

1 **Reviewer 1**
2

3 My overall rating for this paper is “minor revisions”; there are a few places where the text should be
4 clarified. They are however important clarifications – the authors’ title implies that a 4km resolution
5 CTM and 0.11 degree/12.2km resolution driving meteorological model (or even a 4km resolution
6 meteorological model) captures urban meteorology and chemistry. This may not be the case, from
7 some recent work they reference. I wouldn’t say that this invalidates any of the authors’ results, but
8 their conclusions and discussion need to acknowledge that some important meteorological circulation
9 effects may only be captured at finer scales than employed in their study. There are three aspects to
10 this concern, which I outline below, all of which can be addressed by adding appropriate caveats to
11 their text. There are also a number of smaller concerns and spelling/grammar issues which follow,
12 which would clarify and improve the readability of the paper.
13

14 **Larger issues:**
15

16 **(1)** I think it’s important that the authors acknowledge that at least some of these results may be
17 specific to the region they have studied (Ile de France). For example, one of their findings was that
18 model vertical and horizontal resolution (both in the driving meteorology and the resolution of the
19 CTM that made use of that meteorology) had a relatively minor effect on O3. Meanwhile, I’m aware
20 of studies in coastal environments (where the city is located along a large lake or ocean coastline)
21 where the resolution can have a profound effect on the ability of the meteorological model to capture
22 the land-sea (or land-lake) breeze, and that in turn has a significant impact on ozone predictions.
23 Similarly, though the authors note themselves (second sentence of section 2.1) that the region under
24 study is far from the coast and has limited topography, these factors if present could significantly
25 increase the relative impact of horizontal and vertical resolution. In that respect, some caveats should
26 be placed within the text acknowledging the potential limitations of their study towards similar
27 resolution studies in other domains (e.g. in the Abstract, page 4768, line 17, change “are of little
28 effect.” To “are of little effect for the regional/urban domain and models studied here.”
29

30 **Response:** We acknowledge the fact that some of our results related to model resolution are
31 representative to the specific characteristics of our study area. The abstract was revised according to
32 the reviewer’s suggestion: “*The air quality model horizontal and vertical resolution have little effect*
33 *on model predictions for the specific study domain.*” We also insert the following text on page. 4784,
34 line 27: “*Never the less this result could reflect the local area’s characteristics (flat terrain, away*
35 *from the coast) confirming previous studies (Menut et al., 2005; Valari and Menut, 2008). In regions*
36 *with more complex topography or close to the coast the resolution of the meteorological input could*
37 *have a profound effect on the simulated meteorological conditions (Leroy et al., 2014).”*
38

39 **(2)** The authors also need to discuss the accuracy of their meteorological model in the context of
40 resolution – is the lack of a resolution impact of the meteorological model on chemical predictions
41 reflective of resolution not being important (the authors’ conclusion, I think) or reflective of a
42 meteorological model whose performance has not improved in going to higher resolutions? They
43 later (page 4776, lines 18-19) reference previous work that shows that the flat homogeneous
44 topography of the region results in little benefit in going to higher resolution. But what was the
45 highest resolution attempted in the referenced work by Menut et al, 2005, and Valari and Menut,
46 2008)? The highest resolution of the input meteorology used in the authors’ current paper is 0.11
47 degrees or about 12.2km – this is inappropriately coarse to resolve much of the urban circulation (cf.
48 also line 4782, line 2 regarding the CTM itself – I agree, this is very likely, so caveats in the abstract
49 and conclusions need to be added to that effect). In the Leoyer et al paper quoted by the authors,

1 substantial changes in vertical and horizontal transport in an urban environment for a meteorological
2 model occurred mostly in the transition from 2.5km to 1km to even higher horizontal resolutions
3 (e.g., 250m). The resolution impact in the IdF region could thus be substantial, but just not yet seen
4 on the still relatively coarse resolutions used by the authors in this set of tests. 12.2 km meteorology
5 inputted into a 4km CTM won't see much if anything of the urban circulation, and a 4km resolution
6 met model might not see much of that circulation either. The authors' conclusion should be modified
7 to include caveats to that effect, with the message going from, e.g., "horizontal and vertical
8 resolution are not important" to "input horizontal and vertical resolution and CTM resolution were
9 not important for the regional/urban domains and model resolutions studied here. The relative
10 importance of resolution may however increase with further reductions in grid size in both the
11 driving meteorological model (c.f. Leroyer et al, 2014), and the CTM making use of that
12 meteorology."

13
14 **Response:** We agree. The increase of the meteorological model's resolution from 0.5° to 0.1° might
15 not be enough for the chemical model to produce noticeable concentration responses. In Menut et al.,
16 2005 the increase is from 0.5° to 3km and in Valari and Menut, 2008 from 0.5° to 5km. Based on the
17 Leroyer et al., 2014 results this might not be sufficient. Never the less according to Leroyer et al.
18 2014: "*The necessity of using subkilometer meteorological numerical systems remains, however,*
19 *questionable. The improvement in forecasting obtained by going to subkilometer modeling systems,*
20 *with the use of a detailed surface description and of a more physically based (less parameterized)*
21 *representation of atmospheric processes, still has to be rigorously demonstrated".* Despite that it is
22 useful to include such discussion in the text e.g. page. 4784, line 27: "*We note here that the*
23 *refinement in the resolution of the meteorological model from 0.5° to 0.1° may not be sufficient for*
24 *the CTM to simulate noticeable concentration responses. For example Leroyer et al. (2014) (see also*
25 *references therein) observed that substantial changes in vertical and horizontal transport in an*
26 *urban environment occurred mostly in the transition from resolutions of 2.5km to 1km and even*
27 *higher (250m)."* We also revise a sentence in the conclusions section (page 4795, line 4): "*We also*
28 *note the weak sensitivity of modeled concentrations to the increase in the CTM's and the*
29 *meteorological model's horizontal resolution at least for the area and the range of resolutions*
30 *studied here."*

31
32 **(3)** Page 4785, section 4.3. I'm assuming here that this sensitivity test looks at the vertical resolution
33 of the CTM, but there is no corresponding increase in vertical resolution of the driving
34 meteorological data (this should be clarified - mentioned in the first sentence)? This needs to be
35 explicitly made clear in the text. This is another case where the impact of "vertical resolution" needs
36 to be split conceptually in the text between "vertical resolution of the driving meteorology" and
37 "vertical resolution of the CTM". My point here is that a meteorological model with a higher vertical
38 resolution will almost certainly generate different vertical diffusion coefficients, than one with a
39 lower vertical resolution. Increasing the resolution of the CTM may not capture this effect. A caveat
40 to that effect should be in the conclusions and the abstract with regards to the vertical resolution
41 issue. The authors should avoid the use of phrasing like "of the model's vertical resolution" and
42 instead be using "of the CTM's vertical resolution" or "of the driving meteorological model's
43 vertical resolution".

44
45 **Response:** WRF model runs on a 31 vertical layer grid, which we consider highly resolved, this is
46 now added in the modeling setup description. Studying the effect of the vertical resolution of the
47 meteorological model could have been a separate sensitivity test which we didn't chose to study. In
48 any case the WRF meteorology is interpolated to the CTM's (i.e. CHIMERE) vertical grid therefore,
49 technically, increasing the number of vertical layers in CHIMERE from 8 to 12 will result in a

1 refinement of the meteorological input used for the chemical simulation as well. This was added in
2 section 2.4 (description of the sensitivity simulations). With this clarification the phrase added in the
3 abstract (“*The air quality model horizontal and vertical resolution have little effect on model*
4 *predictions for the specific study domain.*”) inevitable covers this -indirect- refinement of the vertical
5 resolution of meteorology. Finally, we changed the ambiguous term ‘model vertical resolution’ into
6 ‘CTM’s vertical resolution’ throughout the manuscript as suggested by the reviewer.

7 8 **Minor issues:**

9
10 **Abstract and conclusions:** needs to be clarified with regards to the resolution of the input
11 meteorology versus the resolution of the CTM used for the modelling.

12
13 **Response:** These additions were implemented, please refer to the responses in the “major issues”
14 section.

15
16 **Page 4768, line 14:** is that supposed to be “meteorological model input resolution”?

17
18 **Response:** No this is the CTM’s horizontal resolution. We have revised the issue in the manuscript.

19
20 **Page 4769, line 20:** another reference here: Kelly et al, 2012, 5367-5390, 2012.

21
22 **Response:** Added.

23
24 **Page 4770, line 19 to line 24:** the sentence “In Markakis (2014)...NO_x-limited conditions.” Could
25 use a little clarification: presumably each didn’t work for the other condition, as well?

26
27 **Response:** We rephrased our sentence to make our point clearer. The sentence now reads: “*In*
28 *Markakis et al. (2014) we showed that ozone formation occurs under a VOC-limited chemical regime*
29 *in the 10-year simulations that used the bottom-up emission inventory. This result is consistent with*
30 *previous studies over the Paris area (Beekman and Derognat, 2003; Beekman and Vautard, 2010;*
31 *Deguillaume et al., 2008). On the contrary, when the regional top-down inventory was used instead,*
32 *ozone formation occurred under a NO_x-limited chemical regime.*”

33
34 **Page 4771, first line:** using RCP6, Kelly et al did this – a laborious process of linking the RCP
35 recommendations with specific industries.

36
37 **Response:** This is indeed an oversight from our part. We revised the manuscript: “*Long-term*
38 *projections are constrained by the evolution of large scale energy supply and demand and the link*
39 *between global and regional scale projections is a laborious task (Kelly et al. (2012)).*”

40
41 **Page 4772, line 13:** The authors should explain why this particular RCP was used in their work,
42 rather than the other RCP scenarios available.

43
44 **Response:** This was an inevitable decision related to the availability of simulations on the larger
45 scale that were used to provide the boundaries to our simulations.

46
47 **Page 4774, line 1:** might be worth mentioning here that the database in question does not explicitly
48 consider point sources.

1 **Response:** We revise “Present-time emissions (as areas sources) are compiled...”

2
3 **Page 4775, line 25:** not clear why the same metrics were not being used for both O3 and PM2.5.
4 Explain.

5
6 **Response:** We use MFB and MFE which are considered more suitable for fine particles evaluation
7 based on EPA guidelines. In Page 4775, lines 6-9 we already include the EPA, 2007 reference.

8
9 **Page 4788, lines 24-25:** This was rather a surprise to me, though I may be used to more detailed
10 emissions inventories in North America. When you say “no major point sources can be found within
11 the urban area” does this mean that “no major point sources exist within the urban area of Paris” or
12 “no major point sources are explicitly included within this inventory within the urban area of Paris”?
13 In contrast, North American emissions inventories in Canada and the USA include tens of thousands
14 of point sources (to the extent that one has to choose criteria for deciding which ones will be selected
15 for plume rise calculations and which have minor enough emissions to be treated as area sources). So
16 a line or two explaining whether the issue here is a lack of data in this inventory for this region, a
17 lack of data in any inventory for this region, or if there really are no major point sources in this large
18 city.

19
20 **Response:** The local emission inventory includes tenths of point source (Page 4774, line 13-15) but
21 the major of these sources are in industrial areas outside the city. We revise: “*Following the*
22 *AIRPARIF post-processing (ANN) all urban emissions are released in the surface layer because*
23 *according to the local point source emission database no major industrial units are found within the*
24 *urban area.*”

25
26 **Page 4789, lines 10-13 and section 4.5 in general.** You might also be interested in having a look at
27 Makar et al, Geoscientific Model Development, 7, 1001-1002, 2014, since there are several points of
28 overlap between the authors’ paper and that one: the reference looks at several stages of emissions
29 improvement and how two different off-line CTMs responded to those changes. Both temporal and
30 spatial changes in emissions were evaluated and the impacts on O3 and PM2.5 predictions evaluated.

31
32 **Response:** The paper is very interesting and we have added a discussion in Page 4778, line 13:
33 “*Makar et al. (2014) investigated the response of modeled concentrations to the refinement of the*
34 *spatial and temporal allocation of input emissions and found that the model was as sensitive to these*
35 *improvements as to the vertical mixing parameterization. Also they conclude that the temporal*
36 *distribution of emissions in particular, could be very important in stable urban atmospheres and that*
37 *this sensitivity is reduced with increased mixing conditions.*” We also add at the end of section 4.5:
38 “*We note here, that recent work has pointed out that the sensitivity of modeled concentrations the*
39 *spatiotemporal resolution of the emission inventory is model-dependent (Makar et al., 2014).*”

40
41 **Page 4791, line 15-16;** see the above discussion; I think that this last sentence currently ending
42 “especially taking into account the large increase of model resolution from 50 to 4km” should be “for
43 the range of meteorological and CTM horizontal resolutions attempted here. A stronger impact of
44 resolution may occur at even higher resolutions (c.f. Leroyer et al, 2014).”

45
46 **Response:** This sentence is under the “effect of CTM’s resolution” section therefore we revise the
47 statement according to that feature alone e.g.: “*We may conclude that the benefit of increasing the*
48 *CTM’s resolution is insignificant for both ozone and PM_{2.5} especially taking into account the large*
49 *refinement attempted here (0.5° to 4km).*” The reviewer’s suggestion as to the possible effect of

1 refining the meteorological model resolution is already added in the text in accordance to the
2 comment number 2 of the “major issues” section.

3
4 **Page 4792, lines 25 through 27** seems to be saying “getting annual emissions totals right has a big
5 impact on O3 results” while lines 12 through 14 seem to be saying “getting the annual emissions
6 totals right has a very minimal impact on O3 results”. This needs to be clarified.

7
8 **Response:** The statement in lines 12-14 regards only urban ozone. This is clarified: “*Considering the*
9 *discrepancies in the inventorying methodologies used to compile the ECLIPSE and the AIRPARIF*
10 *datasets (top-down vs. bottom-up), it is very interesting that the least influential factor to the urban*
11 *ozone responses is the annual emissions totals.*”

12
13 **Page 4793, lines 10 through 12:** the authors may wish to consider and discuss the potential for
14 compensating errors in this regard. That is, (1) the PM2.5 levels in the urban regions are likely
15 mostly controlled by primary emissions; (2) increasing the emissions inventory resolution will
16 defacto concentrate the PM2.5 emissions into a smaller spatial extent of the urban area (the reverse
17 side of the artificial dilution issue that the authors have already discussed); (3) if the emissions totals
18 are themselves biased high, then the resulting error will only become apparent at higher resolution.
19 That is, the conclusion should not necessarily be “the emissions resolution makes the PM2.5 worse”,
20 but “the emissions resolution may be showing us that the emissions totals are too high, and this only
21 becomes apparent at high resolutions”.

22
23 **Response:** “*the emissions resolution makes the PM2.5 worse*”: this is by no means the message we
24 intent to deliver. Table 6 only shows how the REG application would respond if selectively
25 incorporated features of the local application. The message is: “*if the REG application uses the*
26 *coarse inventory totals along with higher resolution modeling and refined spatial allocation of*
27 *emissions this would result in a high overestimation of concentrations*” and that “*the REG*
28 *application has to adopt all emission-related features of local scale to improve model scores*”. We
29 have added a discussion on possible error compensation in the modeling exercise presented in the
30 paper as suggested by both reviewers (see also the 2nd reviewer’s 1st comment).

31
32 **Minor spelling/grammar mistakes to be corrected:**

33
34 **Response:** Spelling mistakes were corrected in the text.

35
36 Page 4768, first sentence of abstract starts with a preposition, better to use “Previous research helped
37 to....spatial scale effect, but our knowledge is limited....”

38
39 Page 4768, line 21: “(same improvement)” should be “(the same improvement)”

40
41 Page 4768, line 23,24: “bias on” should be “bias of”, and “associated to” should be “associated with”

42
43 Page 4769, line 2: “at urban scale” should be “at the urban scale”, ditto, line 24.

44
45 Page 4770, line 2: “By principle” should be “In principle”.

46
47 Page 4770, line 27: “to higher” should be “in higher”. Reductions of what? O3?

48
49 Page 4771, lines 3 to 5: sentence is unclear.

