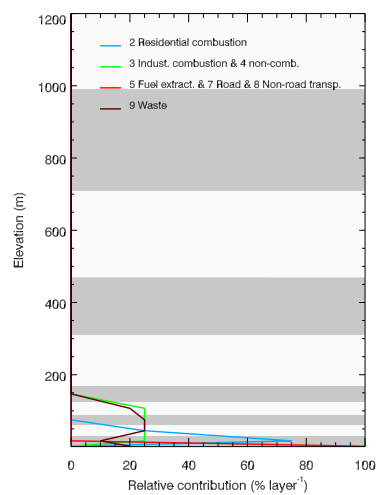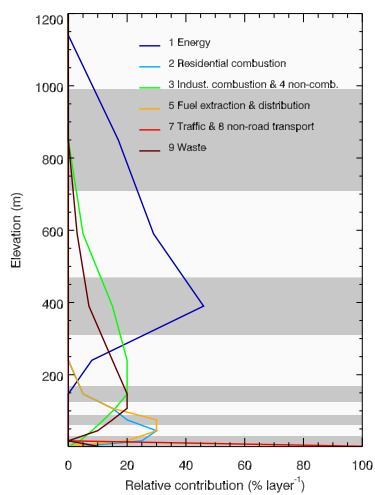
Figure 2. Vertical emission profiles applied for (a) point sources and (b) area/line sources. The alternating gray and white backgrounds denote the vertical layers.
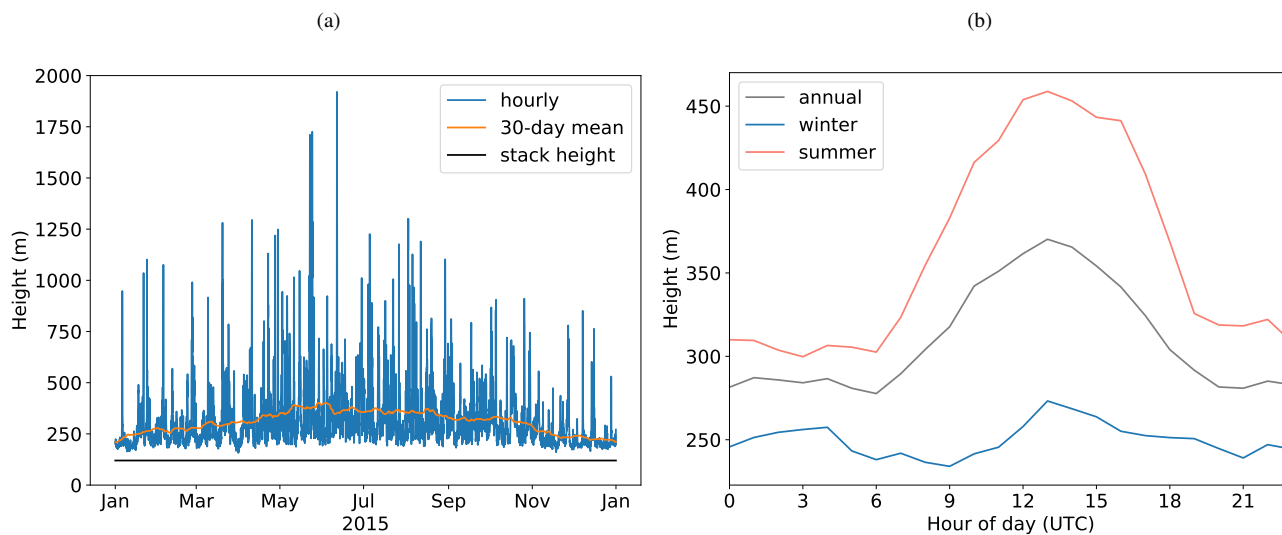
Figure 3. Effective CO_2 emission heights at the power plant Jänschwalde, Germany, based on plume rise calculations. (a) Hourly plume rise in 2015. The orange line is a 30-days moving average. The black solid line denotes the height of the cooling tower (120 m). (b) Mean annual, winter (DJF) and summer (JJA) diurnal cycles.
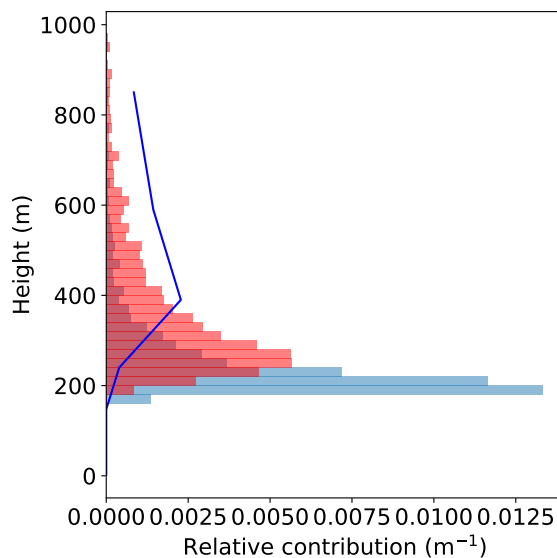

