Response to reviewer comments

We thank reviewers for the critical comments and helpful suggestions. We have taken all these comments and suggestions into account, and they have improved our manuscript considerably. A point-by-point response to reviewers’ comments please found as below.

Responses to Referee 1:

The paper proposes an analysis of the average concentration of CO2, and its growth rate, by comparing several observation networks, and time series filtering. A comparison is also proposed with CO2 concentrations simulated by an atmospheric transport model, after a phase of assimilation of surface observations. A comparison is also made between the average concentration obtained from the surface observation network and the total amount of CO2 in the atmosphere, as simulated by the model after assimilation. Estimating these values is of course important for monitoring atmospheric radiative forcing, both for scientists and policy makers. The study therefore deserves to be published after adding few points for discussion, as detailed below. Overall, the paper is clearly written, but sometimes lacks precision and quantitative values. I think that certain recommendations, such as not extrapolating measurement series, deserve to be more clearly stipulated.

Response: We appreciate the reviewer for commenting on this. We have discussed the extrapolation done in past studies and highlighted the assumption made when extrapolating the measurement period, which could introduce uncertainty in calculating global CO2 concentrations and its growth rate, in the discussion section (Lines 385-397). We have also quantified the difference between the WDCGG method with and without extrapolation, which is about 0.096 ppm in the global CO2 mole fraction, although the extrapolation has a minor effect on the growth rate.

I am concerned by the fact that, every year, slightly different global CO2 estimates emerge from several networks, as detailed in this study. Even if these values differ only slightly (as the results of this study show), I think it’s still a not great to multiply these slightly different estimates. Wouldn’t it be possible to make a recommendation to set up a global reference network to calculate unanimously accepted values? By the way, in addition to estimates based on surface networks, every year we now also see values from networks measuring total columns measured from the ground (TCCON) or from space. This aspect is not discussed at all, but it is conceivable that these measurements could provide a more relevant assessment of the atmospheric global average. Could the model not be used to estimate this contribution from the total column measurements?

Response: We appreciate the reviewer for suggesting this.

The World Meteorological Organization has started a new initiative (Global Greenhouse Gas Watch, GGGW) that aims at establishing a reference network, that will be built on the high-quality observations already performed under the WMO GAW Program that follow consistent good practices and standards. GGGW will fill in critical geographical gaps in such a system. Setting up a global reference network for monitoring global CO2 mole fraction and its growth rate is a complex and challenging task and appropriate network design will be developed as a part of the planned WMO GGGW activities. This study does not aim to provide a solution for
the global reference network, but instead it highlights the importance of station selection and analysis methodologies in monitoring global CO₂ mole fraction and its growth rate.

Based on the existing networks, we recommend the WMO GAW network, which includes not only the background stations (most of NOAA MBL stations) but also stations monitoring CO₂ sink and source from terrestrial ecosystems. We acknowledge that the current monitoring networks have uncertainties, and there is room for improvement in their design. This includes improving geographical coverage to fill observational gaps, ensuring standardization and calibration for consistency across the network, and maintaining long-term operation to increase the number of long-term running stations. Such ongoing efforts by the WMO GAW program are aligned with this goal. Additionally, the WMO has started a new initiative (Global Greenhouse Gas Watch, GGGW) that aims at establishing a reference network.

We have revised our manuscript to make this point clearer (Lines 440-446): “The high-quality observations conducted by the WMO GAW network include not only background stations (most of NOAA MBL stations) but also continental stations. This comprehensive network enables proper global average calculation. Furthermore, the WMO has initiated a new program, Global Greenhouse Gas Watch (GGGW), with the aim of establishing a reference network. This network will be built on the high-quality observations already performed under the WMO GAW program that follows consistent good practices and standards. Although the current monitoring networks have limitations in terms of geographical coverage, data consistency, and long-term measurements, they are well-equipped and have the capacity to effectively represent global surface CO₂ mole fraction and its growth rate and trends in atmospheric CO₂ mass changes.”

