Anonymous Referee #1
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Overview:
Review of “XCO₂ in an emission hot-spot region: the COCCON Paris campaign 2015” by Vogel et al.

Vogel et al. present an analysis from a 2-week field campaign in Paris using the Collaborative Carbon Column Observing Network (COCCON). This network uses 5 FTIR spectrometers in and around Paris. The authors compare upwind and downwind concentrations from these spectrometers and use the CHIMERE-CAMS model to simulate XCO₂ at these sites. The campaign was hampered by poor meteorology and most of the results are from 4 days of measurements. Given this, the authors are unable to draw any major scientific conclusions but the work is nevertheless a nice demonstration of the viability of this kind of network. My main criticisms are that I feel this work is far too long (14 figures and 3 tables) given that another paper describes the construction of the network (Frey et al., AMTD) and the findings are rather limited given the meteorological limitations for this time period. Overall, I think the work should ultimately be published but could use major revisions to better justify the arguments that are novel. There are also a number of formatting and/or grammatical errors that should be addressed.

We have addressed the specific and general comments as well as addressed formatting issues. The text has been clarified and slightly re-focused and 4 figures and 2 tables were moved to the supplement. Original reviewer comments are given in blue with our response in black.

Comments:
I found myself struggling to characterize what I’m actually learning from the paper because the construction of the network is described in Frey et al. AMTD and previous work from this group has already shown the use of a gradient method in Breon et al. and Stauffer et al. To this reviewer, the major contribution is demonstrating that the gradient method can also work for column measurements and, to a lesser extent, that there is substantial uptake from the biosphere in an urban region like Paris. So I think the message of the paper could be better framed.

We did indeed highlight these aspects in the abstract, but we agree that it felt ‘watered-down’ in the main text of the initial manuscript. The manuscript has been streamlined and we have moved 4 figures and 2 tables into the supplement. Some more specific discussions and clarifications were added instead.

Additionally, the manuscript is actually rather weak in the demonstration that the gradient method is working for this region. For example, wouldn’t wind shear also adversely impact the gradient method? If there is wind shear, you may have low-level winds that satisfy the upwind/downwind conditions but mid-to-upper tropospheric (or stratospheric?? Since it’s a column measurement) winds that bring different background conditions. There is little discussion of this (or argument that it is not important in these cases). Are there radiosonde measurements or radar wind profiler data that could be used to demonstrate this?

Our model is based on ECWMF Integrated Forecast System (IFS) weather data, which has been demonstrated to compare fairly well with radiosonde data. We have added hodographs to show the wind-shear in the domain in the supplement (example also below).
As expected there are wind speed and direction changes in higher layers of the atmosphere. However, the wind conditions in higher layers (mid tropo-sphere and stratosphere) are not of key interest for this study as most of the XCO2 variability is driven by the CO2 mixing ratios in the boundary layer (typically well below 650hPa). The distance between upwind and downwind sites in our experiment is below 45km and there are no indications that significant concentration gradients exist in the stratosphere or even the mid/upper troposphere at this length-scale that would be comparable to the horizontal gradients within the PBL. Even in remote regions without strong urban CO2 emissions vertical profiles are dominated by lower levels (below 650hPa) and
compare well with ECMWF-CAMS, which was used in this study as boundary condition (e.g. Membrive et al. 2017; https://doi.org/10.5194/amt-10-2163-2017)

We have added a discussion of wind-shear and why we have made the assumption that it is not a major influence (compared to other factors) here.

Overall, I think the manuscript would be far more useful if the authors were to move much of the discussion to a supplement and focus on the main findings. For example, many of the figures could be combined or reduced:
Thanks for this suggestion we have streamlined the manuscript.

• Figs. 1 and 3 could be combined
Thank you for the suggestion, we actually did try a combined graph before, but it was not easily readable. We have opted to move figure 3 into the supplement now.

• It’s unclear what Figure 4 is supposed to be telling me, Figure 5 seems to show the same data but in a much clearer form
Figure 4 and 5 are indeed the same data, however, Figure 4 allows comparing how similar/different the XCO2 variability is across different sites (spatial variability), while Figure 5 allows to assess the temporal XCO2 variability at an individual sites.

• Figures 8 and 9 could be combined into a 2-panel figure (would facilitate a visual comparison). However they could probably be moved to a supplement since I’m not sure if they’re really necessary. It seems like Figure 10 does a better job of breaking down the contribution from various components (which is actually rather interesting)
Agreed – we have moved Figure 8 into the supplement.

• Table 3 could be in a supplement or cut since the locations are shown in Fig. 1.
Specific comments:
Agreed – we have moved Table 3 into the supplement.

COCCON is in the title, isn’t defined until page 3.
We have added this information in the abstract.

At the beginning of Section 3.1.2 (Page 9), the authors mention that the standard deviation for 1-minute data is 1 ppm. That seems huge given the changes that they’re seeing. Does this mean the error bars on all their data points are ±1 ppm? I suspect there’s something I am missing because that would make me rather skeptical of the results.
This 1-ppm standard deviation is NOT the instrumental precision, but is driven by the variability of atmospheric conditions (in time and space). I.e. all minute data were used to calculate the mean concentration of the campaign period and their variability is 1ppm. We have added text to clarify the issue.

Page 2, Lines 64–67: Just because one single factor doesn’t explain the variations between cities doesn’t necessarily mean it’s uncertain.
This criticism is not completely clear. We did not claim (and don’t want to) that the uncertainty of emissions is caused by the fact that urban emissions cannot be explainable by a simple factor, but we rather wanted to highlight that cities cannot be generalized and need to be investigated individually. We have added wording to make it more clear and avoid confusion.
Page 2, Lines 70–73: Should give references to these other networks. These networks were referenced in line 74. We have moved the references to clarify and we also added other relevant references.

Page 2, Line 74: Urban measurements are representative of a 10000 km$^2$ region?? I’m rather skeptical of that.
The Ile-de-France region (surrounding Grand Paris) is 12’012km$^2$ and the atmospheric GHG and air pollutant composition in this region is often dominated by Parisian emissions, which are constraint by regional observations (Breon et al. 2015 and Staufer et al. 2016). Furthermore, our measurements are conducted in the outskirts of the urban area to not be influenced by local sources only, but to be sensitive to the larger area. We have added text to clarify.

Page 3, Lines 93–95: The recently funded GeoCARB satellite is a geostationary satellite that will have multiple measurements per day.
Thanks, we have clarified this here.

Page 3, Lines 108–110: Again, this applies to LEO satellites but there are upcoming GEO satellites as well.
We specifically mentioned LEOs in line 107 and GEOs in line 112. Where we highlighted that GEOs are not restricted to the LEO time window.

Page 4, Lines 113: Should add the O'Brien et al. AMT (2016) paper because this satellite is actually funded.
We have added the reference to GEOCARB here. (O’Brien was also cited in line 89 of the original manuscript).

Page 4, Line 137: Would be good to flip the order of “airports” and “industrial” because it looks like AIRPARIF just refers to airports (since it starts with AIR).
We have added an explanation here. AIRPARIF is the air quality association that monitors/manages the airshed of Paris. It is not related to the airports agency of Paris (PARIS AEROPORT).

Page 6, Lines 191–195: Impressive!
Removed “impressive” and we will leave it to the readers to judge the performance – sometimes good results get the (co-)authors excited.

Page 7, Line 231: Missing subscript, should be “CO$_2$. Authors should do a search and replace because there are many instances of incorrect subcripting for CO$_2$.
We have corrected all CO$_2$ and XCO$_2$ and FFCO$_2$ to have correct subscripts.

Page 8, Lines 276–293: This nomenclature is very confusing. There are subscripts and superscripts on many variables and some of the variables have multiple letters (e.g., “COs$_2$-model” is not a great variable name). Would be much better if the authors used standard nomenclature from either Rodgers (2000) or the TCCON group. Either would be preferable to the current.
Although we agree that multiple having sub-scripts and super-scripts in equation are suboptimal when considering readability, this is also the case in TCCON publications, e.g. Kuai et al. 2013 (doi:10.5194/amt-6-63-2013). The nomenclature, here, was chosen to be consistent with previous COCCON publications, but we would definitely encourage that a common nomenclature is established across different remote sensing efforts/networks.
Page 8, Line 283: Why is WACCM bolded? Is it supposed to be a matrix (those are the only other bolded terms).
Corrected

Page 9, Lines 316–317: How are these spectra rated? Unclear
The quality of the data for each day was rated according to the overall data availability and similar to the previously published work by Hase et al. 2015 [www.atmos-meas-tech.net/8/3059/2015]. We have added the reference.

Page 9, Lines 328–329: Upwind is higher concentration?? Probably a typo, I think you meant downwind. . .
Yes, thanks for catching this -> corrected

Page 10, Lines 337–338: Are there no other factors?? That seems surprising. Would wind shear or variations in winds, a decreasing anthropogenic source during the day not be able to give decrease? Needs stronger justification w/ data or citations.
We have corrected this sentence as other factors could contribute. However, our simulation attribute the decrease of CO2 to NEE (uptake) within the domain, but the underlying biospheric fluxes could indeed be wrong.

Page 10, Lines 341–344: This doesn’t seem supported by the analysis. I’d like to see a footprint analysis or some other way for this to be justified. . .
Our modelling framework does not provide footprints and running a lagrangian model in addition to CHIMERE is beyond the scope of this work. We have added citations of other studies where XCO2 column footprints were investigated.

