Emission ensemble approach to improve the development of multi-scale emission inventories

Philippe Thunis[1], Jeroen Kuenen[2], Enrico Pisoni[1], Bertrand Bessagnet[1], Manjola Banja[1], Lech Gawuc[3], Karol Szymankiewicz[3], Diego Guizardi[1], Monica Crippa[1,4], Susana Lopez-Aparicio[5], Marc Guevara[6], Alexander De Meij[7], Sabine Schindlbacher[8], Alain Clappier[9]

[1] European Commission, Joint Research Centre, Ispra, Italy
[2] TNO, Department of Air, Climate and Sustainability, Utrecht, The Netherlands
[3] Institute of Environmental Protection – National Research Institute (IEP-NRI), Słowicza 32, 02-170 Warsaw, Poland
[4] Unisystem S.A., Milan, Italy
[5] NILU – Norwegian Institute for Air Research, 2027 Kjeller, Norway
[6] Barcelona Supercomputing Center, Barcelona ~~08034~~, Spain
[7] MetClim, Varese, 21025, Italy
[8] Environnent Agency Austria, Spittelauer Lände 5, 1090 Vienna, Austria
[9] Université de Strasbourg, Laboratoire Image Ville Environnement, Strasbourg, France

Correspondence to: Philippe Thunis (Philippe.THUNIS@ec.europa.eu)

Abstract

Many studies have shown that emission inventories are one of the input with the most critical influences on the results of air quality modeling. Comparing emission inventories among themselves is therefore essential to build confidence in emission estimates. In this work we extend the approach of Thunis et al. (2022) to compare emission inventories by building a benchmark that serves as reference for comparisons. This benchmark is an ensemble that is based on three state-of-the-art EU-wide inventories: CAMS-REG, EMEP and EDGAR. The ensemble-based methodology screens differences between inventories and the ensemble. It excludes differences that are not relevant and identifies among the remaining ones, those that need special attention. ~~In this work, an ensemble inventory (median) is created with the aim of monitoring the status and progress made with the development of Europe-wide inventories. This ensemble inventory also allows comparing a large number of inventories at the same time, foster interactions among emission inventory developers and allow for comparing additional inventories (e.g. bottom-up ones) with all ensemble components. In contrast with other fields of applications (e.g. air quality forecast), this emission ensemble is not necessarily better than any of its components. Although it is not the more accurate inventory, it serves here as a common benchmark for the screening. We focus on differences in terms of country totals, country sectorial share and share of the country emissions to the urban areas for emissions of~~ ~~NO_x, PM2.5, PM coarse, NMVOC, SO_x and NH_3. Because the emission "truth" is unknown, the approach does not tell which inventory is the closest to reality. The methodology rather screens differences between inventories, excludes differences that are not relevant and identifies among the remaining ones, those that are larger than a given threshold, and need special attention. The underlying concept is that above this threshold, differences are so large that one or both inventories must be checked.~~

~~The analysis of the ensemble and the comparison with its individual components highlight a large number of inconsistencies. While two of the three inventories behave more closely to each other (CAMS-REG and EMEP), they yet show inconsistencies in terms of the spatial distribution of emissions. These differences mostly occur for SO_2, PM and NMVOC, for the industrial and residential sectors, and reach a factor 10 in some instances. Necessary improvements have been identified, in particular with EDGAR with the PM emissions from the small-scale combustion sector and SO_2 from the industry and power plant sectors. The comparison with the local inventory for Poland leads to identifying another type of inconsistencies, associated to the sectorial share at country level. This is explained by the fact that some emission sources are omitted in the local inventory due to the lacking of appropriate geographically allocated activity data. The screening process led to identify some sectors and pollutants for which discussion between local and EU-wide emission compilers would be needed in order to reduce the magnitude of the observed differences (e.g. in the residential and industrial sectors). The settings used in this work (e.g. the choice of 150 urban areas or the way sectors are aggregated) are arbitrarily fixed and can easily be adapted for the purpose of other comparisons.~~

We applied the ensemble-based screening to both a EU-wide and a local (Poland) inventory. The EU-wide analysis highlighted a large number of inconsistencies. While the origin of some differences between EDGAR and the ensemble can be identified, their magnitude remains to be explained. These differences mostly occur for SO_2, PM and NMVOC, for the industrial and residential sectors, and reach a factor 10 in some instances. Spatial inconsistencies mostly occur for the industry and other sectors.

At the local scale, inconsistencies relate mostly to differences in country sectorial shares that result from different sectors/activities being accounted for in the two types of inventories. This is explained by the fact that some emission sources are omitted in the local inventory due to lack of appropriate geographically allocated activity data. We identified sectors and pollutants for which discussion between local and EU-wide emission compilers would be needed in order to reduce the magnitude of the observed differences (e.g. in the residential and industrial sectors).

The ensemble-based screening proved to be a useful approach to spot inconsistencies by reducing the number of necessary inventory comparisons. With the progressive resolution of inconsistencies and associated inventory improvements, the ensemble will improve. In this sense, we see the ensemble as a useful tool to motivate the community around a single common benchmark and monitor progress towards the improvement of regional and locally developed emission inventories.

Keywords: emission inventories, quality assurance, quality control, screening, urban emissions, ensemble

1. Introduction

Many studies have shown that emission inventories are one of the inputs with the most critical influences on the results of air quality modeling (Kryza et al., 2015, Zhang et al., 2015). Even more concerning, certain studies have shown that important uncertainties affect emission inventories, which may impeach conclusions based on air quality model results (Trombetti et al., 2018, Markakis et al., 2015). These uncertainties result from the need to compile a wide variety of information to develop an emission inventory. For the many pollutants and activity sectors to cover, the spatial and temporal distribution of emissions is typically based on proxies that can be estimated through different methods.

In Thunis et al. (2022), we showed that comparing emission inventories is an effective way to detect inconsistencies when differences are very large. A methodology was designed to compare two emission inventories, one against the other. This methodology identifies disparities between the two inventories by assessing country totals, their sectorial share and the proportion of the country emissions attributed to the urban areas. In this work, we adhere to the same principle of analyzing differences while introducing a novel ensemble concept to facilitate the simultaneous comparison of a larger number of inventories.

Ensemble of models have widely been used in climate (Kotlarski et al., 2014) and air quality modelling fields throughout the world (Stevenson et al., 2006; Vautard et al, 2009; Marecal et al. 2015; Brasseur et al., 2019) as they generally provide~~ing~~ better and more robust results ~~using a set of model results instead of relying on a unique realization~~. While in some instances, reference values (e.g., measurements) exist against which models can be compared, this is unfortunately not the case for emissions, and hence the emission ensemble is not necessarily better than any of its ~~components~~members. The emission ensemble is therefore not a more accurate inventory. This is, however, not an issue as the ensemble is used here as a common benchmark for comparison. Moreover, our focus is on differences between emission estimates rather than on their absolute values, for which accuracy and robustness is of secondary importance. The underlying concept is that above a certain threshold, differences are so large that one or both inventories can be considered wrong. The choice of this vocabulary, i.e. wrong is intentional and is meant here to foster the process of reviewing the data when differences exceed a given threshold. In other words, a factor 100 difference between inventories for a given sector/pollutant most likely reveals one or more significant errors (or inconsistencies) which are relatively straightforward to identify and must be addressed in either one or both inventories. ~~The emission ensemble is therefore not a more accurate inventory but can serve as a common benchmark to support the assessment of methods to develop spatially resolved emission inventories.~~ ~~Because the emission "truth" is unknown, the approach does not tell which inventory is the closest to reality.~~ The methodology ~~rather~~ screens differences between inventories, excludes differences that are not relevant (i.e., large differences on low emission values are disregarded) and identifies among the remaining ones, those that ~~are larger than a given threshold, and~~ need special attention. ~~The underlying concept is that above this (arbitrary) threshold, differences are so large that one or both inventories can be considered wrong. The choice of this vocabulary, i.e. wrong is intentional and is meant here to foster the process of reviewing the data when differences exceed a given threshold. In other words, a factor 100 between inventory estimates for a given emission most likely reveals one or more huge errors (or inconsistencies) that are relatively straightforward to identify and must be addressed in one or both inventories.~~

~~In Thunis et al. (2022) we designed a methodology to compare two emission inventories, one against the other. This methodology was analysing differences the differences between these two inventories in terms of country totals, country sectorial share and share of the country emissions to the urban areas (i.e. how much of the country total is allocated to the urban area). In this work we follow the same principle to analyse differences but we introduce an ensemble concept to allow comparing a larger number of inventories at the same time.~~

In addition to this key advantage, several other objectives are pursued by introducing the ensemble for EU wide emission inventories, namely~~-~~ (1) to create a unique common benchmark~~, based on state-of-art inventories,~~ to monitor and quantify the current level of agreement ~~associated~~ amongto ~~these inventories~~the ensemble members; (2) to identify and characterize the largest mismatches in terms of pollutant, sector among ~~all~~ them~~ensemble components~~; (3) to foster interactions between EU wide emission inventory developers around identified inconsistencies and (4) to allow for comparing additional inventories (e.g. bottom-up ones) with ~~all~~ the ensemble~~ components in a bilateral approach~~. ~~Because the emission "truth" is unknown, the approach does not tell which inventory is the closest to reality. The methodology rather screens differences between inventories, excludes differences that are not relevant (i.e., large differences on low emission values are disregarded) and identifies among the remaining ones, those that are larger than a given threshold, and need special attention. The underlying concept is that above this (arbitrary) threshold, differences are so large that one or both inventories can be considered wrong. The choice of this vocabulary, i.e. wrong is intentional and is meant here to foster the process of reviewing the data when differences exceed a given threshold. In other words, a factor 100 between inventory estimates for a given emission most likely reveals one or more huge errors (or inconsistencies) that are relatively straightforward to identify and must be addressed in one or both inventories.~~

~~The emission ensemble is also intended as a focal point for inter-comparisons against which bilateral analyses can take place (one inventory against the ensemble), with the aim to improve the benchmark and assessment. The main advantage is to structure the inter-comparison process around a single benchmark, in our case the ensemble, rather than by organizing a series of disconnected inter-comparisons (inventory 1 vs. inventory 2, inventory 2 vs inventory 3…). Finally, it supports discussions among emission compiling teams on the main inconsistencies, methodologies behind compilations, and gain understanding about the main reasons for differences, with a view to resolve them and progressively improve emission inventories.~~

~~When inconsistencies are identified among EU wide inventories, a~~A comparison of the ensemble with local (intended here as national or sub-national) inventories can be indeed helpful, as ~~local scale information is an~~they are independent estimates~~ independent source of information~~, which methods are based on local knowledge and understanding of the activities and processes that result on emissions.

The work is structured as follows. In Section 2, we review the screening methodology proposed in Thunis et al. (2022) and ~~discuss the problematic of introduc~~discuss the construction of~~ing an~~ the ensemble in the frame of this screening approach. In Section~~s~~ 3, we apply the ensemble-based screening approach to ~~the~~ one European-wide inventory ~~components of the ensemble~~ whereas ~~we illustrate~~ in Section 4 we illustrate how this ensemble can then be compared to local inventories in a bilateral manner. For the latter, ~~the Poland~~a local inventory developed for Poland is used. In Section 5, we discuss the main findings from both type of comparisons and conclude in Section 6.

2. Description of the methodology

2.1 Overview of the screening methodology

In this section, we provide a brief summary of the screening method detailed in Thunis et al. (2022). The approach aims at comparing two emission inventories over a series of urban areas over which the consistency is assessed for all sectors and pollutants. Based on gridded ~~yearly~~ annual emission~~ inventorie~~s detailed in terms of~~ emitted~~ pollutants (~~denoted as~~ "p") and sectors of activity (~~denoted as~~ "s"), the data required for each pollutant and sector (~~denoted as a~~ [p,s] couple) are twofold and consist of (1) emissions aggregated over specific urban areas (~~denoted by a~~ lowercase notation $e_{p,s}$) and country scale emissions (~~denoted by an~~ uppercase notation $E_{p,s}$).