It is well-known that the column-averaged dry air mole fraction of CO₂ (xCO₂) and surface CO₂ represent completely different volumes of the atmosphere and offer different coverages. Typically, xCO₂ is lower than surface CO₂, but no study has quantified this difference on a global scale. Although xCO₂ is not directly related to the topic of this paper, it is indeed interesting to quantify and observe this difference. Therefore, in response to the reviewer’s suggestion, we conducted an analysis of TCCON and GOSAT data, comparing them to surface CO₂. However, please note that the analysis and results of comparing surface CO₂ and xCO₂ are not included in this paper, as we think it is beyond the scope of this study. To make this comparison meaningful, a wholly different study of the representative volumes would be required.

According to our analysis (see Fig. R1 and R2), we have found that xCO₂ is consistently lower than surface CO₂. Global TCCON (GFIT) xCO₂ is 0.80-1.64 ppm lower than surface CO₂ (i.e. data from GAW, NOAA, CTE), which is expected because the emissions originating from the surface are mixed with air with lower CO₂ concentration in higher levels of the troposphere. GOSAT (GFIT) xCO₂ has an even lower mole fraction (1.32-2.15 ppm lower), likely due to GOSAT covering a much wider area (flying between 60S and 60N) over both land and sea, and its signal goes through the atmosphere twice. In addition, the seasonality of xCO₂ and surface CO₂ are similar (r>0.99), but the seasonality changes (or amplitude) of xCO₂ are nearly half of those observed in surface CO₂. The xCO₂ growth rate closely matches that of surface CO₂, especially in case of TCCON (r=0.82-0.93, RMSE=0.16-0.26, ME=-0.01-0.06), indicating that changes in column xCO₂ can be well represented by surface observations.
Figure R1. Comparison of globally and locally averaged CO₂ mole fraction (a) and its $G_{ATM}$ (b) from 1980 to 2020. Panel (a) shows the global monthly CO₂ mole fraction from 139 GAW sites (estimated from observations only), 43 NOAA MBL sites and those from 230 sites used in CTE (either from observations or model output). The two $xCO₂$ mole fractions are from TCCON (pink line) and GOSAT (brown line) and analysed using the GFIT method. The red and blue lines show the CO₂ derived from GAW (GFIT) and GAW (WDCGG), respectively. The green and orange lines show the CO₂ derived from CTE_obs (GFIT) and CTE_output (GFIT), respectively. The right y-axis shows their difference from NOAA CO₂ mole fraction, and the dashed lines show the mean of the difference over the available period. Panel (b) compares the corresponding global and local CO₂ growth rate, the legend refers to panel (a). The shadow area shows the uncertainty as 68% confidence interval obtained by the bootstrap analysis.
Figure R2. Pair-wise statistical metrics assess the agreement of monthly global CO$_2$ and xCO$_2$ mole fraction (ppm) and its G$_{ATM}$ (ppm yr$^{-1}$) across various networks and methodologies (see Table 1 and Fig. 4) for the period 2010-2020. Panel (a) presents the Mean Error (ME) quantifying the difference for each pair, focusing on CO$_2$ mole fraction, while panel (b) does the same for G$_{ATM}$. The significance levels of paired t-test for ME are indicated as follows: * p<0.1, ** p<0.05, *** p<0.01. Panel (c) and (d) present the Root Mean Squared Error (RMSE) for CO$_2$ mole fraction and G$_{ATM}$, respectively. Panel (e) and (f) present the Pearson Correlation Coefficient (r) for CO$_2$ mole fraction and G$_{ATM}$, respectively.
In many cases, reference is made to the CO2 concentration at Mauna Loa as a proxy for global CO2. The advantage of relying on 1 or 2 stations (MLO and SPO, for example) is that it avoids the problems of changing the configuration of the global network, and enables a fast calculation. The disadvantage is that you are stuck if the reference station fails. Having said that, I would have been interested to see a comparison of average concentrations and growth rates considering only these 2 stations.