Section 3.13 Page 10: What about wind shear? Were there any radiosondes that indicate the winds are uniform through the column? What about the model? Does that indicate uniform winds throughout the column.
See general comment on wind shear and hodographs

Page 10, Lines 361–365: How representative are the winds at GIF? This could easily be tested in the model, (e.g., look at how variable the winds are over Paris and compare that to the grid cell w/ GIF).
Looking at the hodographs for the lower model levels they seem very consistent across Paris

Page 11, Lines 395–397: Couldn’t you just coarsen the 1km inventory and then do this comparison?
Sorry, this point was not very clear. We cannot compare different anthropogenic emission products in this study as no other ‘inventories’ are available at this resolution. Coarsening IER to the resolution of e.g. EDGAR V4.2 would not still not yield a fair comparison to assess the performance/influence of the spatial disaggregation at the 2x2km2 scale. We have added text to clarify this point.

Page 12, Lines 413–415: How are you directly linking this to NEE? Seems like this needs more justification.
We have added clarification. In our model FFCO2, NEE and BC are transported separately, so we can directly see which source has contributed CO2 to our simulated XCO2 in our domain.

Page 13, Lines 475–476: How is this being assessed? Does the model agree with this (i.e. is the modeled contribution the same at each site)?
Yes, the model predicts similar biogenic contributions and we ASSUME that the influence of rural biogenic fluxes in our domain affect our sites in a similar way. We have added text to clarify.

Page 14, Line 492: Would prefer the authors not use “BC” here, was confusing at first read because of NEE abbreviation right before. Change to … and boundary conditions (i.e. the influence of CO2 being transported into our domain) only….

Page 14, Lines 501–503: Couldn’t the model transport also be wrong? This is a definite yes – the transport model could be wrong. Our reasoning not to assume that this is the dominant factor here is that a.) the wind observations are well reproduced by CHIMERE, see figure 7 and b.) the PIS-RES gradient falls onto the 1:1 line - compared to the 1.7 of MIT-RES. It seems unlikely that the model properly models transport from RES to PIS but fails for RES to MIT as there a no major topographic disturbance in this part of the Paris Basin. We have added a note of caution that the model could be wrong in the discussion, as well as why we think this is not the biggest contributor to the disagreement here.
Anonymous Referee #3  
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General Comments
This manuscript describes a pilot project of five EM27/SUN spectrometers that were deployed for two weeks in the Paris region to investigate CO2 fluxes from that megacity. They also describe a modeling framework to compare to the column measurements and provide some initial comparisons. While it is commendable that the authors are publishing the results of a pilot phase of a project where important details about the instrumentation are shown, their analysis is incomplete. The “background” on the modeled result was very different from the observations, and no hypothesis was given for why that might be the case. Then, the analysis and discussion behind Fig 13 and 14 had several logical gaps and should be completely reconsidered. The conclusions section had several problems and was not well supported by the main text. I do not think it would take that much work to fix these issues, but they are major issues with the analysis and interpretation of the study. Until these are fixed I would not recommend publication. We have significantly streamlined the paper and also expanded on our analysis of the data to fill the gaps identified by the reviewer. Specifically, we have added more explanation/interpretation around Figures 13 and 14 and clarified the text to avoid misunderstandings (e.g. Figure 14 is already the mean daily cycle for selected data i.e. when the site is an upwind site). More explanation in individual improvements and corrections is given below.

Specific Comments

Line 92: remove the word “by”.
Corrected – Thanks.

Line 105: “spectrometers” should be singular, ie “spectrometer”
Corrected - Thanks

Line 111: This sentence is grammatically incorrect and should be reworded slightly. We have reformulated this sentence.

Line 153: remove the word “for”
Corrected - Thanks

Line 154: Add “PM” at the end of the line.
Corrected - Thanks

Line 192: Does instrument 1 have the best agreement with the TCCON instrument? The text here and in the rest of the paragraph indicates that the EM27/SUN measurements CAN be made traceable to the WMO scale, but it doesn’t say IF they were or not. It would be good to explain if they were, and if they were not it is even more important to say that and explain why they were not.
There is nothing special about spectrometer #1, arbitrarily one of the devices has been used as reference here. Any drift of the calibration in either the selected reference spectrometer or of one of the other spectrometers would induce variations in the table entries (correlated between different spectrometers if the reference is drifting, uncorrelated if one of the other spectrometer is drifting)
The measurements can be made traceable to WMO scale to the extent that can be achieved for TCCON. Direct side-by-side comparison with a TCCON spectrometer have been used to estimate the calibration offset of the EM27/SUN wrt TCCON. As the same spectral bands are used for the observations and the instrumental characteristics of both TCCON and EM27/SUN spectrometers are very close to an ideal FTS, it is expected that the calibration between EM27/SUN and TCCON is quite consistent. The paper of Frey et al. provides a good impression of the level of consistency found. (Frey, M., Sha, M. K., Hase, F., Kiel, M., Blumenstock, T., Harig, R., Surawicz, G., Deutscher, N. M., Shiomi, K., Franklin, J., Bösch, H., Chen, J., Grutter, M., Ohyama, H., Sun, Y., Butz, A., Mengistu Tsidu, G., Ene, D., Wunch, D., Cao, Z., Garcia, O., Ramonet, M., Vogel, F., and Orphal, J.: Building the COllaborative Carbon Column Observing Network (COCCON): Long term stability and ensemble performance of the EM27/SUN Fourier transform spectrometer, Atmos. Meas. Tech. Discuss., https://doi.org/10.5194/amt-2018-146

However, we would not want to claim that this data is traceable to WMO, yet as more comparison and methodological work should be done in the future.

Line 328-331: I think the authors got this backwards, the text says that the upwind sites had higher XCO2 than their downwind sites indicating that the FFCO2 from Paris is detectable. I think they meant to switch upwind and downwind.

Yes, indeed we somehow switched this and failed to notice in proof-reading. Thanks for catching this.

Another comment about these lines is that, looking at Fig 4, it looks like RES is lower than MIT for most of the study period, but this switches on May 12/13 and RES is higher than MIT. This makes perfect sense looking at Table 3 where it says that for the first part of the campaign the winds were predominantly from the SW while on the 12/13 the winds were from the NNW and NE. It might be worth explaining this feature in the data here since its interesting.

We have added more discussion on this.

Line 350: The word “northeasterly” is used incorrectly here. Change this to “The two (typically downwind) sites PIS and MIT northeast of Paris show a . . . ”

Corrected

Line 351: What does the word “background” mean in this context? Background could mean several things in this context, so I would encourage the authors to either define what the background is or use a different word here to avoid confusion.

We have removed the word background here as it is indeed confusing and means different things in different communities. We have given more specific explanations when necessary.

Line 356: The word “background” is used again with, I think, a different meaning than was used above. It is also undefined here. I understand exactly what the authors mean, but I think it would be good to provide a little bit of explanation here describing exactly what they mean by background conditions (ie the XCO2 of the air mass entering the urban domain that has been affected by emissions upwind or outside of the domain).

We have removed the word background here as it means different things in different communities – even among the co-authors of this paper. We have added an explanation of the assumption underlying the gradient approach.

Line 360: “are” should be “were”.

Corrected

Line 361: “measurement period” should be “on this day” since the winds do vary over
the whole measurement period, but they were from the SW on this day that is the focus of this figure.
Corrected

Line 362: This sentence needs to be re-written for clarity. How about this: “The observations from GIF showed only minimal differences with RES while the rest of the sites (PIS, JUS, and MIT) had $\Delta$ values of 1 to 1.5 ppm.
Thanks for this suggested - corrected

Line 363: Delete “of most”
Corrected

Lines 355-376: Be careful of the tense in this paragraph and elsewhere. For most of the paper the data was referred to in the past tense, but I noticed that this paragraph is in the present tense. To fix this, change “is” to “was” (etc) when referring to the data throughout the section.
Thanks – this error was corrected (all in past tense, where appropriate).

Lines 405: On this line I initially thought that the authors were drawing a comparison between the modeled XCO2 and the measured XCO2, but after reading the sentence several times I now realize they are just talking about modeled XCO2. It would be good to explicitly say “modeled XCO2” on line 405 to prevent any confusion.
The paragraphs were explicitly separated into measured and modelled to avoid confusion, but we’ve now added more clarifying markers in the paragraph.

Lines 413-415: Might also be worth pointing out that sometimes the NEE flux is slightly positive due to respiration, generally at night.
We’ve added an explanation.

Line 424: I don’t see any shaded areas on Fig 11.
Removed the mention of the grey shaded area as the wind directions are given on top of graph 11.

Lines 420-436 (and Fig 11): After reading this a couple of times and staring at Fig 11, I finally realized that each of the vertical panels represents a different modeled source. The subscripts on the y-axis are very small and are not explained anywhere. It would be good to explain what each of the panels are showing in the figure caption. The authors should also explain in the text that this figure shows the total XCO2 in the top panel, and below that the three panels show the modeled contributions from FFCO2, biological emissions (NEE), and background conditions (BC) respectively.
We have increased the size of the labels and added more explanation in the caption to make the information more accessible to the reader.

Line 440: They should reference Fig 12 at the beginning of this paragraph somewhere.
Figure 12 were references in line 444 now moved to line 440.

Line 441: There is an extra period and spaces on this line.
Removed

Line 445: I would encourage the authors to not the background offset FIRST in this paragraph as that is the most obvious feature. Then, once the offset is noted they can go on to describe the diel cycle and the difference between the sites.
Also, the authors should offer an explanation for why they think their background model is 1-2 ppm off. We have restructured this paragraph and also added the information about the reduced impact of BC when using the gradient approach.

Line 454: The authors should explain here that this comparison is not sensitive to the offset in the BC because it is comparing the modeled upwind with the modeled downwind and the measured upwind with the measured downwind. Also, as a general note, the use of the delta symbol is problematic because of its use in radiocarbon nomenclature. It’s OK if the authors desire to use it, but I would encourage them to find an alternative way of noting this. See reply to comment on line 445.