1
2 Page 4771, line 7: “demands and that” should be “demand. This“
3
4 Page 4771, last line: “f)” should be “(f)”, ditto for “g)” should be “(g)” on the next page.
5
6 Page 4773, line 5: “sulfates, nitrates” should be “sulfate, nitrate”.
7
8 Page 4773, line 14: “vertical” misspelled.
9
10 Page 4773, line 19: “gasses” should be “gases”, I think. Might be British versus American spelling
11 conventions, here.
12
13 Page 4774, line 5: “are available” should be “are also available”
14
15 Page 4776, line 9: “speed was the” should be “speed were the”
16
17 Page 4777, line 9-10: “due to the surface emissions, ozone concentrations in the afternoon peak hour
18 had the second largest sensitivity after meteorology.” Should be “the sensitivity of ozone
19 concentrations in the afternoon peak hour was the second largest after the sensitivity associated with
20 meteorology.”
21
22 Page 4778, line 6: “scale” should be “scales”
23
24 Page 4784, line 3: “to modeled precipitation” should be “to the accuracy of modelled precipitation.”
25
26 Page 4784, line 20: “of the meteorological grid” should be “of the input meteorological grid” (I
27 think).
28
29 Page 4787, line 5: “is in the “ should be “is on the”.
30
31 Page 4790, line 6: “model processes” should be “model process”. Line 18: “Fine particles” should be
32 “Fine particle”
33
34 Page 4791, line 8: “is very little sensitive” should be “has relatively low sensitivity” Page 4791, line
35 24: “same source” should be “same sources”
36
37 Page 4792, line 5: end the sentence with a question mark and put quotations around “what are the
38 main...or at least reduced?”
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1 **Reviewer 2**

2
3 In this study, Markakis et al., conduct a series of sensitivity calculations using WRF and CHIMERE
4 to downscale global inputs (IPSL-CM5A-MR and LMDz-INCA) to 0.44 and 0.11 degrees (for the
5 met) and 4 km (for air quality). The sensitivity calculations conducted are used to assess the response
6 of the modeling system to various different ways of processing inputs (e.g., resolution, which
7 emissions, climate/reanalysis inputs). For those involved in such calculations the manuscript provides
8 useful information.
9

10 **General Comments:**

11
12 **(1)** One issue I found throughout the manuscript is the struggle for what is the appropriate way to
13 evaluate models when conducting such an exercise. There is always a temptation to suggest that the
14 close the model results are to the observations, the better, which is typically true. However, then one
15 has to decide upon a set of metrics, of which there are many. It is not always the case that finer
16 resolution is better (for numerical and other reasons. . . e.g., potential spatial misalignment in inputs).
17 Further, there is the issue of compensatory errors. Thus, one is left with the question of what should
18 be the standard for comparison, and what is the “best” result. This manuscript does well at showing
19 sensitivities, but also discusses bias, with the implicit assumption that a smaller bias means that the
20 simulation is better. However, it could just be that it does a better job at having errors compensate for
21 each other. The manuscript should better deal with this issue.
22

23 **Response:**

24 We added a discussion on possible error compensation on the conclusion section. The text added
25 there reads: “*We note here that PM2.5 levels in the urban regions are likely mostly controlled by*
26 *primary emissions; increasing the emissions inventory resolution will concentrate the PM2.5*
27 *emissions into a smaller spatial extent of the urban area (the reverse side of the artificial dilution*
28 *issue taking place at coarse resolution); if the emissions totals are themselves biased high, then the*
29 *resulting error will only become apparent at higher resolution. Therefore, the emissions resolution*
30 *may be showing that the emissions totals are too high, and this only becomes apparent at high*
31 *resolutions.*”
32

33 **(2)** They should also consider doing a dynamic analysis of the model responses (e.g., the work by
34 Dennis et al., at the US EPA and as part of AQMEII).
35

36 **Response:** We quote the AQMEII Dennis et al. paper: “*This exercise requires historical case studies*
37 *where known emission changes or meteorological changes occurred that could be confidently*
38 *estimated. Dynamic evaluation also requires that these changes have a discernable impact on air*
39 *quality*”. To our best of knowledge we are not familiar with any such studies in IdF that would fit the
40 prerequisites of the dynamical evaluation process.
41

42 **(3)** The authors would also serve the community by digging in to the results of the multiple
43 simulations (and possibly conducting some additional simulations), to provide a more general
44 understanding of the spatio-temporal patterns of model responses. At present, they give specific
45 results for their set up, which may be all they are limited to at present. It would be great if they could
46 better say “our results show that, in general, model resolution will tend to have the following effects
47 on model results: (list of effects, with some indication of spatio-temporal trends). This will likely
48 require looking at distributions of model results. Indeed, as demonstrated by the work from Harvard,
49 when looking at air quality and climate impacts, distributions of air quality responses are very

1 informative (e.g., Wu et al., (2008) JGR, DOI: 10.1029/2007JD008917). This is done, to some
2 degree in Fig. 7, but that does not give a spatio-temporal understanding.

3
4 **Response:** As a general comment the authors would like to note that at this point we do not have the
5 necessary computing power to perform additional simulations. Moreover we understand that the
6 reviewer asks for a finer representation of spatiotemporal variation of the trends. It was in fact our
7 initial decision not to provide too refined information regarding the temporal and spatial patterns of
8 the sensitivities because that would produce a large number of contradicting messages amongst
9 various seasons/areas of the domain. This stems from the fact that (and this is the originality of this
10 work) we discuss results on a local high-resolution domain with very large spatial and temporal
11 gradients of e.g., emissions. We have anticipated that the potential reader of the paper would seek a
12 better understanding of more coarse spatiotemporal surrogates such as the ones we provide: urban,
13 suburban, rural and summer (ozone period for ozone), winter, annual. Regarding the list of effects
14 from the various sensitivities presented, our results reveal a consistent pattern in the trends of
15 modeled ozone (decrease) when refined information is implemented. In contrast for fine particles this
16 is not observed. In the initial version of the manuscript this was not clearly discussed. In the updated
17 version the following discussion has been added to the conclusions section (page 4795, line 4):

18
19 *“Excluding the sensitivities having the smallest impact (roughly less than 2%, see Table 3) we
20 observe a very consistent trend in ozone concentration: daily average and maximum ozone decrease
21 as input data become more refined, namely passing from climate meteorology to reanalysis,
22 increasing the resolutions of the horizontal and vertical CTM grid, of meteorology, of emissions and
23 by using bottom-up emissions and post-processing instead of top-down. This decrease in ozone
24 concentrations, from 2.5% up to 8.3%, is observed mainly in the urban and suburban areas and in
25 all cases stems from enhanced NO_x emission fluxes in the surface-layer leading to titration
26 inhibition. Trends and the underlying changes in emissions are highly variable for PM_{2.5} with
27 increase in PM_{2.5} concentrations that may be as low as 2% or as high as 30% for climate
28 meteorology and resolution of the vertical mesh, meteorology and emissions and also cases where
29 concentration decreases in a wide range of values from 3% up to 34% (annual emissions, model
30 resolution) depending on the season.”*

31
32 We have also added in the abstract: *“In the case of modelled ozone concentrations, the
33 implementation of refined input data results in a consistent decrease (from 2.5% up to 8.3%), mainly
34 due to inhibition of the titration rate by nitrogen oxides. Such consistency is not observed for
35 PM_{2.5}.”*

36
37 (4) One of the most important questions is not addressed by this manuscript: that is, how do these
38 changes (resolution, emissions processing, meteorology) impact how the model responds to
39 emissions changes, e.g., how ozone, PM and NO₂ respond to NO_x, VOC and SO₂ emissions
40 changes.

41
42 **Response:** Actually, we think that this precisely what our paper is all about. With all the sensitivity
43 studies we carried out we try to understand in what way (i.e. by which physical or chemical process)
44 ozone and PM modeling is affected. We will agree though that in some places this is not clear. We
45 have revised the manuscript and added relevant pieces in each sensitivity section. We list here the
46 more striking examples from the manuscript to illustrate this:

47
48 From the sensitivity to vertical resolution:

1 *“Interestingly, the impact of the refinement of the vertical grid on daily averaged O_3 is much*
2 *stronger than on ozone: O_3 , changes by 0.9ppb in the urban and suburban areas. The change in O_3*
3 *is reasonable since in VERT, NO_x emissions are released within a surface layer thinner by 60%*
4 *compared to REF (from 20m to 8m) leading to higher NO_x concentrations.”*

5
6 From the sensitivity to the annual emission totals:

7
8 *“Changes in modeled urban daily average ozone concentrations are small ($|\Delta c| = 0.8\text{ppb}$ or 2.5%)*
9 *with the regional inventory (ECLIPSE annual totals) to tend to increase the bias of the REF run*
10 *(Fig.8a and Table 3). This is due to the fact that when passing from the AIRPARIF to the ECLIPSE*
11 *inventory (see also Fig. 2) NO_x emissions decrease (weakening titration) and NMVOCs increase*
12 *(intensifying production).”*

13
14 From the sensitivity to emission post-processing:

15
16 *“Modeled $PM_{2.5}$ sensitivity is significant for both summer and wintertime ($|\Delta c| = 3.4\mu\text{g}/\text{m}^3$ or 24.8%*
17 *and $4.6\mu\text{g}/\text{m}^3$ or 18.3% respectively) (Table 3). POST wintertime bias is almost two times higher*
18 *than ANN (Fig. 9b). This is because the coarse resolution annual post-processing coefficients weight*
19 *towards allocating more of the annual emissions into the winter period significantly influenced by*
20 *the residential sector emissions which are overstated in the ECLIPSE inventory.”*

21
22 Finally in the conclusions we summarize these effects, please refer to the response of comment 3.

23 **Specific Comments:**

24
25
26 **(1)** The base air quality model was run with 8 layers, with a sensitivity run at 12 layers. Many model
27 are now run, as a base, with significantly more layers (e.g., Simon et al., Environ Sci Technol. 2013
28 Mar 5;47(5):2304-13. doi: 10.1021 use 24 layers). It would have been of interest to have an even
29 higher vertical resolution analysis.

30
31 **Response:** We agree with the reviewer that it would have been interesting to investigate an
32 additional vertical distribution case. Our decision was a compromise between the computational
33 demands and insights of previous work conducted in the region. Menut et al., 2013b studied the
34 same effect (although using an 1-day episode) and found that the observed change of implementing a
35 refined vertical mesh with 20 layers never exceeded $3\mu\text{g}/\text{m}^3$ both for ozone and PM_{10} compared to an
36 8-layer configuration. In any case the increase from 8 to 12 layers might appear small by as we state
37 in the paper: *“VERT implements a 12 vertical σ -p layers instead of 8. The major difference between*
38 *the two configurations (REF vs. VERT) is not the number of layers but the depth of the first model*
39 *layer, which is reduced from 20 to 8 m in VERT.”* A 23 layer configuration would refine the vertical
40 mesh in the boundary layer but the surface layer to which the study of concentrations is the main
41 focus of this paper would be of similar depth. This is important because our work suggests that
42 emissions are the key input in our simulations and these are almost exclusively released in the
43 surface layer.

44
45 **(2)** The article is opaque at times, e.g., “As regards $PM_{2.5}$ modeling . . . regional realization cannot
46 selectively incorporate any combination of local scale features. . .” is tough to parse. “By principle”
47 is not standard English. Likewise, many areas still in need of editing for grammar.

1 **Response:** We have corrected the aforesaid grammatical mistakes. We have also carefully inspected
2 the manuscript to identify errors in grammar and we have revised accordingly.

3
4 **(3)** They should make clear what they mean by “top-down” vs. “bottom-up” emissions inventories.
5 Some people might use “top-down” to refer to using observations.

6
7 **Response:** We already define the top-down methodology in the introduction (page 4770, line 12):
8 “Another key issue is the representativeness of top-down emission inventories over cities. The
9 starting point of these inventories is emission annual totals for families of pollutants at continental,
10 regional or national scale that are temporally and spatially downscaled based on proxies such as
11 land-use and population data, activity-dependent time profiles and chemical speciation to provide
12 gridded hourly emission fields suitable for modeling with CTMs.” For the bottom-up definition we
13 add in page 4770, line 19: “In Markakis et al. (2014) we showed that the implementation of bottom-
14 up emissions (e.g., compiled using source specific activity information) in a decade simulation over
15 Paris...”

16
17 **(4)** Also, the discussion of the weaknesses of various emissions inventory approaches predate
18 Markakis 2010, and should be referenced (e.g., look back at the NARSTO reports, as well as
19 Gilliland, JGR, v. 108).

20
21 **Response:** We referenced the following papers:

22
23 Russell, A., and Dennis, R.: NARSTO critical review of photochemical models and modelling.
24 Atmos. Environ., 34, 2261-2282, 2000.

25
26 Gilliland, A.B., Dennis, R.L., Roselle, S.J., and Pierce, T.E.: Seasonal NH₃ emission estimates for
27 the eastern United States based on ammonium wet concentrations and an inverse modeling method,
28 J. Geophys. Res., 108(D15), 4477, doi:10.1029/2002JD003063, 2003.

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Climate forced air-quality modeling at urban scale: sensitivity to model resolution, emissions and meteorology.

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Abstract

While previous research helped to identify and prioritize the sources of error in air-quality modeling due to anthropogenic emissions and spatial scale effects our knowledge is limited on how these uncertainties affect climate forced air-quality assessments. Using as reference a 10yr model simulation over the greater Paris (France) area at 4km resolution and anthropogenic emissions from a 1km resolution bottom-up inventory, through several tests we estimate the sensitivity of modeled ozone and PM_{2.5} concentrations to different potentially influential factors with a particular interest over the urban areas. These factors include the model horizontal and vertical resolution, the meteorological input from a climate model and its resolution, the use of a top-down emission inventory, the resolution of the emissions input and the post-processing coefficients used to derive the temporal, vertical and chemical split of emissions. We show that urban ozone displays moderate sensitivity to the resolution of emissions (~8%), the post-processing method (6.5%) and the horizontal resolution of the [CTM-air quality model](#) (~5%) while annual PM_{2.5} levels are particularly sensitive to changes in their primary emissions (~32%) and the resolution of the emission inventory (~24%). The air quality model horizontal and vertical resolution have little effect on model predictions for the specific study domain. In the case of modelled ozone concentrations, the implementation of refined input data results in a consistent decrease (from 2.5% up to 8.3%), mainly due to inhibition of the titration rate by nitrogen oxides. Such consistency is not observed for PM_{2.5}. In contrast this consistency is not observed for PM_{2.5}. In addition we use the results of these sensitivities to explain and quantify the discrepancy between a coarse (~50km) and a fine (4km) resolution simulation over the urban

1 area. We show that the ozone bias of the coarse run (+9ppb) is reduced by ~40% by adopting a
2 higher resolution emission inventory, by 25% by using a post-processing technique based on the
3 local inventory (same improvement is obtained by increasing model horizontal resolution) and
4 by 10% by adopting the annual emission totals of the local inventory. The bias ~~on~~of PM_{2.5}
5 concentrations follows a more complex pattern with the positive bias-values associated with~~o~~
6 the coarse run (+3.6µg/m³), increasing or decreasing depending on the type of the refinement.
7 We conclude that in the case of fine particles the coarse simulation cannot selectively incorporate
8 local scale features in order to reduce ~~model-its~~ error.

10 **1 Introduction**

11 Recent epidemiological findings stress the need to resolve the variability of pollutant
12 concentrations at urban scale. The International Agency for Research on Cancer recently
13 classified outdoor air pollution as a “leading environmental cause of cancer deaths” (Loomis et
14 al., 2013) while new findings reveal that living near busy roads substantially increases the total
15 burden of disease attributable to air pollution (Pascal et al., 2013). Research on future projections
16 of air-quality should be addressed primarily at such scale especially given the fact that the efforts
17 to mitigate air-pollution are more intense in areas where the largest health benefits are observed
18 (Riahi et al., 2011).