Response: We appreciate the reviewer for this suggestion. We have added data from these stations (MLO and SPO) to our analysis (see Figure 4 below and in the paper). The data from these stations are derived from the WMO GAW and analyzed using the GFIT method. We found that the local CO2 mole fraction from MLO and SPO clearly differs from the globally averaged values, and their seasonal cycles differ in timing, amplitude, and direction. Even so, the local and global growth rates behave similarly over the long-term period. However, there are evident monthly differences between the local and global growth rates, along with time shifts.

We have revised our manuscript to make this point clearer:

(Lines 245-252): “A common approach to estimate global surface CO2 mole fraction is by using one or two representative sites, such as Mauna Loa (MLO) and South Pole (SPO). The globally averaged monthly surface CO2 mole fractions, derived from the GAW, CTE, and NOAA networks, are significantly (p<0.05) lower by 0.46-0.88 ppm during 1980-2020 (Fig. S1a) and 0.45-1.19 during 2001-2020 (Fig. 5a) than the local CO2 estimates solely based on MLO measurements. Conversely, these global monthly CO2 mole fractions are significantly (p<0.05) higher by 1.91-2.24 ppm during 1980-2020 (Fig. S1a) and 2.21-2.94 during 2001-2020 (Fig. 5a) when compared to local measurements at SPO site. Furthermore, the global seasonal cycle leads the local cycle at MLO by approximately one month (estimated by averaging the time difference between the peaks of their seasonal cycles). In contrast, the local cycle at SPO is not evident and is opposite to the global seasonal cycle (Fig. 4a).”

(Lines 271-274): “Furthermore, over the long-term period of 40 years, the estimated local growth rate at MLO (ME<0.046 ppm yr\(^{-1}\) higher, RMSE<0.272 ppm yr\(^{-1}\), r>0.915) and SPO (ME<0.049 ppm yr\(^{-1}\) lower, RMSE<0.305 ppm yr\(^{-1}\), r>0.888) behaves similarly to the G\(_{\text{ATM}}\) derived from GAW, CTE and NOAA network (Fig. 4b and S1). However, noticeable monthly differences between the local and global growth rates, deviating up to approximately 0.8 ppm yr\(^{-1}\), and time shifts are observed (Fig. 4b).”

(Lines 407-414): “The local growth rate at MLO and SPO generally behaves similarly to the global growth rate derived from the GAW, CTE, and NOAA networks (Fig. 4b and S1). However, the local CO2 mole fraction and its seasonal cycle noticeably differ from global estimates derived from different observational networks. In this regard, the utilization of individual sites for the evaluation of the global average mole fraction and its growth rate is not precise and can only be used for illustration rather than as a substitute for the proper global average calculation. The local observation sites, often situated away from significant local sources and sinks, such as MLO, provide long-term and high-quality data, serving as reference data for global CO2 mole fraction. However, a single observation site cannot capture the CO2
spatial variability, transport, and mixing. To overcome these limitations, global CO₂ trends and variations are best assessed by integrating data from multiple sources and locations.”

Figure 4. Comparison of globally and locally averaged CO₂ mole fraction (a) and its GATM (b) from 1980 to 2020. Panel (a) shows the global monthly CO₂ mole fraction from 139 GAW sites (estimated from observations only), 43 NOAA MBL sites and those from 230 sites used in CTE (either from observations or model output). The two local CO₂ mole fractions are from Mauna Loa (MLO, cyan line) and South Pole (SPO, magenta line) stations, analysed using the GFIT method. The red and blue lines show the CO₂ derived from GAW (GFIT) and GAW (WDCGG), respectively. The green and orange lines show the CO₂ derived from CTE_obs (GFIT) and CTE_output (GFIT), respectively. The right y-axis shows their difference from NOAA CO₂ mole fraction, and the dashed lines show the mean of the difference over the available period. Panel (b) compares the corresponding global and local CO₂ growth rate, the legend refers to panel (a). The shadow area shows the uncertainty as 68% confidence interval obtained by the bootstrap analysis.

Figure 4b shows a maximum divergence of methods over the last few months of 2020, which is confirmed by Table 1, where the U(Gatm) uncertainty in 2020 is about 3 times greater than in previous years. This bias is important in view of the high demand for these estimates in near-real time. You mention the problem in the discussion as a result of a side-effect of the filtering procedures. Could you propose alternative to reduce this side effect?