Concerning, the general note on nomenclature, as radiocarbon is not mentioned in this manuscript and the remote sensing community does not commonly use $\Delta^{14}$C we decided to use this nomenclature as it seemed most appropriate to the (co-) authors.

Line 452-465: I really don’t understand what the significance of Fig 13 is, and this analysis doesn’t make sense to me. I would expect that the observations should only fall on the 1:1 line when the wind direction is directly between the upwind/downwind sites. The fact that most of the observations have a slope close to 1 could alternatively suggest that wind direction doesn’t matter! I would also expect that when the wind direction is from a 90-degree angle to the upwind direction (so that the wind is blowing across the city instead of from one site to the other) that there should be much higher variability and potentially no relationship between the XCO2 at the two sites.

We have added additional elements to better explain figure 13. Fundamentally, the point was to investigate the impact of wind direction on the concentration gradient and if our atmospheric transport model predicts concentrations equally well for all wind conditions. Apparently these messages were lost/unclear. As expected the gradients are strongly positive when MIT and PIS are downwind of Paris and we see negative gradients when they are upwind. Furthermore, the slope of the individual wind directions does not seem to follow the 1:1 line. Especially, northerly winds (blue colors in Fig13 – now Fig 9) seem to have steeper slopes for PIS. This figure also highlights that when trying to assess the impact of Parisian emissions significant gains in signal amplitude can be gained when using data from upwind-downwind situations only. This is not really a ‘surprising’ finding, but we can now quantify how much more signal is seen on average. On the 90 degree question – even in this situation with easterly winds we could expect to see differences. As shown in Figure 1, the RES site is then and MIT is technically upwind, which explains the negative gradient in Figure 13. The gradient for west however is quite different then for Northerly winds as other parts of Paris are then “upwind” of RES.

As the reviewers will find, we have added much more explanation and interpretation around Figure 13 to clarify our interpretation.

Here are a few suggestions for Fig 13. The authors should only plot the data from when the wind is blowing directly from PIS->RES (or MIT->RES) and when it is blowing back from PIS<->RES (or MIT<->RES). There should only be a narrow range of wind direction angles that this comparison should work, maybe 20-30 degrees or something like that. Also, the authors should indicate on the figure, or in the text somewhere what the exact angle it is between the sites in decimal degrees (not just with letters indicating the cardinal directions). Also, the figure caption says that the vertical bars indicate the standard deviation, but they don’t say WHAT it’s the standard deviation of! Is it the standard deviation of the measurements from a range of wind directions? Is it over some time window?
Thanks for the suggestions. We have added exact bearings and distances of all stations relative to RES in table S1 in the supplement. We have kept Figure 13 as it allows seeing the influence of different wind conditions on the concentration gradients. A narrow upwind/downwind window would not allow seeing this and would also remove a lot of the data. We think the color-code in figure 13 allows to see how certain wind conditions are pooled. PIS is optimally upwind of RES at a wind direction of 187 degrees (green) and MIT is optimally upwind for a wind direction of 217 degree (green).

Lines 466-479: This diurnal cycle plot is confusing to me because there are times when the wind is blowing from PIS to RES, and there are times when the wind is blowing in the opposite direction. Shouldn’t the XCO2 be negative when it is blowing in the opposite direction? Wouldn’t it also have no relationship between the sites when there is no upwind/downwind relationship? It would be much better to isolate this comparison to ONLY times when the wind is blowing in the appropriate direction, not during the whole campaign.

This is plot DOES indeed only include data when MIT and PIS are downwind of RES. For the observations (upper panel) only data from days with observations are used. This is also why only so few days (0 to 5) contribute to the mean diurnal cycle (as given by the labels on top panel of Figure 14). In the lower panel all days within the campaign period that fulfill the upwind-downwind requirement are used. We have added text to clarify this.

Line 482: “two-weeks” should be “two-week”.
Corrected

Line 485: What do the authors mean by “easily linked”? This is sloppy language that is easily misinterpreted, especially in the conclusions section. This whole sentence needs to be re-written for clarity so that the wrong impression is not given.
We have reformulated to clarify.

Line 488: The authors don’t actually know what is impacting remote CO2. This should instead say something like “... greatly reduced the impact of background CO2 fluxes.”
Thanks – we have reformulated to clarify. Using the gradient does indeed reduce both the influence of boundary condition CO2 and biogenic fluxes within the CHIMERE domain.

Line 491: the word “significant” has statistical meaning and shouldn’t be used in this instance. Also, “enhanced background” seems incorrect since they never offered a hypothesis about why the background was higher. Actually, just the word “enhanced” should be changed to “higher” or something that is more objective.
We have reformulated and change the wording where appropriate.

Line 492: Here is the word “significantly” again. The authors should use a different word here, like “... also predicts that NEE and BC only has a large impact on XCO2 during a few situations ...”
We have reformulated and change the wording where appropriate.

Lines 491-494: Actually, this whole sentence is problematic and needs revision. The first half of the sentence seems to refer to the discussion surrounding Fig 10 (which is great) but the second half of the sentence referring to upwind and downwind (as it relates to NEE and BC) seems unrelated. If this stays in the text, it needs more detail to explain what the authors were thinking about.
We have reformulated to clarify.
This section refers to Fig 13, and this methodology is flawed since an alternative explanation is that wind direction doesn’t even matter in this data set. Unfortunately, we can’t really follow why the wind directions do not matter in this data set. It seems apparent that southerly winds produce strong positive concentration offsets, while northerly winds cause negative concentration gradients. Looking at the colors they seem to group very consistently implying a strong correlation of wind conditions on $\Delta XCO_2$, both modelled and observed. The slope is not wind dependent, but it is not clear why it should be, as we assume that the model performance should be similar in all wind conditions.

This is wrong. I assume they are referring to Breon et al 2015 Fig 6 where the highest $R = 0.90$ (not 0.91). Also, this was a straight measured/modeled mole fraction comparison, whereas the analysis in Fig 13 is supposedly the upwind/downwind measured/modeled gradient. Even if the analysis in Fig 13 were done correctly, this would be a different metric for model evaluation and should not be compared with Breon et al 2015. Its comparing apples (measured/modeled in-situ mole fractions) and oranges (measured/modeled GRADIENTS BETWEEN SITES across a city during a 2-week period with a lot of wind direction changes).

Concerning Breon et al. 2015: that study also compared measured and modelled in-situ CO2 gradients between sites across the city for a period of 4 weeks see figures 8, S4 and section 4.4. (https://www.atmos-chem-phys.net/15/1707/2015/acp-15-1707-2015.pdf). R2 for GIF and GON gradients is 0.91 as previously mentioned in the manuscript. The sites chosen in our COCCON campaign are also similar to the locations used in Breon et al. 2015. Some of the (co-)author of this study are indeed (co-)authors of Breon et al. 2015 and although it might not be exactly an ‘apples to apples’ comparison when comparing in-situ CO2 gradients to XCO2 gradients we wanted to cite this study to show that we are attempting to applying an analogous approach.

This whole section needs to be redone after the analysis in Fig 13 is fixed. Also, the speculation about model dispersion is not based on anything and therefore it has no place in the paper unless the authors care to actually try to do some analysis to quantify it.

We have re-worked this section. The hypothesis that the dispersion is an issue is indeed only based on experience using the CHIMERE model and the fact that the model resolution is limited to 2x2km2 which leads to numerical diffusion that is very likely larger then real dispersion. In a recent study, model simulations of CO2 in the boundary layer of Paris at 5m x 5m resolution where performed using MICRO-SWIFT-SPRAY. A lot of heterogeneity is visible (the Figure below is a 2x2 km2 pixel of downtown Paris: Jardin du Luxembourg to Jussieu with the Seine River in the top of domain).
Such localized sources and plumes are immediately dispersed within a 2x2km² CHIMERE cell.

Line 509-511: I actually agree with this statement, but it’s exactly the opposite of what the analysis in this paper shows. Fig 11 shows that the biospheric flux in a gradient sense is small (less than 1ppm almost all the time).
We have clarified that the influence of the biosphere on XCO2 is a major factor but not as important when considering ΔXCO2 here. The point we want to make is that biospheric fluxes are not always important, however there are periods when they cannot be ignored, even when using a gradient approach.

Line 514: They forgot the word “not”. It should be “… and underlying fluxes could NOT be investigated here.”
Corrected

Line 522: I would disagree that they have demonstrated that the modeling framework is “suitable”. They have provided some initial modeling results from a pilot test field campaign and the modeling framework will need a lot of work before it can be usefully applied to interpret fluxes.
We have reformulated to highlight that we also see this as a step forward towards our inversion system.

Figure 3: The x and y axes should be labeled longitude and latitude.
We’ve moved the figure into the supplement and made it bigger (also the lat-long labels)
Figure 7 (top): the y-axis scale could be 0-10 instead of 0-16.
Figure 8: The acronyms MACC is not defined anywhere in the manuscript.
Thanks – we’ve added this information in the caption
Figure 10: In the caption the authors should add “(BC)” so that the reader knows that the legend entry “CHIMERE BC only” means background conditions.
Corrected
XCO₂ in an emission hot-spot region: the COCCON Paris campaign 2015

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Abstract. Providing timely information on urban Greenhouse-Gas (GHG) emissions and their trends to stakeholders relies on reliable measurements of atmospheric concentrations and the understanding of how local emissions and atmospheric transport influence these observations. Portable Fourier Transform Infra-Red (FTIR) spectrometers were deployed at 5 stations in the Paris metropolitan area to provide column-averaged concentrations of CO₂ (XCO₂) during a field campaign in spring of 2015, as part of the Collaborative Carbon Column Observing
Here, we describe and analyze the variations of XCO$_2$ observed at different sites and how they changed over time. We find that observations upwind and downwind of the city centre differ significantly in their XCO$_2$ concentrations, while the overall variability of the daily cycle is similar, i.e., increasing during night-time with a strong decrease (typically 2-3 ppm) during the afternoon.