19 Climate and atmospheric composition are related through a series of physical and chemical
20 mechanisms and atmospheric feedbacks. A significant portion of the published literature on this
21 issue uses global scale models to focus on the impact of climate on tropospheric ozone at global
22 or regional scales (Brasseur et al., 1998; Liao et al., 2006; Prather et al., 2003; Szopa et al., 2006;
23 Szopa and Hauglustaine, 2007). More recent studies have integrated advanced chemistry
24 schemes capable of resolving the variability of pollutant concentrations at regional scale, which
25 spans from several hours up to a few days, with chemistry transport models (CTMs) (Colette et
26 al., 2012, 2013; Forkel and Knoche, 2006, 2007; Hogrefe et al., 2004; Katragkou et al., 2011;
27 Kelly et al., 2012; Knowlton et al., 2004; Lam et al., 2011; Langner et al., 2005, 2012; Nolte et
28 al., 2008; Szopa and Hauglustaine, 2007; Tagaris et al., 2009, Zanis et al., 2011). Global models
29 with a typical resolution of a few hundreds of kilometers and regional CTMs used at resolutions
30 of a few tens of kilometers – and their parameterization of physical and chemical processes make
31 them inadequate for modeling air-quality at urban scale (Cohan et al., 2006; Forkel and Knoche,

1 2007; Markakis et al., 2014; Sillman et al., 1990; Tie et al., 2010; Valari and Menut, 2008; Valin
2 et al., 2011; Vautard et al., 2007).

3 The challenge we face is how to model climate forced atmospheric composition with CTMs at
4 fine resolution over urban areas, where emission gradients are particularly sharp, without
5 introducing large errors due to emissions² and meteorology related uncertainties as well as to
6 CTMs numerical resolution. In the absence of plume--in--grid parameterization, emissions in
7 CTMs are instantly mixed within the volume of model grid-cells before chemical reaction
8 transport and mixing take place. When the volume of these cells is large compared to the
9 characteristic time scale of these processes, sub-grid scale errors occur such as over-dilution of
10 emissions leading to unrealistic representation of urban scale chemistry such as ozone titration.
11 The resolution of meteorological modeling is another issue: Leroyer et al. (2014) argue that only
12 high-resolution meteorological modeling can correctly capture the urban heat island, also Flagg
13 and Taylor (2011) showed that high-resolution modeling is very much dependent on the
14 resolution of the surface layer input data.

15 Another key issue is the representativeness of top-down emission inventories over cities. The
16 starting point of these inventories is ~~emission~~ annual totals for families of pollutants at
17 continental, regional or national scale that are temporally and spatially downscaled based on
18 proxies such as land-use and population data, activity-dependent time profiles and chemical
19 speciation to provide gridded hourly emission fields suitable for modeling with CTMs. It has
20 been shown that these inventories cannot adequately portray the plethora and complexity of the
21 anthropogenic emissions over large cities (Gilliland et al., 2003; Markakis et al., 2010, 2012;
22 Russell and Dennis, 2000). In Markakis et al. (2014) we showed that ozone formation occurs
23 under a VOC-limited chemical regime in the 10-year simulations that used the bottom-up
24 emission inventory. This result is consistent with previous studies over the Paris area (Beekmann
25 and Derognat, 2003; Beekmann and Vautard, 2010; Deguillaume et al., 2008). On the contrary,
26 when the regional top-down inventory was used instead, ozone formation occurred under a NO_x-
27 limited chemical regime. Such a discrepancy is critical when mitigation scenarios are
28 investigated because they may lead to controversy when studying the ozone response in the
29 future. As shown in Markakis et al. (2014) regional scale modeling and the use of top-down
30 emissions can result to higher future reductions than the urban scale modeling using bottom-up
31 emissions. Other challenges stem from the fact that emission projections are mostly based on

1 scenarios developed to represent changes at global scale and are rarely suited for assessment at
2 regional let alone urban scales. Long-term projections are constrained by the evolution of large
3 scale energy supply and demand and the link between global and regional scale projections is a
4 laborious task (Kelly et al. (2012)).

5 The major caveat of simulating regional scales at high resolution is the enormous computational
6 demands and that is particularly relevant to climate studies where the simulated periods extend
7 over several decades. To fill the gap between regional and city-scale assessments we need to
8 combine in a single application the advantages of each scale; on one hand the high spatial
9 coverage (but with low resolution) and on the other a good representation of emissions over
10 cities. To achieve this goal we need to understand the major sources of error and their respective
11 impact on climate forced atmospheric composition simulations at urban scale.

12 This study builds on the previous work of Markakis et al. (2014) where a qualitative comparison
13 was accomplished between an urban (local) and a regional scale simulation over Paris. The aim
14 of the present study is to disentangle modeling errors of climate forced air-quality atmospheric
15 composition—studies over finer scales due to different factors such as emission and
16 meteorological input as well as the CTM’s horizontal and vertical resolution. We use as
17 reference run a 10yr long simulation (1996-2005) over the Ile-de-France region in France (IdF)
18 at 4km resolution, using the high-resolution (1km) bottom-up emission inventory of the region’s
19 environmental agency (AIRPARIF, 2012). Boundary conditions for this run are taken from a
20 regional scale simulation at 0.5° over Europe, where the ECLIPSE top-down emissions were
21 used (Klimont et al., 2013, 2015). We carry out several sensitivity tests to quantify the impact of
22 an envelope of effects such as a) meteorology from a climate model versus reanalysis data; b) the
23 spatial resolution of the meteorological input; c) the air-quality model vertical resolution,
24 especially close to the surface; d) bottom-up versus top-down emissions; e) AIRPARIF versus
25 EMEP post-processing information (temporal, vertical and chemical split) of emissions to
26 provide appropriate fluxes on the air-quality modeling mesh grid f) the resolution of the emission
27 input g) the CTM’s horizontal resolution. We aim to point out the most influential parameters of
28 model configuration to help improving regional scale climate change assessments.

29 30 **2 Materials and methods**

31 **2.1 Meteorological and air-quality models’ setup**

1 The IdF region is located at 1.25–3.58° east and 47.89–49.45° north with a population of
2 approximately 11.7 million, more than two million of which live in the city of Paris (Fig. 1). The
3 area is situated away from the coast and is characterized by uniform and low topography, not
4 exceeding 200 m above sea level.

5 In order to simulate air-quality in the study region we employ a dynamical downscaling
6 approach: at first the IPSL-CM5A-MR global circulation model (Dufresne et al., 2013) is used to
7 derive projections of the main climate drivers (temperature, solar radiation etc.) using the RCP-
8 4.5 dataset of greenhouse gas emissions (van Vuuren et al., 2011). Global climate output is
9 downscaled with the Weather Research and Forecasting (WRF) mesoscale climate model
10 (Skamarock and Klemp, 2008) over Europe at a 0.44° horizontal resolution grid (details on these
11 simulations can be found in Kotlarski et al. (2014)). For the purpose of the sensitivities presented
12 in the paper we also employ meteorology driven by ERA reanalysis data at two resolutions; 0.11°
13 and 0.44° (Vautard et al., 2013). The vertical resolution of the meteorological input consists of
14 31 σ -p layer extending to 500hPa.

15 Pollutant concentrations at global scale are modeled with the LMDz-INCA chemistry model
16 (Hauglustaine et al., 2004, 2013) forced with RCP-4.5 emissions. These concentration fields are
17 downscaled at regional scale with the CHIMERE (2013a version) off-line chemistry-transport
18 model (<http://www.lmd.polytechnique.fr/chimere>) in two steps: initially at 0.44° resolution grid
19 (~50 km) over Europe (EEA, 2104) and subsequently at 4km resolution over the IdF region. The
20 nesting scheme is presented in Fig. 1. CHIMERE is a cartesian mesh-grid model including gas-
21 phase, solid-phase and aqueous chemistry, biogenic emissions modeling ~~depending on~~
22 ~~meteorology~~ with the MEGAN model (Guenther et al., 2006), dust emissions (Menut et al.,
23 2005) and resuspension (Vautard et al., 2005). Gas-phase chemistry is based on the MELCHIOR
24 mechanism (Lattuati, 1997) which includes more than 300 reactions of 80 gaseous species. The
25 aerosols model species are sulfates, nitrates, ammonium, organic and black carbon and sea-salt
26 (Bessagnet et al., 2010) and the gas-particle partitioning of the ensemble
27 Sulfate/Nitrate/Ammonium is treated by the ISORROPIA code (Nenes et al., 1998) implemented
28 on-line in CHIMERE. CHIMERE is been benchmarked in the past in a number of model inter-
29 comparison experiments (see Menut et al. (2013a) and references therein).

30 For the reference run at urban scale (hereafter REF), we use the same model setup as in Markakis
31 et al. (2014): the modeling domain has a horizontal resolution of 4 km and consists of 39 grid

1 cells in the west-east direction, 32 grid cells in the north-south direction and 8 σ -p hybrid vertical
2 layers from the surface (999hPa) up to approximately 5.5 km (500hPa) with the surface layer
3 being 25m thick. The configuration of the reference run represents the best compromise between
4 local scale emission data and the high computational demand of a long-term simulation at fine
5 resolution.

6

7 **2.2 Climate and emissions**

8 The RCP-4.5 long-term scenario of greenhouse gases, used as global scale predictor of present-
9 time climate, displays a 20% GHG emission reduction for Europe, constant population at about
10 575 million inhabitants and mid-21st century change in global radiative forcing by 4 W/m²,
11 increasing to 4.5 W/m² by 2065 and stabilizing thereafter. The RCP-4.5 also includes century-
12 long estimates of air pollutant emissions and, including aerosols and was used to drive the
13 ~~global scale~~-LMDz-INCA simulations at the global scale.

14 The regional scale simulations for the present-time (2010) employ an emission database
15 developed in the framework of the ECLIPSE (Evaluating the Climate and Air Quality Impacts of
16 Short-Lived Pollutants) project (Klimont et al., 2013; 2015) implementing emission factors from
17 GAINS (Amann et al., 2011). Present-time emissions (as areas sources) are compiled by the
18 International Institute for Applied Systems Analysis (IIASA) and as regards Europe they include
19 the results of the work undergone in the UNECE Convention on Long-Range Transboundary Air
20 Pollution (CLRTAP). The emission estimates are available at a 0.5° x 0.5° resolution grid.

21 Present-time (2008) emission estimates for the IdF region are also available in hourly basis over
22 a 1km resolution grid. This emission inventory is compiled by the Ile-de-France environmental
23 agency and combines a large quantity of city-specific information (AIRPARIF, 2012) based on a
24 bottom-up approach. The spatial allocation of emissions is either source specific (e.g., locations
25 of point sources) or completed with proxies such as high-resolution population maps and a
26 detailed road network. The inventory includes emissions of CO, NO_x, Non-methane Volatile
27 Organic Compounds (NMVOCs), SO₂, PM₁₀ and PM_{2.5} with a monthly, weekly and diurnal -
28 source specific- temporal resolution. Emissions from point sources are inputted as area emissions
29 in the model and the grid cells containing those sources adopt a vertical distribution across model
30 layers which varies in time—dependent ~~from~~ several meteorological variables such as
31 temperature and wind inputted in a plume-rise algorithm (Scire et al., 1990). Consequently the

1 distribution of emissions among different activity sectors reveals that in the IdF region the
2 principal emitter of NO_x, on annual basis, is the road transport sector (50%), for NMVOCs the
3 use of solvents (50%) and for fine particles the residential sector (37%). The raw data of the 1km
4 resolution emissions were aggregated to the 4km resolution modeling grid.

6 **2.3 Data and metrics for model evaluation**

7 Model results from the different sensitivity runs are compared against observational data for O₃,
8 NO, NO₂ and PM_{2.5}. Pollutant concentrations measured at 29 sites of the air-quality network of
9 AIRPARIF (17 urban, 4 suburban and 8 rural) are compared to first-layer modeled
10 concentrations on the grid-cells containing the corresponding monitor sites. To benchmark model
11 performance we use the skill score S which is based on the equations of Mao et al. (2006):

$$13 \quad S = \frac{1}{2} \left(1 - \left| \frac{BIAS}{MGE} \right| + \left| \frac{MGE}{RMSE} \right| \right) \quad (1)$$

14
15 where MGE represents the absolute mean gross error and RMSE the root mean square error. A
16 skill score close to 1 is indicative of an unbiased model with no significant errors present, but in
17 the case of biased results this rating masks the information on the magnitude of the bias and the
18 corresponding error. For this reason, alongside S , we employ the mean normalized bias (MNB)
19 and mean normalized gross error (MNGE) as regards ozone evaluation and the Mean Fractional
20 Bias (MFB) and Mean Fractional Error (MFE) as regards PM_{2.5} (EPA, 2007).

21 We extract these metrics from the daily concentration values and not the decade average bearing
22 in mind that this is not typical for runs forced by climate simulations but for operational forecast
23 evaluation. We should note here, that it is reasonable to expect lower scores than those achieved
24 in operational forecast analysis due to the presence of climate biases (Colette et al., 2013; Menut
25 et al., 2013a). As in Markakis et al. (2014) we aim to evaluate our simulations by utilizing
26 metrics that are time averaged on a scale finer than a climatological one.

28 **2.4 Description of the sensitivity simulations**

29 Through a number of test cases we study the ability of the model ~~in-to predict~~ predicting present-
30 time decadal air-quality with respect to emission and meteorological input as well as the CTM's
31 horizontal and vertical resolution. For that purpose we conduct five sets of 10yr long simulations

1 (1996-2005) over a 4km resolution grid covering the IdF region (see Table 1). In all our
2 comparisons we use as a measure of sensitivity of modeled ozone and PM_{2.5} the absolute
3 difference between the mean of daily averaged concentrations ($|\Delta c|$) as well as the absolute
4 change in the skill score S . For ozone we also compare the MNB, MNGE and for PM_{2.5} the MFB
5 and MFE. All scores are calculated to represent an average of all urban, suburban or rural
6 stations. For PM_{2.5} for which only observations from urban stations are available we represent
7 the results for summer, winter and in annual basis ~~for the~~of urban stations.

8 The first sensitivity case focuses on the climate bias due to the meteorological forcing. It is well
9 established that ozone and certain particulate matter species are sensitive to temperature changes
10 (Fiore et al., 2012; Im et al., 2011, 2012; Jacob and Winner, 2009; Megaritis et al., 2014). Menut
11 et al. (2003) using an adjoint model studied the sensitivity of ozone concentrations at the
12 afternoon peak to numerous model processes and inputs for a typical summer episode in Paris
13 and found that temperature and wind speed were the most influential parameters to the observed
14 changes. For our test we utilize meteorological input that stems from a WRF run employing
15 ERA40 reanalysis data over a 0.44° resolution regional scale grid (ERA05) and compare with the
16 REF simulation utilizing climate model meteorology. Both configurations share identical
17 emission inventories (AIRPARIF) and vertical resolution (8 σ -p layers). Modeled meteorological
18 fields are further interpolated over the 4km-resolution IdF grid for the air-quality simulation. We
19 note here, that interpolating the 0.44° resolution meteorology over the 4km resolution CHIMERE
20 grid adds a source of uncertainty in modeled pollutant concentrations, but due to the flat
21 topography of the area and as shown in previous research studies in the same region, increasing
22 the resolution of the meteorological input does not improve model performance (Menut et al.,
23 2005; Valari and Menut, 2008). To study the impact of the resolution of the input meteorology
24 here, we conduct a second sensitivity run where meteorological input stems from a WRF
25 simulation using ERA40 reanalysis data over a finer resolution mesh with grid spacing of 0.11°
26 (ERA01) and compare with the ERA05 run.

27 The third sensitivity test addresses the issue of the CTM's vertical resolution (VERT). A
28 previous sensitivity analysis conducted with the same air-quality model showed only small
29 changes in modeled ozone and PM₁₀ concentrations over the IdF region due to increase in the
30 CTM's vertical resolution (Menut et al., 2013b). On the other hand Menut et al. (2003) showed
31 that vertical diffusivity was one of the most influential parameters to the observed daily peak

1 concentrations of ozone for a typical summertime episode in IdF. Here, we undertake a similar
2 analysis but in a climate modeling framework, where enhanced meteorological bias is expected.
3 VERT implements a 12 vertical σ -p layers instead of 8. The major difference between the two
4 configurations (REF vs. VERT) is not the number of layers but the depth of the first model layer,
5 which is reduced from 20 to 8 m in VERT. We note that because the WRF meteorology
6 (resolved in 31 layers) is interpolated to the CTM's vertical grid, technically, increasing the
7 number of vertical layers in CHIMERE from 8 to 12 will result in a refinement of the
8 meteorological input used for the chemical simulations as well.

