



Recommendation on whether CO source of CO₂ needs to be accounted for in CO₂ inversions

MILESTONE 3

Author(s):	Audrey Fortems-Cheiney
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Introduction

One of the main objectives of EYE-CLIMA is to support European and national policies through its top-down emissions estimation methodology and verification of national greenhouse gas inventories (NGHGs). There is indeed a growing interest in top-down verification methods, linking emissions to changes in atmospheric concentrations, and the use of atmospheric observations to verify NGHGs has been highlighted in the 2019 refinement of the IPCC Guidelines. EYE-CLIMA has a strong focus on improving the accuracy of regional inversions for CO₂ fluxes, especially in Europe with the goal to support Europe's Green Deal policies.

Different sources of uncertainties can all interact in regional inversions, including misrepresentation of i) the a priori knowledge from fluxes, ii) initial and boundary conditions and iii) the atmospheric chemistry and transport modelling. For instance, the inversions generally do not take the atmospheric source of CO₂ from the oxidation of CO into account, assuming this source is negligible.

This document describes sensitivity tests performed to assess the impact of the atmospheric CO source of CO₂ on the CO₂ concentrations and on the CO₂ land ecosystem fluxes estimated from the inversions in the frame of the Task T3.8 of the EYECLIMA project. Such assessment has been made with simulations for the year 2019 and inversions for the month of June 2019, when the posterior estimates at the 0.5° spatial resolution identify a CO₂ peak uptake (Deliverable D3.1) and is presented in this deliverable.



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1. Inverse modelling system and experimental framework

The inversion system relies on the coupling between the variational mode of the Community Inversion Framework (CIF, Berchet et al., 2021), the regional chemistry transport model (CTM), CHIMERE (Menut et al., 2013; Mailler et al., 2017) and the adjoint of this model (Fortems-Cheiney et al., 2021b).

1.1. Configuration of the regional CHIMERE chemistry-transport model for the simulation of CO and CO₂ over Europe

The CHIMERE domain for Europe covers latitudes 31.75-73.75°N and longitudes 15.25°W -34.75°E with a 0.5°×0.5° horizontal resolution and 17 vertical layers up to 200 hPa. Meteorological forcing for CHIMERE is generated using operational forecasts from the Integrated Forecasting System (IFS) of the European Centre for Medium Range Weather Forecasting (ECMWF).

The reference configuration used here for the inversion of the CO₂ land ecosystem fluxes in Europe is the one developed for the preliminary inversions documented in Deliverable D3.1 in the frame of the EYECLIMA project. Two model set-ups have been used. First, considering CO₂ as a passive tracer (the reference) in which only the atmospheric transport modeling components have been used (the chemistry modeling components were deactivated), and second considering CO₂ as a labile tracer with the chemistry modeling components activated. The chemical scheme used here is MELCHIOR-2, with more than 100 reactions, including 24 for inorganic chemistry (Lattuati, 1997; Derognat et al., 2003), describing the CO chemistry.

The different components needed for the simulations of CO and CO₂ are described in the following sections.

1.2. CO and CO₂ boundary conditions

Initial, lateral and top conditions for CO₂ concentrations at the boundaries of the model and at the simulations initial times are generated from the latest Copernicus Atmospheric Monitoring Service (CAMS) global CO₂ inversions (v22r1) assimilating surface data (Chevallier et al., 2005; Chevallier et al., 2010). This global inversion product is also used to complement the vertical columns of CO₂ above the top boundary of CHIMERE when comparing the model to XCO₂ observations.

Initial and boundary conditions for several key gaseous species responsible for the oxidation capacity of the lower atmosphere (e.g., CO, NO, NO₂, O₃, H₂O₂, HCHO, etc) were specified using monthly climatological data from LMDz-INCA global model (Szopa, 2008).