The c~~C~~onsistency between emissions in both inventories is assessed around three aspects: (1) the total pollutant emissions assigned at country level; (2) the way these country emissions are ~~shared~~ distributed ~~in terms of~~across sector~~ of activity~~ and 3) the way country~~ scale~~ emissions are distributed spatially, and therefore, allocated to ~~to~~ main~~the~~ urban areas. To address these three aspects, we decompose the ratio of the known pollutant-sector emissions for each city as follows:

$$\frac{e_{p,s}^1}{e_{p,s}^2} = \frac{\frac{e_{p,s}^1}{E_{p,s}^1}}{\frac{e_{p,s}^2}{E_{p,s}^2}} * \frac{\frac{E_{p,s}^1}{\bar{E}_p^1}}{\frac{E_{p,s}^2}{\bar{E}_p^2}} * \frac{\bar{E}_p^1}{\bar{E}_p^2} \qquad (1)$$

where $\bar{E}_p$ represents the country scale emissions summed over all sector for a given pollutant. Superscripts refer to the two inventories used for the screening. Equation (1) is an identity where all terms are known from input quantities, i.e. the city and country scale emissions detailed in terms of pollutants and sectors. The three terms on the right-hand side of the identity provide information on ~~the urban share~~spatial distribution (~~denoted as~~ *FAS*, ~~for~~ Focus Area Share*)*, on the country sectorial share (~~denoted as~~ *LSS*, ~~for~~ Large Scale Sectorial share*)* and on the country pollutant totals (~~denoted as~~ *LPT*, ~~for~~ Large scale P~~p~~ollutant Total*)*.

For convenience, we rewrite equation (1) in logarithm form as:

$$log\left(\frac{e_{p,s}^1}{e_{p,s}^2}\right) = log\left(\frac{\frac{e_{p,s}^1}{E_{p,s}^1}}{\frac{e_{p,s}^2}{E_{p,s}^2}}\right) + log\left(\frac{\frac{E_{p,s}^1}{\bar{E}_p^1}}{\frac{E_{p,s}^2}{\bar{E}_p^2}}\right) + log\left(\frac{\bar{E}_p^1}{\bar{E}_p^2}\right) \qquad (2)$$

Which can be rewritten as equation (3) with simplified notations:

$$\hat{e} = \widehat{FAS} + \widehat{LSS} + \widehat{LPT} \qquad (3)$$

where the hat symbol (^) indicates that quantities are expressed as logarithmic ratios. These three quantities ~~are at~~form the basis of the screening methodology and serve as input information for ~~the~~ a graphical representation ~~as well~~that facilitates the interpretation of the results.

~~Because~~ As the number of [*p*,*s*] points under screening, ~~equal~~ equivalent to the product of the number of pollutants ~~by~~ and ~~the number of~~ sectors ~~itself~~ further multiplied by the number of urban areas (i.e. $N \times N_p \times N_s$), may become overwhelming, we ~~proceed~~ adopt a series ~~with a number~~ of steps to concentrate the ~~that help focusing the~~ screening on priority aspects. First, we restrict the screening to emissions that are relevant, i.e. large enough ~~(in practice the condition > × is tested for each (p,s) couple with a user threshold parameter set by the user, γ_t)~~. As shown in Thunis et al. (2022), this exclusion step ~~with γ_t =0.5~~ leads to eliminating a large fraction of the [*p*,*s*] couples from the screening process (between 80 and 90%). Second, we flag, among the remaining ~~relevant~~ emissions, only those for which inventory ~~differences in~~ emission ratios are larger than a given threshold (β_t).

~~Differences originate from methodological choices but also from errors generated during the inventory compilation process.~~ When differences are small, it is not possible to tell whether they originate from methodological choices or from errors. We refer to these small differences as "uncertainty". Although very large differences may result from methodological choices as well (e.g., inclusion or not of particulate matter condensable emissions for the residential sector), they are more likely to be associated to errors. Given the magnitude of the differences, it will in most cases be possible to identify one best value out of the two inventory estimates, even though the true emissions are unknown. These large differences are named "inconsistencies". In the proposed screening methodology, a β_t threshold of 2 (free parameter) is introduced to distinguish inconsistencies from uncertainties.

As a follow-up step, all [p,s] couples that remain after the relevance ~~(>)~~test and inconsistency detection steps ($\beta_{p,s} > \beta_t$), are used to calculate an "Emission Consistency Indicator (ECI)" as follows:

$$ECI = \max_{\{relevant\ emissions\}} \frac{\log(\beta_{p,s})}{\log(\beta_t)} \tag{4}$$

The ECI quantifies the maximum difference among all relevant [p,s], normalized by the inconsistency level (β_t). It therefore quantifies the ratio between the maximum inconsistency and the assumed level of uncertainty. A value of ECI less than one means that all differences are considered as uncertainty (in other words none of the inventory can be identified as best performing). Together with the ECI, which quantifies this maximum difference, we associate the percentage of inconsistent [p,s] with respect to the total number of relevant data, to provide information on the number of detected inconsistencies. ~~To facilitate the screening process, these concepts are displayed graphically.~~

Finally, we prioritise inconsistencies following the LPT – LSS – FAS hierarchy. In other words, if large scale inconsistencies are spotted for LPT, they are flagged as the priority, regardless of the magnitude of inconsistencies calculated for LSS and/or FAS. If no inconsistency is flagged for LPT, the same holds for LSS regardless of the level of inconsistency calculated for FAS.

Consequently, the inconsistency flagged as priority might not be the largest inconsistency. This hierarchy is motivated by the fact that addressing large scale inconsistencies will lead to potentially resolving several ~~many~~ issues ~~at small scale~~ at once (e.g. all urban areas within a given country). Inconsistencies are counted when the individual terms in equation (3) are larger than the threshold β_t but also when the indicators sums (i.e., $\widehat{FAS} + \widehat{LSS} + \widehat{LPT}, \widehat{LSS} + \widehat{LPT}$) exceed this threshold.

It is important to note that the ~~approach~~ method follows a bottom-up approach, i.e., we assess the three types of inconsistencies for each city, pollutant and sector. This means that the same LPT inconsistency ~~are~~ is counted for all cities within a given country or for all sectors for a given pollutant. Similarly, a LSS inconsistency is counted for each city belonging to the same country. While this might be seen as double counting of some inconsistencies, the approach allows comparing local vs country scale indicators.

2.2 Construction of an ensemble as reference

This work aims at applying ~~the~~ a novel ensemble concept to extend the Thunis et al. (2022) methodology to several inventories. The ensemble is calculated from EU-wide inventories that have been developed and regularly updated over several years within the EU[1]. While either the mean or the median of these inventories could be used to calculate the ensemble, we choose ~~here~~ to use the median as it has been shown to be a more robust indicator compared to the ~~than the~~ mean (Riccio et al. 2007). Indeed, if one of the inventories is a strong outlier (i.e., much larger or much smaller values), the mean would be strongly influenced by these extreme values and would differ from the values of ~~the majority~~most of the inventories. On the other hand, the median is not affected by extreme values and therefore takes a value closer to the values taken by ~~the majority~~most of the inventories. It therefore remains further away from outliers, which become easier to identify.

In this work, the ensemble is created from three state-of-the-art Europe wide inventories: CAMS-REG (Copernicus Atmospheric Monitoring Service), EMEP and EDGAR.

EDGAR is a comprehensive global emission inventory providing country and sector specific greenhouse gas and air pollutant emissions from 1970 up to date. EDGAR is becoming a global reference for anthropogenic emissions, in particular contributing to the IPCC AR6 (Sixth Assessment Report) and to the annual UNEP emissions gap reports (UNEP2023) tackling global climate change issues. In the context of air pollution, EDGAR is also widely used by air quality modellers, playing an important role as gap-filling inventory in the Hemispheric Transport of Air Pollution mosaic compilation. Emissions are computed using a consistent methodology for all world countries, following the IPCC Guidelines (IPCC 2006, 2019) and EMEP/EEA Guidebook (EMEP/EEA, 2016, 2019) for greenhouse gases (GHGs) and air pollutants, respectively. Emissions are calculated for all anthropogenic sectors outlined by the IPCC excluding Land Use, Land Use Change and Forestry. This computation utilizes international statistics and default emission factors complemented with state-of-the-art information. Subsequently, annual

[1] Note that EDGAR is designed as a global inventory but we consider here its European coverage only in this analysis and refer to it as a European wide inventory

emissions specific to each sector and country are downscaled globally at 0.1x0.1 degree employing a multitude of spatial proxies. Comprehensive insights into the EDGAR methodology and the underlying assumptions regarding the spatial data used for downscaling national emissions are available in several scientific publications (Janssens-Maenhout et al. 2015, 2019; Crippa et al. 2018, 2021; Crippa et al. 2020; Oreggioni et al. 2022). Additionally, the yearly emission data are further disaggregated into monthly emissions to further support atmospheric modellers in capturing the seasonality of anthropogenic emissions (Crippa et al. 2020).

CAMS-REG version 5.1 is an emission inventory developed as part of CAMS to support European scale air quality modelling (Kuenen et al. 2022). The inventory builds on the officially reported emission data to EMEP in the year 2020, which are complemented by other sources where reported data are not available or deemed of insufficient quality. The data are spatially distributed consistently across the entire domain at a resolution of 0.05x0.1 degrees (latitude-longitude). The spatial distribution takes into account specific point source emissions as reported in the European Pollutant Release and Transfer Register (EPTR2022) to correctly represent point source emissions to the extent possible. The emissions are provided in GNFR (Gridded Nomenclature For Reporting) format. The emission dataset is used in support of the CAMS regional modelling activities, but is also publicly available to support air quality assessment at European level. CAMS-REG-v5.1 is an update of version 4.2 that includes official national emission submissions for the year 2020.

The EMEP-GNFR emissions (Mareckova et al., 2017), based on 2017 reporting, are compiled within the "UNECE co-operative programme for monitoring and evaluation of the long-range transmission of air pollutants in Europe", or also known as EMEP. EMEP is a scientifically based and policy driven programme under the Convention on Long-range Transboundary Air Pollution (CLRTAP) for international co-operation, that has the final aim of solving transboundary air pollution problems. Emissions are built from officially reported data provided to CEIP (Centre of Emission Inventory and Projection by the Member States in Europe) and follow the EMEP/EEA guidebook guidelines (EMEP/EEA 2019) to define the annual totals. The emissions are gap-filled with gridded TNO data from CAMS and EDGAR. The dataset consists of gridded emissions for SO_x, NO_x, NMVOC, NH_3, CO, $PM_{2.5}$, PM_{10} and PMcoarse at 0.1° x 0.1° resolution. More information on the emissions and where to download can be found in the User Guide (https://emep-ctm.readthedocs.io/en/latest/) and in Mareckova et al., (2017). The EMEP domain covers the geographic area between 30°N-82°N latitude and 30°W-90°E longitude.

Based on these three inventories, the ensemble ~~(see details in following section) and~~ is defined on a yearly basis (here 2018)~~by taking values of the year of interest.~~ . Urban ($e_{p,s}$) and country emissions ($E_{p,s}$) for the selected year are required as input. Independent ensemble values for *E* and *e* are defined for each pollutant-sector couple [p,s] ~~for each $E_{p,s}$ and $e_{p,s}$~~ as the median of the three inventory values. For a given area, the urban and country scale emission ensembles for a given year read as:

$$e_{p,s}^{ens} = median\left\{e_{p,s}^{CAMS}, e_{p,s}^{EMEP}, e_{p,s}^{EDGAR}\right\}$$
$$E_{p,s}^{ens} = median\left\{E_{p,s}^{CAMS}, E_{p,s}^{EMEP}, E_{p,s}^{EDGAR}\right\} \quad (5)$$

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Note that this calculation implies that $e_{p,s}^{ens}$ and $E_{p,s}^{ens}$ might not belong to the same inventory for a given area and pollutant-sector couple [p,s]. It is also worth mentioning that should one inventory pollutant-sector value behave as an ~~outlier,~~outlier; its value will not be selected in the ensemble.

As the three emission inventories are characterised by different grid resolutions and sector aggregations, harmonisation is required to construct the ensemble. This is done in 2 steps:

- by grouping the initial emission categories into common categories based on the GNFR classification (NFR-I, 2023 and Table 1 in supplementary material). The original GNFR sectors have been aggregated in 5 categories: road transport (F), residential (C), power plants (A), industry (B) and others. The latter category includes fugitive emissions (D), solvents (E), shipping (G), aviation (H), off-road transport (I), waste (J) and agriculture (K-L).
- by aggregating gridded emissions on common polygons that delineate the area covered by an urban area or by a country. Urban area emissions ($e_{p,s}$) ~~s~~ are ~~defined~~calculated ~~as~~over functional urban areas (FUA, OECD 2012), ~~for which emissions ($e_{p,s}$) are obtained by aggregating grid cell values over these areas. The FUA is~~ composed of a core city plus its wider commuting zone, consisting of the surrounding travel-to-work areas. About 150 FUAs across Europe are selected for this screening. Details on these ~~cities~~urban areas are provided in Thunis et al. (2018). The larger scale emissions ($E_{p,s}$) are defined at country level, level at which emissions are initially reported for these emission inventories.