9 The fourth sensitivity case estimates the discrepancy in modeled ozone and PM_{2.5} concentrations
10 between two runs where emission totals stem from different inventories, namely the local
11 AIRPARIF inventory and the ECLIPSE regional-scale dataset. In Menut et al. (2003) it was
12 shown that the sensitivity of ozone concentrations in the afternoon peak hour due to surface
13 emissions was the second largest after the sensitivity associated with meteorology. In Markakis
14 et al. (2014) we compared the two approaches as for their ability to correctly represent ozone
15 photo-chemical production under typical anticyclonic summer conditions and also found
16 important differences. In the present work we push the analysis a step further and quantify model
17 response to the emission input over longer timescales. For this purpose we compile a new 4km
18 resolution emission dataset over the IdF domain (ANN) in which annual emission fluxes match
19 the ECLIPSE emissions (0.5° resolution) but are downscaled spatially and temporally to obtain
20 4km-resolution and hourly emissions based on the local scale information implemented in the
21 bottom-up approach of the AIRPARIF emission inventory. The same approach is applied on the
22 chemical speciation of the inventory's pollutants to obtain emissions for all the species required
23 ~~by the CTM's chemical mechanism for the air quality simulation chemical mechanism~~. Therefore
24 the only difference amongst the two runs stem from the use of different annual quantified
25 emission fluxes for the region (Table 1). To give a sense of the discrepancies between the two
26 inventories over ~~the IdF region~~ we compare the annual domain-wide fluxes of NO_x, NMVOCs
27 and PM_{2.5} (Fig. 2). NMVOCs emissions are considerably higher in the ECLIPSE inventory while
28 NO_x emissions are lower than AIRPARIF. In terms of photochemical ozone production, this
29 makes ~~the ECLIPSE inventory~~ more favourable of NO_x-limited conditions than the bottom-up
30 AIRPARIF inventory, which is consistent with the findings of Markakis et al. (2014). Fine
31 particles emissions are 2.4 times more in ECLIPSE, which probably stems from the use of a

1 population proxy to spatially allocate wintertime emissions from wood-burning. We note here,
2 that the interest of comparing the two emission inventories is strictly to quantify the added value
3 of implementing local scale information in city-scale climate studies and not by any means to
4 compare qualitatively the two datasets. It ~~is~~ should be made clear that ECLIPSE dataset is not
5 meant to accurately represent emissions at such fine scales.

6 In the fifth sensitivity case we study the impact of the post-processing methodology e.g., the
7 process followed in order to split the annual emission totals into hourly emission fluxes for all
8 the species and vertical layers required by the air-quality model. Menut et al. (2012a) showed
9 that model performance improves when time-variation profiles developed on the basis of
10 observations are applied for the temporal allocation of emissions instead of the EMEP
11 coefficients. Mailler et al. (2013) found that model results are highly sensitive to the coefficients
12 used for the vertical distribution of emissions. Makar et al. (2014) investigated the response of
13 modeled concentrations to the refinement of the spatial and temporal allocation of input
14 emissions and found that the model was as sensitive to these improvements as to the vertical
15 mixing parameterization. Also they conclude that the temporal distribution of emissions in
16 particular, could be very important in stable urban atmospheres and that this sensitivity is
17 reduced with increased mixing conditions. For ~~this-our~~ test emission totals must match between
18 the two emission datasets. We compile a new emission dataset (POST) where the ECLIPSE
19 annual totals are spatially (both horizontally and vertically) and temporally downscaled on the
20 4km-resolution IdF grid. This procedure is based on coefficients extracted from the ECLIPSE
21 post-processed inventory which in turn derive from the EMEP model. Comparing between the
22 POST and ANN runs (Table 1) we can model the impact on pollutant concentrations of
23 integrating a bottom-up approach in regional emission modeling ~~on pollutant concentrations~~.

24 Finally the impact of model horizontal resolution is a crucial issue for air-quality modeling. As
25 regards urban ozone there are plentiful studies on the effect of model resolution refinement with
26 an overall tendency to show improvement of the model's quality when increasing resolution
27 from about 30-50km to 4-12km (Arunachalam et al., 2006; Cohan et al., 2006; Tie et al., 2010;
28 Valari and Menut, 2008). On the other hand reports are scarce for fine particles: Pungler and
29 West. (2013) show that increasing the resolution from 36km to 12km improved the 1h daily
30 maximum concentrations but not the daily average, Stroud et al. (2011) reported better
31 agreement of fine particles of organic origin with measurements from a modeling exercise at a

1 2.5km resolution domain over a 15km resolution domain while Queen and Zhang. (2008) also
2 show improvement but their results include the effect of increasing the resolution of the
3 meteorological input as well. Valari and Menut. (2008) showed that the impact of the resolution
4 of emissions on modeled concentrations of ozone may be higher than the model resolution itself.
5 ~~This~~ese question has not yet been raised in the framework of climate driven atmospheric
6 composition modeling at the local scale. In our study we disentangle the impact of the resolution
7 of the emission dataset ~~used as input for the air quality simulation~~ from the effect of model
8 resolution itself by conducting two more tests. In the first test we employ the 0.5° resolution
9 simulation (REG hereafter) from which all aforementioned simulations take their boundary
10 conditions. We also compile the AVER database which uses as a starting point the modeled
11 concentrations at 4km resolution from the POST run spatially averaged over the 0.5° grid-cells of
12 the REG resolution mesh. REG vs. AVER (see Table 1) can provide information on the
13 influence of model resolution while comparing AVER against POST provides the sensitivity to
14 the resolution of the emission inventory ~~only~~.

16 **3 Model evaluation**

17 **3.1 Evaluation of present-time meteorology**

18 There are three WRF simulations involved in the study: i) climate model driven meteorology
19 downscaled from a global scale climate model (MET_CLIM); ii) meteorology from reanalysis
20 datasets at 0.5° resolution (MET_ERA05) and iii) meteorology downscaled from reanalysis data
21 at 0.11° (MET_ERA01). In this section we present a short evaluation of these datasets comparing
22 model results against surface observations from seven meteorological monitoring sites existing
23 in the domain. We note here, that from these monitors only one is located inside the highly
24 urbanized city of Paris. A thorough evaluation of the reanalysis dataset in Europe may be found
25 in Menut et al. (2012b).

26 The mean wintertime (DJF) and summertime (JJA) modeled and observed daily average values
27 are compared for four different meteorological variables relevant for air-quality, namely 2m-
28 temperature, 10m-wind speed, relative humidity and total precipitation (Table 2). A strong
29 positive bias is observed in modeled wind speed for both MET_CLIM and MET_ERA05
30 meteorology especially during the winter period. Such a bias, consistent with previous studies
31 (see e.g., Jimenez et al. (2012) for WRF or Vautard et al. (2012) for other models), is expected to

1 enhance pollutants' dispersion and lead to less frequent stagnation episodes. The bias is stronger
2 for the MET_CLIM dataset than for the MET_ERA05. A systematic wet bias in both
3 summertime and wintertime precipitation is observed for the two datasets. This can significantly
4 reduce PM concentrations through rain scavenging (Fiore et al., 2012; Jacob and Winner, 2009).
5 MET_ERA05 fields provide a better representation of precipitation especially in wintertime
6 where the bias is reduced by a factor of more than 2 compared to MET_CLIM. Summertime
7 temperature is adequately represented in the climate dataset whereas a wintertime weak cold bias
8 (-0.3°C) is observed. A strong hot bias during the winter is found for the reanalysis meteorology.
9 A warmer climate can increase ozone formation through thermal decomposition of PAN
10 releasing NO_x (Sillman and Samson, 1995). RH is generally well represented in both cases.
11 Finally we notice that the finer resolution reanalysis dataset (MET_ERA01) is not able to reduce
12 the observed domain-wide biases of the coarse meteorological run with the exception of specific
13 locations such as the Montsouris station in Paris where the bias in wintertime precipitation and
14 wind speed bias is reduced by 22% and 40% respectively.

15

16 **3.2 Evaluation of the reference simulation (REF)**

17 Mean modeled daily surface ozone and the daily maximum of 8-hour running means (MD8hr)
18 are compared against surface measurements in urban, suburban and rural stations (Fig. 3a). The
19 results presented are averaged over the ozone period (April-August). We also use odd oxygen
20 $O_x = O_3 + NO_2 - 0.1 * NO_x$ (Sadanaga et al., 2008) as an indicator of the efficiency of the model to
21 represent photochemical ozone build-up. Contrary to O₃, the concentration of O_x is conserved
22 during the fast reaction of ozone titration by NO and is therefore, a useful metric for the
23 evaluation of the photochemical ozone build-up by ruling out titration near high NO_x sources
24 (Vautard et al., 2007).

25 The model performs well in the urban areas capturing the mean daytime ozone levels (bias
26 +1.8ppb) while O_x is also accurately represented with an underestimation of only 4.1%,
27 illustrating the efficiency of the model to reproduce both daytime formation and titration of
28 urban ozone. The bias in daytime average is smaller and less than 1ppb. The O_x bias in daily
29 averages is similar to the daytime one, suggesting underestimation of nighttime titration. This is
30 consistent with other studies using CHIMERE (Szopa et al., 2009; Van Loon et al., 2007;

1 Vautard et al., 2007; Szopa et al., 2009). Model benchmark ratings show a high skill score (0.78)
2 while MNB and MNGE are +20.6 and 38.9 respectively.

3 We observe an overestimation of mean daytime suburban ozone (+5ppb). The small bias in O_x
4 (+0.6ppb) suggests that the problem stems from the representation of local titration and more
5 specifically daytime titration; the daily average ozone bias drops to +3.9ppb while O_x is
6 accurately represented in this case (-0.2ppb). Suburban stations present the lowest skill score
7 (0.63) compared to urban and rural. Model performance over rural stations is adequate, with an
8 overestimation in mean daily ozone of 8.2% (bias=+2.8ppb) and a good skill score (0.73). ~~We~~
9 ~~The identified~~ two major downwind locations in the IdF domain ~~and found that they~~
10 ~~represent~~ which present the lowest biases (less than 0.1ppb and 1.1ppb for the south-west and
11 north-east directions respectively). The bias of the daytime average reaches +2.1ppb.

12 Ozone daily maxima in the urban and rural stations are underestimated by 10% (-4.2ppb) and 7%
13 (-3.2ppb) respectively but we consider the magnitude of the underestimation small given the
14 climate framework of the simulation. Daily average ozone is better represented than daily
15 maxima, highlighting model sensitivity to accumulated errors (Valari and Menut, 2008).
16 Modeled peak concentrations are particularly sensitive to temperature compared to the daily
17 averages as shown in Menut et al. (2003). This could also be due to the fact that 4km is still an
18 insufficient model resolution.

19 The evaluation of $PM_{2.5}$ (Fig. 3b) shows a good representation of daily average levels during
20 wintertime where the highest annual concentrations are presented (bias less than $1\mu g/m^3$). In
21 annual basis the bias is also small while a larger underestimation is predicted for the summertime
22 season (bias= $2.8\mu g/m^3$). The latter can be due to underestimation of summertime emission fluxes
23 (resuspension emissions are not considered in our simulations) and underestimation of secondary
24 organic aerosols formation (Hodzic et al., 2010; Markakis et al., 2014; Solazzo et al., 2012). The
25 overestimation in wind and precipitation also contributes to the observed PM underestimation.
26 Wintertime and annual statistics show a high skill score. Interestingly in wintertime and in
27 annual basis the site located in downtown Paris presents the lowest bias ($<0.3\mu g/m^3$). Overall
28 the results indicate that the fine scale setup is able to predict the main patterns of ozone and fine
29 particle pollution in the area.

30

31 4. Sensitivity cases

4.1 Sensitivity to climate model driven meteorology (REF vs. ERA05)

~~The goal of t~~This case study ~~is to~~ estimates the discrepancy between an air-quality model run where regional meteorology is downscaled with WRF from reanalysis data (ERA05) and a simulation where meteorology is downscaled from a global scale climate model (REF). The wet bias in MET_CLIM meteorology is significantly reduced with meteorology from reanalysis data (Sect. 3.1). This is expected to have a significant role in the modeled PM concentrations. Another influential factor is the colder bias found in summertime temperature in the MET_ERA05 dataset. This ~~may could~~ lead to decreased ~~in the~~ reaction rates, less biogenic emissions and consequently to less ozone. The lower bias in 10m wind speed under MET_ERA05 is bound to ~~increase surface concentrations through~~ ~~lead to less reduced~~ dispersion ~~and higher surface concentrations~~. We also compare the average modeled boundary layer height (PBL) for the summer and winter periods between the two datasets: PBL is reduced by 5% and 12% in summer and winter respectively ~~(not shown)~~ when reanalysis data are used instead of climate model output. This may result in less dilution of emissions and therefore higher surface concentrations for primary emitted species, such as PM and NO_x.

Comparing the results of the two air-quality model runs for ozone (Fig. 4a and Table 3) we find only a small sensitivity ~~of ozone~~ to using meteorology from a climate model or reanalysis data over all three types of monitor sites; ~~urban, suburban and rural~~ ($|\Delta c| \sim 1$ ppb or 3.4%). ~~suggesting a~~ ~~The~~ small improvement of model performance with the reanalysis dataset ~~(ozone decreases through higher NO_x emissions following the PBL scheme described above)~~ ~~is due~~ ~~which stems from to~~ the fact that titration is more realistically represented in ERA05 (the difference is O_x between the two runs is negligible). The response of urban daily maximum values to the meteorological dataset is also negligible ($|\Delta c| = 0.1$ ppb or 0.3%).

Wintertime PM_{2.5} concentrations, on the contrary show a large sensitivity to the meteorological dataset. The change in the daily average ~~concentrations~~ is 3.1 μg/m³ (17.6%) while summertime levels remain unchanged (Table 3). Focusing on the annual averages, the small underestimation observed in the REF run turns into small overestimation in the ERA05 run ($|\Delta c| = 1.4$ μg/m³ or 9.4%). The use of ~~the~~ reanalysis data leads to a strong overestimation of wintertime concentrations (Fig. 4b), which stems directly from the reduction (and improvement) of precipitation by a factor of 2 in the meteorology from reanalysis. This leads to the conclusion that the small bias observed in the REF simulation during wintertime (Fig. 4b) could be due

1 model error compensation such as unrealistically high precipitation and possible inhibition of
2 vertical mixing or overestimation of wintertime emissions. The scores suggest a slight
3 deterioration in model performance when passing from meteorology from a climate model to
4 reanalysis meteorology in both winter and summer but improvement when focusing on the
5 annual statistics.

6 We conclude that using climate model driven meteorology has a small impact on modeled ozone
7 whereas larger sensitivity is observed for wintertime PM_{2.5} levels due to the accuracy of modeled
8 precipitation.

9

10 **4.2 Sensitivity to the resolution of the meteorological input (ERA01 vs. ERA05)**

11 Here we model the sensitivity of modeled ozone and PM_{2.5} concentrations to the resolution of the
12 meteorological input (Fig. 5 and Table 3). Daily average ozone shows a very weak response over
13 urban and rural sites ($|\Delta c| < 0.4$ ppb or $< 0.8\%$) and daily urban maxima improve slightly with the
14 ERA01 run ($|\Delta c| = 0.4$ ppb or 1%). At the suburban area the impact, though small ($|\Delta c| = 1.4$ ppb or
15 4.3%), is definitely higher than over urban or rural sites. O_x change at the suburban area (not
16 shown) is much weaker compared to ozone ($|\Delta c| \sim 0.5$ ppb or 1.2%) showing that the increase in
17 the resolution of meteorology has an impact on the representation of ozone titration leading to
18 improved model performance. The skill score over suburban sites increases by 9% while NMB
19 improves by 22% from 26.1 in ERA05 to 20.3 in ERA01. Interestingly, the response of suburban
20 ozone to the resolution of the meteorological input is the strongest modeled sensitivity for this
21 variable amongst all ~~the~~ studied cases.