1.3. Estimates for the CO emissions

The CO emissions are based on the TNO-GHGco version 3, updated from the TNO inventory documented in Kuenen et al. (2014) and in Super et al. (2020), based on country emission reporting to the European Monitoring and Evaluation Program (EMEP)/Center on Emission Inventories Projection (CEIP). The TNO-GHGco-v3 inventory maps CO emissions at a 6×6 km² horizontal resolution. It combines emissions from area sources, set at the surface, and from point sources. Emissions from point sources, mainly from the energy production and the industrial sectors, are distributed on the first eight vertical model layers in CHIMERE depending on the typical injection heights provided in the TNO inventory, based on Bieser et al. (2011). In the TNO inventory, annual and national budgets are disaggregated in space based on proxies of the different sectors (Kuenen et al., 2014). Temporal disaggregation is based on temporal profiles provided per GNFR sector code with typical month to month, weekday to weekend and diurnal variations (Ebel et al., 1994, Menut et al., 2011).



It should be noted that CO emissions from fires and CO biogenic emissions are not considered, as in Fortems-Cheiney et al. (2023). In addition to CO, the chemical scheme MELCHIOR-2 needs emissions from other species, such as non-methane volatile organic compounds (NMVOCs). The anthropogenic emissions for NMVOCs are obtained from the EMEP inventory (Vestreng et al., 2005). The fixed biogenic NMVOC emissions are derived from the MEGAN model (Guenther et al., 2006). The different emission products have been aggregated at the $0.5^\circ \times 0.5^\circ$ horizontal resolution of the CHIMERE grid.

1.4. Prior estimates of the CO₂ land ecosystem fluxes and other CO₂ surface fluxes

The principle of the inversion is to correct *a priori* estimates of the net land ecosystem flux maps later referred as “prior” fluxes. Here, we have used the “CRUERA” prior NBP, from a European scale simulation run in the frame of the EYE-CLIMA project, with a dedicated forcing at the spatial resolution of 0.125° . This prior is described in the Deliverable D2.3 of the EYECLIMA project. We have aggregated these prior fluxes at the $0.5^\circ \times 0.5^\circ$ horizontal resolution of the CHIMERE grid.

The other component of CO₂ fluxes from ocean and anthropogenic activities are fixed throughout the inversion. We have aggregated these fluxes at the $0.5^\circ \times 0.5^\circ$ horizontal resolution of the CHIMERE grid.

The anthropogenic fossil fuel and biofuel CO₂ emissions (Gerbig and Koch, 2023) are derived from the TNO-GHGco version 3 inventory, as for CO.

The estimate of sea/ocean fluxes within the CHIMERE domain is based on a hybrid product combining the coastal ocean flux estimates from the University of Bergen and a global ocean estimate from MPI-BGC-Jena (Rödenbeck et al., 2014; McGrath et al., 2023). The data is provided from 2005 to 2020 at a $0.125^\circ \times 0.125^\circ$ horizontal resolution and at daily temporal resolution.

1.5. Observations

The inversion assimilates the relatively high-resolution satellite total column CO₂ mole fraction (XCO₂) observations from the OCO-2 NASA-JPL mission (the v11 dataset). The OCO-2 satellite carries high-resolution spectrometers that return high-precision measurements of reflected sunlight received within the CO₂ and O₂ bands in the short-wave infrared spectrum (Crisp et al., 2012) and flies on a 705 km sun-synchronous orbit with a 16-day (233 orbits) ground track repeat cycle. The nominal footprint of the OCO-2 ground pixels is $1.29 \times 2.25 \text{ km}^2$ (across \times along track) at nadir, with a cross-track swath width of about 10 km. We only consider “good” retrievals as identified by the XCO₂ quality flag of the product.