Figure 4. Normalized histogram of plume rise at the power plant Jänschwalde in winter (blue) and summer (red). The blue line shows the EMEP standard profile for power plants for comparison. Areas below the curves integrated along the vertical axis are normalized to 1.
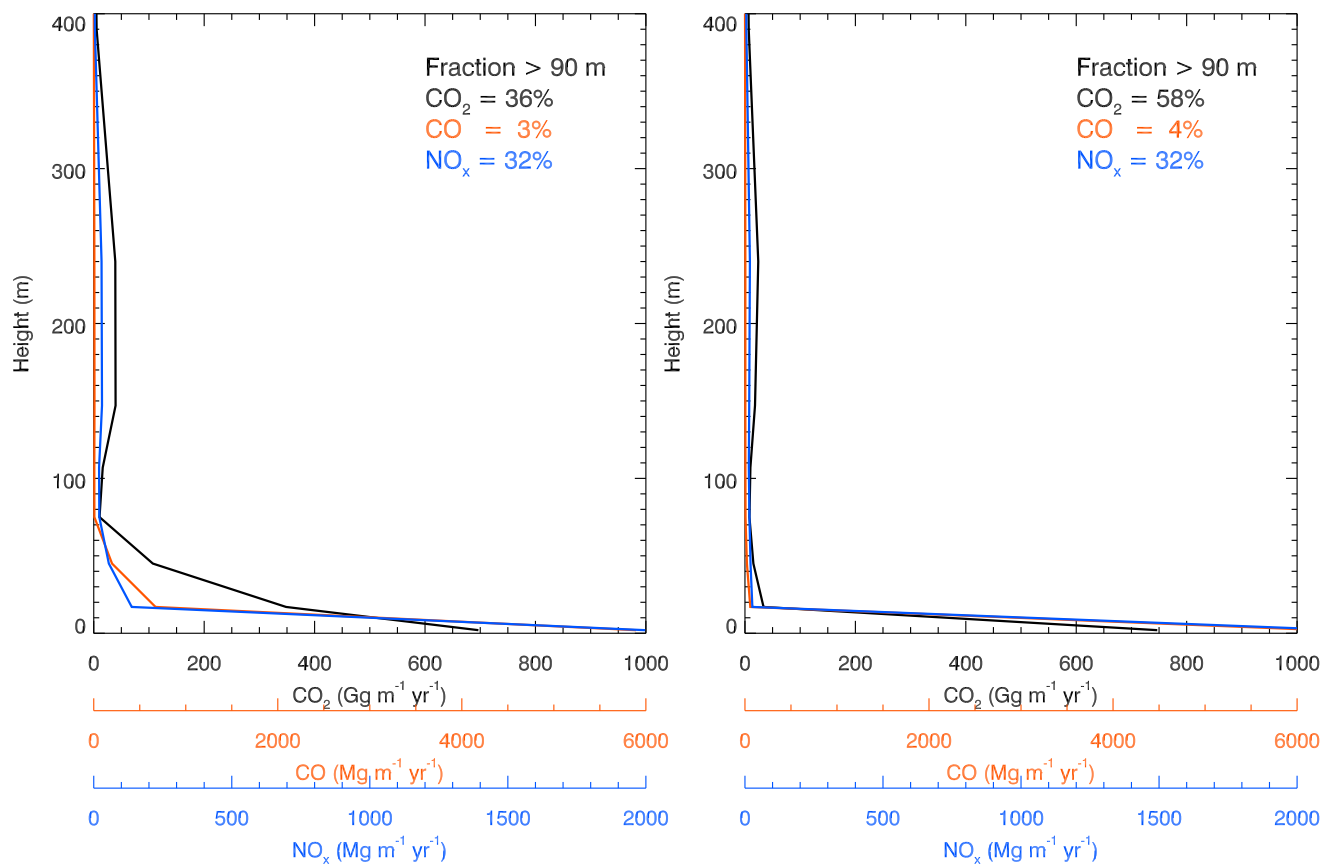
(a) February