22 Weak sensitivities are modeled for PM_{2.5} (Table 3) during summertime ($|\Delta c| = 0.3$ μg/m³ or 3.4%)
23 and on annual basis ($|\Delta c| = 0.6$ μg/m³ or 4%), but stronger during the winter season ($|\Delta c| = 1.3$
24 μg/m³ or 6.8%). In fact, wintertime statistics suggest that model bias actually increases with the
25 refinement of the ~~input~~ meteorological grid as a consequence of the reduced modeled
26 precipitation (less scavenging), ~~and wind speed (weaker dispersion) and PBL by 20% (weaker~~
27 ~~dispersion)~~ in MET_ERA01 compared to the climate model driven meteorology (Sect. 3.1).
28 Again, this points to the same error compensation scheme described in the REF vs. ERA05
29 comparison (Sect. 4.1).

30 We conclude that the resolution of the meteorological input has a small impact on modeled
31 ozone while moderate sensitivity is observed for suburban ozone and wintertime PM_{2.5}. Never

1 the less this result could reflect the local area's characteristics (flat terrain, situated away from
2 the coast) confirming previous studies (Menut et al., 2005; Valari and Menut, 2008). In regions
3 with more complex topography or close to the coast the resolution of the meteorological input
4 could have a profound effect on the simulated meteorological conditions (Leroyer et al., 2014).
5 We note here that the refinement in the resolution of the meteorological model from 0.5° to 0.1°
6 may not be sufficient for the CTM to simulate noticeable concentration responses. For example
7 Leroyer et al. (2014) (see also references therein) observed that substantial changes in vertical
8 and horizontal transport in an urban environment occurred mostly in the transition from
9 resolutions of 2.5km to 1km and even higher (250m).

11 **4.3 Sensitivity to the resolution of the CTM's vertical grid (REF vs. VERT)**

12 This study addresses the impact of the resolution of the CTM's vertical mesh and more
13 specifically of the thickness of the first CTM layer, on modeled ozone and $PM_{2.5}$ concentrations
14 (Fig. 6). Mean daily ozone is practically insensitive to the refinement of the vertical mesh at the
15 urban, suburban and rural areas (Table 3). Similarly, maximum ozone at the urban area changes
16 by only 0.5ppb (1.4%) with increased bias in the VERT run. Changes in summertime and annual
17 modeled $PM_{2.5}$ concentrations are also small, while the wintertime daily average shows some
18 weak sensitivity ($|\Delta c| = 0.5 \mu\text{g}/\text{m}^3$ or 2.2%). Scores are hardly affected.

19 Interestingly, the impact of the refinement of the vertical grid on daily averaged O_x is much
20 stronger than on ozone: O_x changes by 0.9ppb in the urban and suburban areas. The change in O_x
21 is reasonable since in VERT, NO_x emissions are released within a surface layer thinner by 60%
22 compared to REF (from 20m to 8m) leading to higher NO_x concentrations. That should normally
23 affect titration which is the driver of urban ozone concentrations. The fact that ozone remains
24 insensitive to the change in NO_x concentrations suggests that some other modeled processes
25 counteracts titration. To further investigate this issue we study the change in dynamical processes
26 such as vertical mixing and dry deposition. We extract the vertical diffusion coefficient K_z (m^2/s)
27 and dry deposition rates (g/m^3) for ozone, NO_2 and $PM_{2.5}$ for all grid cells that include an urban
28 monitor site and looked how modeled sensitivities change as a function of these parameters
29 (Fig.7).

30 NO_2 concentrations increase with the refinement of the first vertical layer of the CTM for all
31 vertical mixing conditions (Fig. 7a). However it is only under low vertical mixing ($1 < K_z < 5$

1 m²/s) that ozone sensitivity becomes positive (Fig. 7b). Under stronger turbulence ($K_z > 5$ m²/s),
2 the 12-layer setup leads to higher first-layer NO₂ concentrations (stronger titration) leading to
3 negative values for ozone sensitivity (such conditions account for the 93% of the simulated
4 period). On the other hand the refinement of the vertical mesh primarily affects NO₂ deposition
5 rates which accelerate by 14.3% but leaving ozone deposition rates unaffected. We may assume
6 that under low mixing conditions, the increased deposition rate of NO₂ slows down the increase
7 in NO₂ concentration due to the emission effect and dynamical processes become more
8 influential than titration. As a result the surface layer is enriched in ozone by getting mixed with
9 air from higher atmospheric layers (Menuet et al., 2013b).

10 For almost the entire K_z range, PM_{2.5} concentrations increase with VERT (Fig. 7c). This is due to
11 the fact that emissions are released in smaller volumes as discussed above. On the other hand,
12 here too, the refinement of the vertical resolution of the CTM₂ enhances deposition rate. These
13 two conflicting effects explain the small impact of the CTM's vertical resolution on PM_{2.5}
14 concentrations.

15 We conclude that bBoth ozone and PM_{2.5} sensitivities to the refinement of the vertical mesh are
16 small. Our analysis suggests that in both cases this is the result of two competing processes,
17 either titration against vertical mixing (ozone) or emission versus deposition (PM_{2.5}). Although
18 in the Ile-de-France area (low topography) the overall effect is insignificant, it may not be the
19 case in other regions with more complex topography.

21 **4.4 Sensitivity to the annual emission totals (REF vs. ANN)**

22 This case study compares modeled concentrations between two runs where annual emission
23 totals stem from either the AIRPARIF inventory (REF) or the ECPLISE dataset (ANN). Changes
24 in modeled urban daily average ozone concentrations are small ($|\Delta c| = 0.8$ ppb or 2.5%) with the
25 regional inventory (ECLIPSE ~~annual totals~~) to tend to increase the bias of the REF run (Fig. 8a
26 and Table 3). This is due to the fact that when passing from the AIRPARIF to the ECLIPSE
27 inventory (see also Fig. 2) NO_x emissions decrease (weakening titration) and NMVOCs increase
28 (intensifying production). This is also seen in the weaker sSensitivity is weaker for of O_x (0.4 ppb
29 or 1%) suggesting that the main reason for the improvement brought about by the use of the local
30 inventory (REF run) is due to a better representation of the ozone titration process. At the
31 suburban area, the sensitivity is larger ($|\Delta c| = 1.1$ ppb or 3.2%) and of the same order of magnitude

1 as the sensitivities to climate model driven meteorology and to the resolution of the
2 meteorological input. The weaker change in suburban O_x ($|\Delta c|=0.1\text{ppb}$ or 0.3%) suggests that
3 this area benefits more than the urban area from the improvement in the titration process. The
4 skill score associated to the REF run is also higher by 8% (Fig. 8a). Changes in daytime averages
5 at both urban and suburban areas are similar to those in the daily averages suggesting that
6 modeled sensitivity stems mainly from daytime titration. Rural ozone is practically unaffected
7 ($|\Delta c|=0.3\text{ppb}$ or 1%). It is noteworthy that the absolute change in modeled ozone concentrations
8 is ~~on~~in the order of 1ppb or less despite the large differences in ozone precursors' emissions
9 between the local and the regional inventory.

10 Changes in the daily average fine particle concentrations in summertime, wintertime and in the
11 annual basis daily average are much stronger than ozone ($|\Delta c|=4.1\mu\text{g}/\text{m}^3$ or 33%, $6.6\mu\text{g}/\text{m}^3$ or
12 33.8% and $5.5\mu\text{g}/\text{m}^3$ or 31.9% respectively). $PM_{2.5}$ concentrations modeled with the ANN run are
13 significantly higher than those modeled with the REF run (Fig. 8b). Wintertime bias in ~~the~~ ANN
14 ~~run~~ reaches $\pm 5.8\mu\text{g}/\text{m}^3$ showing that fine particle emissions from the ECLIPSE inventory are
15 overestimated (see also Fig. 2). The main source of primary wintertime $PM_{2.5}$ emissions over the
16 IdF region as well as in Paris in the ANN run is wood burning (see discussion in Sect. 2.4),
17 which is unrealistic for a city like Paris and stems directly from the use of the population proxy
18 to spatially allocate national totals over the finer scale. This is consistent to the fact that the
19 summertime bias in the ANN run is much lower ($+1.4\mu\text{g}/\text{m}^3$). In fact, in this case the ANN bias
20 is even smaller than the REF bias ($-2.8\mu\text{g}/\text{m}^3$) enhancing our hypothesis that summertime fine
21 particle emissions in the AIRPARIF inventory are underestimated (see also Sect. 2.1). The ~~REF~~
22 skill score in REF is higher than in the ANN ~~score~~ in wintertime and lower in summertime.

23 We conclude that ozone sensitivity to the annual emission totals is low but strong for fine
24 particles.

25 26 **4.5 Sensitivity to emission post-processing (ANN vs. POST)**

27 Here we use identical annual totals but two different methods for their vertical and temporal
28 allocation to obtain hourly fluxes over the 4km-resolution domain ~~and~~as well as different
29 matrices for their chemical speciation. The ANN dataset uses the AIRPARIF bottom-up
30 approach whereas the EMEP methodology is applied ~~on~~to the POST dataset. To compile the
31 ANN inventory we had to extract the post-processing coefficients of the bottom-up inventory

1 and apply them on the ECLIPSE annual totals. This procedure though was not emission source-
2 sector oriented and this inconsistency definitely affects model results. On the other hand the
3 post-treatment of ~~the (sectoral)~~ raw emissions in large-scale applications are typically based on
4 sectoral coefficients that don't link back to the same quantified emissions either. For example in
5 the regional application used this study (REG) the ~~per SNAP-sectoral~~ ECLIPSE raw emissions
6 ~~quantified in SNAP level~~ are treated with ~~the respective sectoral coefficients~~ ~~SNAP-level EMEP~~
7 ~~information~~ that stems from the EMEP inventory having ~~a very~~ different synthesis of sub-SNAP
8 sources from that of ECLIPSE. Therefore when we compare ANN with POST we consider that
9 what we observe is the bias of this inconsistency in regional modeling. The question raised is:
10 what is the benefit of adopting a bottom-up post-processing for regional scale air-quality
11 modeling?.

12 The effect on ozone ~~concentrations~~ over the urban area is ~~considered~~ moderate ($|\Delta c|=1.9\text{ppb}$ or
13 6.4%) (Fig. 9a and Table 3). Model bias is reduced from $+4.5\text{ppb}$ in POST to $+2.6\text{ppb}$ in ANN.
14 Ozone sensitivity in this case, is twice as high as the sensitivity to climate model driven
15 meteorology and even higher compared to the impact of annual totals. The ANN simulation is
16 able to increase the skill score by 14% and reduce MNB by 26%. The low O_x sensitivity suggests
17 that discrepancies are mainly due to a better representation of ozone titration. Suburban and rural
18 ozone is practically insensitive to the post-processing technique. Even if emission totals are ~~the~~
19 ~~identical same~~ between the two configurations, ozone concentrations over the urban area are
20 lower in the ANN run than in the POST run because ~~the~~ ANN has more ground-layer NO_x
21 emissions than POST enhancing ozone titration. This stems from the fact that the annual
22 emission totals are allocated in the CTM's vertical layers very differently. Following the
23 AIRPARIF post-processing (ANN) all urban emissions are released in the surface layer because
24 according to the local point source emission database no major industrial units are found within
25 the urban area. On the contrary, the regional scale post-processing (POST) does not resolve the
26 urban from the suburban and rural areas, where industrial zones are located and assigns only
27 70% of the total NO_x emissions ~~over~~ in Paris in the first model layer.

28 Another important piece of information ~~of the post-processing of emissions regards their~~ ~~id the~~
29 diurnal variation ~~of emissions~~. Although the time scale of a climate forced run largely exceeds
30 the hourly basis we aim to illustrate how important ~~can~~ the choice of the diurnal patterns ~~can~~ be
31 to the final modeled concentrations. Fig. 10a shows the average diurnal variation of modeled and

1 observed urban ozone for ANN and POST (for the modeled fields we use the grid cells of the
2 monitoring sites). The two downscaling approaches compared here, apply different diurnal
3 profiles on emissions to provide hourly fluxes. Between 10:00LT and 15:00LT, ANN
4 underestimates ozone concentrations due to too much NO emissions, enhancing titration and this
5 is maximized in the local peak (15:00LT) where NO concentrations are overestimated by a factor
6 of 2 (not shown). The daily maximum concentration shows the highest sensitivity in the emission
7 post-treatment among all the presented cases ($|\Delta c|=2.2$ ppb). This is consistent with Menut et al.
8 (2003) who also found that the afternoon peak concentrations at a typical summertime episode in
9 Paris are very sensitive to the NO emissions change. In the evening (after 15:00LT) ANN
10 deviates from the observations faster than POST ~~from the observations~~ because the afternoon
11 peak in traffic emissions is more pronounced in the AIRPARIF diurnal profile compared to that
12 used in the ECLIPSE processing which represents an average situation of anthropogenic sources
13 hence a smoother variation. These results indicate that the diurnal variability of modeled ozone
14 over the urban area is very sensitive to the choice of the diurnal profile. But in the climate
15 concept where hourly values are timely too short to take into account, the sensitivity is
16 considered moderate as seen in Table 3.

17 Modeled PM_{2.5} sensitivity is significant for both summer and wintertime ($|\Delta c|=3.4\mu\text{g}/\text{m}^3$ or
18 24.8% and $4.6\mu\text{g}/\text{m}^3$ or 18.3% respectively) (Table 3). POST wintertime bias is almost two times
19 higher than ANN (Fig. 9b). This is because the coarse resolution annual post-processing
20 coefficients weight towards allocating more of the annual emissions into the winter period
21 significantly influenced by the residential sector emissions which are overstated in the ECLIPSE
22 inventory. A late afternoon peak is modeled with ANN accounting for the traffic emissions,
23 whereas PM_{2.5} evening levels modeled with the POST run (after 20:00LT) are related to the
24 residential heating activity (Fig. 10b).

25 What we can conclude is that in a climate forced – air quality framework the model response for
26 daily average ozone by 6.2% is rather small considering the significant differences that the two
27 post-processing approaches prescribe for the vertical distribution of emissions and their diurnal
28 variation. Fine particle concentrations are much more sensitive to the applied emission post-
29 processing technique. We note here, that recent work has pointed out that the sensitivity of
30 modeled concentrations the spatiotemporal resolution of the emission inventory is model-
31 dependent (Makar et al., 2014).

1

2 **4.6 Sensitivity to the emission inventory resolution (POST vs. AVER)**

3 Here, we quantify the effect of the resolution of the emission input ~~from the impact of model~~
4 ~~resolution~~. Results show that in the urban areas this sensitivity is the most influential amongst all
5 tests presented in this paper with ozone changes reaching $|\Delta c|=2.8\text{ppb}$ or 8.3% (Fig. 11a). The
6 change in daily average O_x is smaller but comparable ($|\Delta c|=1.2\text{ppb}$ or 2.9%) suggesting that
7 ozone titration is not the only model process that is affected by the increase in the resolution of
8 the emission dataset. The skill score and MNB improve significantly in the POST run (Table 3).
9 Ozone precursors' emissions from urban sources are mixed with the lower emissions from the
10 surrounding suburban and rural areas inside the large cells of the coarse mesh-grid (AVER). This
11 leads to lower titration rates and therefore, higher ozone levels. Therefore the increase in the
12 resolution of the emission input leads to a reduced positive bias from +7.3ppb (AVER) to
13 +4.5ppb (POST). AVER overestimates ozone peaks by 0.8ppb while POST underestimates them
14 by -1.2ppb. The sensitivity of ozone concentration at the hour of the afternoon peak is linked to
15 NO_x concentration at the same hour, which reaches a local maximum due to the evening rush
16 hour (see also Sect. 4.5). Suburban and rural ozone is less sensitive than urban ($|\Delta c|=0.7\text{ppb}$),
17 with scores practically unchanged (Table 3).