Although the biases in OCO-2 over the ocean acquired in glint mode have been substantially reduced since the initial version 7 (O'Dell et al., 2018), Chevallier et al. (2019) claimed that the assimilation of OCO-2 ocean observations still produced unrealistic results in their global atmospheric inversions. Consequently, they are not considered in this study. After this selection, all individual observations are assimilated and compared to their corresponding horizontal grid-cells in CHIMERE (i.e. to the CHIMERE CO₂ vertical column in this horizontal grid cell), defined for a given observation as that containing the centre of the ground projection of the OCO-2 pixel at the observation time: there is no aggregation of the observations at the model resolution. As described in Section 3.1.2, the CAMS global CO₂ inversions are used to complement the vertical columns of CO₂ above the top boundary of CHIMERE when comparing the model to XCO₂ observations. To make suitable comparisons between simulations and satellite observations, the vertical profiles of CO₂ mole fraction in the corresponding atmospheric columns of the model simulations are first interpolated on the satellite CO₂ retrieval levels (with a vertical mass-conserving interpolation on pressure levels). Then, the appropriate simulated XCO₂ values are computed using both the OCO-2 averaging kernels and prior estimates provided in the OCO-2 retrieval product. As an example, the average of the OCO-2 observations for the month of June 2019 is presented



in Figure 1a while the average of the simulated XCO₂ values activating the CO atmospheric source of CO₂ corresponding to these observations are presented in Figure 1b.

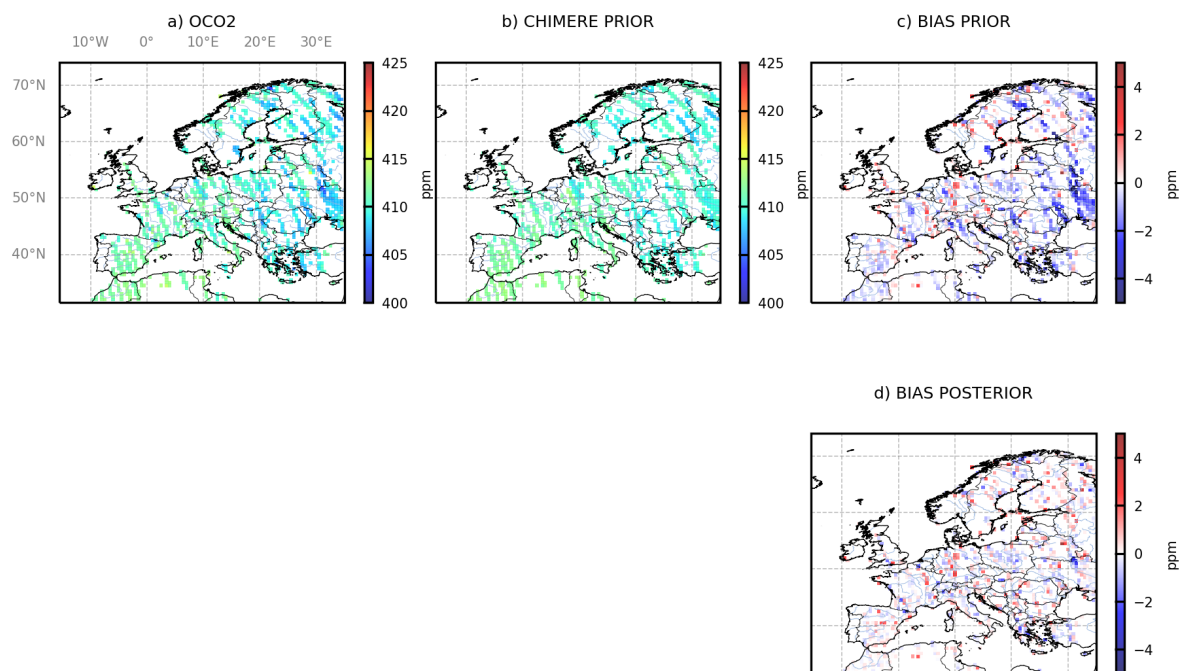


Figure 1. Comparison between the OCO-2 XCO₂ observations and the corresponding CHIMERE XCO₂ simulations activating the CO atmospheric source of CO₂ for the month of June 2019: averages over the month of the XCO₂ values per grid cell of the model (observations, prior and posterior simulations, and differences), in ppm.

1.6. Experiments

The different simulations or inversions for the year 2019 or for the month of June 2019 performed in this study, using ORCHIDEE CRUERA as NEE prior inventory and using OCO-2 satellite observations, are respectively presented in Table 1 and in Table 2.