In terms of pollutants, we consider NOx, NMVOC, PM2.5, PMco (coarse PM, calculated as the difference between PM10 and PM2.5 emissions), SO2 and NH3.

The ~~proposed~~ approach then consists in comparing ~~each~~ a given inventory with the ensemble to identify inconsistencies. It is important to note that while ~~This generalization of Thunis et al. (2022) leads to the same kind of conclusions where~~the approach ~~inconsistencies most~~ likely highlight errors in the ~~flagged~~ inventory under screening, ~~but~~ it is however not possible to exclude that the inconsistency originates from the ensemble (i.e., be present in all other inventories). Despite this inconveniency, the method remains an efficient way to identify, among the large amount of data from several inventories, those that are most likely to be problematic and therefore need to be verified in priority.

~~This generalization of Thunis et al. (2022) leads to the same kind of conclusions where inconsistencies most likely highlight errors in the flagged inventory, but it is however not possible to exclude that the inconsistency originates from the ensemble (i.e., be present in all other inventories). Despite this inconveniency, the method remains an efficient way to identify, among the large amount of data from several inventories, those that are most likely to be problematic and therefore need to be verified in priority.~~

3. Application to EU-wide inventories

~~3.1 Input data~~

~~The screening methodology is applied to three state of the art inventories: CAMS-REG v5.1, EDGAR v.6.1 (Crippa et al. 2022) and EMEP (2022 gridding) that cover emissions for Europe for the main air pollutants. Urban areas are defined as functional urban areas (FUA, OECD 2012) for which emissions ($e_{p,s}$) are obtained by aggregating grid cell values over these areas. The FUA is composed of a core city plus its wider commuting zone, consisting of the surrounding travel-to-work areas. About 150 FUAs across Europe are selected for this screening. Details on these cities are provided in Thunis et al. (2018). The larger scale emissions ($E_{p,s}$) are defined at country level, level at which emissions are initially reported for these emission inventories.~~

~~In terms of pollutants, $E_{p,s}$ and $e_{p,s}$ include the following: NO_x, NMVOC, $PM_{2.5}$, PM_{co} (coarse PM, calculated as the difference between PM_{10} and $PM_{2.5}$ emissions), SO_2 and NH_3, whereas sectors are based on the Gridded Nomenclature For Reporting (GNFR) classification (NFR-I, 2023 and Table 1 in supplementary material). The original GNFR sectors have been aggregated in 5 categories: road transport (F), residential (C), power plants (A), industry (B) and others. The latter category includes fugitive emissions (D), solvents (E), shipping (G), aviation (H). off-road transport (I), waste (J) and agriculture (K-L). The reference year for all three inventories is 2018. Finally, the threshold to distinguish relevant from non-relevant emissions as well as the threshold to distinguish uncertainties from inconsistencies are set to 0.5 and 2, i.e., $\gamma_t=0.5$ and $\beta_t=2$.~~

~~CAMS-REG version 5.1 is an emission inventory developed as part of the Copernicus Atmosphere Monitoring Service (CAMS) to support European scale air quality modelling (Kuenen et al. 2022). The inventory builds on the officially reported emission data to EMEP in the year 2020, which are complemented by other sources where reported data are not available or deemed of insufficient quality. The data are spatially distributed consistently across the entire domain at a resolution of 0.05x0.1 degrees (lat-lon). The spatial distribution takes into account specific point source emissions as reported in the European Pollutant Release and Transfer Register (EPTR2022) to correctly represent point source emissions to the extent possible. The emissions are provided in GNFR format. The emission dataset is used in support of the CAMS regional modelling activities, but is also publicly available to support air quality assessment at European level. CAMS-REG-v5.1 is an update of version 4.2 (which is extensively described in Kuenen et al. 2022), the main difference being the latest version based on the official submissions of national emission inventories in the year 2020.~~

~~EDGAR is a global emission inventory providing country and sector specific greenhouse gas and air pollutant emissions from 1970 till nowadays. EDGAR is becoming a global reference in the field of anthropogenic emissions, in particular contributing to the IPCC AR6 and to the yearly UNEP emissions gap report (UNEP2021) tackling global climate change issues. In the context of air pollution, EDGAR is also widely used by air quality modellers and in particular is used as gap-filling inventory in the context of the Hemispheric Transport of Air Pollution mosaic compilation. Emissions are computed using a consistent methodology for all world countries,~~

~~following the IPCC Guidelines (IPCC 2006, 2019) and EMEP/EEA Guidebook (EMEP/EEA, 2016, 2019) for greenhouse gases (GHGs) and air pollutants, respectively. Emissions are computed for all IPCC anthropogenic emitting sectors, with the exception of Land Use, Land Use Change and Forestry, making use of international statistics and default emission factors complemented with state-of-the-art information. Annual sector and country specific emissions are then downscaled over the globe at 0.1x0.1 degree resolution making use of hundreds of spatial proxies. Details about the EDGAR methodology and the assumptions behind the spatial data used to downscale national emissions are available in several scientific publications (Janssens-Maenhout et al. 2015, 2019; Crippa et al. 2018, 2021; Crippa et al. 2020; Oreggioni et al. 2022). Annual emission data are further disaggregated into monthly emissions to further support atmospheric modellers in simulating the seasonality of anthropogenic emissions (Crippa et al. 2020).~~

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~~The EMEP-GNFR (Gridded Nomenclature For Reporting) emissions (Mareckova et al., 2017), based on 2017 reporting, are compiled within the "UNECE co-operative programme for monitoring and evaluation of the long-range transmission of air pollutants in Europe", or also known as EMEP. EMEP is a scientifically based and policy driven programme under the Convention on Long-range Transboundary Air Pollution (CLRTAP) for international co-operation, that has the final aim of solving transboundary air pollution problems. Emissions are built from officially reported data provided to CEIP (Centre of Emission Inventory and Projection by the Member States in Europe) and follow the EMEP/EEA guidebook guidelines (EMEP/EEA 2019) to define the annual totals. The emissions are gap-filled with gridded TNO data from Copernicus Atmospheric Monitoring Service (CAMS) and EDGAR. The dataset consists of gridded emissions for SO_x, NO_x, NMVOC, NH_3, CO, $PM_{2.5}$, PM_{10} and PMcoarse at 0.1° x 0.1° resolution. More information on the emissions and where to download can be found in the User Guide (https://emep-ctm.readthedocs.io/en/latest/) and in Mareckova et al., (2017). The EMEP domain covers the geographic area between 30°N-82°N latitude and 30°W-90°E longitude.~~

~~As these three emission inventories are characterised by different grid resolutions and sector aggregations, harmonisation is required prior to the screening process for a meaningful comparison. This has been done in 2 steps:~~

- ~~by grouping the initial emission categories to common categories, based on GNFR sectors;~~
- ~~by aggregating gridded emissions on common polygons, representing cities and countries.~~

~~After this process, emissions inventories can be easily compared among each other.~~

~~3.2 Results~~

The first objective of the ~~emission~~ ensemble-based screening is to systematically monitor and quantify ~~the current level of~~existing uncertainties and~~/~~ inconsistencies ~~associated~~ within~~to~~ EU-wide inventories. It aims to~~, and~~ identify the sources of discrepancies in ~~where large differences come from, in~~ terms of pollutant, sector and location. To perform this task, we ~~apply the screening methodology by~~ compar~~ing~~e bilaterally each of the three inventories to the ensemble and ~~report the results~~present the findings in -Figure 1~~Figure 1~~ (~~(top~~ left). This figure provides for all ensemble members an overview of ~~In this figure, only~~existing inconsistencies~~ are shown~~, i.e.~~,~~ for emissions that are relevant (i.e., large enough values) and ~~that are relevant (i.e., large enough values) for which differences~~that differ from -the ensemble ~~between inventories are larger~~by more than a factor 2 ($\beta_t = 2$). Each inconsistent emission [p, s] is represented by a point that has larger-scale emissions ($\widehat{LSS} + \widehat{LPT}$) as abscissa and spatial distribution of emissions ($\widehat{FAS}$) as ordinate. The sum of these two terms is equal for points that lie on "−1" slope diagonals. The diamond shape (in the middle of the diagram) delineates the inconsistency limits. Therefore, each [p, s] point lying outside this shape is an inconsistency. In this diamond diagram, shapes are used to differentiate activity sectors, while colors indicate pollutants. The size of the symbol is proportional to the relevance of the emission contribution. Finally, we use symbol filling to distinguish the type of inconsistencies (i.e., LPT, LSS, and FAS).~~. Symbols are used to differentiate inventories while colours are used to distinguish sectors.~~ We refer to Thunis et al. (2021) for details.

The summary report (bottom part of Figure 1~~the top-left figure~~) provides overview information about inconsistencies. More than 21% (number within brackets beside the ECI indicator) of the relevant emission~~s~~ ratios show inconsistencies. The ECI indicator is equal to 132, meaning that the largest inconsistency is more than two orders of magnitude larger than the level associated to uncertainties. ~~In our case, t~~The EDGAR inventory is flagged for two thirds of them (the total number of inconsistencies, denoted as NI is 227 out of 357), with the largest part of them associated to industry for SO_2 and PM_{co} (see numbers within brackets besides the sectors/pollutants in the bottom legend: Figure 1). Most of the i~~I~~nconsistencies are obtained within the~~mostly originating from the urban~~ allocation of emissions at urban scale~~process~~ (218), although ~~but~~ an important number of them also ~~originates~~ occur at country scale (LSS+LPT=80+59). The diagram also shows that EDGAR reports larger residential and industrial emissions at country level (yellow squares on the right of the X-axis). It is important to remember that flagging one particular inventory does not necessarily indicates that this inventory is the problematic one. But this flagging means that this inventory and/or the others show an important inconsistency for that city, pollutant and sector which requires further checking.

In addition to providing a useful summary that details the current state of variability, the diagram can also serve as basis to monitor progress, through the ECI indicator and associated percentage.

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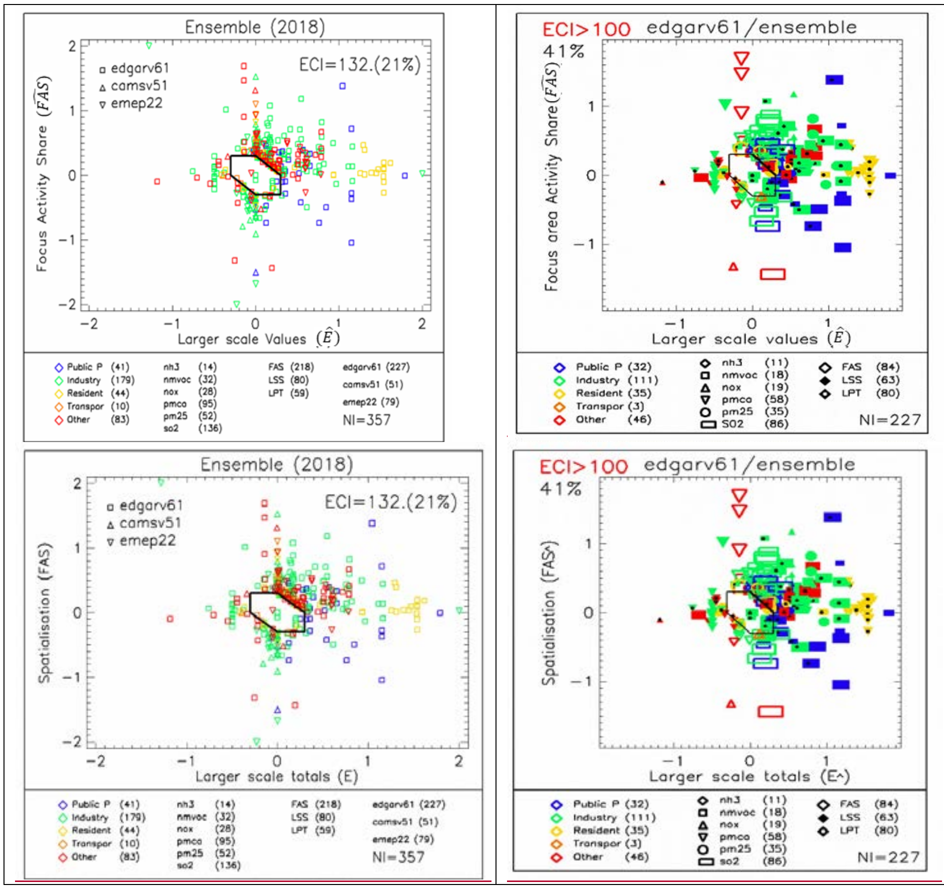