18 Fine particle concentrations are also very sensitive to the resolution of the emission input,
19 especially in wintertime ($|\Delta c|=7.1\mu\text{g}/\text{m}^3$ or 30%), with higher concentrations modeled with the
20 refined emission inventory in POST (Table 3). ~~Similarly to ozone t~~This is because in the coarser
21 inventory represented here by AVER, emissions in the high emitting areas in the city are
22 smoothed down and diluted when averaged with emissions of the less polluted outer areas.

23 We conclude that the resolution of the emission input is the most influential factor from all the
24 studied cases, even more than model resolution itself. $\text{PM}_{2.5}$ showed higher sensitivity than
25 ozone concentrations. The non-linear nature of ozone chemistry suggests that it is important for
26 the ozone precursor emissions to be concentrated correctly to the high emitting areas such as the
27 urban centres.

28

29 **4.7 Sensitivity to model horizontal resolution (AVER vs. REG)**

30 Here, we study the sensitivity of ozone and $\text{PM}_{2.5}$ concentrations to ~~model the CTM's~~ horizontal
31 resolution. We compare the simulations ~~of at~~ two different spatial resolutions, the AVER run

1 (averaged over the grid-cells of the coarser grid) and the REG simulation on a grid of 0.5°
2 resolution (Fig. 12). REG₂ models higher ozone concentrations than AVER over the urban area
3 ($|\Delta c|=1.7\text{ppb}$ or 4.7%). As discussed above, NO_x emissions in the REG simulation are lower than
4 in REF due to dilution in the coarser grid cells leading to lower ozone titration rates. Suburban
5 and rural ozone has ~~relatively~~ low sensitivity to model resolution ($|\Delta c|=0.5\text{ppb}$ or 1.4% and
6 0.2ppb or 0.5% respectively) because photochemical build-up occurs at larger time and space
7 scales compared to titration and the refinement of the model grid does not ~~provide-increase~~
8 ~~performancemuch new information to the modeling~~. This confirms the results in Markakis et al.
9 (2014). The effect on modeled PM_{2.5} is very small with concentrations slightly higher over the
10 finer mesh grid as a result of the lower primary emissions in REG.

11 We may conclude that the benefit of increasing the CTM's resolution is insignificant for both
12 ozone and PM_{2.5} especially taking into account the large refinement attempted here (0.5° to
13 4km).

14 15 **5 Sources of error in regional climate forced atmospheric composition** 16 **modeling**

17 In this paper we utilize simulations at two spatial scales: at urban scale over a grid of 4km
18 resolution using the AIRPARIF bottom-up inventory of anthropogenic emissions (REF) and a
19 regional scale run at 0.5° resolution where emissions stem from the ECLIPSE top-down
20 inventory (REG). Both realizations implement identical climate driven meteorology (at 0.44°
21 resolution) and an 8-layer vertical mesh therefore are susceptible to the same sources of error due
22 to climate model driven meteorology, the resolution of the meteorological input and the
23 resolution of the CTM's vertical grid. However the remaining biases presented in Table 3 over
24 urban areas e.g., the emissions resolution, the model horizontal resolution, the annual quantified
25 fluxes and the post-processing method concern mainly the REG run. As regards ozone REG has
26 a positive bias of 9ppb over the city of Paris while the bias of REF is only +1.8ppb (Fig. 13a).
27 The question we raise is “what are the main sources of uncertainty in regional scale climate
28 driven air-quality simulations and how these could be eliminated or at least reduced?”.

29 With this study we are able to identify the source of the excess of $|\Delta c|=7.2\text{ppb}$ of ozone modeled
30 with the REG run compared to REF (Table 4); 26.4% ($|\Delta c|=1.9\text{ppb}$) is related to the post-
31 processing of the annual emissions totals which are based on the EMEP factors, 11.1%

1 ($|\Delta c|=0.8\text{ppb}$) to the annual emission totals in the ECLIPSE inventory, 23.6% ($|\Delta c|=1.7\text{ppb}$) to
2 coarse model resolution and 38.9% ($|\Delta c|=2.8\text{ppb}$) to the coarse resolution of the ECLIPSE
3 emission inventory.

4 Considering the discrepancies in the inventorying methodologies used to compile the ECLIPSE
5 and the AIRPARIF datasets (top-down vs. bottom-up), it is very interesting that the least
6 influential factor to the urban ozone response is the annual emissions totals. It seems that the
7 regional simulation would not benefit much from the integration of the local annual totals alone
8 but a more important gain would stem from the application of the AIRPARIF post-processing
9 methodology. The added value from both these factors would reduce the positive bias of REG by
10 2.7ppb. Even largest improvement comes through the better spatial representation of ozone
11 precursors emissions in the local emission inventory ($|\Delta c|=2.8\text{ppb}$) leading to more faithful
12 titration process; O_x levels are very close in REF and REG (Fig. 13a). It could therefore argued
13 that without increasing model resolution of which the gain would reach only 1.7ppb, the REG
14 simulation would benefit significantly by simply integrating the aforementioned local scale
15 information.

16 The difference in modeled ozone between REF and REG is much smaller over the suburban area
17 ($|\Delta c|=2.4\text{ppb}$) and the most influential factor to this difference is the annual emission totals
18 covering 45.8% of this difference. Finally as regards ozone one important result of this study is
19 that in the climate-air quality framework modeled concentrations from a coarse resolution run,
20 well agree with the much more intensive (in terms of computational time) fine resolution run and
21 the bias is considered of small magnitude (Fig. 13a). This is because the formation of rural ozone
22 is a slower process than in urban areas and comparable to the characteristic transport time of
23 precursor's pollutants to the coarse grid cell.

24 Focusing on the wintertime $\text{PM}_{2.5}$ concentrations where the largest annual levels are observed,
25 these are better simulated with the REF run with a bias of $-0.8\mu\text{g}/\text{m}^3$ and a high skill score of
26 0.78 compared to a strong positive bias of $+3.6\mu\text{g}/\text{m}^3$ and a skill score of 0.68 with the REG run
27 (Fig. 13b). We should remind here that both runs suffer from a strong wet bias reducing
28 significantly $\text{PM}_{2.5}$ concentrations (see also Sect. 3.1). Contrary to ozone, where information
29 from the local scale improves in all cases model performance, the resolution of the emission
30 inventory seems to deteriorate the modeling performance of $\text{PM}_{2.5}$ with increase in the bias by
31 $7.1\mu\text{g}/\text{m}^3$. This only means that if the emission totals from ECLIPSE are used over Paris in the

1 coarse REG application then refining the resolution will only accumulate additional emissions in
2 the city augmenting the modeled concentrations. The remaining features have also a positive
3 effect; model resolution reduces the bias by $0.4\mu\text{g}/\text{m}^3$, annual emission totals by $6.6\mu\text{g}/\text{m}^3$ and
4 post-processing of the annual totals by $4.5\mu\text{g}/\text{m}^3$. This essentially means that the regional
5 realization cannot selectively incorporate any combination of local-scale features in order to
6 improve performance as in the case of ozone. But the results indicate that by simply integrating a
7 bottom-up post-processing technique would result in an overall bias of the regional application
8 of $-0.9\mu\text{g}/\text{m}^3$.

10 **6 Conclusions**

11 In the present paper we assess the sensitivity of ozone and fine particle concentrations with
12 respect to emission and meteorological input with a 10yr long climate forced atmospheric
13 composition simulation at fine resolution over the city of Paris.

14 As a general observation our study shows that overall ozone response is considered low to
15 moderate while $\text{PM}_{2.5}$ concentrations were generally very sensitive for the presented cases. The
16 largest sensitivity in modeling the average daily ozone concentrations was observed in the urban
17 areas primarily due to the resolution of the emission inventory ($|\Delta c|=2.8\text{ppb}$ or 8.3%) and
18 secondly to the post-processing methodology applied on the annual emission totals ($|\Delta c|=1.9\text{ppb}$
19 or 6.2%). These sensitivities are attributed to changes in the titration process. When post-
20 processing coefficients were derived from the bottom-up AIRPARIF inventory instead of EMEP,
21 too much ozone titration takes place at the hour of the ozone peak and the sensitivity of daily
22 maximum reached its highest value among all the studied cases ($|\Delta c|=2.2\text{ppb}$ or 5.8%). It is
23 interesting that despite the fact that ozone precursor's emissions are very different between the
24 bottom-up and the top-down inventories, ozone sensitivity to the annual totals was shown to be
25 very small ($|\Delta c|=0.8\text{ppb}$ or 2.5%). Also modeled ozone is fairly insensitive to the use of climate
26 model or reanalysis meteorology. Finally all cases of suburban and rural ozone both for average
27 and maximum concentrations showed a sensitivity of less than 5%.

28 Regarding $\text{PM}_{2.5}$ concentrations, amongst all the presented factors, the emissions related were
29 those ~~are~~ shown to be the most influential. The corresponding sensitivity to the use of annual
30 emission totals from a top-down and a bottom-up inventory reached 33% in summer, 33.8% in
31 winter and 31.9% for the daily average ~~annual~~ concentrations. This is connected to the

1 downscaling methodology applied ~~on~~in the regional-scale totals ~~of~~in the ECLIPSE inventory;
2 uUsing population as proxy for their spatial allocation, leads to overestimation of particle
3 emissions from wood-burning over the Paris area. Large sensitivity was also shown due to the
4 resolution of the emission inventory (20.3% in the summer, 30% in the winter and 24.2% in
5 annual basis) because the coarser inventory smoothens the sharp emission gradients over the
6 urban area leading to less primary emissions. Fine particle concentrations were also sensitive to
7 the applied emission post-processing technique (22.1% in summer and 16.7% in winter). Only
8 wintertime PM_{2.5} concentrations were significantly affected by the meteorological related
9 sensitivities; by 17.6% due to the use of meteorology from reanalysis instead of climate (mainly
10 because the prescribed changes in modeled precipitation) and by 6.8% due to refinement of the
11 meteorological grid.

12 Both ozone and PM_{2.5} are little sensitive to the CTM's vertical resolution (changes of less than
13 2.2%). Nevertheless we provide evidence that this low sensitivity may be the result of
14 counteracting factors such as ozone titration, dry deposition and vertical mixing, too much
15 dependent on local topography to be able to generalize for other regions. We also note the weak
16 sensitivity of modeled concentrations to the increase in the CTM's and the meteorological
17 model's horizontal resolution at least for the area and the range of resolutions studied here.

18 Excluding the sensitivities having the smallest impact (roughly less than 2%, see Table 3) we
19 observe a very consistent trend in ozone concentration: daily average and maximum ozone
20 decrease as input data become more refined, namely passing from climate meteorology to
21 reanalysis, increasing the resolutions of the horizontal and vertical CTM grid, of meteorology, of
22 emissions and by using bottom-up emissions and post-processing instead of top-down. This
23 decrease in ozone concentrations, from 2.5% up to 8.3%, is observed mainly in the urban and
24 suburban areas and in all cases stems from enhanced NO_x emission fluxes in the surface-layer
25 leading to titration inhibition. Trends and the underlying changes in emissions are highly variable
26 for PM_{2.5} with increase in concentrations that may be as low as 2% or as high as 30% for climate
27 meteorology and resolution of the vertical mesh and also cases where concentration decreases in
28 a wide range of values from 3% up to 34% (annual emissions, model resolution) depending on
29 the season.

30 To fill the gap between regional and city-scale assessments we have to combine in a single
31 application the advantages of regional and local scale applications; the low resolution (but high

1 spatial coverage) from one hand and the good representation of emissions (but limited area of
2 coverage) on the other. The results of this study move towards that goal and can be used in order
3 to identify the main sources of error in regional scale climate forced air-quality modeling over
4 the urban areas. These biases could be taken into account ~~used~~ in policy relevant assessments.
5 The difference in modeled daily average ozone between the local and regional application over
6 the urban areas ($|\Delta c|=7.2\text{ppb}$) is attributed to several sources of error: 38.9% is related to the
7 resolution of the emission inventory, 26.4% stems from the post-processing of national annual
8 emission totals, 23.6% is due to model resolution (4km or 0.5°) and 11.1% is associated to the
9 annual emissions ~~totals~~ used as starting point for the compilation of the anthropogenic emission
10 dataset. Although the greatest benefit in the regional-scale modeling seems to come through the
11 increase in the resolution of the emission inventory, simpler actions may be also meaningful,
12 such as the integration of the locally developed annual totals and the downscaling coefficients
13 derived from the existing bottom-up modeling systems which combined could ~~reduc~~ing the bias
14 of the regional application by 37.5%. We note here that $\text{PM}_{2.5}$ levels in the urban regions are
15 likely mostly controlled by primary emissions; increasing the emissions inventory resolution will
16 concentrate the $\text{PM}_{2.5}$ emissions into a smaller spatial extent of the urban area (the reverse side of
17 the artificial dilution issue taking place at coarse resolution); if the emissions totals are
18 themselves biased high, then the resulting error will only become apparent at higher resolution.
19 Therefore, the emissions resolution may be showing that the emissions totals are too high, and
20 this only becomes apparent at high resolutions.

21 As regards $\text{PM}_{2.5}$ modeling our study shows that the regional realization cannot selectively
22 incorporate any combination of local-scale features in order to improve performance as in the
23 case of ozone. The simulation at regional scale (REG) predicts an excess of $3.6\mu\text{g}/\text{m}^3$ during
24 wintertime compared to the fine scale simulation (REF) showing a bias of $-0.8\mu\text{g}/\text{m}^3$ and this is
25 attributed to the allocation of wood-burning emissions over the Paris area. Therefore, the most
26 influential factor for $\text{PM}_{2.5}$ modeling is the resolution of the emission input (REG-
27 REF= $+7.1\mu\text{g}/\text{m}^3$). But the implementation of the refined emission resolution of the local
28 inventory alone would not benefit the regional simulation (which would increase the overall bias
29 to $10.7\mu\text{g}/\text{m}^3$), neither the implementation of the annual emissions of the bottom-up inventory
30 alone (REG-REF= $-6.6\mu\text{g}/\text{m}^3$) which would generate an overall negative bias of $3\mu\text{g}/\text{m}^3$. A

1 simpler action would be to integrate the post-processing bottom-up technique (REG-REF=-
2 $4.5\mu\text{g}/\text{m}^3$) giving an overall bias in REG of $-0.9\mu\text{g}/\text{m}^3$.

3

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1 Table 1. Parameterization of the different sets of simulations presented in the paper. Changes
 2 with respect to the REF case are marked in red. Changes with respect to a simulation other than
 3 REF are marked in green.

	Annual emission totals ^a	Air-quality model resolution	Emission inventory resolution	Emission post-processing ^b	climate/reanalysis meteorology and resolution	Number of layers in air-quality model
REF	AIRPARIF	4km	4km	Bottom-up	RCP-4.5 (0.44°)	8
REG ^c	ECLIPSE	0.5°	0.5°	Top-down	RCP-4.5 (0.44°)	8
Sensitivity simulation						
ERA05	AIRPARIF	4km	4km	Bottom-up	ERA (0.44°)	8
ERA01 ^d	AIRPARIF	4km	4km	Bottom-up	ERA (0.11°)	8
VERT	AIRPARIF	4km	4km	Bottom-up	RCP-4.5 (0.44°)	12
ANN	ECLIPSE	4km	4km	Bottom-up	RCP-4.5 (0.44°)	8
POST ^e	ECLIPSE	4km	4km	Top-down	RCP-4.5 (0.44°)	8
AVER ^f	ECLIPSE	4km	0.5°	Top-down	RCP-4.5 (0.44°)	8

4 ^a The resolution of the emission inventory of AIRPARIF is 1km (aggregated to 4km for the
 5 purpose the local simulations) and the ECLIPSE inventory 50km.

6 ^b Temporal, vertical allocation and chemical speciation.

7 ^c This simulation is used as boundary conditions for all local scale simulations.

8 ^d The ERA01 simulation is compared with the ERA05 not with the REF.

9 ^e The POST simulation is compared with the ANN not with the REF.