(b) August

Fraction > 90 m
CO_2 = 36%
CO   = 3%
NO_x = 32%

Fraction > 90 m
CO_2 = 58%
CO   = 4%
NO_x = 32%

Figure 5. Monthly mean vertical emission profiles of CO_2, CO and NO_x over Berlin in (a) February and (b) August.
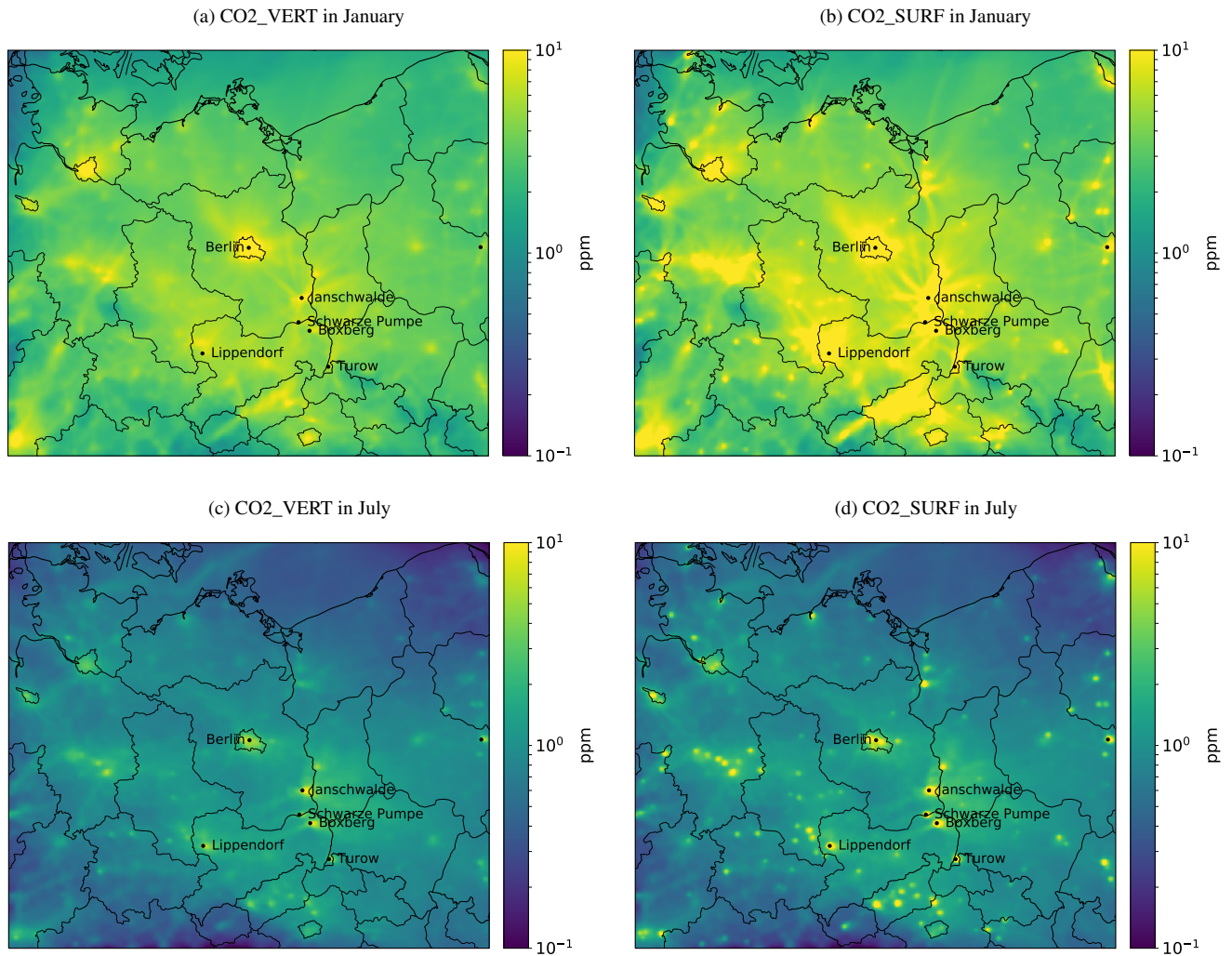
Figure 6. Mean afternoon (14-16 LT) near-surface CO_2 in (top) January and (bottom) July contributed by all anthropogenic sources in the domain. (Left) Tracer CO2_VERT released according to realistic vertical profiles. (Right) Tracer CO2_SURF released only at the surface.