Table 1. Description of the simulations performed in this study for the year 2019.

Name	Experiment	Chemistry CO-CO2
FWD-ref	Forward simulation	no
FWD-CO/CO2	Forward simulation	yes

Table 2. Description of the inversions performed in this study for the month of June 2019.

Name	Experiment	Control of the initial conditions	Control of the lateral conditions	Chemistry CO-CO2
INV-ref	inversion	yes	yes	no
INV-CO/CO2	inversion	yes	yes	yes

2. Results

2.1. Computational cost

The computational cost of the different experiments performed in this study are presented in Table 3. Forward simulations have been performed for the entire year 2019. Inversions have been performed for the month of June 2019.

While an inversion takes only 10 hours with 10 processors when CO₂ is considered as a passive tracer, an inversion takes more than 8 days when the CO atmospheric source of CO₂ is considered with the MELCHIOR-2 chemistry scheme (Table 3). This computational cost could be reduced in the future by having a highly simplified linear chemical scheme in which CO oxidation to CO₂ is a first order process using pre-calculated fields of radical hydroxyl OH. In addition, technical developments on CHIMERE, in particular by porting the code to GPU-compatible environment, are currently in progress to allow future series of reference long-term inversions in D3.4. It would also help to perform inversions considering the CO/CO₂ chemistry, if relevant. However, further optimizations over the requested memory are still needed to accommodate carrying out adjoint simulations.

Table 3. Description of the experiments performed in this study and associated computational cost.

Experiment	Period	Activating the CO atmospheric source of CO ₂	Total Computational cost
FWD-ref	Year 2019	no	2 hours
FWD-CO/CO ₂	Year 2019	yes	3 days
INV-ref	June 2019	no	10 hours
INV-CO/CO ₂	June 2019	yes	8 days

2.2. Comparison between simulations of the CO₂ mole fractions activating or not the CO atmospheric source of CO₂

Prior simulations with and without activating the CO atmospheric source are compared in Figure 2, and result in differences of less than 0.01% in the CO₂ simulated concentrations at the surface. The prior simulations with and without activating the CO atmospheric source are also compared to measurements of CO₂ mole fraction from the European Obspack compilation of atmospheric carbon dioxide data from ICOS and non-ICOS European ground based continuous measurement stations for the period 1972-2022 called “obspackco2466GLOBALVIEWplusv8.02023-03-30” (ICOS RI et al., 2023). Comparisons between the bias and root mean square (RMS) differences (denoted RMS errors, i.e., RMSE) between the time series of measured and simulated concentrations are shown in Tables 4 and 5 and in Figure 3. When taking the hourly assimilated observations of all the stations into account, **the mean monthly biases between simulated versus measured are very similar, whether the CO atmospheric source of CO₂ is activated or not, both at the yearly scale (mean bias of 1.23 ppm against 1.24 ppm, Table 4) or at the monthly scale for the month of June 2019 (mean bias of 1.54 ppm against 1.56 ppm, Table 5).** The mean monthly RMSE between simulated versus measured CO₂, of about 6 ppm at the yearly scale and 4.99 ppm for the month of June 2019, are equal whether the CO atmospheric source of CO₂ is activated or not. These similarities in terms of bias and RMSE are seen for all the stations (Figure 3).



These first analysis only show a very slight impact of the activation of the CO atmospheric source of CO₂ on CO₂ simulated concentrations at the surface. It also does not show a clear improvement of the comparison with surface measurements.

Table 4. Statistics on the performance of the CHIMERE CTM compared to independent surface mole fraction measurements, for the year 2019. Mean prior of determinant coefficient (r^2), Root mean squared error (RMSE) and bias, taking into account all the assimilated surface stations.