10 ^f This is not a standalone simulation. Concentrations modeled at 4km resolution with the POST
 11 run are averaged spatially to match the cells of REG (0.5° resolution simulation). AVER results
 12 are compared to REG to quantify the effect of model resolution and with POST to quantify the
 13 effect of the resolution of the emission inventory.

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1 Table 2. Observed and modeled daily average meteorological variables over the Ile-de-France
 2 region. MET_CLIM dataset stems from a climate model and MET_ERA05, MET_ERA01 from
 3 reanalysis data at 0.5° and 0.1° resolution respectively. Absolute model bias is given in
 4 parenthesis.

Variable	<i>Obs</i>	<i>MET_CLIM</i>	<i>MET_ERA05</i>	<i>MET_ERA01</i>
<i>Summer (JJA)</i>				
T2 (°C)	19.19	19.14 (-0.05)	18.28 (-0.91)	18.19 (-1.0)
WS10 (m/s)	2.9	4.0 (+1.1)	3.8 (+0.9)	3.8 (+0.9)
RH (%)	69.1	68.1 (-1.0)	68.3 (-0.8)	67.3 (-1.8)
PRECIP (mm/day)	0.076	0.108 (+0.032)	0.097 (+0.021)	0.098 (+0.022)
<i>Winter (DJF)</i>				
T2 (°C)	4.3	4.0 (-0.3)	6.0 (+1.7)	5.8 (+1.3)
WS10 (m/s)	3.6	6.2 (+2.6)	5.7 (+2.1)	5.5 (+1.9)
RH (%)	85.0	80.3 (-4.7)	79.7 (-5.3)	79.5 (-5.5)
PRECIP (mm/day)	0.069	0.112 (+0.043)	0.089 (+0.02)	0.087 (+0.018)

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1 Table 3. Absolute difference (and percentage in parenthesis) between daily averaged ozone (ppb)
 2 and PM_{2.5} (µg/m³) from two climate forced air-quality runs. The most influential factor for each
 3 sensitivity test is marked in bold.

Ozone	Urban	Suburban	Rural
Climate meteo (REF vs. ERA05)	1.0 (3.4%)	1.1 (3.2%)	0.9 (2.5%)
Meteo. resolution (ERA05 vs. ERA01)	0.2 (0.6%)	1.4 (4.3%)	0.3 (0.8%)
Vertical resolution (REF vs. VERT)	0.3 (1.2%)	<0.1 (0.2%)	<0.1 (1.5%)
Annual emis. totals (REF vs. ANN)	0.8 (2.5%)	1.1 (3.2%)	0.3 (1.0%)
Emission post-proc. (ANN vs. POST)	1.9 (6.4%)	0.1 (0.4%)	<0.1 (0.02%)
Emission resolution (POST vs. AVER)	2.8 (8.3%)	0.7 (1.9%)	0.2 (0.5%)
Model resolution (AVER vs. REG)	1.7 (4.7%)	0.5 (1.4%)	0.2 (0.5%)
PM_{2.5}	Summer	Winter	Annual
Climate meteo (REF vs. ERA05)	<0.1 (0.05%)	3.1 (17.6%)	1.4 (9.4%)
Meteo. resolution (ERA05 vs. ERA01)	0.3 (3.4%)	1.3 (6.8%)	0.6 (4.0%)
Vertical resolution (REF vs. VERT)	<0.1 (0.3%)	0.5 (2.2%)	<0.1 (0.2%)
Annual emis. totals (REF vs. ANN)	4.1 (33.0%)	6.6 (33.8%)	5.5 (31.9%)
Emission post-proc. (ANN vs. POST)	3.4 (24.8%)	4.5 (18.3%)	0.2 (0.7%)
Emission resolution (POST vs. AVER)	2.1 (20.3%)	7.1 (30.0%)	4.3 (24.2%)
Model resolution (AVER vs. REG)	0.4 (4.1%)	0.4 (1.9%)	0.7 (0.5%)

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1 Table 4. Top row presents the coarse resolution application (REG) model bias of the April-
 2 August average urban ozone and wintertime urban PM_{2.5}. Subsequently, marked with italics the
 3 signals -measured as the absolute concentration change from REG- of several refinements such
 4 as increase of resolution (model or emissions) and adaptation of annual quantified fluxes and
 5 post-processing of a bottom-up inventory. The individual signals sum up to the absolute bias
 6 found under the fine resolution simulation (REF).

Ozone	Ozone (ppb)	PM_{2.5} (µg/m³)
REG (50km)	+9.0	+3.6
Model resolution	<i>-1.7</i>	<i>-0.4</i>
Emissions resolution	<i>-2.8</i>	<i>+7.1</i>
Annual emission totals	<i>-0.8</i>	<i>-6.6</i>
Emissions post-processing	<i>-1.9</i>	<i>-4.5</i>
REF (4km)	+1.8	-0.8

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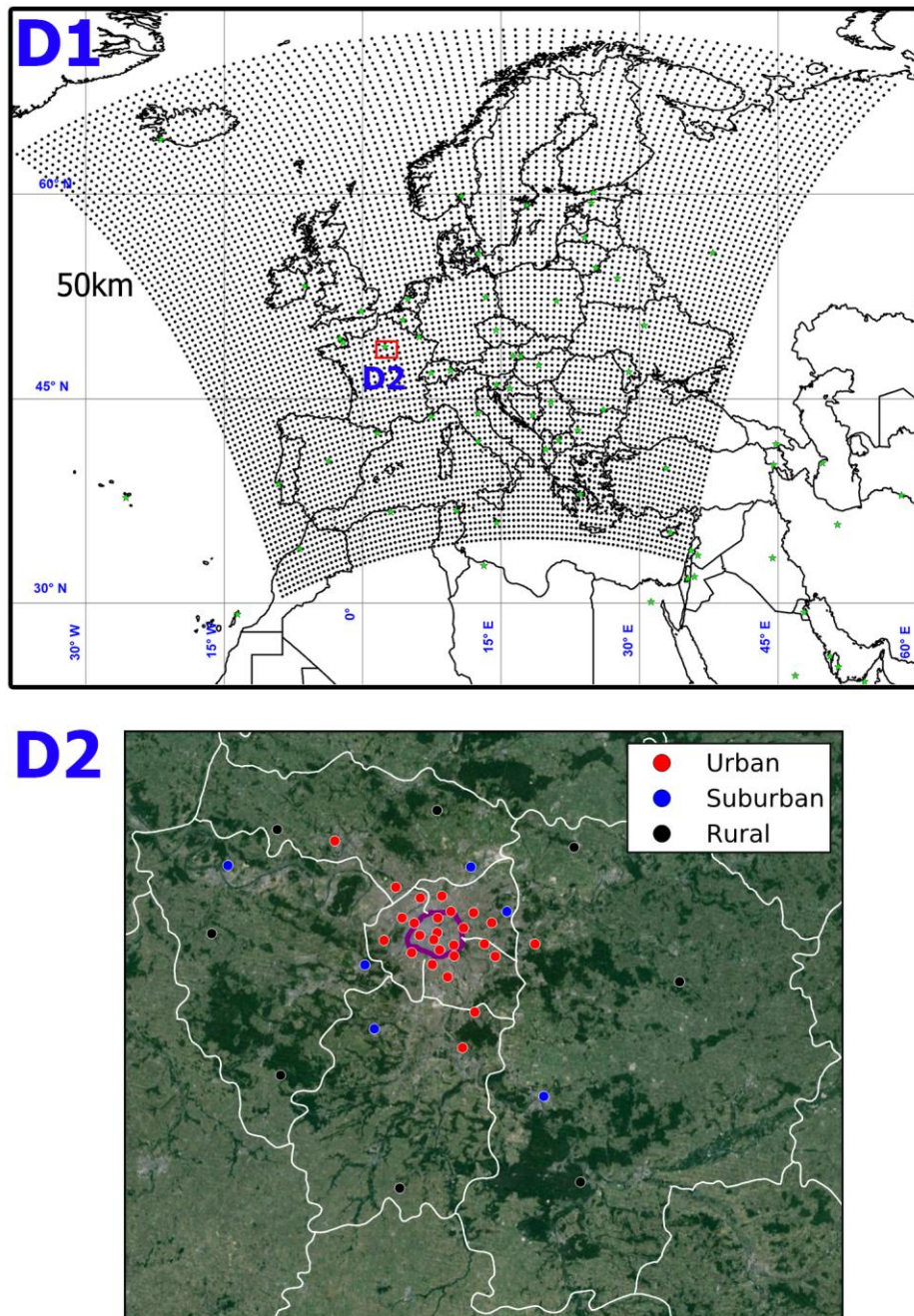
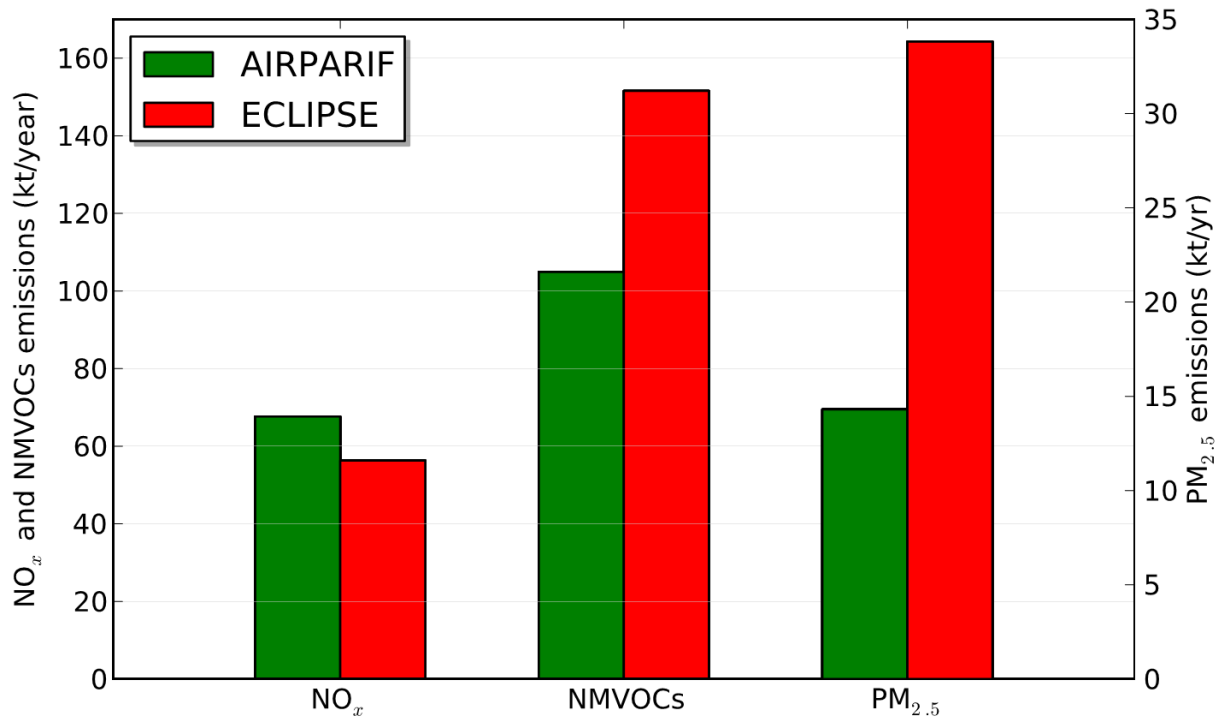


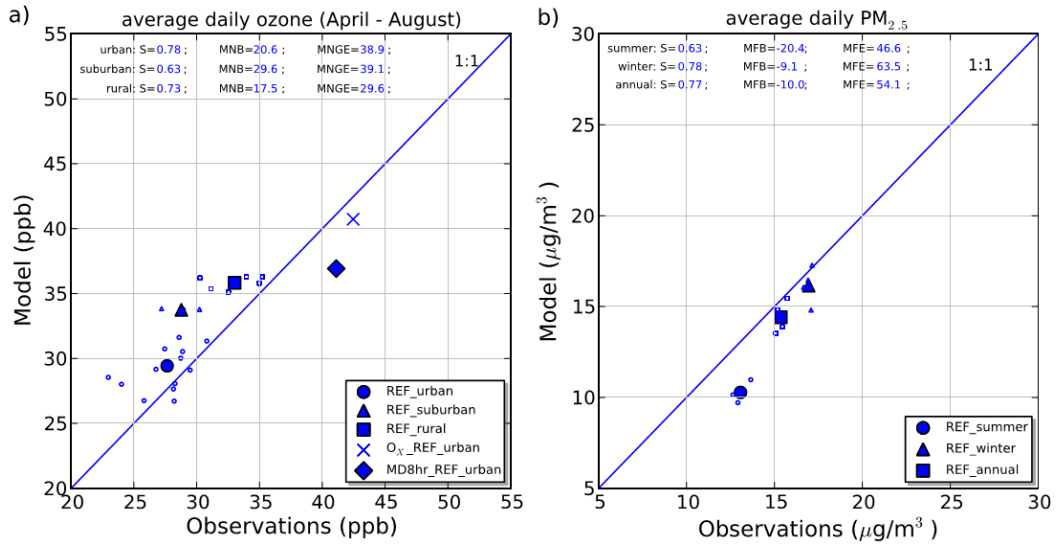
Figure 1. Overview of the coarse (D1 having 50km resolution) and local scale (D2, illustrated by the red rectangle having 4km resolution) simulation domains. In D2 the city of Paris is located in the area enclosed by the purple line. Circles correspond to sites of the local air-quality monitoring network (AIRPARIF) with red for urban, blue for suburban and black for rural.



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Figure 2. Domain-wide annual emissions of NO_x, NMVOC (left-axis) and PM_{2.5} (right-axis) from the local (bottom-up) and the regional (top down) inventory (summed across the vertical column).

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Figure 3. Panel a: Scatter plots and scores of daily average ozone concentrations at urban, suburban and rural stations from the REF simulation. Odd oxygen (O_x) and daily maximum values at urban locations are also shown. Panel b: daily average PM_{2.5} concentrations in wintertime (DJF), summertime (JJA) and on annual basis over urban stations.

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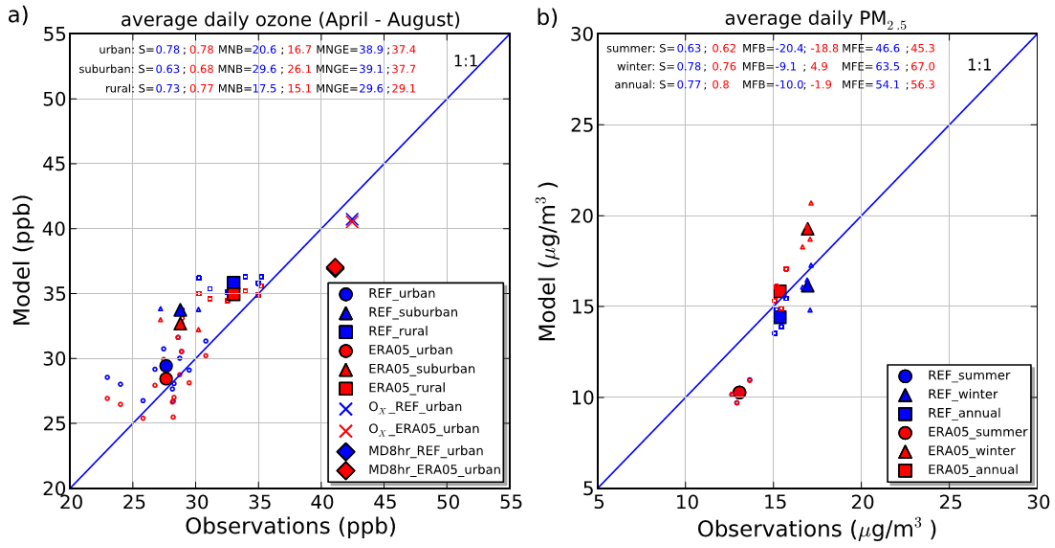
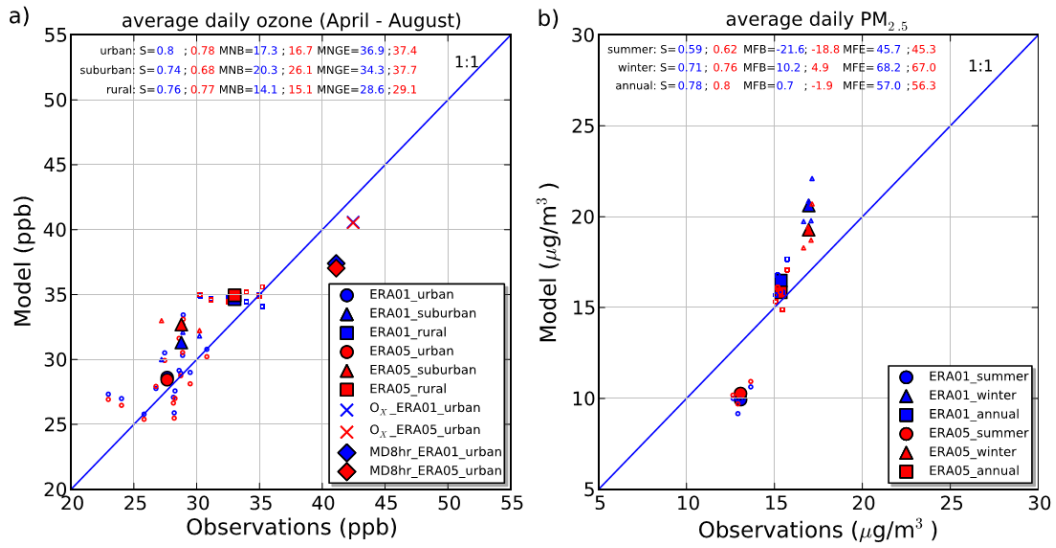


Figure 4. Scatter plots and scores for the sensitivity test on climate model driven meteorology for ozone and PM_{2.5}.