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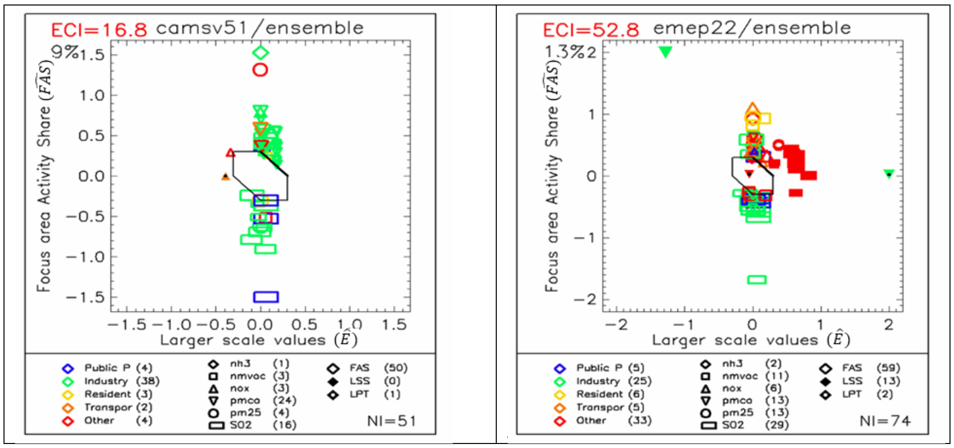


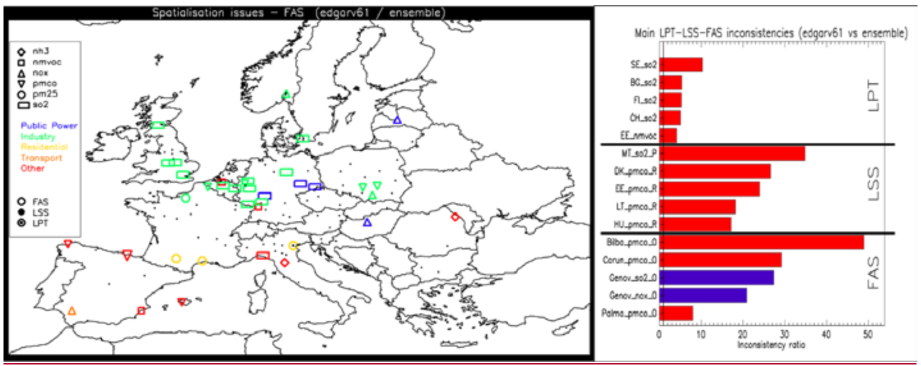
Figure 1: Overview diamonds. The ~~top~~ left diagram shows the comparison of the three ensemble ~~components~~members (CAMS-REG, EDGAR, EMEP) with the ensemble for 2018. The ~~three~~ right~~following~~ picture~~s~~ isolate the bilateral comparison between EDGAR and the~~of each~~ ensemble ~~component with the ensemble~~. Symbols and colours are as specified in the legend. Please note that ~~these~~ symbols/colors differ between the right and left figures~~for the top-left panel, compared to the three others~~. In ~~all~~ both diagrams, only inconsistencies are displayed. For visualization purposes, we limit the axis to a factor 2 in terms of magnitude (from -2 to 2) and bound the ECI to 100 (e.g. values of ECI larger than 100 are plotted with a value of 2). Numbers within bracket in the bottom legend are the total number of inconsistencies for a given pollutant, sector or type.

The ensemble-based screening methodology also serves as a benchmark to compare individual inventories. It is applied here (Figure 1 - right) to one of the three state of the art inventories used to build the ensemble, EDGAR v.6.1 (Crippa et al. 2022). Results for the two other ensemble members: CAMS-REG v5.1 and EMEP (2022 gridding) are discussed in the supplementary material (Section 1).

~~A bilateral comparison of each inventory against the ensemble provides additional information.~~

~~For EDGART, t~~the ECI (>100) indicates that the maximum inconsistency is at least a factor 100 larger than the estimated level of uncertainty ~~(a factor 2 in our case, a value below which differences are assumed to result from uncertainties and small errors, see Section 2.1)~~. Moreover, about 41% of the relevant emission points ~~(large enough emissions)~~ show an inconsistency ~~(difference larger than a factor 2)~~. As indicated ~~by~~ in the overview table, these 41% amount to 227 inconsistencies (NI) ~~that~~ which are shared into about 35% within the spatial distribution of emissions (FAS=84) ~~originating from the urban share~~ and 65% ~~originating from~~at country scale ~~issues~~ (LPT+LSS=83+80). Most of the inconsistencies are identified, as ~~, mostly~~ for SO_2, PM_{co} and $PM_{2.5}$ from the industry sector, in line with the findings of De Meij et al. (2023). ~~.~~There are also an important number of inconsistencies related to the ~~"~~other~~"~~ (46), residential (35) and public power sectors (32). In general, for all inconsistencies, EDGAR estimates are larger than those represented by the ~~the~~ ensemble ~~ones~~ (all points on the right and/or top of the diagram).

To prioritize the inconsistency analysis, ~~In Figure 2~~Figure 2 (right side) shows ~~we identify~~ the largest ~~inconsistencies~~differences ~~(right side) for each of the three right-hand-side terms in equation (3), i.e.~~for LPT (country pollutant total), LSS (country sectorial share) and FAS (spatial distribution~~isation~~), which are also identified on the map~~,~~ (left on Figure 2).~~most important inconsistency for each city (left side) as well as the largest inconsistencies (right side) for each of the three right-hand-side terms in equation (3), i.e. LPT (country pollutant total), LSS (country sectorial share) and FAS (spatialisation).~~



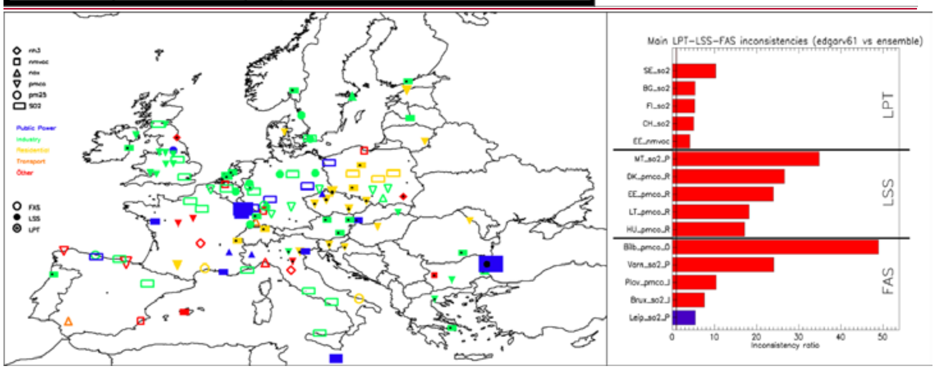


Figure 2: Left~~:~~: Main inconsistencies spotted at urban scale for EDGAR when compared to the ensemble. Only the main spatial inconsistency (FAS) for each city is plotted. See explanation of symbols on the top left of the figure. Right: Major LPT (top 5), LSS (middle 5) and FAS (lower 5) inconsistencies. The two first letters indicate the country code for LSS and LPT whereas the 4 first city letters are given for FAS. Red shading indicates an overestimation and blue shading an underestimation for the EDGAR inventory~~Main inconsistencies spotted at urban scale for EDGAR when compared to the ensemble (2018). Only the main inconsistency for each city is plotted. See explanation of symbols on the top left of the figure. Right: Major LPT (top 5), LSS (middle 5) and FAS (lower 5) inconsistencies. The two first letters indicate the country code for LSS and LPT whereas the 4 first city letters are given for FAS. Red shading indicates an overestimation and blue shading an underestimation for the EDGAR inventory.~~

Th~~ese~~e following ~~figures point to the following~~ main issues can be extracted from Figure 2 for EDGAR:

- Inconsistencies in SO_2 country totals (LPT) are notably observed in Sweden (factor 10), Bulgaria, Finland and Switzerland (factor 5). In the case of Sweden and Finland, we could identify that the main difference comes from the industry sector, ~~and especially from~~particularly the pulp, paper and print sub-sector, for which the inclusion of black liquor

Formatted: Bulleted + Level: 1 + Aligned at: 0" + Indent at: 0.25" use for energy purposes in EDGAR ~~is the main factor for differences~~[2]~~. EDGAR activity data related to the black liquor statistics~~ need to be revised. For Bulgaria, the SO_2 total is dominated by the public power sector for which the activity data, sourced from IEA energy balances, ~~is~~ subject to regular updates, influenc~~ing~~e the magnitude of the differences. According to the Bulgarian Informative Inventory Report ~~IIR~~ (IIR) of emissions in 2022 ~~for Bulgaria~~, SO2 emissions are regularly updated with measurements, which is not the case ~~of~~ for the EDGAR emissions estimat~~ions~~es, explaining part of the differences. Work is in progress to update SO_2 abatement measures in EDGAR. Another issue that can explain these inconsistencies relates to the ~~application of~~ different emission factors applied for SO_2 that are based on the sulphur content of fuels, usually not reported regularly by countries, values which are ~~used~~ integral to~~in~~ CAMS-REG and EMEP[3]. As a follow-up of this analysis, ~~In EDGAR~~ the SO_2 emission factors for the power sector in EDGAR ha~~s~~ve been revised taking into account the limits established by the implementation of the large Combustion Directive (Directive 2001/80/EC).

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~~• Slightly different is the situation in the industry sector where SO_2 emission factors for some fuels need to be revised.~~

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- A larger sectorial share (LSS) at the country level for SO_2 in Malta for Public Power (factor 30), for residential PMco emissions in Denmark, Estonia (above a factor 20) and Lithuania and Hungary (about a factor 10) is found. The large differences in the residential sector ~~between EDGAR and the other inventories based on country reported values~~ is ~~linked~~ related to ~~the estimate of~~ biomass burning emissions, both in terms of technology allocation and emission factors applied. Given the large differences with the ensemble, the review of the EDGAR methodology led to the indication that ~~The~~ EDGAR estimates needed to be updated ~~need to be updated~~, especially in terms of technology allocation. This adjustment is important to accurately reflect the current technological structure within that sector. Although the filter on low emission values (relevance test) is applied, it is not effective in the case of Malta because it is a small country where national totals are composed of few power plants only. The large LSS ratios obtained there are not significant as the values estimated for the power plant sector appear to be very small.

- A few large inconsistencies also appear at the local scale (FAS) due to the use of different proxies to spatially distribute emissions. The largest inconsistencies occur for the other sector (likely originating from the waste treatment installations). ~~This is the case for PMco for the "other" sector in Bilbao (factor 50).~~ This can probably be explained by the approach followed in EDGAR for the waste sector for which all emissions are distributed over a few

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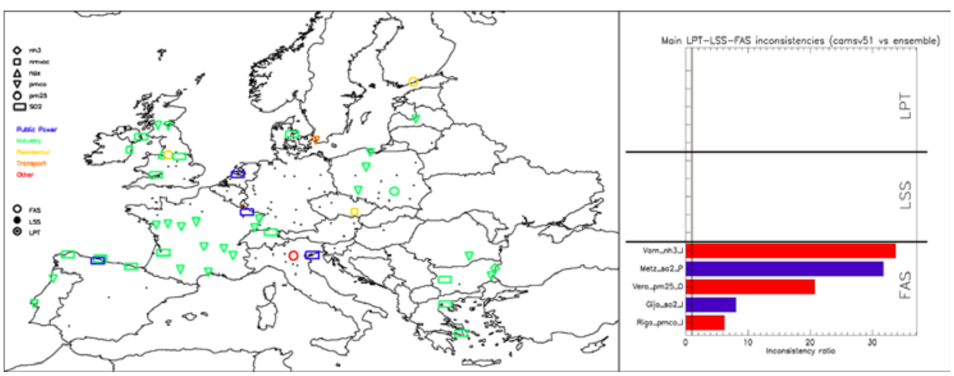
~~[2] In Sweden (IIR 2022), the use of black liquor is not applied for energy purposes, whereas in Finland IIR 2022 a revised methodology for the estimation of SOx-NOx emissions has been performed which resulted in lower country-specific emission factors.~~

[3] The default EMEP/EEA Guidebook 2019 emission factor for SO_2 are w/o abatements and only for 1% mass sulphur content for coal and oil and 0.01 g/m3 for gas (EMEP/EEA guidebook 2019).

locations only, using E-PRTR locations for landfilling and incineration and population in case of missing information. This results in large differences ~~among~~ with other inventories due to the proportion of the emissions being placed within the city area (see Figure ~~1~~ 7 and following in supplementary material, section 3). A ~~sA~~ similar issue appears in many north west European cities ~~Varna f~~for SO_2 for public power (~~factor >20~~green rectangles in the left Figure). Work is in progress to update the spatial allocation of the public power and waste sectors emissions (personal communication M. Crippa 2023).