Response: We appreciate the reviewer for commenting on this. We acknowledge the limitation of end-effect in the GFIT and WDCGG methods. The end-effect not only affects the last few
months but also has an influence on the first few months. For instance, there is a big shift at the beginning of 2014 in TCCON xCO₂ growth (Fig. R1b), which is due to two new stations joining TCCON in 2014. This implies that caution is warranted when using short-run or recently joined stations for global analysis when using the current methods (e.g. WDCGG, NOAA, and GFIT).

We have two ideas for reducing these end-effect:

a. Using machine learning to extrapolate the smoothed trend for ONE year before and after; this extrapolated portion is used exclusively for calculating local mole fraction and growth rate, BUT it is not included in the global or zonal average.

b. Similar as previous idea, the short period extrapolation is derived from bias-correcting the data comes from models (e.g. CTE) using site observations.

Overall, we think that reducing the end-effect is beyond the scope of this study. Perhaps we will consider a short follow-up paper for this.

We have revised our manuscript to make this point clearer (Lines 380-384): “Therefore, Conway et al. (1994) suggested that the growth rate curves for the last 6 months should be viewed with caution. Reducing the end-effect requires further study, such as using machine learning or bias-correction methods to extrapolate the smoothed trend for a short period (e.g. one year) before and after. This extrapolated portion is used exclusively for calculating local mole fraction and growth rate, while it is not included in the global or zonal average, as it could introduce additional uncertainty.”

Data filtering: In addition to station selection, it was not clear for me if you apply a filter on the day/night periods. Mountain stations like Manua Loa are traditionally selected only during the night, while continental stations on the plains are generally selected during the day to increase the representativeness of the time series. In the discussion you mention the higher concentrations when adding continental sites, but clearly the offset will be quite strongly different if you include or not the night time CO₂ accumulation in continental surface stations. Could you elaborate on this aspect?

Response: We appreciate the reviewer for commenting on this. We agree that analyzing the CO₂ data during the day/night period could lead to different results. However, in this study, we are using monthly data for analysis, and in this sense, the CO₂ variation during the day plays a minor role in our analysis.

We have mentioned in the manuscript (Lines 144-148): “This study synchronizes monthly CO₂ records with the fitting and filter method developed at the NOAA Global Monitoring Laboratory (Thoning et al., 1989, Conway et al., 1994), without extrapolation. The station selection and CO₂ averaging method are kept the same as in the WDCGG method (Text S1). This method will be referred to as the GFIT method and will be compared to the WDCGG method without extrapolation.”

Few more specific comments:

Title ‘Global mean surface CO₂’: For temperature measurements the elevation is normalized (e.g 10 m above ground level). This is not the case for CO₂, for which we rather avoid measuring the concentration close to the surface. Consequently, it would be more accurate to refer to the ‘global marine boundary layer CO₂’
Response: By using the term 'Global mean surface CO₂,' we refer to the mean CO₂ concentration within the planetary boundary layer (PBL), which extends from the Earth’s surface up to a few hundred/thousand meters in height. The CO₂ within PBL exchanges with land, vegetation, urban areas, and water. In this study, we do not solely focus on the CO₂ exchange with the marine boundary layer (MBL), which has the stable atmospheric conditions and interacts specifically with the ocean. Many sites in GAW and CTE network are continental (Fig. 1), thus using the term “global MBL CO₂” in this study is not appropriate.

We have revised our manuscript to make this point clearer (Lines 91-92): “The global mean surface CO₂ refers to the mean CO₂ mole fraction within the planetary boundary layer, which extends from the Earth’s surface up to a few hundred or thousand meters in height.”

Lines 47-48: please provide a reference for the conversion GtC yr⁻¹ to ppm yr⁻¹

Response: The number of the growth rate refers to Friedlingstein et al., (2022). The conversion from GtC to ppm is using the factor of 2.124, which refers to Ballantyne et al. (2012).

