



WP4 COORDINATING EFFORTS ON ATTRIBUTING CO₂ EMISSIONS FROM IN-SITU MEASUREMENTS

2nd General Assembly

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Roadmap from the 2015 CO₂ report

The abovementioned limitations can be overcome by collecting two types of atmospheric measurements:

- Dense atmospheric CO₂ measurements, with detailed, high resolution, coverage of emissions hotspots. With their global and high revisit coverage, satellite observations of column integrated CO₂ (XCO₂) at high spatial resolution can meet this requirement, e.g., using space-borne sensors with imaging capabilities (Chapter-3).
- In-situ measurements of specific tracers in the atmosphere that allow the separation of fossil CO₂ from other sources. The best of these tracers is ¹⁴C (radiocarbon) in atmospheric CO₂, which is nearly a direct tracer of emissions (Turnbull et al., 2014) but is so far expensive to measure because of its very low abundance. Fossil fuels do not contain any radiocarbon: when CO₂ from their combustion is released to the atmosphere, it dilutes with CO₂ from other sources that contain ¹⁴C, which creates a measurable depletion of the ¹⁴C isotope composition in air masses containing fossil CO₂. The accuracy at which the fossil fuel component of atmospheric CO₂ can be determined from ¹⁴C measurements of CO₂ in air samples is about 1 ppm with Accelerator Mass Spectrometry. Carbon monoxide is also a tracer of combustion processes. In

many environments, it is a reas though subject to uncertainty in differences in fuel type and con

CO₂ HUMAN EMISSIONS

We recommend to build urban monitoring networks for selected European large cities. Results from city-scale inversions of CO₂ data from urban networks will be used to evaluate independently satellite-based city-scale emission estimates. ¹⁴C measurements should be deployed a set of approximately 50 atmospheric CO₂ monitoring stations across the European continent, with higher density over regions with high emissions. Results from regional- and continental-scale inversions of in-situ ¹⁴C and CO₂ measurements will be used to evaluate independently satellite-based emission estimates.

C) Emerant Commission

 CU_2

Sense Copenicus Linear environment

WP Objectives

- Explore the practical implications of distinguishing between anthropogenic vs. biogenic CO₂ fluxes when using CO₂ satellite imagery.
 - Do we need accompanying surface measurements?
 - Anthropogenic = fossil fuel emissions, and also non-fossil waste burning, biofuels, etc.
- Focus on the optimization of the space-time sampling of ¹⁴CO₂, CO and APO at the surface.
 - Network design
 - 4 regional modelling frameworks (EMPA, CEA/LSCE, MPI-BGC, NILU)
 - 1 inventory provider (TNO)
 - 1 expert in APO (UEA) and 1 expert in surface flux measurements and citizen data (CMCC)
- + link with WP3 (ULUND)

Survey of existing European in-situ network (D4.1, lead MPI-BGC)

Documented existing *in situ* measurements, their frequency, and uncertainty for:

- carbon dioxide (CO₂)
- carbon monoxide (CO),
- the ratio of radiocarbon in carbon dioxide ($\Delta^{14}CO_2$)
- and atmospheric potential oxygen (APO)

Documented currently-operating urban flux tower sites in Europe, to support the planned work of CMCC to assess the use of these data to solve and/or validate the attribution problem.

• Potentially valuable for the definition of time factors

CO₂ HUMAN EMISSIONS



20° E

10[°] E

10° W

Ground-based in situ measurement sites

ICOS Class 1 ICOS Class 2 Non-ICOS sites

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D4.2: High-resolution scenarios of CO₂ and CO emissions (lead TNO)