Figure 7. Difference in mean afternoon (14-16 LT) near-surface CO_2 between tracers CO2_SURF and CO2_VERT in (a) January and (b) July.

Figure 8. Mean diurnal cycles of the tracers CO2_VERT and CO2_SURF in January and July.

Figure 9. Mean 11:30 LT ~~column mean~~ dry air mole fractions (XCO$_2$) in (top) January and (bottom) July contributed by all anthropogenic sources in the domain. (Left) Tracer CO2_VERT released according to realistic vertical profiles. (Right) Tracer CO2_SURF released only at the surface.
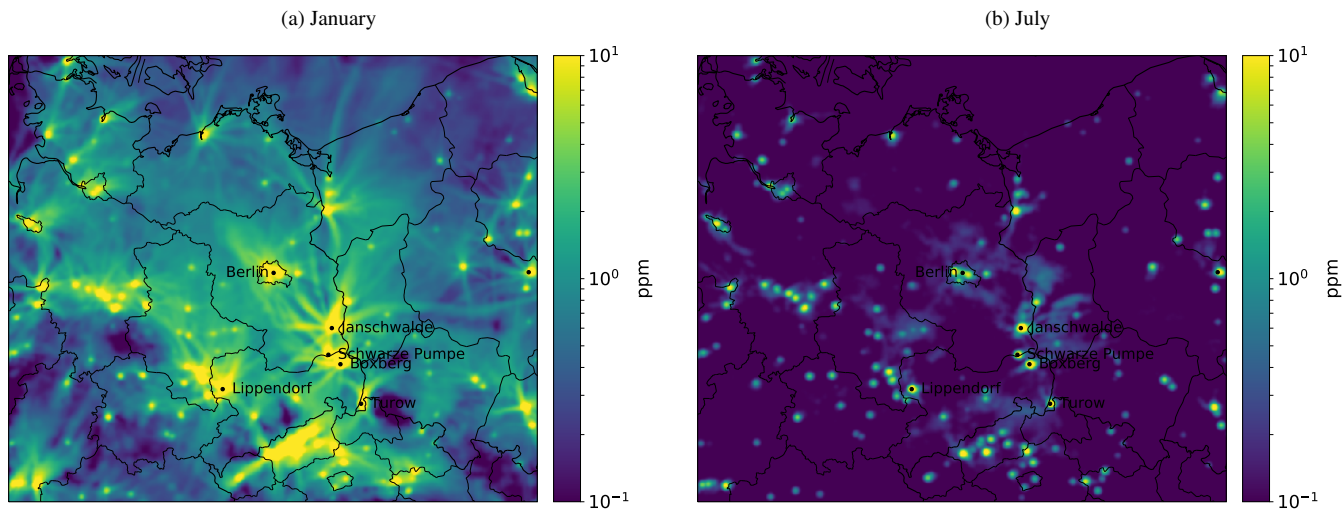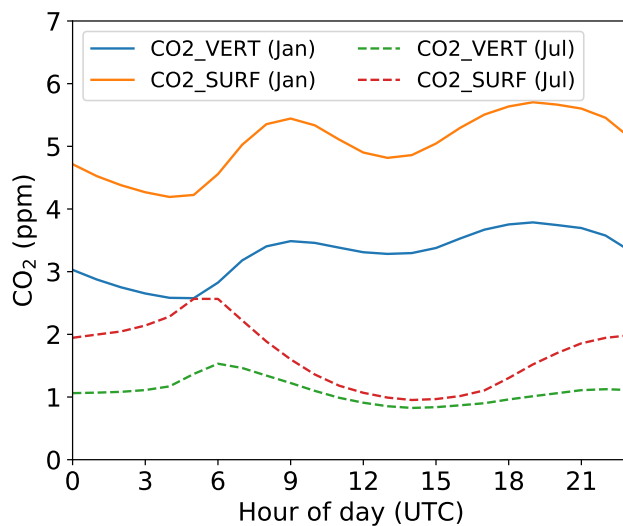