An atmospheric transport model framework (CHIMERE-CAMS) was used to simulate XCO$_2$ and predict the same behaviour seen in the observations, which supports key findings, e.g., that even in a densely populated region like Paris (over 12 Million people), biospheric uptake of CO$_2$ can be of major influence on daily XCO$_2$ variations. Despite a general offset between modelled and observed XCO$_2$, the model correctly predicts the impact of the meteorological parameters (e.g. wind direction and speed) on the concentration gradients between different stations. Looking at the local gradients of XCO$_2$ for upwind and downwind station pairs, which is found to be less sensitive to changes in XCO$_2$ regional background boundary conditions and biogenic fluxes within the domain and, we find the model-data agreement significantly further improved. Our modelling framework indicates that the local XCO$_2$ gradient between the stations is dominated by the fossil fuel CO$_2$ signal of the Paris metropolitan area. This further highlights the potential usefulness of XCO$_2$ observations to help optimise future urban GHG emission estimates.

Introduction

Atmospheric background concentrations of CO$_2$ measured since 1958 in Mauna Loa, USA, have passed the symbolic milestone of 400 ppm (monthly mean) as of 2013 [Jones 2013]. Properly quantifying fossil fuel CO$_2$ emissions (FFCO$_2$) can contribute to define effective climate mitigation strategies. Focussing our attention on cities is a critical part of this endeavour as emissions from urban areas are currently estimated to represent from 53 % to 87 % of global FFCO$_2$, depending on the accounting method considered, and are predicted to increase further [IPCC-WG3 2014, IEA 2008, Dhakal 2009]. As stated in the IPCC 5th assessment report, “current and future urbanisations trends are significantly different from the past” and “no single factor explains variations in per-capita emissions across cities and there are significant differences in per capita greenhouse gas (GHG) emissions between cities within a single country” [IPCC-WG3 2014]. Therefore, findings in one city cannot often not be simply used and extrapolated to other urban regions. Furthermore, the large uncertainty of the global contribution of urban areas to CO$_2$ emissions today and in the future is why a new generation of city-scale observing and modelling systems are needed.

In recent years, more and more atmospheric networks have emerged that observe GHG concentrations using the atmosphere as a large-scale integrator, for example in Paris, (France) (e.g., Bréon et al. 2015, Xuereb-Remy et al. 2018), Indianapolis, (USA) (e.g., Turnbull
et al. 2015, Lauvaux et al. 2016), Salt Lake City, USA (Strong et al. 2011, Mitchell et al. 2018), Heidelberg, Germany (e.g. Levin et al. 2011, Vogel et al. 2013) and Toronto, Canada (e.g. Vogel et al. 2012). The air measured at in-situ ground-based stations is considered to be representative of surface CO₂ fluxes of a larger surrounding area (1 km²-10000 km²), i.e. the emissions of the Greater Paris Area dominate the airshed of the Ile-de-France (ca. 12'000 km²) (e.g., Bréon et al. 2015, Xueref-Remy et al. 2018). If CO₂ measurements are performed both up-wind and downwind of a city, the concentration gradient between the two locations is influenced by the local net emission strength between both sites and atmospheric mixing (Xueref-Remy et al. 2018, Bréon et al. 2015, Turnbull et al. 2015). To derive quantitative flux estimates, measured concentration data are typically assimilated into numerical atmospheric transport models which calculate the impact of atmospheric mixing on concentration gradients for a given flux space-time distribution. Such a data assimilation framework implemented for Paris with three atmospheric CO₂ measurement sites (Xueref-Remy et al. 2018) previously allowed deriving quantitative estimates of monthly emissions and their uncertainties over one year (Staufer et al. 2016).

Space-borne measurements of the column-average dry air mole fraction of CO₂ (XCO₂) are increasingly considered for the monitoring of urban CO₂. This potential was shown with OCO-2 and GOSAT XCO₂ measurements, even though the spatial coverage and temporal sampling frequency of these two instruments were not optimized for FFCO₂ (Kort et al., 2012, Janardanan et al. 2016, Schwandner et al. 2017), while other space-borne sensors dedicated to FFCO₂ and with an imaging capability are in preparation (O’Brien et al., 2016, Broquet et al. 2017). Important challenges of satellite measurements are that they are not as accurate as in-situ ones, having larger by-systematic errors, while the XCO₂ gradients in the column are typically 7-8 times smaller than in the boundary layer. Another difficulty of space-borne imagery with passive instruments is that they will only sample city XCO₂ plumes during clear sky conditions for geostationary satellites and with an additional constraint to observations at around mid-day for low-earth orbiting satellites.

The recent development of a robust portable ground-based FTIR (Fourier Transform InfraRed) spectrometer as described in Gisi et al. (2012) and Hase et al. (2015) (EM27/SUN, Bruker Optik, Germany) greatly facilitates the measurement of XCO₂ from the surface, with better accuracy than from space and with the possibility of continuous daytime observation during clear sky conditions. Typical compatibility (uncorrected bias) of the EM27/SUN retrievals of the different instruments in a local network is better than 0.01 % (i.e. ` 0.04 ppm) after a careful calibration procedure and a harmonized processing scheme for all spectrometers (Frey et al. 2015). The Collaborative Carbon Column Observing Network (COCCON) (Frey et al. 2018) intends to offer such a framework for operating the EM27/SUN. This type of spectrometers.
therefore represents a remarkable opportunity to document XCO₂ variability in cities as a direct
way to estimate FFCO₂ [Hase et al. 2015] or in preparation of satellite missions.
When future low-Earth-orbit operational satellites with imaging passive spectrometers of
suitable capabilities to invert FFCO₂ will sample different cities, this will likely be limited to clear
sky conditions and at a time of the day close to local noon. Increasing the density of the
COCCON network stations around cities will allow to evaluate those XCO₂ measurements and
to monitor XCO₂ during the early morning and afternoon periods, which will not be sampled
with satellites low-earth orbit satellite, except from geostationary orbit, which can also have
other benefits, those time-periods can however be observed and could be compared to
ground-based measurements [e.g. Butz et al., 2015, O’Brien et al. 2016].
This study focuses on the measurements of XCO₂ from ground based EM27/SUN
spectrometers deployed within the Paris metropolitan area during a field campaign in the
spring of 2015, and modelling results. This campaign can be seen as a demonstration of the
COCCON network concept applied to the quantification of an urban FFCO₂ source. Several
spectrometers were operated by different research groups, while closely following the
common procedures suggested by Frey et al. [2015]. The paper is organised as follows. After
the instrumental and modelling setup descriptions of section 2, the observations of the field
campaign and the modelling results will be presented in section 3. Results are discussed in
section 4 together with the study conclusions.

2 Methods and materials
2.1 Description of study area and field campaign design
During the COCCON field campaign (April 28th to May 13th, 2015) five portable FTIR
spectrometers (EM27/SUN, Bruker Optik, Karlsruhe, Germany) were deployed in the Parisian
region (administratively known as Île-de-France) and within the city of Paris. The campaign
was conducted in early spring as the cloud cover is typically low in April and May and the time
between sunrise and sunset is more than 14 hours.
The Paris metropolitan area houses over 12 million people, with about 2.2 million inhabiting
the city of Paris. This urban region is the most densely populated in France with ~1000
inhabitants/km² and over 21000 inhabitants/km² for the city of Paris itself [INSEE 2016 -
https://www.insee.fr/fr/statistiques]. The estimated CO₂ emissions from the metropolitan
region are 39 Mt/year, according to the air quality association (AIRPARIF), that monitors the
airshed of Greater Paris, with on-road traffic emissions, and residential and the tertiary (i.e.
commercial) sector as are the main sources (accounting for over 75 %), and minor
contributions from other sectors such as industrial sources and airports [AIRPARIF
[https://www.airparif.asso.fr/en/, AIRPARIF 2016]. It was crucial to understand the spatial
distribution of these CO₂ sources to optimally deploy the COCCON spectrometers. To this end
a 1 km emission model for France by IER (Institut fuer Energiewirtschaft und Rationelle Energieanwendung, University of Stuttgart, Germany) was used as a starting point [Latoska 2009]. This emission inventory is based on the available activity data such as, e.g., traffic counts, housing statistics, or energy use, and the temporal disaggregation was implemented according to Vogel et al. [2013]. In brief, the total emissions of the IER model were re-scaled to match the temporal factors for the different emission sectors according to known national temporal emission profiles.

To quantify the impact of urban emissions on XCO₂, the FTIR instruments were deployed along the dominant wind directions in this region in spring, i.e., southwesterly [Staufer et al. 2016], in order to maximize the likelihood to capture upwind and downwind air masses (see Figure 1). The two southwesterly sites (GIF and RES) are located in a less densely populated area, where emissions are typically lower than in the city centre, where the station JUS is located. The data in Fig. 1 show that the densest FFCO₂ emission area extends northwards and eastwards. The two Northwesterly sites (PIS and MIT) were placed downwind of this area.

All instruments were operated manually and typically started for operations around 7-8 am local time from which they continuously observe XCO₂ until 5-6 pm.