The ensemble-based comparison highlights an important number of inconsistencies at country level. While the two other ensemble members (EMEP and CAMS-REG) use (but to different extents) officially reported emissions and therefore rely on similar total emissions per country, EDGAR estimates emissions in an independent bottom-up approach, starting from activity levels and emissions factors from international agencies and bodies (Crippa et al., 2018, Oreggioni et al. 2022). This difference in approach can explain a large number of inconsistencies identified for EDGAR but some of them are very large, especially for SO_2 and PM in the industrial sector. For this particular sector, estimates mostly come from the LPS and E-PRTR databases in EMEP/CAMS-REG, with emissions being mostly based on measurements or facility-level estimates. Such information is not used in EDGAR, where estimates are based on fuel consumption and emission factors that are very general and not plant specific.

~~For CAMS-REG, the ECI (=16.8) indicates that the largest inconsistency is around a factor 15 larger than the estimated level of uncertainty. About 9% of the relevant emission points show an inconsistency larger than a factor 2. As indicated by the overview table, these 9% amount to 51 inconsistencies that are almost all related to urban share issues (50), mostly for PM_{co} and SO_2 from the industry sector.~~



~~*Figure 3: same as Figure 2 but for CAMS-REG*~~

~~Figure 3 points to the following main issues:~~

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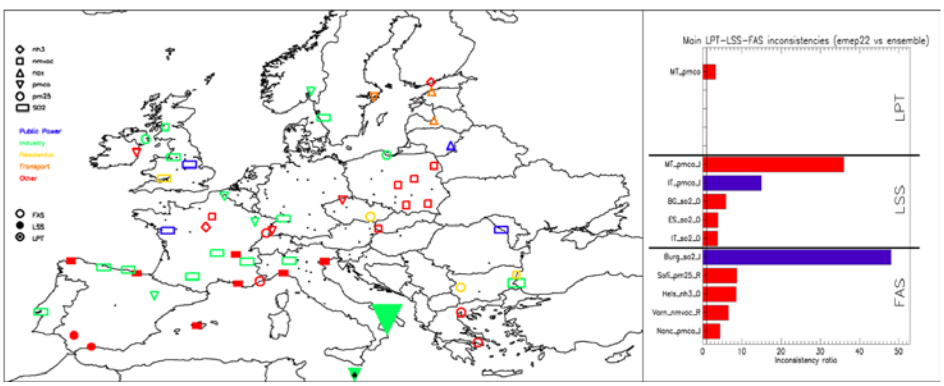
- ~~All major inconsistencies are related to the choice of spatial proxies (FAS) and occur in particular in Varna for industrial NH_3, in Metz for SO_2 from power plant and in Verona for PM2.5 from the “other” sector. These three inconsistencies exceed a factor 20. Note also that these inconsistencies are either over- or under-estimations (red and blue color bars, respectively). In Bulgaria, the largest industrial point source in E-PRTR (68% of the country total) is located near Varna, hence the high emissions there. The large differences among inventories occur due to the proportion of these emissions being placed within the city area (see Figure 2 in supplementary material). The same explains the differences for SO_2 in Metz for the power plant sector or for PM2.5 in Verona for the “other sector”.~~

- ~~Although of lower importance, inconsistencies are also spotted for industrial PMco emissions in France and are systematic in several cities across the country. The same occur for industrial SO_2 emissions in the UK and in Spain. The diamond plot shows that while PMco has larger estimates in the CAMS-REG inventory, the opposite is true for SO_2. A likely explanation for the differences in SO_2 emissions is that their attribution to point sources is done only for those included in point source reporting (E-PRTR). Smaller sources which are below the threshold for E-PRTR reporting are distributed as diffuse sources to industrial zones (land cover class). This may lead to over-allocation in some urban areas.~~

~~For EMEP, the ECI (52.8) indicates that the maximum inconsistency is about a factor 50 larger than the estimated level of uncertainty. About 13% of the relevant emission points show an inconsistency. As indicated by the overview table, these 13% amount to 74 inconsistencies that are mostly related to the spatial share of the emissions (FAS=59), mostly for SO_2 (29), and in a lesser extent to $PM_{2.5}$ (13), PM_{co} (13) and NMVOC (11) originating from the “other” sector (33), but also from the industry (25) sectors.~~



~~*Figure 4: same as Figure 2 but for EMEP*~~

~~Figure 4 points to the following main issues:~~

- ~~One inconsistency only is spotted at country total level (LPT) for the PMco industrial emissions in Malta (factor 3). Similarly to what reported for EDGAR for Malta, the low-~~

~~emission filter is not efficient to remove these small (not relevant) emissions, given the small size of the country~~

- ~~A series of inconsistencies are associated with the sectorial share at country level (LSS). The largest is observed for PMco industrial emissions in Malta (factor >30) and add up to the inconsistency at country total level previously highlighted. The same inconsistency, although as underestimation (blue shaded bar in figure 4), occurs in Italy with a factor 15. LSS inconsistencies also occur for SO_2 emissions from the "other" sector in countries like Bulgaria, Spain and Italy (between a factor 3 and 6)~~
- ~~Regarding inconsistences related to spatial proxies, one large one (factor >50) is flagged in Burgas for SO_2 emissions from the industry sector (see Figure 3 in supplementary material). This type of inconsistencies also occur in a lesser measure in other cities and similarly to CAMS-REG, are likely explained by the precision of their attribution as point sources.~~

4. Application to local inventories: a case-study over Poland

4.1 ~~Input data~~The high resolution Poland emission inventory

The ensemble-based screening methodology also serves as a benchmark to compare local inventories. In this section, it is applied to ~~In this section, we use~~ the ~~local~~ inventory for Poland ~~and compare it to the Europe wide ensemble~~.

The Central Emission Database (CED) is a local emission inventory designed for Polish national air quality modelling. The CED is based on source location and provides accurate resolution-free data, which can be gridded depending on the requested target resolution for different computational grid configurations over Poland (typically 2.5 km over the entire country and 0.5 km for agglomeration zones). The majority of data is processed with respect to its exact geographical ~~localisation~~location. ~~The intention behind CED is to include documented emission sources in Poland. Since the inventory is fairly new (the first version was ready in 2019),~~ ~~p~~Priority ~~was~~ is given to the most critical sectors, like residential combustion (described in detail in Gawuc et al., 2021) and road transport. The road transport data presented in this paper (~~topical~~ relative to~~for~~ 2019) was based on traffic models for the major roads in the country. Emissions on minor roads were distributed using the residue values taken from subtracting emission on major roads from the national totals. The c~~C~~urrent methodology ~~(topical for 2022)~~ is based on smartphone car navigation app which provides GPS data on road traffic and annual average car speed.

One of the essential components of CED is the "National database on greenhouse gases and other substances emission" (so-called national database – NB). NB consists of information on installations and sources' location responsible for emission into the atmosphere. NB has similarities to E-PRTR, but unlike it, it covers all emission sources regardless of type, power or production level. Registered NB users provide information on emission volumes resulting directly from the exploitation of their installations, as well as ancillary processes, which may cause fugitive emissions. ~~NB users may rely on direct stack measurements (continuous or periodic) in case of more significant emitters.~~ To be applied for CED and air quality modelling, the reported data is categorized into SNAP (Selected Nomenclature for Air Pollution) and converted to GNFR if needed (Table 1, supplementary material).

NB is a basis for GNRF A (public power), B (industry), D (fugitive), E (solvents), and J (waste) emission estimations contributing to CED. Two approaches are applied to evaluating CED data. Firstly, as part of each modelling stream (i.e., operational air quality forecast, annual air quality assessment, station representativeness analysis), a comprehensive evaluation is undertaken (station-by-station time series for over 100 monitoring sites for each pollutant). Moreover, spatial patterns of the increments calculated in the assimilation procedure let to identify and improve the assumptions behind CED. The database is updated every year and there is a continuous attempt to improve emission estimates both – for total load and spatial distribution of sources. Modelling results helped to identify missing sources (e.g. resuspension, underestimated agriculture sector, domestic water heating).
All sectors in CED are constantly improved using the best available activity data.

Note that the CED reference year (2019) differs from the ensemble one (2018). Inconsistencies are however generally large enough to justify explanations other than those originating from the difference in terms of reference year.

~~The comparison between CED and ensemble data is performed on 14 cities, 5 sectors and 6 pollutants, leading to 420 emission ratios being tested. Among these 420 available data, 84 only remain after the relevance test ($\gamma_t > 0.5$). These 84 [p,s] points serve as basis to identify inconsistencies ($\beta_t > 2$).~~

~~Note that although the year of comparison differs (2018 for the ensemble vs. 2019 for the Polish emission data), inconsistencies are generally large enough to justify explanations other than those originating from the difference in terms of reference year.~~
~~We first assess how well the Europe wide emission ensemble components agree over Poland and identify the main inconsistencies from a EU-wide perspective. In a second step, we use local information to (1) help solving the inconsistencies identified at European level and (2) identify additional inconsistencies between the ensemble and the local inventory.~~

4.2 ~~Results~~Comparison of the CED inventory to the ensemble

The ensemble-based screening applied to Poland is performed for 14 cities (see city locations in Figure 5), 5 sectors and 6 pollutants, leading to 420 emission ratios being tested.

Before proceeding with the screening of the local data, we first analyse the level of consistency among EU-wide inventory over Poland (Figure 3 is a zoom of Figure 1 over Poland). Among the 420 available data, 84 remain after the relevance test ($\gamma_t > 0.5$). These 84 [p,s] points serve as basis to identify inconsistencies ($\beta_t > 2$). Inconsistencies occur for about 13% of the relevant [p,s] points, with a maximum inconsistency (ECI) 2.5 times larger than the assumed level of uncertainty. As seen from the overview table, most of the issues are related to the EDGAR (20) and EMEP (6) inventories, in particular to the residential sector for EDGAR, to the industry sector for CAMS-REG and to the other sector for EMEP. Additional details are provided in the supplementary material (Section 2).

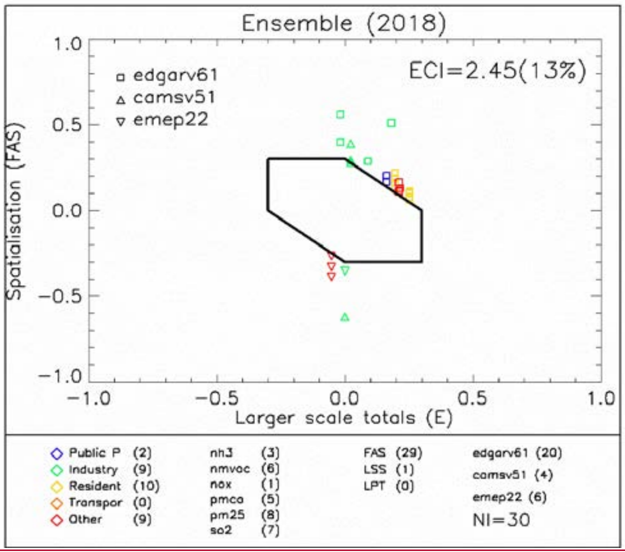


Figure 3: Overview diamonds. The diagram shows the comparison of the three ensemble members (CAMS-REG, EDGAR, EMEP) with the ensemble inventory over Poland. Symbols and colours are as specified in the legend. In all diagrams, only inconsistencies are displayed.