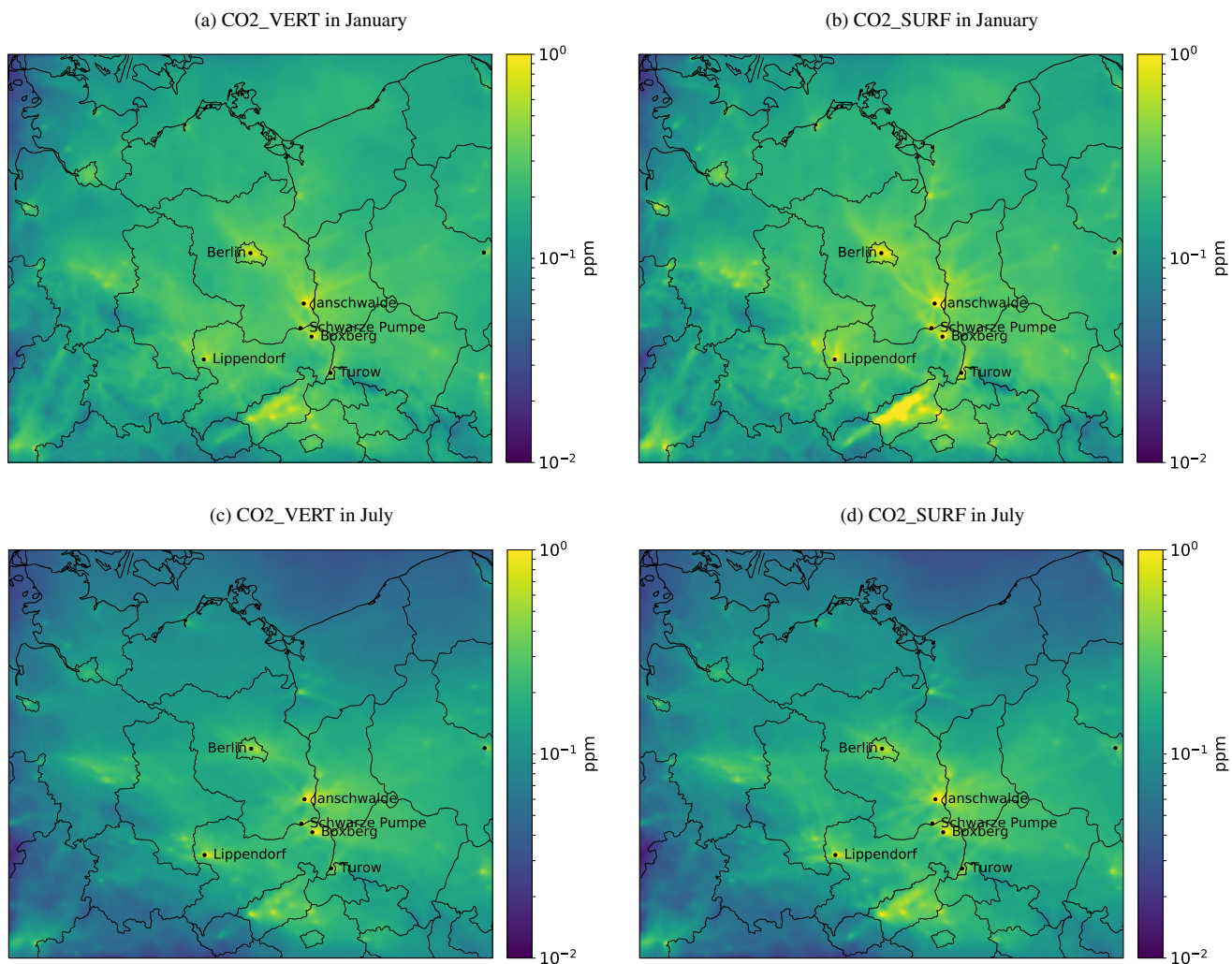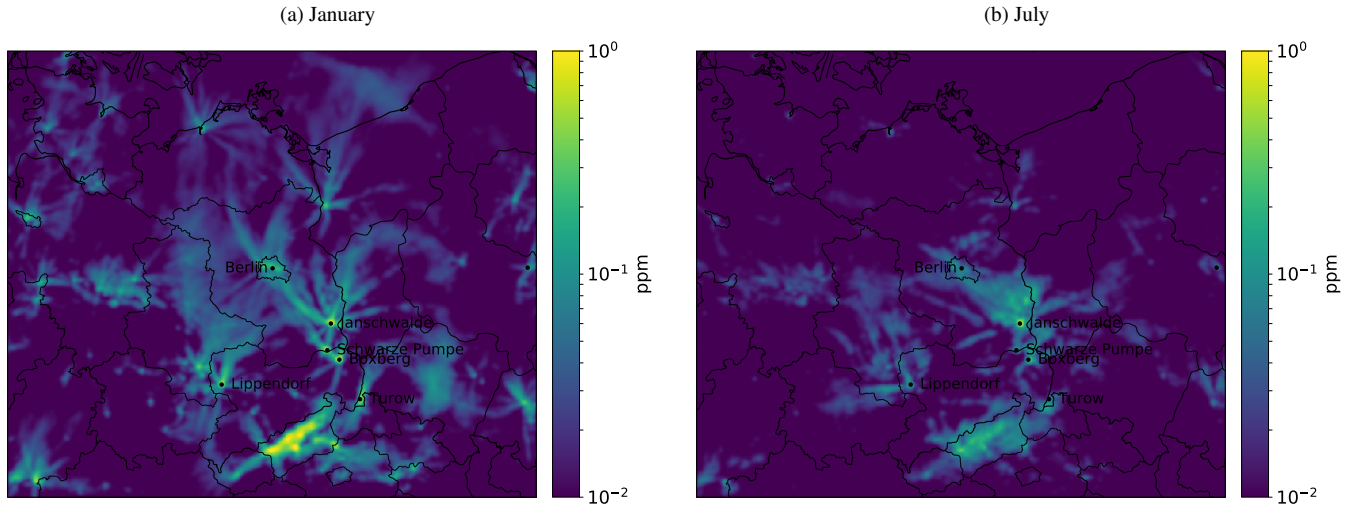
Figure 10. Difference in 11:30 LT ~~column mean~~ column-averaged dry air mole fractions (XCO_2) between tracers CO2_SURF and CO2_VERT in (a) January and (b) July.
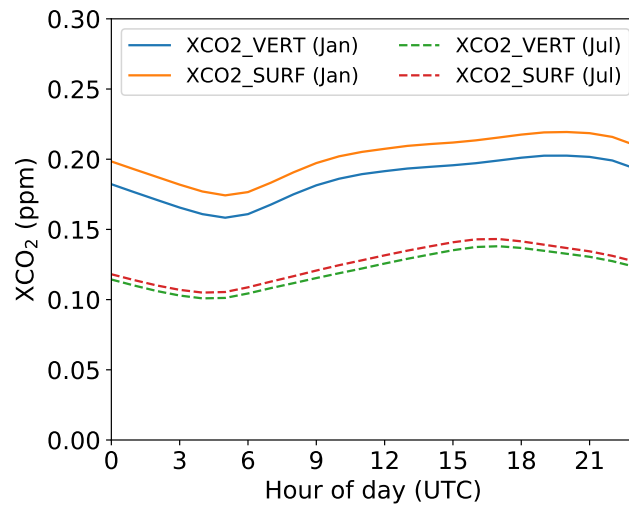