2.2 Instrumentation, calibration, and data processing

The EM27/SUN is a portable FTIR spectrometer which has been described in detail in, e.g., Gisi et al. [2012] and Frey et al. [2015]. Here, only a short overview is given. The centre piece of the instrument is a Michelson interferometer which splits up the incoming solar radiation into two beams. After inserting a path difference between the beams, the partial beams are recombined. The modulated signal is detected by an InGaAs detector covering the spectral domain from 5000 to 11000 cm⁻¹ and is called an interferogram. As the EM27/SUN analyzes solar radiation, it can only operate in daylight sunny conditions. A Fourier transform of the interferogram generates the spectrum and a DC correction is applied to remove the background signal and only keep the AC signal (see Keppel-Aleks et al. [2007]). A numerical fitting procedure (PROFFIT code) [Schneider and Hase et al., 2009] then retrieves column abundances of the concentrations of the observed gases from the spectrum. The single-channel EM27/SUN is able to measure total columns of O₂, CO₂, CH₄ and H₂O. The ratio over the observed O₂ column, assumed to be known and constant, delivers the column-averaged trace gas concentrations of XCO₂, XCH₄ in µmol / mol dry air, with a temporal resolution of one minute. XCO₂ is the dry air mole fraction of CO₂, defined as XCO₂ = Column[CO₂] / Column[Dry Air]. Applying the ratio over the observed oxygen (O₂) column reduces the effect of various possible systematic errors; see Wunch et al. (2011).

In order to correctly quantify small differences in XCO₂ columns between Paris city upstream and downstream locations, measurements were performed with the five FTIR instruments side
by side before and after the campaign, as we expect small calibration differences between the
different instruments due to slightly different alignment for each individual spectrometer. These
differences are constant over time and can be easily accounted for by applying a calibration
factor for each instrument. Previous studies showed that the instrument specific corrections
are well below 0.1 % for XCO2 [Frey et al. 2015, Chen et al. 2016] and are stable for individual
device. The 1-sigma precision for XCO2 is in the order of 0.01 % - 0.02 % (< 0.08 ppm) e. g.
calibration measurements for this campaign were performed in Karlsruhe w.r.t. the Total
Carbon Column Observing Network (TCCON) [Wunch et al. 2011] spectrometer at the
Karlsruhe Institute of Technology (KIT), Germany for 7 days before the Paris campaign
between April 9th and 23rd, and after the campaign on May 18th until 21st.

Figure 2 S1 (left panel) shows the XCO2 time series of the calibration campaign, where small
offsets between the instruments raw data are visible. As these offsets are constant over time,
a calibration factor for each instrument can be easily applied; actually these are the calibration
factors previously found for the Berlin campaign [Frey et al. 2015]. These factors are given in
Table 21, where all EM27/SUN instruments are scaled to match instrument No. 1. The
calibrated XCO2 values for April 15th are shown in Fig. 2 S1 (right panel). None of the five
instruments that participated in the Berlin campaign show any significant drift; in other words,
the calibration factors found one year before were still applicable. This is an impressive good
demonstration of the instrument stability stated in section 2.2, especially as several
instruments (Nos. 1, 3, 5) were used in another campaign in Northern Germany in the
meantime. The EM27/SUN XCO2 measurements can also be made traceable to the WMO
international scale for in-situ measurements by comparison with measurements of a collocated
TCCON spectrometer from the TCCON. TCCON instrumentsare calibrated against in-
situ standards by aircraft and aircore measurements [Wunch et al. 2010, Messerschmidt et al.
2012] performed using the WMO scale.

During the campaign and for the calibration measurements we recorded double-sided
interferograms with 0.5 cm⁻¹ spectral resolution. Each measurement of 58 s duration consisted
of 10 scans using a scanner velocity of 10 kHz. For precise timekeeping, we used GPS
sensors for each spectrometer.

In-situ surface pressure data used for the analysis of the calibration measurements performed
at KIT have been recorded at the co-located meteorological tall tower. During the campaign,
a MHD-382SD data-logger recorded local pressure, temperature and relative humidity at each
station. The analysis of the trace gases from the measured spectra for the calibration
measurements has been performed as described by Frey et al. [2015]. For the campaign
measurements we assume a common vertical pressure-temperature profile for all sites,
provided by the model, so that the surface pressure at each spectrometer only differs due to
different site altitudes. The 3-hourly temperature profile from the European Centre for Medium-Range Weather Forecasts (ECMWF) operational analyses interpolated for site JUS located in the centre of the array was used for the spectra analysis at all sites. The individual ground-pressure was derived from site altitudes and pressure measurements performed at each site. Before and after the Paris campaign, side by side comparison measurements were performed with all 5 EM27/SUN spectrometers and the TCCON spectrometer operated in Karlsruhe at KIT. All spectrometers were placed on the top of the IMK office building North of Karlsruhe. The altitude is 133 m above sea level (a.s.l.), coordinates are 49.09° N and 8.43° E. The processing of the Paris raw observations (measured interferograms) were performed as described by Gisi et al. [2012] and Frey et al. [2015] for the Berlin campaign: spectra were generated applying a DC correction, a Norton-Beer medium apodization function and a spectral resampling of the sampling grid resulting from the FFT on a minimally sampled spectral grid. PROFFWD was used as the radiative transfer model and PROFFIT as the retrieval code.

2.3 Atmospheric transport modelling framework

We used the chemistry transport model CHIMERE (Menut et al., 2013) to simulate CO₂ concentrations in the Paris area. More specifically, we used the CHIMERE configuration over which the inversion system of Bréon et al. [2015] and Staufer et al. [2016] was built to derive monthly to 6-hour mean estimates of the CO₂ Paris emissions. Its horizontal grid, and thus its domain and its spatial resolution, are illustrated in Figure S23. It has a 2 × 2 km² spatial resolution for the Paris region, and 2 × 10 km² and 10 × 10 km² spatial resolutions for the surroundings. It has 20 vertical hybrid pressure-sigma (terrain-following) layers that range from the surface to the mid-troposphere, up to 500 hPa. It is driven by operational meteorological analyses of the ECMWF Integrated Forecasting System, available at an approximately 15 × 15 km² spatial resolution and 3 h temporal resolution.

In this study the CO₂ simulations are based on a forward run over April 25th - May 12th 2015 with this model configuration; we do not assimilate atmospheric CO₂ data and so no inversion for surface fluxes was conducted. In the Paris area (the Île-de-France administrative region), hourly anthropogenic emissions are given by the IER inventory, see section 2.1. The anthropogenic emissions in the rest of the domain are prescribed from the EDGAR V4.2 database for the year 2010 at 0.1° resolution [Olivier and Janssens-Maenhout et al., 2012]. In the whole simulation domain, the natural fluxes (the Net Ecosystem Exchange: NEE) are prescribed using simulations of C-TESSEL, which is the land-surface component of the ECMWF forecasting system [Boussetta et al., 2013], at a 3 hourly and 15 × 15 km² resolution. Finally, the CO₂ boundary conditions at the lateral and top boundaries of the simulation domain and the simulation CO₂ initial conditions on April 25th 2015 are prescribed using the CO₂
forecast issued by the Copernicus Atmosphere Monitoring Service (CAMS, http://atmosphere.copernicus.eu/) at a ~15 km global resolution [Agustí-Panareda et al., 2014]. The CHIMERE transport model is used to simulate the XCO₂ data. However, since the model does not cover the atmosphere up to its top, the CO₂ fields from CHIMERE are complemented with that of the CAMS CO₂ forecasts from 500 hPa to the top of the atmosphere to derive total column concentrations. The derivation of modelled XCO₂ at the sites, involves obtaining a kernel-smoothed CO₂ profile of CHIMERE and CAMS and vertical integration of these smoothed profiles, weighted by the pressure at the horizontal location of the sites.

The parametrisation used to smooth modelled CO₂ profiles approximates the sensitivity of the EM27/sun CO₂ retrieval is a function of pressure and sun elevation. Between 1000 hPa and 480 hPa, a linear dependency of the instrument averaging kernels on solar zenith angle (θ) is assumed with boundary values following Frey et al. [2015]:

\begin{align}
(1a) \quad k(480 \text{ hPa}) &= 1.125 \\
(1b) \quad k(1000 \text{ hPa}) &= 1.0 + 0.45 s^3
\end{align}

where \( s = \theta/90^\circ \). Approximate averaging kernels are obtained by linear interpolation to the pressure levels of CHIMERE and CAMS, respectively. If \( p > 1000 \text{ hPa} \), \( k \) is linearly extrapolated. Above 480 hPa (\( p < 480 \text{ hPa} \)), the averaging kernels can be approximated by

\begin{align}
(2) \quad k(u,s) &= 1.125 - 0.6 u^3 - 0.4 u s^3
\end{align}

where \( u = (480 \text{ hPa} - p) / 480 \). The kernel-smoothed CO₂ profile, \( CO₂_{\text{model}} \), is obtained by

\begin{align}
(3) \quad CO₂_{\text{model}}^2 = K CO₂_{\text{model}}^a + (I - K) CO₂^a
\end{align}

where \( CO₂_{\text{model}} \) is the modelled CO₂ profile by CHIMERE or CAMS, \( I \) the identity matrix and \( K \) is a diagonal matrix containing the averaging kernels \( k \). The a priori CO₂ profile, \( CO₂^a \), is provided by the Whole Atmosphere Community Climate Model (WACCM) model (version 6) and interpolated to the pressure levels of CHIMERE and CAMS. \( CO₂_{\text{model}}^a \) is the appropriate CO₂ profile to calculate modelled XCO₂ at the location of the sites. For a given site, the simulated XCO₂ data are thus computed from the vertical profile of this site as:
where $p_{surf}$ is the surface pressure, $p_{top CHIM} = 500$ hPa the pressure corresponding to the top boundary of the CHIMERE model, and CO$_2_{CHIM}$ and CO$_2_{CAMS}$ are the smoothed CO$_2$ concentrations of CHIMERE and CAMS respectively. For comparison we also calculated XCO$_2$ at a lower spatial resolution with the CAMS data alone as:

$$XCO_2_{CAMS} = \int_{p_{surf}}^{p=0 \text{mbar}} CO_2_{CAMS} dp$$

3 Results and discussion

3.1 Observations

3.1.1 Meteorological conditions and data coverage/instrument performance.

During the measurement campaign (April 28th until May 13th, 2015), meteorological conditions were a major limitation for the availability of XCO$_2$ observations. Useful EM27/SUN measurements require direct sunlight and low wind speeds typically yield higher local XCO2. Most of the time during the campaign conditions were partly cloudy and turbid, and so successful measurements at high solar zenith angle (SZA) were rare. Therefore, the data coverage between April 28th and May 3rd is limited (see Table 32). As is typical for spring periods in Paris, the temperature and the wind direction vary and display less synoptic variations than in winter. The dominant wind directions were mostly northeasterly at the beginning of the campaign and mostly southeasterly during the second half of the campaign. We find that the wind speeds during daytime nearly always surpass 3 m s$^{-1}$, which has been identified by Breon et al. [2015] and Staufer et al. [2016] as the cut-off wind speed above which the atmospheric transport model CHIMERE performs best in modelling CO$_2$ concentration gradients in the mixed layer.