We have revised our manuscript to highlight this (Lines 50-51): “..., leaving a net increase of 5.0 ± 0.2 GtC yr⁻¹ of CO₂ in the atmosphere, corresponding to an atmospheric CO₂ mole fraction increase of 2.4 ± 0.1 ppm yr⁻¹ (the conversion factor comes from Ballantyne et al. (2012)).”


Line 75: “… hundreds of stations coordinated by WMO GAW: really ?”

Response: We appreciate the reviewer for noting this. Yes, the WMO GAW coordinates the surface-based observational network, which includes GAW Global (31 stations) and Regional (about 470 stations) stations. However, the number of stations currently coordinated by WMO GAW might change over time. Detailed information and the status of GAW stations can be found in the GAW Station Information System (GAWSIS, http://gawsis.meteoswiss.ch). It should be noted though that measurement programs at GAW stations can be very different and just a small percentage of these stations are involved in the measurements of greenhouse gases. The GAW program covers greenhouse gases, reactive gases, stratospheric and total ozone, aerosols, total atmospheric deposition, and solar UV radiation.

We have rephrased the sentence (Lines 78-79): “… that together form a network of hundreds of stations coordinated by WMO GAW [http://gawsis.meteoswiss.ch].”

Line 84: “… i.e. the full troposphere (up to ~8-15 km altitude) and the stratosphere or the regions of the world with substantial observational gaps”

Response: We have rephrased the sentence (Lines 86-88): “However, the two aforementioned approaches neither represent the parts of the atmosphere with low CO₂ mole fraction levels (i.e., the full troposphere, up to ~8-15 km altitude, and the stratosphere), nor do they cover the regions of the world with substantial observational gaps.”
Line 125: “CTE compare well…”: could you more precise?

Response: We have rephrased the sentence (Lines 131-133): “…, and the comparison of CTE CO₂ product to the other data assimilation systems used in GCP shows good agreement (within 0.8 ppm at all latitude bands) (Friedlingstein et al., 2022)”

Figure 4 & Line 249: Same trend with and without continental sites. Figure 4A shows that CTEobs-NOAA differences change quite markedly before and after 2000, which is not the case for GAW-NOAA scenarios. In particular, strong winter differences emerge with the development of continental stations. It's a little disturbing that this fairly clear shift between the two networks isn't reflected in long-term trends. I imagine that the difference would be seen on the trend of annual concentrations, but not when looking at annual growth rates, as this is a transient change. Could you discuss this issue?

Response: We appreciate the reviewer for commenting on this. Yes, we can see the differences emerge with the development of continental stations when we look at the trend of annual concentrations (see Figure S4 below). The slope of the CTE trend (1.859) is clearly higher than that of the NOAA trend (1.832) due to more available continental measurements in the CTE network. The slope of the GAW (1.838) is slightly higher than the NOAA trend because GAW has more continental stations than NOAA but not as many as the CTE network (See Figure 1 in the paper).

The reason for different trends in global CO₂ mole fraction and global CO₂ growth rate is:

The global CO₂ mole fraction trend refers to the long-term change in the overall atmospheric CO₂ level over time, which is influenced by various factors, e.g. fossil fuels burning, deforestation, volcanic activity, and more. On the other hand, the global CO₂ growth rate trend reflects the change in the rate of CO₂ increase in the atmosphere, which is influenced by the balance between CO₂ emissions and uptake (e.g. by terrestrial ecosystem).

We have revised our manuscript to make this point clearer (Lines 275-280): “The trend analysis reveals that with development of continental sites, the slope of the trend of annual global CO₂ mole fraction changes from NOAA network (1.832 ± 0.029 ppm yr⁻¹) to CTE network (1.859 ± 0.029 ppm yr⁻¹) during 1980-2020 (Fig. S4). However, the G_ATM increased steadily at a rate of 0.030 ± 0.002 ppm per year each year from 1980 to 2020 (Fig. 6a), based on the observations from the three networks (i.e. GAW, CTE and NOAA). This implies that over long-term period (here 40 years), the networks with and without continental sites exhibits the same trend of the G_ATM and has little effect on the transient change in the rate of CO₂ increase in the atmosphere.”
Figure S4 shows the trends of global CO$_2$ mole fraction for the GAW network (red line), the CTE network (green line) and the NOAA network (black line) during the whole period 1980-2020. The cycles show the annual CO$_2$ mole fraction, respectively.