Experiment	r^2	RMSE (ppm)	Bias (ppm)
FWD-ref	0.53	5.91	0.31
FWD-CO/CO2	0.53	5.91	0.32

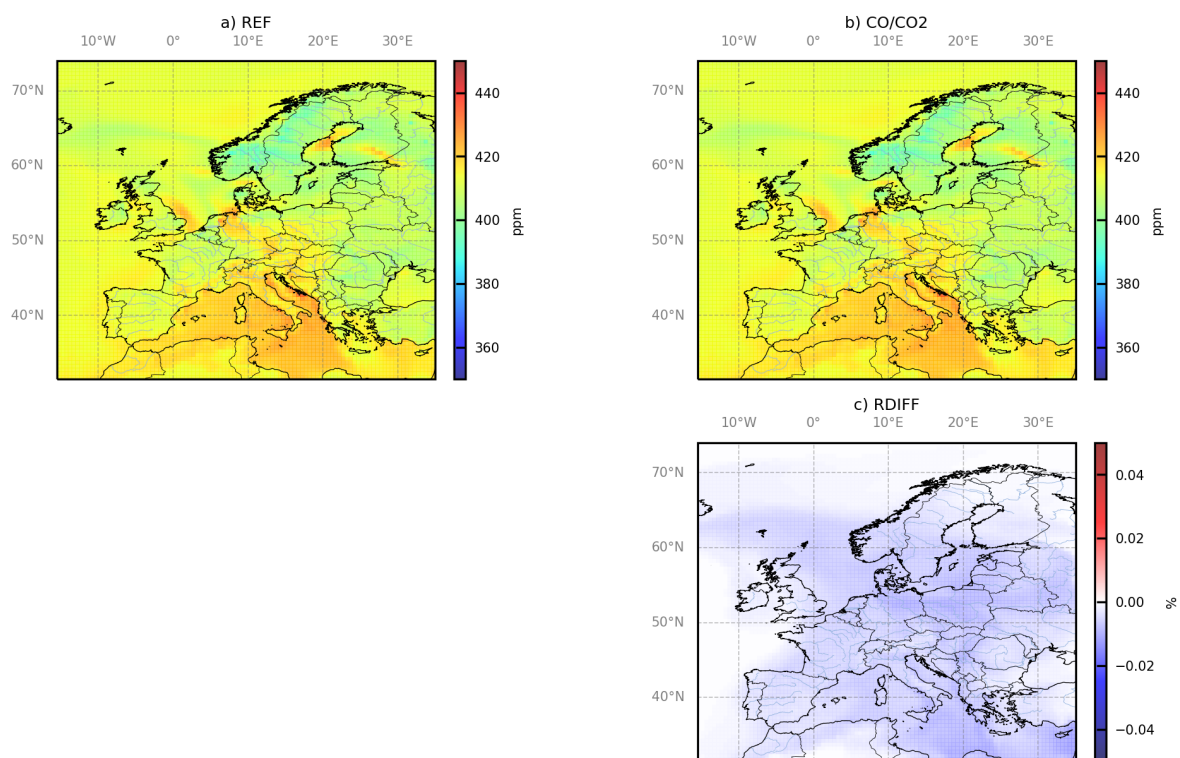


Figure 2. CHIMERE CO₂ simulated concentrations at the surface, activating the CO atmospheric source of CO₂ (b) or not (a), in ppm and c) Relative differences, in %. Illustration for the 15th of June 2019 at 10 am.