Despite some periods with unfavourable conditions, more than 10,000 spectra were retrieved among the five deployed instruments. The quality of the spectra for each day was rated according to the overall data availability and consistent with Hase et al. (2015). The best measurement conditions prevailed for the period between May 7th and May 12th.

3.1.2 Observations of XCO$_2$ in Paris

The observed XCO$_2$ in the Paris region for all sites (10415 observations) ranges from 397.27 to 404.66 ppm with a mean of 401.26 ppm (a median of 401.15 ppm). The strong atmospheric variability of XCO$_2$ across Paris and within the campaign period is reflected in the standard deviation of 1.04 ppm for 1-minute averages. We find that all sites exhibit very similar
diurnal behaviours with a clear decrease of XCO$_2$ during daytime and a noticeable day-to-day variability as seen in Figure 42. This is to be expected as they are all subject to very similar atmospheric transport in the boundary layer height and to similar large-scale influences, i.e., surrounding with stronger natural fluxes or air mass exchange with other regions at synoptic time scales. However, observed XCO$_2$ concentrations at the upwind-downwind sites for our network remain clearly higher from sites that are downwind-upwind of Paris (see Figure 42). The shifting dominant wind conditions also explain why the site RES and GIF are lowest in the beginning of the campaign and higher on May 12$^{th}$ and 13$^{th}$ after meteorological conditions changed. This indicates that the influence of urban emissions is detectable with this network configuration under favourable meteorological conditions. By comparing the different daily variations in Fig. 53, it is apparent that the day-to-day variations observed at the two southwesterly (typically upwind) sites GIF and RES are approximately 1 ppm, with both sites exhibiting similar diurnal variations throughout the campaign period. This can be expected as their close vicinity would suggest that they are sensitive to emissions from similar areas and to concentrations of air masses arriving from the southwest.

The typical decrease in XCO$_2$ found over the course of a day is about 2 to 3 ppm. This decrease can only be driven by (natural) sinks of CO$_2$, which can be expected to be very strong as our campaign took place after the start of the growing season in Europe for most of southern and central Europe [Roetzer and Chmielewski 2001]. The observations at the site located in Paris (JUS) displays similarly low day-to-day variations and a clear decrease in XCO$_2$ over the course of the day. The latter feature indicates that even in the dense city centre, XCO$_2$ is primarily representative of a large footprint like in other areas of the globe [Keppel-Aleks 2011] and supports the findings of [Keppel-Aleks, 2014;Belikov et al. (2017)] concerning the footprints for the Paris and Orleans TCCON sites. Thus, our total column and is not observations are less critically affected by local emissions than in-situ measurements [Breon et al. 2015, Ammoura et al. 2016]. It is also apparent that the decrease in XCO$_2$ (the slope) during the afternoon for April 28$^{th}$ and 29$^{th}$ as well as May 7$^{th}$ and 10$^{th}$ is noticeably smaller than at other days during this campaign. As XCO$_2$ is not sensitive to vertical mixing, this has to be caused by different CO$_2$ sources and sinks acting upon the total column arriving at JUS.

The two northeasterly (typically downwind) sites PIS and MIT northeast of Paris show a markedly larger day-to-day spread in their general XCO$_2$ levels background as well as strongly changing slopes for the diurnal XCO$_2$ decrease. For these sites the exact wind direction is critical as they can be downwind of the city centre that has a much higher emission density or less dense suburbs (see Fig. 1).

3.1.3 Gradients in observed XCO$_2$
In order to focus more on the impact of local emissions and less on that of background CO₂ fluxes from outside of our urban domain influences in our analysis of XCO₂, we choose to study the spatial gradients (∆) between different sites. Fundamentally, this approach assumes that regional and large-scale fluxes have a similar impact on XCO₂ for the sites within our network, due to the close proximity of sites and the smoothing of remote emission signals due to atmospheric transport by the time the air-mass arrives in our domain. Ideal conditions were sampled during May 7th, with predominantly southwesterly winds, and on May 10th with southerly winds. We can see in Fig. 6 that all sites were, on average, elevated compared to RES, chosen as reference here as it was upwind of Paris during the measurement period those days. The hodographs for both days also indicate that the wind fields were consistent across Paris (see Figure S3). The observations from GIF only showed only minimal differences with RES, while the rest of the sites (PIS, JUS and MIT) had Δ values of increases between PIS, JUS and MIT and RES to reach 1 to 1.5 ppm. During southwesterly winds, MIT is downwind of most of the densest part of the Paris urban area, and JUS is impacted by emissions of neighborhoods to the southwest. The site of PIS is still noticeably influenced by the city centre but, as can be seen in Fig. 1, we likely do not catch the plume of the most intense emissions but rather from the suburbs. On May 10th, with its dominant southerly winds, the situation was markedly different. While GIF was still only slightly elevated, the XCO₂ enhancement at MIT is significantly lower and quite similar to JUS for large parts of the day. The highest ∆XCO₂ can be observed at PIS, again typically ranging from 1 to 1.5 ppm. As seen in Fig. 1, PIS is directly downwind of the densest emission area, while MIT is only exposed to CO₂ emissions from the eastern outskirts of Paris. It is also important to note that the impact of the local biosphere that is assumed to cause the strong decrease in XCO₂ during the day is not seen on both days for these spatial gradients. For a more comprehensive interpretation of these observations the use of a transport model (as described in section 2.3) is necessary.

### 3.2 Modelling

#### 3.2.1 Model performance

Before interpreting the modelled XCO₂ we need to evaluate the performance of the chosen atmospheric transport model framework as described in section 2.3. Comparing it to meteorological observations (wind speed and wind direction) at GIF, we find that CHIMERE predicts these variables well throughout the duration of the campaign (see Figure S4). Changes in wind speed direction and speed are reproduced with a slight overestimation at low wind-speeds (>1 m/s). Besides the meteorological forcing, the model performance can also be expected to depend on the chosen model resolution. Therefore, we compared XCO₂ at JUS calculated based on the coarser resolution atmospheric transport and flux framework...
CAMS (15 km), and the higher resolution emission modelling input for the framework based on CHIMERE (2 km) for the inner domain and on CAMS boundary conditions (see Fig. S28). We find that the coarser model displays similar inter-daily variations, but that the high-resolution model modifies the modelling results on shorter time-scales. We find that the afternoon XCO₂ decreases are often more pronounced in CHIMERE. Only the high-resolution will be considered and referred to in the following.

The impact of using different flux maps (fossil fuel CO₂ and biosphere models) on the modelled XCO₂ can unfortunately not be explicitly investigated here as only one high-resolution (1 km) emission product available for fossil fuel CO₂ was available for this study region (see section 2.3) and other global emission products are usually not intended for urban-scale studies.

3.2.2 Modelled XCO₂ and its components

The modelled XCO₂ for the five sites (Fig. 95) co-evolves over the period of the campaign with occurrences of significant differences. This was already seen with the measurements, but the model allows looking at the full time series. The model reveals clear daily cycles of XCO₂, with an accumulation during nighttime and a decrease during daytime. Despite a good general agreement of modelled XCO₂ at all sites for, e.g., the timing of daily minima and their synoptic changes, differences in XCO₂ are observed between the sites for many days. Typically the northeasterly sites (PIS, MIT) show an enhancement in modelled XCO₂ compared to the southwesterly sites (GIF, RES).

To understand the synoptic and diurnal variations of the modelled XCO₂, we analyzed the contribution of different sources (and sinks) of CO₂, namely the net ecosystem exchange (NEE), the fossil fuel CO₂ emissions (FFCO2), and the boundary conditions (BC), i.e., the variations of CO₂ not caused by fluxes within our domain (the example of JUS is given in Fig. 106; see the supplement for the other sites). The day-to-day variability of modelled XCO₂ is dominated by changing boundary conditions and coincides with synoptic weather changes. As the CO₂ emitted from the different sources is transported in the model as independent tracers, the strong daily decrease in XCO₂ can be directly linked to NEE, which leads to a decrease of ~1 ppm (but up to 4 ppm) during the day, but can also cause positive enhancements during nighttime driven by biogenic respiration. The XCO₂ from fossil fuel emissions causes significant enhancements compared to the background, but is often compensated by NEE. During short periods, fossil fuel emissions can however lead to enhancements of up to 4 ppm.