- Task Objective: Construction of ten 1×1 km² scenarios of CO₂ and CO emissions associated with anthropogenic activities in Europe over a full year (2015).
- Closely related to WP2 and the production of a new European emission inventory at 6×6 km² for year 2015 which is the basis for the 1×1 km² zoom version for WP4. The baseline 1×1 km² inventory was delivered to the partners on 28 January 2019.
- Quantification of uncertainties in 4 key parameters for this baseline: activity data, emission factors, spatial distribution proxies and temporal distribution proxies.
 - Each of these key parameters has an uncertainty function, which is being included in a covariance matrix.
 Within a Monte Carlo simulation TNO creates an ensemble (N=10) by drawing random samples from this matrix and calculates emission maps for each ensemble member. This creates a set of possible solutions in the emission space, reflecting the uncertainties in the underlying parameters.
- Deliverable report D4.2 submitted for review 08/03/2019; family of grids ready end of March.

Examples of individual grids





Individual maps by sector combined yield the new high resolution 1×1 km² emission grid for WP4. (Point sources at exact location). This dataset is the basis for the family of 10 grids

Uncertainties HR gridded CO₂ and CO emission data: Method used



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I. Uncertainties in activity data and emission factors

- For activity data (incl. biomass¹):
- For CO₂ emission factors (incl. biomass¹):
- For CO emission factors:

¹Biomass low coverage in NIRs

²Indicated range assumed for 95% confidence interval

Average from National Inventory Reports (NIRs) Average NIRs Literature range² (EEA Guidebook, BREFs)

| Sector | | Range found in reported uncertainties (%) | | |
|----------------------------|---------|---|---------------------|-----------------|
| | | Activity data | Emission factors | |
| | | | CO2 | CO down/up |
| Public power | | 2 - 6 | 2 - 5 | 30-50 / 50-200 |
| Industry | | 3 - 4 | 2 - 6 | 20-70 / 40-300 |
| Other stat. comb. | Fossil | 4 - 10 | 2 - 14 | 30-90 / 40-150 |
| | Biomass | 10 - 20 ¹ | 5 - 10 ¹ | 70-90 / 300-800 |
| Road transport (incl. bio) | | 3 - 5 | 2 - 5 | 30-60 / 40-250 |
| Other mobile sources | | 5 - 10 | 2 - 4 | 50-70 / 200-300 |
| | | | | |
| Distribution | | Normal | Normal | Logn. & norm. |

II. Uncertainties in spatial proxy use

Uncertainty in spatial distribution:

- European Pollutant Emission Register (EPRTR) point source location and emission Neglected for now
- Proxy quality (e.g. population, roads, industrial areas) Neglected for now
- Proxy representativeness for linked activity, at individual cell level:
 - 1. Spatial pattern

Combined in 1 numerical indicator

2. Cell values

Spatial uncertainty of 60 most important Sector – Proxy combinations estimated:

Low uncertainty, e.g.

• Household gas use – Population

Medium uncertainty, e.g.

• Iron & steel industry – Older proprietary plant capacity databases

High uncertainty, e.g.

- Gas refining CORINE general industrial area
- Residual emissions Default proxy (e.g. population)

III. Uncertainties in temporal variations

- Fixed monthly, daily and hourly fractions per source sector (GNFR definition) based on long term averages
- In reality, temporal distribution is more irregular and shows large regional differences
- Uncertainty has been assessed by comparing fixed fractions to observations
- E.g. residential combustion in the Netherlands (below)



This comparison suggests an uncertainty of 20 – 50% in the averaged monthly and daily fractions

All GNFR sectors were analysed in a similar manner

Result: uncertainty in total CO₂ and CO emissions per source sector before adding uncertainty of spatial and temporal distribution

Shown error bars indicate range in 10 random draws

CO data preliminary due to difficult drawing from lognormal distributions



¹⁴CO₂ emissions (CEA/LSCE)

- Simulation of C and ¹⁴C cycles (Wang, 2016)
 - Nuclear ¹⁴CO₂ emissions from yearly data in TBq/yr (Zazzeri et al., 2018).
 - Biogenic δ¹⁴CO₂ from extrapolation of products based on ORCHIDEE-MICT simulations (Wang, 2016). These coefficients will be applied to the new VPRM CO₂ biogenic fluxes.
 - Biofuel under the hypothesis of no lag between growth, harvest and burning.
 - Cosmogenic ¹⁴CO₂ neglected.
- Inversion : control of the CO₂ fluxes
 - Ant., GPP, respiration and the corresponding δ¹⁴CO₂