~~Figure 5 displays a zoom of Figure 1 over Poland, focusing on Europe-wide inventories only. Inconsistencies (Figure 5 top left) occur for about 13% of the relevant [p,s] points, with a maximum inconsistency (ECI) 2.5 times larger than the assumed level of uncertainty. As seen from the overview table, most of the issues are related to the EDGAR (20) and EMEP (6) inventories, in particular to the "residential" sector for EDGAR (Figure 5 top right), to the industry sector for CAMS-REG (Figure 5 bottom left) and to the "other" sector for EMEP (Figure 5 bottom right).~~

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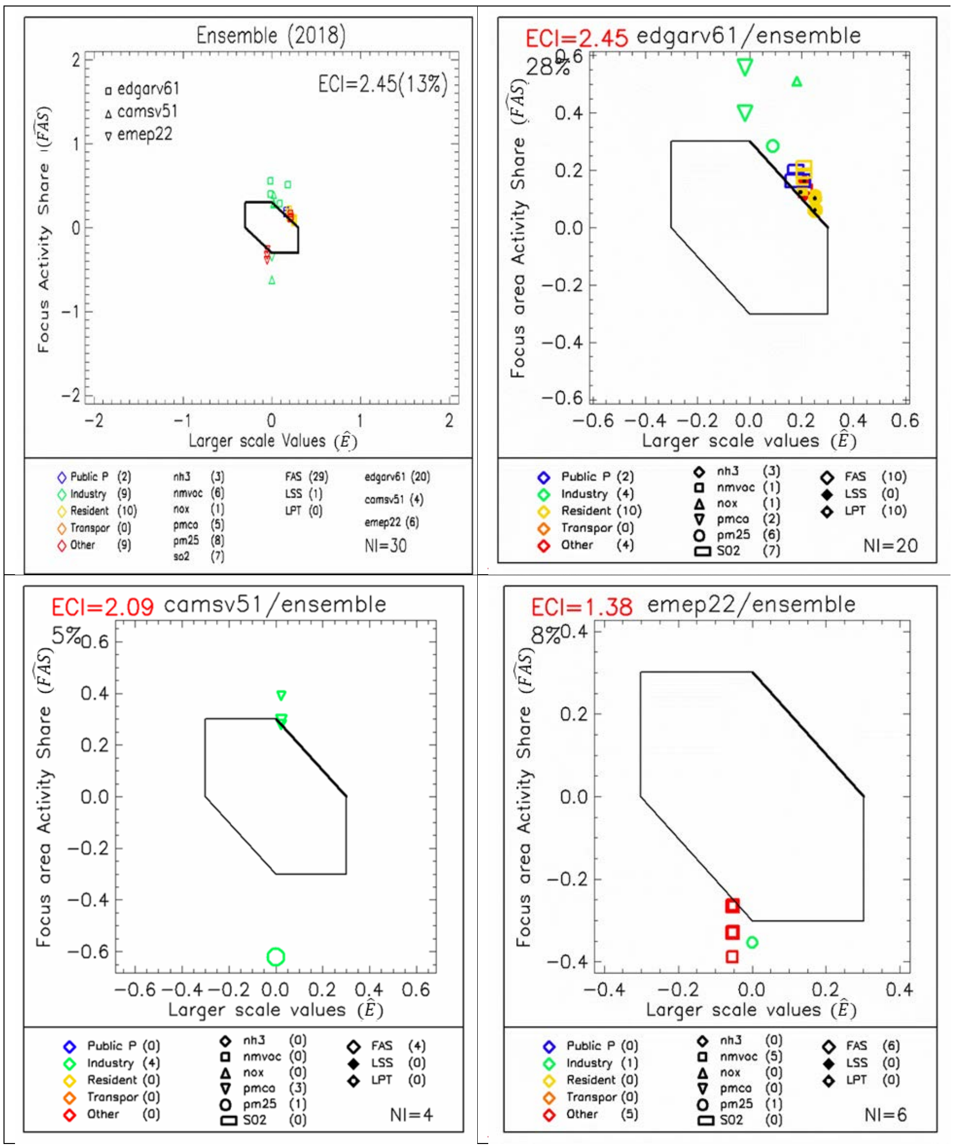
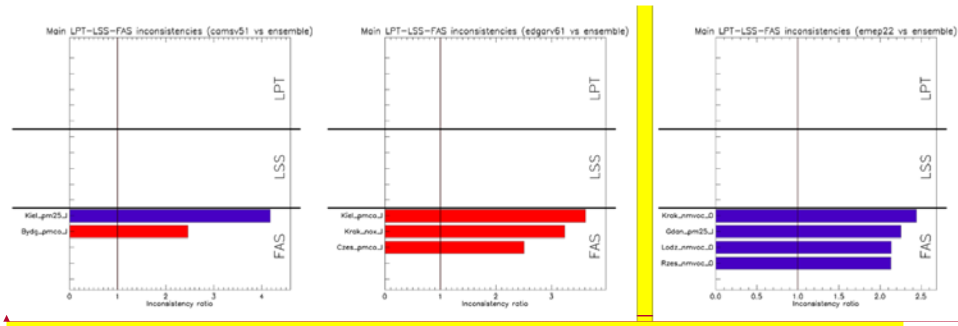


~~*Figure 5: Overview diamonds. The top-left diagram shows the comparison of the three ensemble components (CAMS-REG, EDGAR, EMEP) with the ensemble inventory over Poland. The three following pictures isolate the bilateral comparison of each ensemble component with the Ensemble. Symbols and colours are as specified in the legend. Please note that these symbols/colors differ for the top-left panel, compared to the three others. In all diagrams, only inconsistencies are displayed.*~~



~~Figure 6: Major inconsistencies (up to 5 per category) for LPT, LSS and FAS for CAMS-REG (left), EDGAR (middle) and EMEP (right). Red and blue shadings indicate an overestimation or underestimation of the individual inventory with respect to the ensemble, respectively.~~

~~For EDGAR, while Figure 5 indicates a comparable share between country and urban scale inconsistencies, these country inconsistencies appear because the sum of LPT and LSS is larger than the threshold of 2 while their individual values remain below this threshold. This is why no country scale issues appear in Figure 6. The largest (factor 3) urban scale issues (FAS) are identified for the industrial sector for PMco in Kielce and Czestochowa and for NOx in Krakow. Gridded data for PMco/Kielce are shown in Figure 4 of the supplementary material. While the industrial locations are quite similar with those of EMEP and CAMS-REG, EDGAR emission estimates are much larger in the case of EDGAR.~~

~~For EMEP, inconsistencies are all related to the urban share of the emissions (FAS) with factors slightly larger than 2 for the “other” sector NMVOC emissions in the cities of Krakow (Figure 5 supplementary material), Lodz and Rzeszow and for the PM2.5 industry emissions in Gdansk. Here again, the localization of the main emission sources is similar with EDGAR and CAMS-REG, the EMEP estimates are significantly lower.~~

~~Similar to EMEP, all inconsistencies in CAMS-REG are related with the spatial share of emissions. The largest inconsistencies occur for industrial emissions of PM2.5 in Kielce (Figure 6 supplementary material) and of PMco in Bydgoszcz. In both cases, CAMS-REG distributes its emissions over more locations with a higher intensity.~~

~~EDGAR also shows different values in the residential sector for PM2.5 at country level. Explanations for such differences are linked with the fact that no emissions are allocated to biomass technologies in EDGAR, and that emission factors for some fuels are very different. For example, the EDGAR emission factor for other bituminous fuel allocated to small boilers is nearly the double of the default values. On the other hand, the values reported for Poland (2020) for both coal and biomass emission factors are well below default values, increasing the difference with the EDGAR estimation. Note that these emission factors have been significantly revised in the Poland 2022 submission, which will be reflected in future EMEP and CAMS-REG inventories.~~

~~Next, we check whether the local inventory flags similar and/or other issues.~~ The overview diamond diagram (~~in~~ Figure 4~~Figure 7~~ - left) shows the comparison of the CED local inventory

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with the ensemble. It ~~displays a comparison of the local (CED) and Europe wide ensemble (CAMS-REG, EDGAR, EMEP) inventories for all relevant sector-pollutant points for all cities in Poland.~~ indicates that o~~O~~ut of the 420 emission ratios being tested, only 73 are associated to relevant emissions among which 49 (i.e. 67%) are identified as inconsistencies. The consistency indicator (ECI) is around 14, indicating that the maximum inconsistency is larger than the assumed level of uncertainty by a factor 14. The summary table (at bottom of the diamond, Figure 4) points to the residential and "other" sectors as the main issues with NMVOC and $PM_{2.5}$ in terms of pollutants. Most inconsistencies originate at country level, ~~in majority~~and mostly ~~in terms~~related to the ~~of~~ country sectorial share.

PM residential emissions are systematically larger in CED than in the ensemble for $PM_{2.5}$, whereas smaller for PM_{co}. This can be partially explained by the inclusion of condensable in CED (not included in EU-wide ensemble). Note that including or not condensable results more than doubles total PM2.5 emissions over Poland due to the importance of residential wood combustion emissions. Note that in this case, the CED inventory likely performs better than the ensemble, highlighting the fact that ensemble estimates are not necessarily more accurate. Despite this, inconsistencies are flagged and paths for improvements are identified.

Relatively less important but yet about a factor between 2 ~~to~~and 5, ~~similar~~ low values occur for ~~the Power plant~~ SO_2 emissions from power-generation sector (blue rectangles, Figure 4) ~~figure 7 left)~~. As n~~N~~one of the three~~3~~ Europe-~~-~~wide inventory shows an inconsistency for th~~ese~~is sector~~s~~/pollutant, this~~s~~ indicat~~ing~~es a general issue between local and ~~all~~ EU-wide inventories. Th~~e~~is ~~difference~~ might be explained by the fact that CED is solely based on NB, supplied directly with users' data, while Europe wide inventories (EMEP) likely include additional emissions as they are based on overall fuel sales. In addition, point source emissions from E-PRTR may be different from point source emissions used in national inventories~~, which is also the case for Poland and may be therefore another source of inconsistency.~~

The transport and industry sectors show the lowest number of inconsistencies, which is observed by few points related to those sectors in the diagram (Figure 4 left). While this is expected for transport which is a diffuse source, this is surprising for the industry as this sector was the main source of inconsistencies at Europe wide level (see Figure 3).

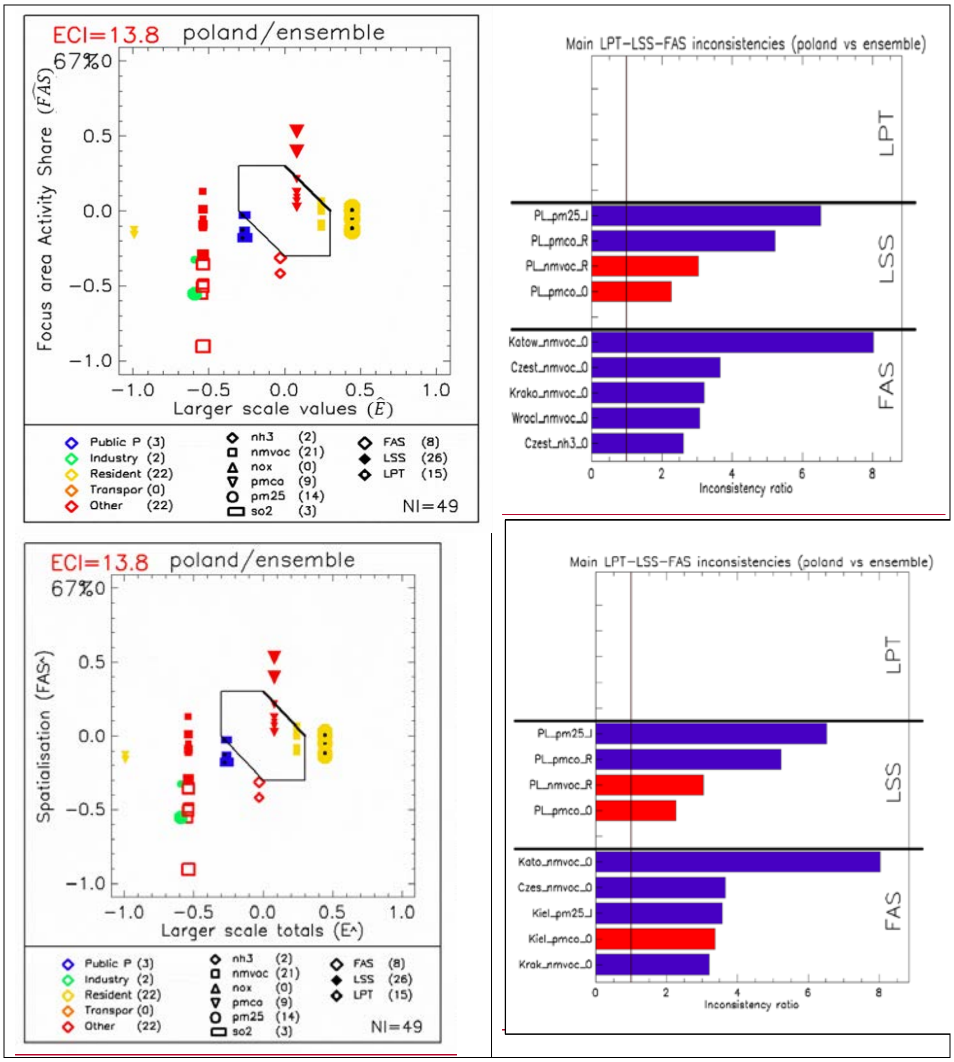

Figure 4: Diamond comparison of the local Polish vs ensemble inventory (left) and comparison of the ensemble top-down ~~components~~members vs the ensemble restricted to the Polish territory.