Figure 11. Mean diurnal cycles of ~~column mean~~ column-averaged dry air mole fractions (XCO_2) of the tracers CO2_VERT and CO2_SURF in January and July.
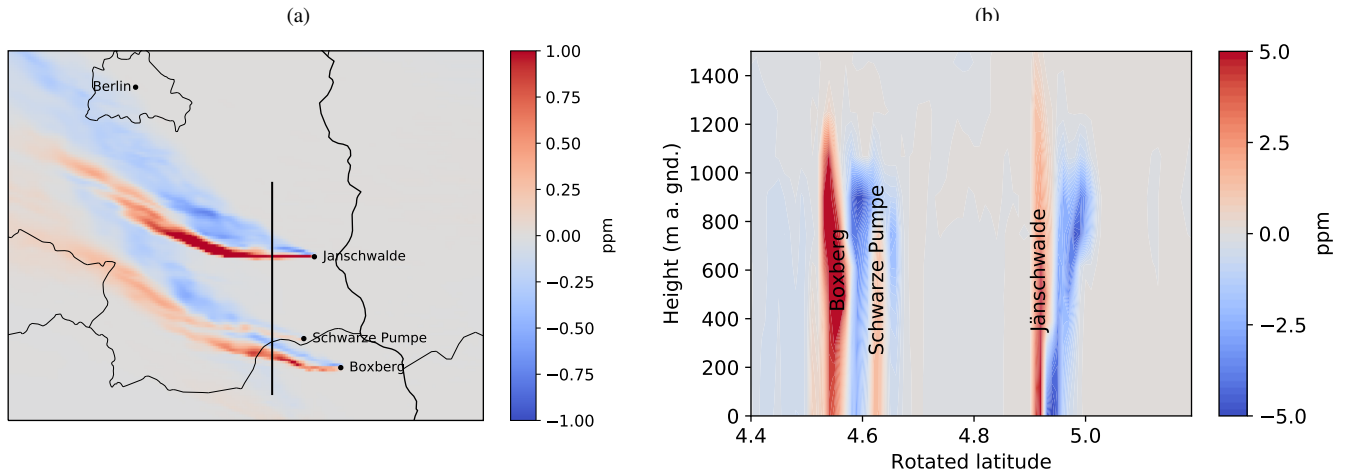
Figure 12. Difference on 2 July 2015 11:00 LT between power plant CO_2 tracer with explicit plume rise (CO2_PP-PR) and released according to standard EMEP SNAP-1 profile (CO2_PP-EMEP). (a) Map of the difference in ~~column mean~~ column-averaged dry air mole fractions XCO_2. (b) Vertical cross-section of the difference in CO_2 along the transect indicated in (a). The length of this cross-section is about 100 km.