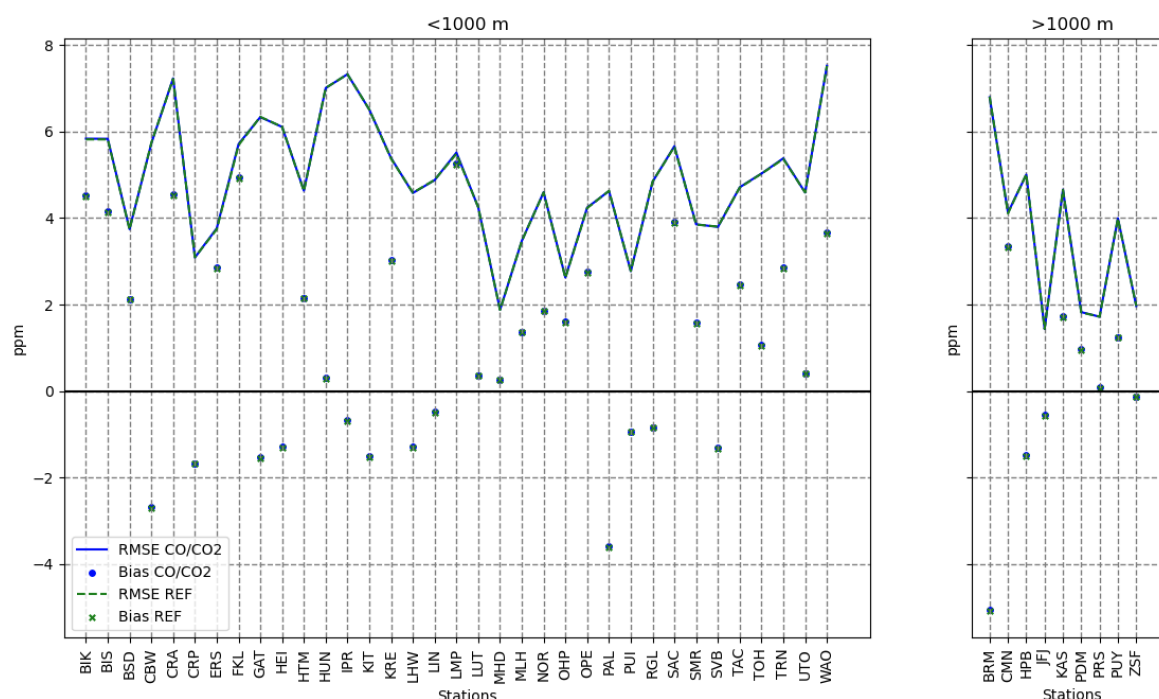


Figure 3. Forward simulation activating (blue) or not (green) the CO atmospheric source of CO₂ mean bias (dots) and RMSE (solid lines) at observation site with altitude lower (left) and higher (right) than 1000m, in ppm, for the month of June 2019.

Fit to the assimilated OCO-2 observations

The reduction of the misfits between the simulation and the assimilated observations due to the corrections applied by the CIF-CHIMERE inversions to their prior estimate of the NEE is illustrated in Figure 1d. The prior misfits between these observations and the prior simulation are strongly decreased, with a reduction of the bias reaching about 78% and 77%, respectively with the simulations activating the CO atmospheric source of CO₂ or not, in June 2019.

Corrections to the NEE fluxes therefore seem to produce a clear improvement in the fit of simulated mole fractions to the satellite observations. **The bias between the OCO-2 observations and the simulation is reduced in a similar way with the simulations with and without activating the CO atmospheric source of CO₂.**

2.3. Evaluation of the posterior emissions with independent surface measurements

The evaluation of the NEE posterior fluxes –was made by comparing posterior simulations with independent surface measurements and is illustrated in Figure 4 and in Table 5, which present statistics for all the available stations in June 2019. The corrections applied by the CIF-CHIMERE inversion to the prior estimate of the NEE from the OCO-2 satellite observations mainly result in an improvement of the comparison with surface measurements.

For example, when taking the hourly surface observations of all the stations into account, the increase (from the prior to the posterior simulations) of the monthly r^2 coefficient between simulated versus measured CO₂ is about 15% whether the CO atmospheric source of CO₂ is considered or not (Table 5). The mean biases between simulated versus measured CO₂ are also strongly decreased by about 96% whether the CO atmospheric source of CO₂ is considered or not (Table 5). **Corrections to the NEE fluxes with a system considering the atmospheric source of CO₂ results in a similar comparison of simulated**

mole fractions with surface measurements as a system not considering the atmospheric source of CO₂.