Figure 4 (right) highlights the priorities for the analysis. At country scale, the largest inconsistency occurs for the industrial share ~~for~~ of PM2.5 (factor 6 larger in the Polish inventory, LSS, Figure 4), ~~for~~ for PMco and NMVOC from the residential sector by a ~~(~~factor 5 lower and ~~factor~~ 3 larger in the Polish inventory, respectively~~)~~ ~~as well~~, as well as for~~for~~ PMco from the other sector (factor 3 lower in the Polish inventory). ~~To support analyses on the country level, we~~

~~present a comparison of EMEP and CED country totals per pollutant for each GNRF sector analyzed as well as some explanations for these differences (Table 2, supplementary material).~~ In the case of PM2.5, the difference can be explained by the fact that the reports provided to NB are based on user-specific permits which specify the list of pollutants to be reported whereas in EU wide inventories, emissions are generally calculated using official EMEP/EEA emission factors.

A comparison of EMEP and CED country totals per pollutant and GNRF sector is available in Table 2 of supplementary material.

~~In the case of NMVOC emissions, EMEP has higher values for all the sectors, with the exception of residential combustion (GNFR C). The issue therefore originates from the sectorial share at country level.~~

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At the local scale (Figure 5~~FAS~~), the spatial allocation of ~~the~~ NMVOC emissions for the other sector leads to important differences in cities like Katowice (factor 8, Figure 4 – right), Czestochowa and Krakow. A similar situation is found for PM in Kielce. We see from Figure 4~~Figure 7~~ that this issue occurs ~~is general~~ for ~~all~~ many cities in the southern part of Poland. The large differences spotted in some cities (e.g. Kielce) ~~for the "other" sector~~ are likely caused by emissions from heaps and excavations. While in CED, emissions from these sources are accounted for, only emissions from brown coal excavations (part of NFR 1B1a) are included in the EMEP inventory. ~~These could explain the identified differences between the local scale and Europe-wide ensemble inventory.~~ Hence, including all heap and excavations emissions in ~~EMEP (and consequently in CAMS-REG)~~EU-wide inventories~~y~~ would be advisable.

~~Relatively less important but yet about a factor 2 to 5, similar low values occur for the Power plant SO_2 emissions (blue rectangles figure 7 left). None of the 3 Europe wide inventory shows an inconsistency for these sectors/pollutants indicating a general issue between local and all EU-wide inventories. The difference might be explained by the fact that CED is solely based on NB, supplied directly with users' data, while Europe wide inventories (EMEP) likely include additional emissions as they are based on overall fuel sales. In addition, point source emissions from E-PRTR may be different from point source emissions used in national inventories, which is also the case for Poland and may be therefore another source of inconsistency.~~

~~Another general issue is related to the PM residential emissions for which the Polish inventory values are systematically larger than the ensemble ones for $PM_{2.5}$ and smaller for PM_{co} which can be partially explained by inclusion of condensable in CED. The EDGAR inventory differs from the ensemble in a similar way and is therefore closer to the CED inventory values. Although the magnitude of this inconsistency is less than previously mentioned ones, the size of the symbols in the diamond diagram (Figure 7 left) indicate that the amount of PM2.5 emission is important for that sector. The difference may be (partially) explained by the fact that the EMEP and CAMS-REG inventories rely on versions of the official reported national inventories from Poland that did not yet consider condensables as part of the PM2.5 emissions from small combustion. In the 2022 submission, this was included and resulted in more than doubling of~~

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~~total PM2.5 emissions from Poland as a whole. This will be included in future versions of CAMS-REG and EMEP.. This is further addressed in the Discussion section.~~

~~The transport and industry sectors show the lowest number of inconsistencies (few points related to those sectors in the diagram). While this is expected for transport which is a diffuse source, this is surprising for the industry as this sector was the main source of inconsistencies at Europe wide level. It is connected with the fact that the Polish EMEP reports, unlike CED, are based not only on data provided by the users of NB.~~

~~The priority inconsistencies for each city are highlighted in Figure 8, and they are mostly related to NMVOC for the "other" sector and PM for the residential sector. This is probably partially a consequence of the processes behind the spatial allocation in the European-wide inventories. While EU-wide inventory compilers distribute country totals obtained from bulk national statistics, population density is often used as a spatial proxy. In this context, the resolution-free design of CED inventory might be a paradoxical limitation here since the exact geographical location of emission sources is prioritized, and some activities are very tough to allocate. For example, coating applications (2D3d) which are responsible for >63 Mg of NMVOC emissions (2018) in Poland, might be omitted in CED due to a lack of reliable spatial data in case they are not provided by NB users in full. Yet another issue is that this pollutant is not being monitored *in-situ* in Poland (and many other countries), which also hampers the interpretation of emission data.~~

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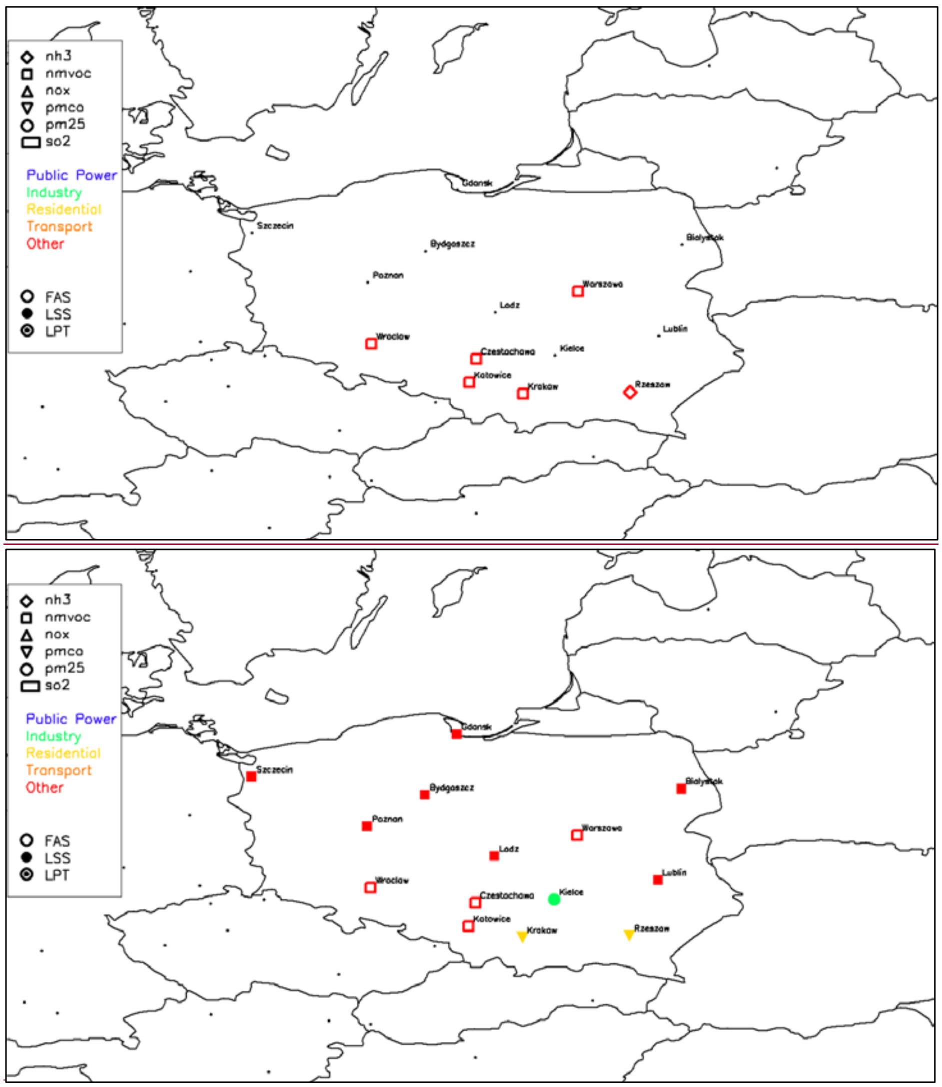

Figure 5: overview of inconsistencies for the comparison between local emission inventory in Poland and the Europe wide emission inventory ensemble

In conclusion, the comparison of the Polish inventory with the ensemble mostly spots issues that are related to a difference in terms of sectorial share at country level, explained by the accounting of different sources in the two types of inventories. A similar argumentation can explain part of the large discrepancies observed in some cities. Most of the issues occur for the residential and ~~“~~"other~~”~~" sectors and mostly for PM and NMVOC. Although the number of inconsistencies may seem large, many of these are similar for all cities.

~~Another reason likely to explain why spatialisation i~~Inconsistencies in the spatial distribution of the emissions are relatively minor. This is ~~is related~~due to the fact that EMEP reports for Poland, used in two out of three EU-wide inventories in the ensemble, ~~,~~are gridded by Polish experts, utilizing spatial proxies based on CED activity data for several sectors like~~. This is the case in particular for~~ stationary combustion ~~(GNFR C)~~, road transport ~~(GNFR F),~~ and livestock ~~(GNFR K).~~ (~~The~~ last updated ~~was done~~ in 2021, ~~(~~Bebkiewicz et al. 2022).

5. Added value and limitations of the ensemble approach~~Discussion~~

European wide inventories are not totally independent of each other. Interlinkages between the CAMS-REG, EDGAR and EMEP inventories ~~have consequences for the comparison~~ exist. For example, the link between EMEP ~~is linked to~~and CAMS-REG is ~~in~~ that (1) ~~it (1)~~ both inventories rely on country reported data and ~~(2)~~ may use the same spatial proxies ~~in case~~when country do not report ~~or the quality of the reported data is poor~~. EMEP is also linked to EDGAR as it uses in some cases EDGAR distribution as a proxy for gridding in case a Party is not reporting ~~or the quality of the reported data is poor~~ (CEIP2022). Consequently, these interlinkages hide some of the inconsistencies, when all inventories behave similarly. It is however expected that repeated screenings lead to improvements and to a progressive convergence among inventories, hence reducing the number of flagged inconsistencies.

In our work, the number of members of the ensemble is limited to three. This would be an issue if the goal were to obtain more accurate and robust results with the ensemble. In such a case, the more members, the more robust the results of the ensemble. Our goal is however different and consists in creating a benchmark for comparison. Rather than looking at absolute values, we assess differences (between an inventory and the ensemble), for which the accuracy and robustness of the absolute values is of secondary importance.