3.2.3 Modelled ΔXCO₂ gradients and its components
To be able to assess the impact of local sources and reduce the influence of NEE and BC on the modelled signals, we analyse the XCO$_2$ gradient (i.e. station-to-station difference) with RES being taken as reference. In Fig. 11, we compare $\Delta$ in the top panel, and its components, i.e. fossil fuel CO$_2$, biogenic CO$_2$ and CO$_2$ transported across the boundary of the domain (boundary conditions: BC), along a south-north direction. For the modelled $\Delta$ we can see that MIT shows a positive value during the campaign period whenever the predominant wind direction was southwesterly (grey shaded areas). We also find that $\Delta$ between JUS and RES was both negative and positive during the campaign, and predominantly negative between MIT and JUS. When split into FF CO$_2$, BC and NEE components, we can clearly see that the total $\Delta$ is dominated by FF causing XCO$_2$ offsets of up to 4 ppm, but more typically 1 ppm gradients are observed. Gradients can also change rapidly (within a few hours) if the wind direction changes, for example on May 1st and May 12th.

This highlights the fact that, during such conditions, we cannot assume a simple upwind-downwind interpretation of our sites. As expected, the contributions from BC and NEE are generally greatly reduced when analysing $\Delta$XCO$_2$. The most important impact of NEE on the XCO$_2$ gradients of -1 ppm and +1 ppm can be seen on May 8th and May 11th, respectively. This means that, despite greatly reducing the impact of NEE on average, the contribution of NEE cannot be fully ignored. BC is an overall negligible contribution to $\Delta$XCO$_2$, even though it reaches -0.4 ppm on May 11th.

3.3 Model data and observations comparison

3.3.1 XCO$_2$

A comparison of modelled and observed XCO$_2$ is of course limited to the relatively short periods when observations are available. Over these periods we can also see a general issue in reproducing the background XCO$_2$ for each day in the model as observed XCO$_2$ is significantly lower by typically between 1 to 2 ppm. As our CO$_2$ boundary conditions were from a forecast product, this is not unexpected, as already small issues in estimating carbon uptake (or emissions) at the European scale can have such an impact on the boundary conditions. However, we observe that the main features, like daily cycles and synoptic changes of the modelled and observed XCO$_2$ are comparable, as seen in Figure 8. The daytime variations are well reproduced by the model and the general relative concentrations between sites are preserved, e.g., the highest values for XCO$_2$ at MIT are on May 9th and highest XCO$_2$ for PIS are later on May 10th and May 11th as seen in Figure 12. We can also see a general issue in reproducing the background XCO$_2$ for each day in the model as observed XCO$_2$ is significantly lower by typically between 1 to 2 ppm. We also see that the timing of the daily minima is not fully covered in the observed data as it typically
happens after sunset and cessation of biosphere uptake. To reduce the impact of uncertainties of the boundary conditions on our analysis a gradient approach was tested.

### 3.3.2 $\Delta XCO_2$

Due to the prevailing southeasterly wind conditions, we can compare XCO$_2$ at the typical downwind sites (PIS, MIT) relative to the mostly upwind sites (RES, GIF) and expect elevated XCO$_2$ downwind. Furthermore, we can expect to see negative gradients for opposing wind conditions, i.e. northwesterly. For other wind conditions, the concentration difference is not determined by emissions between the station pairs, but rather by the areas upwind of the sites, (see Figure 1). We find that the $\Delta XCO_2$ of PIS relative to RES generally falls along the 1:1 line with a slope of 1.07±0.09 with a Pearson’s R of 0.8. Negative $\Delta XCO_2$ values, seen in Fig. 139, are associated with meteorological conditions when winds come from northerly or easterly directions, i.e., the roles of normal upwind and downwind sites are reversed. For wind perpendicular to the direct line of sight for (PIS, RES) the concentration enhancements are small and harder to interpret and no slope was calculated. The gradient of XCO$_2$ MIT relative to RES has a significantly lower range for modelled XCO$_2$ while the observed range of XCO$_2$ is similar to PIS. The slope of observed to modelled $\Delta XCO_2$ for upwind-downwind (or downwind-upwind conditions) is 1.72±0.06 with a Pearson’s R of 0.96. This points to a significant underestimation of the impact of urban sources, on the MIT-RES gradient, which is also especially visible in the more negative $\Delta XCO_2$ during northerly wind conditions. This could indicate that the spatial distribution of our emissions prior should be improved, i.e., emissions in the eastern outskirts/suburbs are likely underestimated in the IER emissions model. The low modelled $\Delta XCO_2$ could also be due to overestimated horizontal dispersion in the model, which seems less likely. Again the model does not predict concentration differences well for perpendicular wind conditions. When comparing the mean modelled daily cycle of the days with south-westerly wind conditions and when observations exist with the mean diurnal cycle for all days within the field campaign period when MIT and PIS can be considered downwind of RES, we find that the days with observations do not significantly differ from those without observations (see Fig. 1410). An investigation of typical diurnal variations of modelled $\Delta XCO_2$ can only be performed to a limited degree with the observational data available for suitable wind conditions. Within the large uncertainties, the modelled and observed $\Delta XCO_2$ agree throughout the day. When analysing the modelled $\Delta XCO_2$ components we also find that the observed daytime increases of $\Delta XCO_2$ are driven by CO$_2$ added by urban FF CO$_2$ burning and that the impact of FF is significantly higher at PIS (up to 1 ppm) then at MIT site (0.5 ppm) in the model when both sites are downwind of Parisian emissions. Our observations indicate that both sites have strong diurnal variations. As given that the most important biogenic sinks
in our domain, can be expected to be found in the rural parts surrounding Paris; we would expect the biogenic contribution to be similar at both sites (as predicted by the model). This is another point towards indication that the impact of FF emissions on the MIT site is larger than predicted by our modelling framework.

Different $\Delta$XCO$_2$ diurnal variations can be found for other upwind-downwind site pairs, but they are all systematically driven by the locally-added CO$_2$ from FFCO$_2$.

5 Conclusion and Outlook

For the two-weeks field campaign we demonstrated the ability of a network of five EM27/SUN spectrometers, placed in the outskirts of Paris, to successfully track the XCO$_2$ changes due to the urban plume of the city. However, we also found that XCO$_2$ cannot be simply interpreted in the context of local emissions easily linked to local emissions as, even in such a densely populated area, XCO$_2$ is still significantly influenced by natural CO$_2$ uptake during the growing season. Understanding the area influencing XCO$_2$ and/or the use of suitable atmospheric transport models seems indispensable to correctly interpret atmospheric XCO$_2$ variations.

Using a gradient approach, i.e., analysing the difference between XCO$_2$ measured at upwind and downwind stations, greatly reduced the impact of remote CO$_2$ sinks/boundary condition, that reflect fluxes outside the domain and biogenic fluxes within the domain. Overall, the XCO$_2$ variability modelled using our ECMWF-CHIMERE system with IER (1 x 1 km$^2$) emissions data was found to be comparable with the observed variability and diurnal evolution of XCO$_2$, despite a significantly enhanced background for modelled XCO$_2$. Our modelling framework, run at a 2 x 2 km$^2$ resolution over Paris also predicts that NEE-biogenic fluxes and boundary conditions (i.e. the influence of CO$_2$ being transported into our domain) have only very small impact on $\Delta$XCO$_2$; only significantly noticeable impacts impacting $\Delta$XCO$_2$ during a few situations, specifically when meteorological conditions changes made the concept of ‘upwind’ and ‘downwind’ not applicable. When comparing modelled and measured $\Delta$XCO$_2$ we find strong correlations (Pearson’s R) of 0.8 and 0.96 for PIS-RES and MIT-RES, respectively. This can be considered as an excellent degree of correlation as even a model simulation which used optimised fluxes, based on surface observation, showed correlations of 0.91 for its posterior results [Breon et al. 2015]. The offset between model and observations also diminished for $\Delta$XCO$_2$ and the slope found between observed and modeled PIS-RES gradient is statistically in accordance with a 1:1 relationship (1.07±0.09). However, the slope of the MIT-RES XCO$_2$ gradient of 1.72±0.06 suggests that the emission model could potentially be improved; as it seems unlikely that the general atmospheric transport in the model is the key issue as both site pairs would be subject to very similar winds. Another potential source of error that needs to be investigated is unless this is a bias/underestimation.
of $\Delta$XCO$_2$ could be caused by the limited model resolution, is caused by overestimated dispersion in the model. It also seems rather likely that a 2x2km$^2$ model the dispersion would cause a general spreading of emission plumes point source emissions and not systematically underestimate emissions impacts from less densely populated, parts of Île-de-France. The data also confirm previous results by models that XCO$_2$ gradients caused by a megacity do not exceed 2 ppm, which supports the previous requirement for satellite observations of less than 1 ppm precision on individual soundings, and biases lower than 0.5 ppm (Ciais et al. 2015). The gradients are mainly caused by the transport of FF CO$_2$ emissions but, interestingly, during specific episodes, a significant noticeable contribution comes from biogenic fluxes, suggesting that these fluxes cannot always be neglected even when using gradients.

Unfortunately, the duration of the campaign was relatively short, so that an in-depth analysis of mean daily cycles or the impact of ambient conditions (traffic conditions, temperature, solar insolation, etc.) on the observed gradient and underlying fluxes could not be investigated here. Hence, future studies in Paris and elsewhere should aim to perform longer-term observations during different seasons, which will allow better understanding changes in biogenic and anthropogenic CO$_2$ fluxes. A remotely-controllable shelter for the EM27/SUN instrument is currently under development [Heinle and Chen, 2017]. This will considerably facilitate the establishment of permanent spectrometer arrays around cities and other sources of interest. Nevertheless, our study already indicates that such observations of urban XCO$_2$ and $\Delta$XCO$_2$ contain original information to understand local sources and sinks and that the modelling framework used here is suitable a step forward to support their detailed interpretation in the future. An improved model will also be able to adjust or better model the background conditions and potentially use this type of observations to estimate local CO$_2$ fluxes using a Bayesian inversion scheme similar to the existing system based on in-situ observations for Paris [Stauffer et al. 2016].