Table 1. Stack parameters used for plume rise calculation for six largest power plants in the model domain.

Name	Longitude (° E)	Latitude (° N)	Stack height (m)	Effluent temperature[1] (K)	Volume flux[1] (m^3 s^{-1})
Jänschwalde	14.458	51.837	120[2]	322	790
Lippendorf	12.372	51.184	175[2]	322	790
Schwarze Pumpe	14.354	51.538	141[2]	322	790
Boxberg	14.569	51.418	155[2]	322	790
Turów	14.911	50.948	150[1]	416	159
Pątnów	18.238	52.303	114[1]	447	59

[1] average parameters by fuel and plant capacity taken from Pregger and Friedrich (2009)

[2] https://de.wikipedia.org/wiki/Kühlturm (last access: 28 Aug 2018)

Table 2. Vertical emission profiles for point sources for different SNAP categories as fraction of the total emitted mass per vertical layer. Layers are denoted by their lower and upper limits (m above ground).

SNAP	Description	0–4	4–30	30–60	60–90	90–125	125–170	170–310	310–470	470–710	710–990
1	Energy industry	0.00	0.00	0.00	0.00	0.00	0.00	0.08	0.46	0.29	0.17
2	Non-industrial combustion	0.05	0.25	0.30	0.20	0.15	0.05	0.00	0.00	0.00	0.00
3	Combustion in manufacturing industry	0.00	0.04	0.08	0.12	0.16	0.20	0.20	0.15	0.05	0.00
4	Production processes	0.00	0.20	0.30	0.30	0.15	0.05	0.00	0.00	0.00	0.00
5	Extraction/distribution of fossil fuels	0.00	0.20	0.30	0.30	0.15	0.05	0.00	0.00	0.00	0.00
6	Product use	0.50	0.50	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
7	Road transport	1.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
8	Non-road transport	1.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00
9	Waste treatment	0.10	0.00	0.10	0.15	0.20	0.20	0.15	0.07	0.03	0.00
10	Agriculture	1.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00

Table 3. Vertical emission profiles for area sources for different SNAP categories as fraction of the total emitted mass per vertical layer. Layers are denoted by their lower and upper limits (m above ground). Emissions in SNAP 1 are released exclusively from point sources.

SNAP	Description	0–4	4–30	30–60	60–90	90–125	125–170	170–310	310–470	470–710	710–990
1	Energy industry	-	-	-	-	-	-	-	-	-	-
2	Non-industrial combustion	0.00	0.75	0.25	0.00	0.00	0.00	0.00	0.00	0.00	0.00
3	Combustion in manufacturing industry	0.00	0.25	0.25	0.25	0.25	0.00	0.00	0.00	0.00	0.00
4	Production processes	0.00	0.25	0.25	0.25	0.25	0.00	0.00	0.00	0.00	0.00
9	Waste treatment	0.20	0.10	0.25	0.25	0.20	0.00	0.00	0.00	0.00	0.00
5-8,10	Other categories	1.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00	0.00

Table 4. Statistics of near surface CO_2 and column averaged dry air mole fractions XCO_2 in the model domain for the tracers CO2_VERT and CO2_SURF. Differences are presented in terms of (CO2_SURF-CO2_VERT)/CO2_VERT.

Tracer	Month	Hour	Mean (ppm)	Median (ppm)	25% (ppm)	75% (ppm)	95% (ppm)
CO2_VERT	January	14-16 LT	3.34	3.13	2.35	4.00	5.76
CO2_SURF	January	14-16 LT	4.78	3.91	2.76	5.36	9.40
Difference			43%	25%	17%	34%	63%
CO2_VERT	July	14-16 LT	0.84	0.78	0.51	1.02	1.61
CO2_SURF	July	14-16 LT	0.96	0.80	0.54	1.09	1.81
Difference			14%	3%	6%	7%	12%
XCO2_VERT	January	11:30 LT	0.193	0.185	0.145	0.230	0.317
XCO2_SURF	January	11:30 LT	0.208	0.194	0.149	0.243	0.355
Difference			7.7%	4.9%	2.8%	5.7%	12.0%
XCO2_VERT	July	11:30 LT	0.125	0.110	0.067	0.170	0.253
XCO2_SURF	July	11:30 LT	0.131	0.112	0.068	0.173	0.276
Difference			4.8%	1.8%	1.5%	1.8%	9.1%