Table 5. Statistics on the performance of the CHIMERE CTM compared to independent surface mole fraction measurements, before and after the inversions, for the month of June 2019. Mean prior and posterior of determinant coefficient (r^2), Root mean squared error (RMSE) and bias, considering all the assimilated surface stations.

	r^2		RMSE (ppm)		Bias (ppm)	
Experiment	prior	post	prior	post	prior	post
INV-ref	0.21	0.25	4.85	5.08	0.79	-0.17
INV-CO/CO ₂	0.21	0.25	4.85	5.08	0.78	-0.18

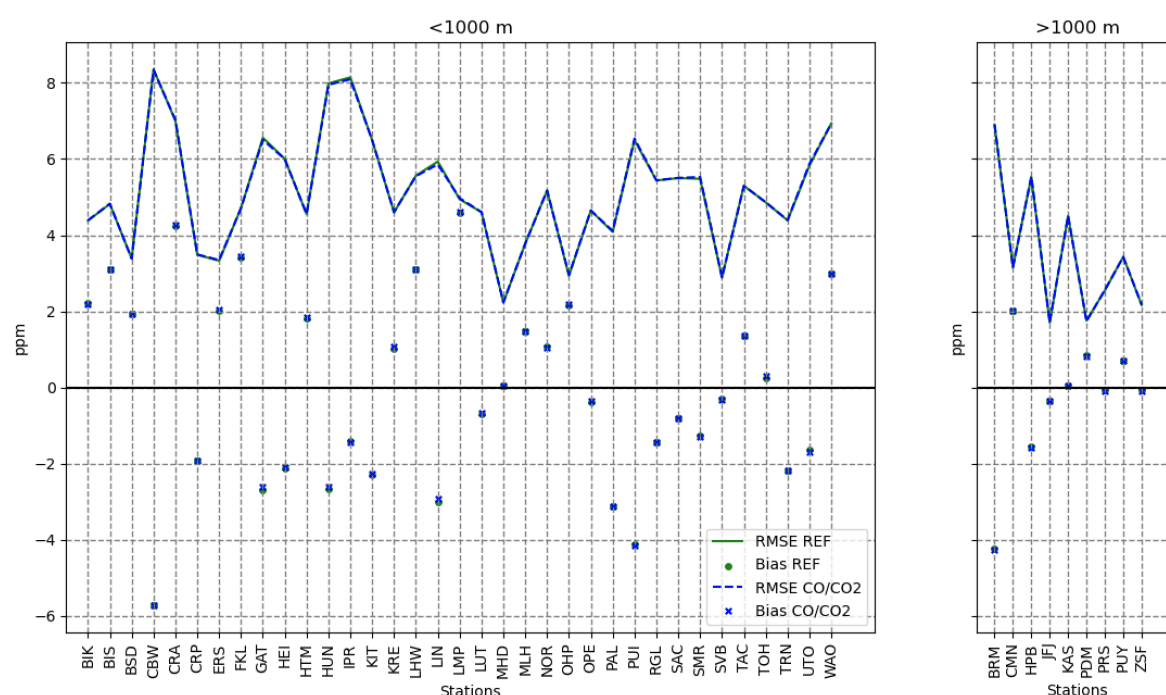


Figure 4. Evaluation of the posterior simulation activating the CO atmospheric source of CO₂ (blue) or not (green) mean bias (dots) and RMSE (solid lines) at independent surface measurements, in ppm, for the month of June 2019.

2.4. EU27+3 NEE budget and spatial variability of the corrections applied to the prior terrestrial ecosystem fluxes from surface observations

Both the inversions, considering the CO atmospheric source of CO₂ or not, result in a NEE budget of about -0.29 PgC for the EU-27+3 area (including United Kingdom, Switzerland and Norway in addition to EU-27) from the ORCHIDEE CRUERA prior estimates of about -0.23 PgC, for the month of June 2019.

Figure 5 presents maps of the corrections provided by the inversions to the ORCHIDEE CRUERA prior estimates when assimilating OCO-2 satellite observations in June 2019, whether the CO atmospheric source of CO₂ is considered or not. The patterns of the corrections are very similar, with differences not exceeding 5%, whether considering the CO atmospheric source of CO₂ or not. Considering the CO

atmospheric source of CO₂ or not, therefore, does not seem to have an impact on the spatial corrections applied to the prior terrestrial ecosystem fluxes, at least at the monthly scale in summer.