We expect that the previous successful collaboration in the framework of the Paris campaign will mark the permanent implementation of COCCON as a common framework for a French-Canadian-German collaboration on the EM27/SUN instrument. The acquisition of additional spectrometers is planned by several partners.

**Author contribution**

FRV, MF, FH, IXR, MKS, PCh, PJ, YVT, CJ, TB, QT and JO, supported the field campaign and contributed data to this study.

MF, FH, FRV, JS, GB and PCI planned the fieldwork and modelling activities for this study.

JS, GB, FC, and FRV performed the CHIMERE modelling, provided modelling data input and/or analysed the output data.
MF, FH and FRV processed and analysed the EM27Sun data.

FRV, MF, JS, FH and PCi wrote sections of the manuscript and created figures and tables.

All authors reviewed, edited and approved the manuscript.

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Xueref-Remy, I., Dieudonné, E., Vuillemin, C., Lopez, M., Lac, C., Schmidt, M., Delmotte, M., Chevallier, F., Ravetta, F., Perrussel, O., Ciais, P., Bréon, F.-M., Broquet, G., Ramonet, M., Spain, T. G., and Ampe, C.: Diurnal, synoptic and seasonal variability of atmospheric CO2 in the Paris megacity area, Atmospheric Chemistry and Physics, 18(5), pp.3335-3362. https://doi.org/10.5194/acp-2016-218, accepted in ACP.
Figure 1. CO$_2$ emissions in the Île-de-France region according to the IER emission inventory. Measurement sites are indicated by red crosses.
Figure 2. Sample parallel measurements of the five EM27/SUN instruments in Karlsruhe for raw data (left panel) and the data with the applied correction (right panel) taken on April 15th, 2015.
Figure 3. Modelling domain and numerical grid configuration of CHIMERE with a zoom on the Ile-de-France region at 2 x 2 km².
Figure 42. Time series of observed XCO₂ in the Parisian region for all five sites (all valid data of ~1 minute averages).
Figure 35. Time series of observed XCO₂ in the Parisian region sorted by station.
Figure 46. Observed spatial gradients of XCO$_2$ for May 7$^{th}$ (southwesterly winds) and May 10$^{th}$ (southerly winds).
Figure 7. Comparison of modelled and observed wind speeds and directions at the Gif-Sur-Yvette measurement site.
Figure 8. Comparison of modelled XCO$_2$ from ECWMF-CAMS (15 x 15 km$^2$) with CHIMERE simulation (inner domain, 2 x 2 km$^2$) for JUS.
Figure 59. Modelled XCO$_2$ for all stations.
Figure 106: Time series of XCO2 and related fluxes for JUS. The top panel provides a comparison of modelled total XCO2 and XCO2 variations due to changes in boundary conditions (BC only). The lower panel shows the contribution of the different flux components, namely fossil fuel CO2 emissions and biogenic fluxes.
Figure 117. Modelled XCO₂ gradients for each station relative to RES are given in the top panel with its contributing components in the panels below. Total $\Delta$XCO₂ (top), the fossil fuel contribution $\Delta$XCO₂,ff (second from top), the biogenic contribution $\Delta$XCO₂,bio (third from top) and the influence of the boundary conditions, $\Delta$XCO₂,BC (bottom). The dominant wind conditions for each day given at the top of the figure and days without observations due to precipitation are in red.
Figure 128. Comparison of modelled (solid lines) and observed hourly averaged XCO$_2$ (symbols) with standard deviations as error bars.
Figure 139. Comparison of modelled and observed hourly averaged $\Delta$XCO$_2$ for gradients between PIS and RES (left) and MIT and RES (right), with standard deviations of the minute values of the hourly mean as vertical bars and the points color coded by wind direction from 0 to 359 degrees.
Figure 14. Comparison of modelled (black) and observed mean daily cycle (blue) of hourly averaged ∆XCO₂ of PIS with RES (top left) and of MIT with RES (top right) during the campaign when RES can be considered as upwind site. Labels on top of the upper figures denote the number of days contributing to the mean. The mean daily cycle for all days within the campaign period when PIS and MIT are downwind of RES is given in light grey and the modelled contribution of different CO₂ sources/sinks to the mean daily cycle for days with observations for the two sites is given in the (bottom panels).
<table>
<thead>
<tr>
<th>Location</th>
<th>ID</th>
<th>Lat (deg)</th>
<th>Lon (deg)</th>
<th>Position</th>
</tr>
</thead>
<tbody>
<tr>
<td>Piscop</td>
<td>PIS</td>
<td>49.019</td>
<td>2.347</td>
<td>20 km NNW of JUS</td>
</tr>
<tr>
<td>Mitry-Mory</td>
<td>MIT</td>
<td>48.984</td>
<td>2.626</td>
<td>25 km NW of JUS</td>
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<tr>
<td>Jussieu</td>
<td>JUS</td>
<td>48.846</td>
<td>2.356</td>
<td>Paris city centre</td>
</tr>
<tr>
<td>Saulx-les-Chartreux</td>
<td>RES</td>
<td>48.688</td>
<td>2.284</td>
<td>20 km SSW of JUS</td>
</tr>
<tr>
<td>Gif-Sur-Yvette</td>
<td>GIF</td>
<td>48.708</td>
<td>2.148</td>
<td>20 km SW of JUS</td>
</tr>
</tbody>
</table>

Table 1. Location of FTIR measurement instruments during the field campaign
Table 21. Normalisation factors for the five EM27/SUN instruments derived during measurements before and after the Paris field campaign. Values in parentheses are standard deviations. Measurements of instrument 1 were arbitrarily chosen as reference from which the others were scaled. The calibration factors from a previous field campaign in Berlin [Hase et al. 2015] are also shown. Calibration factors between the two field campaigns agree well within 0.02 % (~0.08 ppm) for all instruments.
<table>
<thead>
<tr>
<th>Date</th>
<th>No. of observations</th>
<th>Quality</th>
<th>Wind speed (m s$^{-1}$)</th>
<th>Wind direction</th>
</tr>
</thead>
<tbody>
<tr>
<td>28 Apr 2015 (Tu)</td>
<td>MIT GIF PIS RES JUS</td>
<td>++</td>
<td>4</td>
<td>W</td>
</tr>
<tr>
<td>29 Apr 2015 (We)</td>
<td></td>
<td>+</td>
<td>5</td>
<td>SW-W</td>
</tr>
<tr>
<td>04 Mai 2015 (Mo)</td>
<td>194 85 96 163 83</td>
<td>+</td>
<td>6</td>
<td>S-SE</td>
</tr>
<tr>
<td>05 Mai 2015 (Tu)</td>
<td>77 27 85 185 92</td>
<td>+</td>
<td>8</td>
<td>S-SW</td>
</tr>
<tr>
<td>06 Mai 2015 (We)</td>
<td>81 88 87 139 0</td>
<td>+</td>
<td>8</td>
<td>SW</td>
</tr>
<tr>
<td>07 Mai 2015 (Th)</td>
<td>169 313 252 286 238</td>
<td>+++</td>
<td>3</td>
<td>SW</td>
</tr>
<tr>
<td>09 Mai 2015 (Sa)</td>
<td>179 0 181 289 149</td>
<td>++</td>
<td>6</td>
<td>W</td>
</tr>
<tr>
<td>10 Mai 2015 (Su)</td>
<td>325 478 362 542 282</td>
<td>++++</td>
<td>3</td>
<td>S</td>
</tr>
<tr>
<td>11 Mai 2015 (Mo)</td>
<td>410 431 251 298 413</td>
<td>++++</td>
<td>3</td>
<td>SSW</td>
</tr>
<tr>
<td>12 Mai 2015 (Tu)</td>
<td>324 222 230 326 203</td>
<td>+++</td>
<td>4</td>
<td>NNW</td>
</tr>
<tr>
<td>13 Mai 2015 (We)</td>
<td>159 18 182 28 56</td>
<td>+</td>
<td>4</td>
<td>NE</td>
</tr>
</tbody>
</table>

Table 32. Summary of all measurement days with the number of observations at each of the sites, Mitry Mory (MIT), Gif Sur Yvette (GIF), Piscop (PIS), Saulx-les-Chartreux (RES), Jussieu (JUS), the overall quality ranking of each day according to the number of available observations—and temporal coverage (with classification from poor to great: +, ++, ++++, ++++), and the ground-level wind speed and direction.
<table>
<thead>
<tr>
<th>Site</th>
<th>Total Obs</th>
<th>Mean (ppm)</th>
<th>STD (ppm)</th>
<th>Quartile1 (ppm)</th>
<th>Median (ppm)</th>
<th>Quartile3 (ppm)</th>
</tr>
</thead>
<tbody>
<tr>
<td>RES</td>
<td>2616</td>
<td>401.11</td>
<td>0.93</td>
<td>400.44</td>
<td>400.88</td>
<td>401.96</td>
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<tr>
<td>GIF</td>
<td>1888</td>
<td>401.05</td>
<td>0.92</td>
<td>400.94</td>
<td>400.94</td>
<td>401.58</td>
</tr>
<tr>
<td>JUS</td>
<td>1803</td>
<td>401.33</td>
<td>1.17</td>
<td>401.31</td>
<td>401.31</td>
<td>402.04</td>
</tr>
<tr>
<td>PIS</td>
<td>1904</td>
<td>401.62</td>
<td>0.95</td>
<td>401.03</td>
<td>401.59</td>
<td>402.44</td>
</tr>
<tr>
<td>MIT</td>
<td>2207</td>
<td>401.26</td>
<td>1.15</td>
<td>401.11</td>
<td>401.11</td>
<td>401.95</td>
</tr>
</tbody>
</table>

Table 4. Statistics of observed XCO₂ 1-minute averages for all sites