~~Part of the inconsistencies regarding Europe wide inventories are related to inconsistent values at country scale. The comparison of EU-wide inventories highlights an important number of large inconsistencies at country scale between EDGAR and the other two inventories: CAMS-REG and EMEP. While the two latter use (but to different extents) officially reported emissions and therefore rely on similar total emissions per country, EDGAR estimates emissions in an independent manner, starting from activity levels and emissions factors from international agencies and bodies (Crippa et al., 2018, Oreggioni et al. 2022). While this difference in approach can explain a large number of inconsistencies identified for EDGAR, some of them are very large, especially for SO_2 and PM in the industrial sector. For this particular sector, estimates mostly come from the LPS and E-PRTR databases in EMEP/CAMS-REG, which emissions are mostly based on measurements or facility-level estimates. Such information is not used in EDGAR where estimates are based on fuel consumption and emission factors that are very general and not plant specific. The screening analysis allowed identifying some of the causes behind these differences (e.g. outdated sources and/or emission factors) that need to be improved in EDGAR.~~

~~EU-wide, spatial inconsistencies mostly occur for the industry and “other” sectors. Inconsistencies associated with EMEP and CAMS-REG mostly appear for the “other” and industry sectors, mainly pointing to issues related to spatialisation, i.e. to urban activity shares. The fact that the largest inconsistencies are found for sectors where point sources play a major role was expected. Indeed, while a diffuse sector like transport may be distributed quite differently, outliers would not appear as strongly as for point sources. A likely explanation for the differences in SO_2 emissions is that their attribution to point sources is done only for those included in point source reporting (E-PRTR). Smaller sources which are below the threshold for E-PRTR reporting are distributed as diffuse sources to industrial zones (land cover class). This may lead to over-allocation in some urban areas.~~

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~~Local and EU-wide inventories are based on different emission estimation methodologies that lead to inconsistencies in terms of sectorial share at country level. The reasons for inconsistencies between local and European-wide inventories lays in different emission estimation methodologies dictated by the primary purpose of these inventories. Based on statistical data, commonly available in many countries, European-wide inventories rely on general downscaled procedures to spatialize emissions, procedures that put a limit on the final spatial resolution that can be reached for the inventory. On the contrary, local inventories like CED are based on a bottom-up processes where the location and details of each source are known. While we would therefore intuitively expect differences between local and European-wide inventories to be driven mainly by spatialisation aspects, this is not always the case in our analysis. Inconsistencies indeed relate mostly to differences in country sectorial shares that result from different sectors/activities being accounted for in the two types of inventories. This is particularly true for sectors like residential, industry or “others”. As a result, for industry (GNFR B), significant differences are noted for NMVOC, PM_{10} and $PM_{2.5}$ (Table Supplementary material 2). For the residential sector, the main issue with European-wide inventories is the use of a generic approach for spatialisation over Europe, that neglects national and most important subnational differences in the fuel energy mix. This is better captured in CED because of the proxies that are based on local knowledge (see details in Gawuc et al., 2021). In the case of NMVOC in GNFR C, there are two possible reasons behind higher values in CED than the ensemble. First, the larger share of coal in fuel mixes. Second, the higher values in emission factors used in CED (see Table 2 in Gawuc et al. 2021).~~

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~~Another reason likely to explain why spatialisation inconsistencies are minor is related to the fact that EMEP reports for Poland, are gridded by Polish experts, utilizing spatial proxies based on CED activity data for several sectors. This is the case in particular for stationary combustion (GNFR C), road transport (GNFR F), and livestock (GNFR K). The last update was done in 2021 (Bebkiewicz et al.2022).~~

~~Many possible reasons for differences between local and Europe-wide inventories exist. In the case of Poland, another possible source of inconsistencies between European-wide and local Polish inventory is a consequence of how the Polish NB operates and under what rules. Any given “user of the environment” is obliged to report emissions caused by a specific industrial/chemical process for which his/hers “permit to use the environment” is issued. The pollutants and GHG list that must be reported to NB differs among chemical/industrial processes~~

~~altering "users of environment" obligations. Emission from NB data is not taken into account by the Polish National Statistical Office directly and the primary source of Europe wide inventories activity data relies on national statistics. Furthermore, while the Polish EMEP reports are partially based on NB and partially on original methodology (additional emission values) causing disagreements with NB, CED directly adopts emission values reported to NB without additional changes. This issue will be further investigated among CED and Polish EMEP compilers.~~

~~Yet another issue is that in the case of specific installations registered in NB, reports might be based on direct stack measurements or actual condition of installations while the top-down approach accounts only for general resources/fuel consumption. The advantage of NB over top-down approaches is its sensitivity to temporal variability since reporting users are aware of any changes in fuel or other resources quality they consume, rapid changes in production volumes, new technologies used, newly mounted stack filters, etc. Those small changes might not be captured in full in bulk national statistics, commonly based on fuel sales. Finally, it must be commented that in the case of NB, the possible accidental "human factor" might be a source of additional errors since reports are done manually via the online system. Despite some automatic checking algorithms and manual expert evaluation, discrepancies are possible.~~

~~Finally it is interesting to note that the comparison between local and European-wide inventories lead to additional inconsistencies than when the comparison is limited to Europe wide inventories.~~

~~Uncertainties related to the screening methodology.~~ As emission inventories are characterized by different grid resolution and sector aggregations, ~~some~~ harmonization is required prior to the screening process for a meaningful comparison. Conversion to a common grid resolution might result in point sources shifted by one grid cell and be in the urban area in one inventory and not in another, although having the same geographical coordinates in both inventory. However, ~~the~~ city specific diamond diagrams can be used to check if this issue occurs.

While it is more effective for inventory teams to meet and compare approaches in detail to understand and correct differences between inventories, this can be challenging at times, especially in the absence of a specific project to support the work. It must however be noted, that in many instances the reporting of an inconsistency, especially when it is very large, leads to a generally straightforward identification of the underlying cause without requiring too detailed information regarding the inventories.

The settings used in this work, e.g. the choice of 150 urban areas or the way sectors are aggregated are arbitrarily fixed. The method allows for flexible choices and could be applied to other areas than urban (e.g. complex industrial areas or intensive agriculture land) to assess the consistency with respect to other types of emissions. In terms of sectors, a further disaggregation of the other sector will be performed in future to better understand where inconsistencies originate from.

~~In this sense, the limited number of inventories (3) to create the ensemble is not a real issue in this work although it should be kept in mind when analysing the details.~~

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6. Conclusions

The ~~-~~approach presented in this work ~~is intended as~~ supports the~~a~~ screening ~~tool to~~and flagging of inconsistencies among inventories, through the construction of an ensemble benchmark~~and support the assessment of methods to estimate and spatially distribute emissions~~. ~~Only differences that are above a user-defined threshold are detected while smaller differences are disregarded. This threshold reflects the limit between uncertainties and small errors on one side, for which no emission inventory can be estimated to be the best because true emissions are unknown, and larger differences on the other side for which we know that at least one inventory has an error. Given the magnitude of the difference, in most cases this error is likely easy to identify and that improvement in one or both inventories can be made, despite no real value is known.~~ This ensemble

~~In this work, we~~is created ~~an ensemble inventory (median) with the aim of~~to monitor~~ing~~ the status and progress made with the development of Europe-wide inventories, but also ~~.~~ ~~Introducing an ensemble also~~to facilitate ~~allows~~ the comparison~~ng~~ among ~~many~~ inventories ~~at the same time~~ in a relatively simple manner. ~~and foster the interactions between emission inventory developers around the identified inconsistencies. In contrast with other fields of applications (e.g. air quality forecast), this emission ensemble is however not necessarily better than any of its components. While it is not the more accurate inventory, it serves here as a common benchmark for the screening.~~
~~In this sense, the limited number of inventories (3) to create the ensemble is not a real issue in this work although it should be kept in mind when analysing the details.~~
The analysis of the EU-wide ensemble and the comparison with its individual ~~components~~members highlighted a large number of inconsistencies. While two out of the ~~of the~~ three inventories constituting the ensemble behave more closely to each other (CAMS-REG and EMEP), ~~as to a large extent both inventories use emissions submitted to the CLRTAP as input data,~~ they yet show inconsistencies in terms of the spatial distribution of emissions. ~~While t~~The origin of some differences between these inventories and EDGAR can be identified but~~,~~ their magnitude remains to be explained. These differences mostly occur for SO_2, PM and NMVOC, for the industrial and residential sectors, and reach a factor 10 in some instances. The results of the screening ~~results~~ provided useful information that allowed identifying necessary improvements on the estimation of air pollutants emissions, in particular for EDGAR, with the PM emissions from the small-scale combustion sector and SO_2 from the industry and power plant sectors. S

patial inconsistencies mostly occur for the industry and other sectors. The fact that the largest inconsistencies are found for sectors where point sources play a major role was expected. Indeed, while a diffuse sector like transport may be distributed quite differently, outliers would not appear as strongly as for point sources.

The application of the ensemble-screening approach to~~comparison with~~ the local inventory for Poland leads to identifying another type of inconsistencies. While we would intuitively expect differences between local and European-wide inventories to be driven mainly by the spatial distribution of the emissions, this is not always the case in our analysis. Inconsistencies indeed relate mostly to differences in country sectorial shares that result from different sectors/activities

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being accounted for in the two types of inventories. ~~. While one of the main differences between pan-European and local inventories lies in the way emissions are spatialized, the identified inconsistencies do not relate to this spatialisation process but are rather associated to the sectorial share at country level.~~ This~~ese~~ can ~~be also~~be explained by the fact that ~~there are different sources of data to calculate emission in local inventory than in the European ones. In local inventory~~ some emission sources are omitted in the local inventory due to lack of ~~the~~ appropriate geographically allocated activity data.~~, whereas are available on country level e.g. industrial production. The screening~~We identified ~~some~~ sectors and pollutants for which discussion between local and EU-wide emission compilers would be needed in order to reduce the magnitude of the observed differences (e.g. in the residential and industrial sectors mostly for NMVOC, PM2.5 and PM10).

It is also interesting to note that the comparison at local and European-wide scale lead to different types of inconsistencies. ~~The latter point is key. While it is more effective for inventory teams to meet and compare approaches in detail to understand and correct differences between inventories, this can be challenging at times, especially in the absence of a specific project to support the work. It must however be noted, that in many instances the reporting of an inconsistency, especially when it is very large, leads to a generally straightforward identification of the underlying cause without requiring too detailed information regarding the inventories.~~

~~The settings used in this work, e.g. the choice of 150 urban areas or the way sectors are aggregated are arbitrarily fixed. The methods allows for flexible choices and could be applied to other areas than urban (e.g. high emission industrial or intensive agriculture areas) to assess the consistency with respect to other types of emissions. In terms of sectors, a further disaggregation of the "other" sector will be performed in future to better understand where inconsistencies occur.~~

~~The ensemble is not meant to be a static entity. It will evolve as inconsistencies are progressively discussed and solved. An ensemble is therefore associated with reference inventory versions as well as with a reference year. The ECI and other statistics are provided to monitor progress and point to potential improvements. In this sense the ensemble represents a useful tool to motivate the community around a single common benchmark and monitor progress towards the improvement of regional and locally developed emission inventories. It also ensures that improvements become permanent, as forgotten improvements would indeed be flagged again by the system.~~

While the comparison to one local inventory is presented in this work ~~for~~ as an example, these comparisons can be systematized to improve the quality of the ensemble.

The ensemble is not meant to be a static entity. It will evolve as inconsistencies are progressively discussed and solved and emission inventories get improved. The ensemble is therefore associated with reference inventory versions as well as with a reference year. In this sense the ensemble represents a useful tool to motivate the community around a single common benchmark and monitor progress towards the improvement of regional and locally developed

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Table of abbreviations

CAMS-REG	Copernicus Atmospheric Monitoring Services - Regional
CED	Central Emission Database
CEIP	Centre of Emission Inventory and Projection
CLRTAP	Convention on Long-range Transboundary Air Pollution
CO	Carbon Oxides
ECI	Emission Consistency Indicator
EEA	European Environment Agency
E-PTR	European Pollutant Release and Transfer Register
EU	European Union
FAS	Focus Area Share
FUA	Functional Urban Area
GHG	GreenHouse Gases
GNFR	Gridded Nomenclature For Reporting
GPS	Global Positioning System
IIR	Informative Inventory Report
IPCC – AR6	Intergovernmental Panel on Climate Change - Sixth Assessment Report
LPT	Large-scale Pollutant totals
LSS	Large-scale Sectorial Share
NMVOC	Non-Methane Volatile Organic Carbons
NFR	Nomenclature For Reporting
NH3	Ammonia
NOX	Nitrogen Oxides
OECD	Organisation for Economic Co-operation and Development
NB	National dataBase
PM	Particulate matter
PM2.5	Particulate matter with diameter less than 2.5 µm
PM10	Particulate matter with diameter less than 10 µm
SNAP	Selected Nomenclature for Air Pollution
SO2	Sulfur Oxides
UNECE	United Nations Economic Commission for Europe
UNEP	United Nations Environment Program

Code and data availability.
Supporting data and source code are available at: “Philippe Thunis. (2023). Supporting data for the publication "Emission ensemble approach to improve the development of multi-scale emission inventories" [Data set]. Zenodo. https://doi.org/10.5281/zenodo.7940402”
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Author contributions.
PT and AC contributed to the study conception and design. Material preparation, data collection and analysis were performed by PT, EP, ADM, JK, MB, LG, KS, and AC. All authors reviewed the manuscript. All authors read and approved the final manuscript.

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Competing interests. The authors declare that they have no conflict of interest.

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