(Wang, 2016)

Modelling APO: fossil fuels (UEA and TNO)

 $APO (per meg) = \frac{O_2}{N_2} - \alpha_L \times CO_2 = \left(\frac{ocean_{O_2} - ((\alpha_F - \alpha_L) \times ff_{CO_2}) + (\alpha_L \times ocean_{CO_2})}{S_{O_2}} - \frac{ocean_{N_2}}{S_{N_2}}\right) \times 10^6$

- $\alpha_F = O_2:CO_2$ ratios assigned, either directly using values from the literature (but not many exist) or from assigned H:C ratios for each TNO inventory fuel type, using values from the literature (based on the fact that for most fuels, H:C ratios and $O_2:CO_2$ ratios are strongly correlated).
- Assumptions: for each TNO inventory fuel
 - Sulphur and nitrogen content has negligible impact for most fuels.
 - Fuels are burnt completely, and any CO produced is short-lived in the atmosphere and converted to CO₂ relatively quickly (Keeling et al. 1988).
 - Energy production is nearly proportional to O₂ consumption across a wide range of fuels, based on relatively similar energetic efficiencies (NHHV, in kcal/mole) across different types of fuels (Keeling 1988)
- Assign uncertainties to O_2 :CO₂ ratios that account for the above assumptions (work still ongoing).

Modelling APO: biosphere (UEA)

 Uncertainty in α_L (oxidative ratio of terrestrial biospheric exchange) will be assessed using existing European atmospheric O₂ and CO₂ data.

Used O₂ and CO₂ data from across Europe (ZOT, BIK, MHD, WAO, GOE), and excluded periods that were likely influenced by fossil fuel by using APO, leaving data that are representative of mostly natural sources.

- Diurnal and seasonal variability in α_L ?
- None found
- Is $\alpha_L = -1.1$ valid for Europe?

Use -1.07 +/- 0.04 mol/mol (1 sigma SD, normal distribution) instead of -1.1.

• Is there a gradient in α_L between north and south, or east and west?

None found, but geographical ranges of available datasets are limited (latitude: ~51-60 deg N; longitude: ~10 deg W to 90 deg E)

COSMO-GHG forward simulations (EMPA)







Modelling the Benelux plume (CEA/LSCE)

In-situ observations only (red points)

Impact of the assimilation of satellite and in-situ observations

- The satellite pass mainly decreases the uncertainty for the regions within its field of view
- The uncertainty reduction is usually < 25%
- Negative correlations associated to relatively high posterior uncertainties are indicative of difficult separation between the different flux components, but they remain small here.

Correlations between posterior uncertainties in the anthropogenic and biogenic morning budgets

-1.00 -0.75 -0.50 -0.25 0.00 0.25 0.50 0.75 1.00

Uncertainty reductions between prior and posterior morning anthropogenic budgets



Uncertainty reductions between prior and posterior morning biogenic budgets







Satellite pass (hashed area)







0.0 2.5 5.0 7.5 10.0 12.5 15.0 17.5 20.0 Uncertainty Reduction (%) Both In-Situ and Satellite







Modelling the Oslo downwind plume of CO₂ (NILU)



FLEXPART simulation at 1 km horizontal resolution during 3 hours starting 2017-01-01 at 00:00. With the URBES inventory (a). With the ODIAC inventory (b).

Main plans for the rest of Y2

•D4.3: Attribution problem configurations (report, lead MPG, M21)

- Prior ocean APO flux estimates
- Finalize set-ups for CO₂, CO, ¹⁴C and APO (model configurations, input data, statistical hypotheses)
- First consolidated OSSEs



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