Line 253: “red and blue lines”: actually, it is red and green on Figure 5

Response: Thanks, we have corrected it in the manuscript (Lines 285).

Line 256: earlier detection of GAtm change: can you quantify how much earlier?

Response: We appreciate the reviewer for suggesting this. We have estimated the start of CO$_2$ growth rate increase/decrease for the three strong El Niño events (i.e. 1987-1988, 1997-1998 and 2014-2016) (Table S2 below and in the paper), and further quantified the earlier detection of GAtm change in the observation network with more continental stations. We found that the GAW and CTE networks detect the change in CO$_2$ growth rate for the El Niño events approximately 1-2 months earlier (Table S2).

The quantification is done by smoothing the trend of CO$_2$ growth rate using a Gaussian filter (with sigma=1.96, Figure S9 below and in the paper), which aids in finding the local extrema (i.e. the start of CO$_2$ growth rate increase/decrease).
We have rephrased the sentence (Lines 282-288): “During three strong El Niño events, which are marked as grey bands in Fig. 6b, the \( G_{\text{ATM}} \) derived from the GAW and CTE network (red and green lines) begins to increase approximately 1-2 months (Table S2) earlier before the El Niño events (marked as blue circles in Fig. 6b) and reaches its peak approximately 1-2 months (Table S2) earlier during the El Niño events (marked as orange circles in Fig. 6b), compared to the \( G_{\text{ATM}} \) derived from the NOAA network (black line). This suggests that continental sites can aid in the early detection of \( G_{\text{ATM}} \) changes resulting from changes in biogenic emission or uptake.”

Table S2. displays the estimates of CO\(_2\) growth rate increase/decrease dates for the three strong El Niño events (i.e. 1987-1988, 1997-1998 and 2014-2016). These estimates are calculated from the smoothed trend of CO\(_2\) growth rate based on CTE, GAW and NOAA networks (Fig. S9).

<table>
<thead>
<tr>
<th>El Niño 1987-1988</th>
<th>Trough (( G_{\text{ATM}} ) starts increasing)</th>
<th>Peak (( G_{\text{ATM}} ) starts decreasing)</th>
</tr>
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<tbody>
<tr>
<td>Date</td>
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</tr>
<tr>
<td>CTE</td>
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</tr>
<tr>
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<th>El Niño 1997-1998</th>
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<th>Peak (( G_{\text{ATM}} ) starts decreasing)</th>
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<table>
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<tr>
<th>El Niño 2014-2016</th>
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<td>NOAA</td>
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</table>

Figure S9. presents the smoothed trend of CO\(_2\) growth rate for each month during 1980-2020. The trends (depicted in Figure 6b) are smoothed by using a Gaussian filter (with \( \sigma = 1.96 \)). The dots represent the local extrema, which aid in identifying the start of CO\(_2\) growth rate increase/decrease.

Line 306-307: ‘The NOAA network tracks atmospheric CO\(_2\) change better’: I would rather say that the result based on the NOAA network comes closer to the CTE estimate.

Response: We have rephrased the sentence (Lines 349-352): “The CO\(_2\) IAV based on the NOAA network exhibits a slightly closer relationship (\( r = 0.938 \)) with the CTE atmospheric CO\(_2\)
mass estimates than the GAW (r=0.861) and CTE (r=0.812) networks. This finding is consistent with the long atmospheric residence time and well-mixed nature of CO₂ in the NOAA network.”

Lines 405-407: The conclusion need to be rephrased for clarity.

Response: We have rephrased the sentence (Lines 448-452): “Continuous monitoring of atmospheric CO₂, based on the current GAW network together with reliable global data integration methods, provides essential information. This includes understanding trends in atmospheric CO₂ concentration, assessing the impacts of past policies, identifying high-emission areas, informing climate models, forecasting future scenarios, and raising public awareness. Policymakers can rely on this information to support their efforts in mitigating the global warming.”