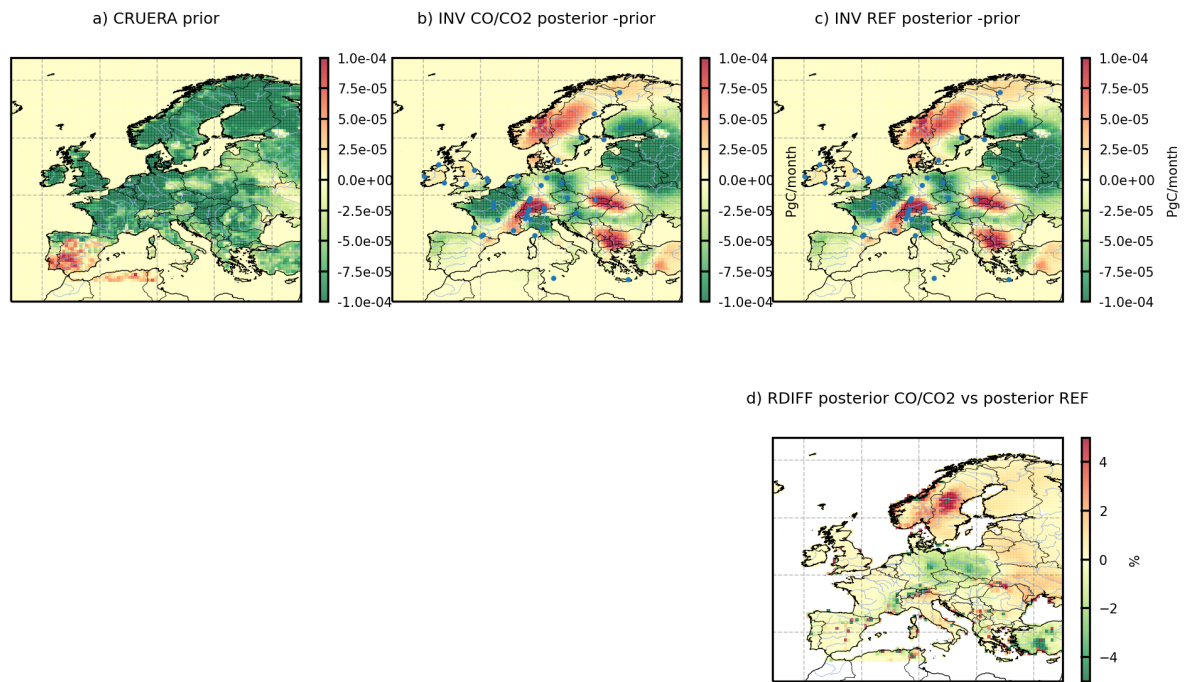


Figure 5. a) ORCHIDEE CRUERA prior estimates and maps of the corrections provided by the inversions to the prior when assimilating OCO-2 satellite observations, considering the CO atmospheric source of CO₂ (b) or not (c), in PgC/month, in June 2019. d) Relative differences between the posterior NEE fluxes considering the CO atmospheric source of CO₂ or not. Blue circles indicate the location of the surface stations used for the evaluation of the NEE posterior estimates.

Conclusions

Several conclusions can be made:

- The computational cost of the inversions is much higher when the CO atmospheric source of CO₂ is activated versus not activated. This computational cost could be reduced in the future by using pre-calculated OH and treating CO oxidation to CO₂ as first order process, or by porting the CHIMERE code to GPU-compatible environment.
- The mean monthly biases between simulated and measured CO₂ mole fractions are very similar both at the yearly scale and at the monthly scale for the month of June 2019 whether the CO atmospheric source of CO₂ is activated or not.
- The spatial corrections applied to the prior terrestrial ecosystem fluxes are similar in June 2019 when the posterior estimates at the 0.5° spatial resolution identify a CO₂ peak uptake, whether the CO atmospheric source of CO₂ is activated or not.

At this stage, we therefore recommend not considering the CO atmospheric source of CO₂ for the inversions of CO₂ land ecosystem fluxes.

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