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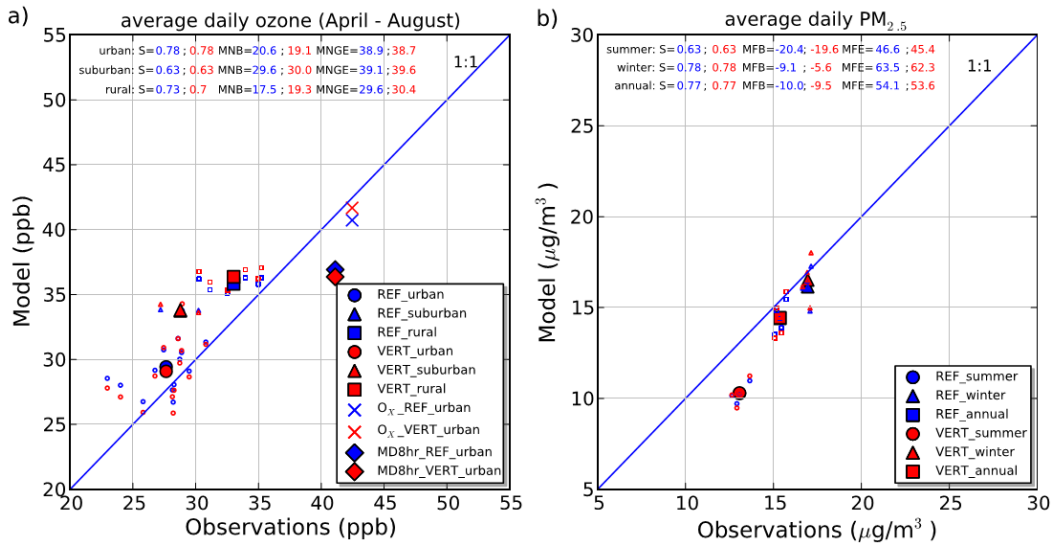


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Figure 5. Scatter plots and scores for the sensitivity test on the resolution of meteorology for ozone and PM_{2.5}.

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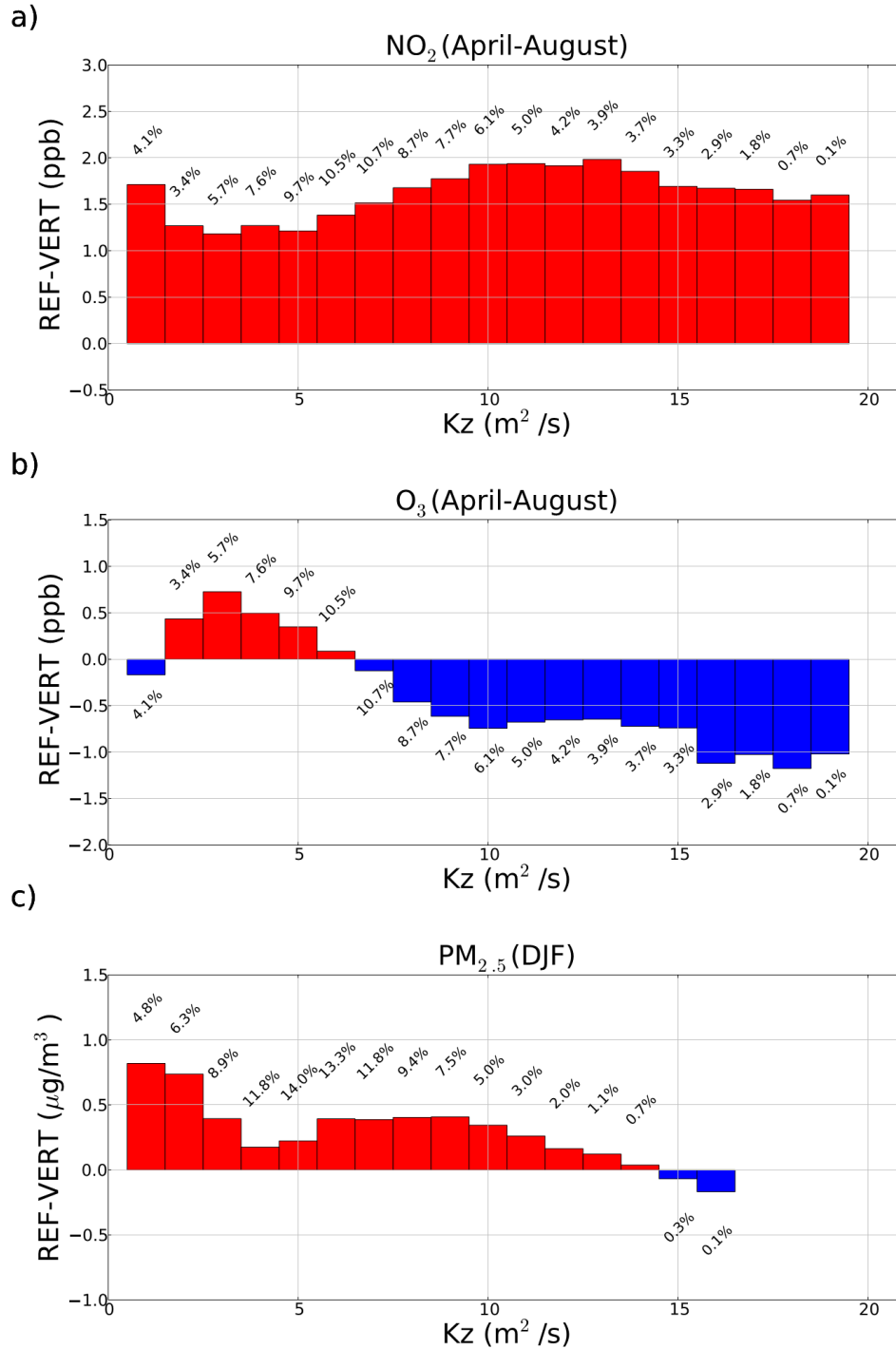
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Figure 6. Scatter plots and scores for the sensitivity test on the ~~model-CTM's~~ vertical resolution for ozone and PM_{2.5}.

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Figure 7. Difference in average daily simulated NO₂ (a), ozone (b) and PM_{2.5} (c) concentrations between VERT (12 vertical layers) and REF (8 vertical layers) at urban areas per range of K_z (bins of 1 m²/s). Positive differences indicate that the refined vertical mesh leads to increased pollutant concentration and vice versa. The occurrence of sensitivity values within each K_z range is also provided.

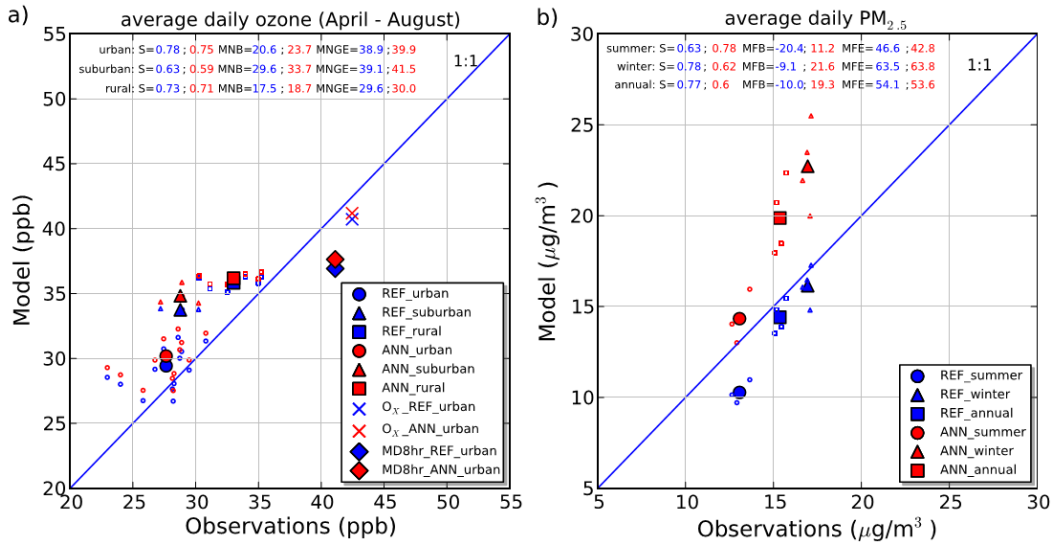
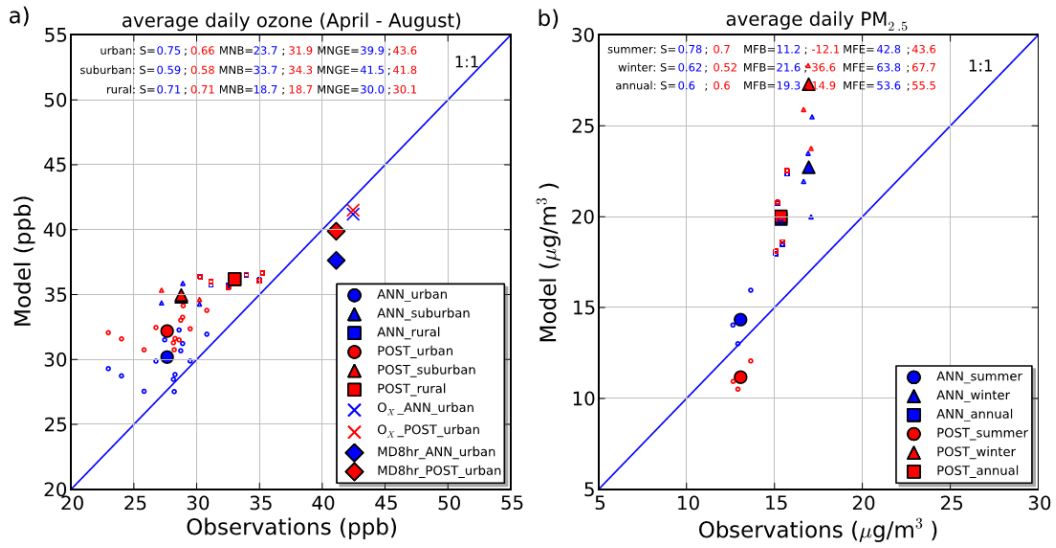


Figure 8. Scatter plots and scores for the sensitivity test on the annual emission totals for ozone and PM_{2.5}.

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Figure 9. Scatter plots and scores for the sensitivity on the post-processing (temporal analysis and chemical speciation) technique applied on the annual emission totals for ozone and PM_{2.5}.

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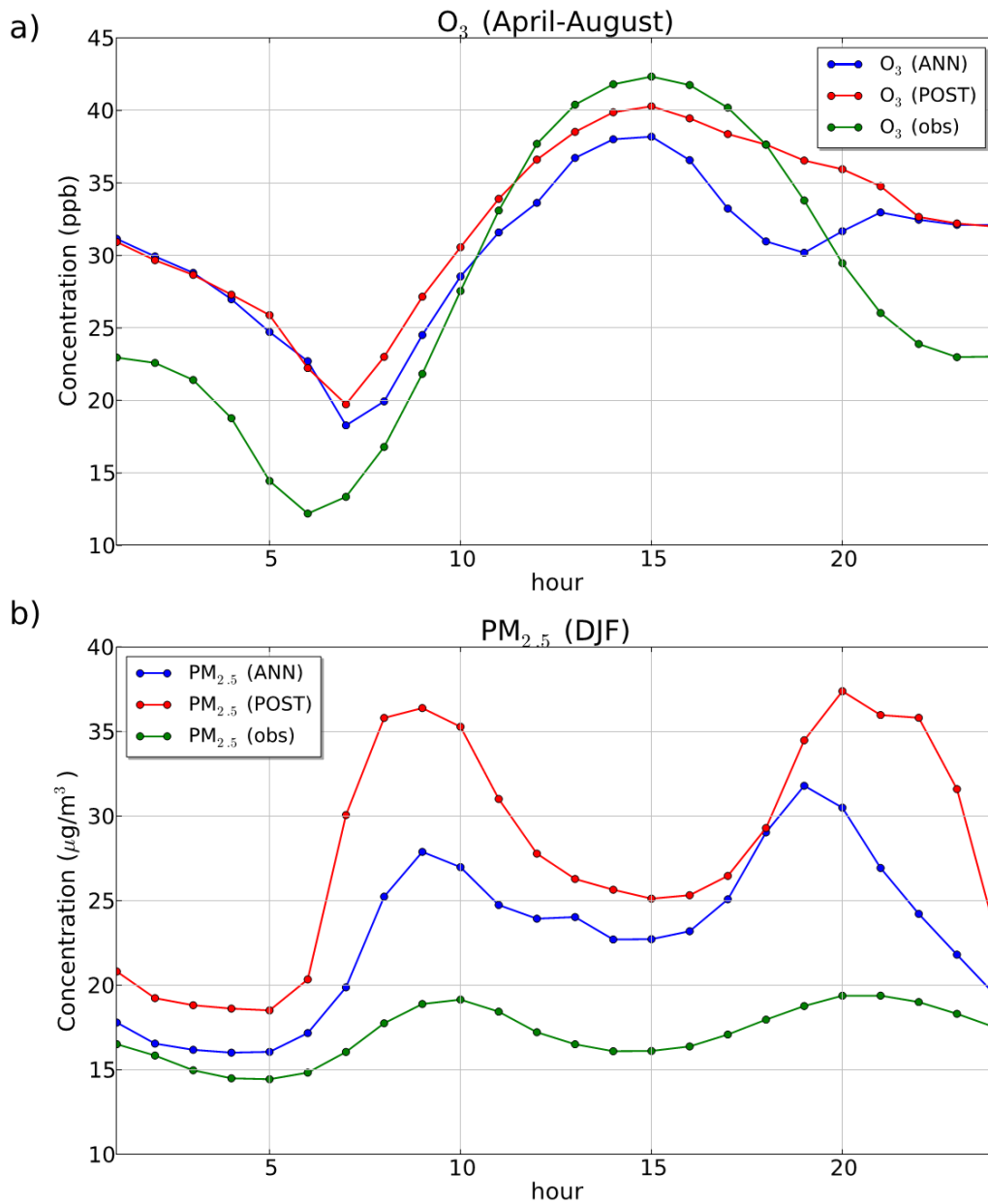
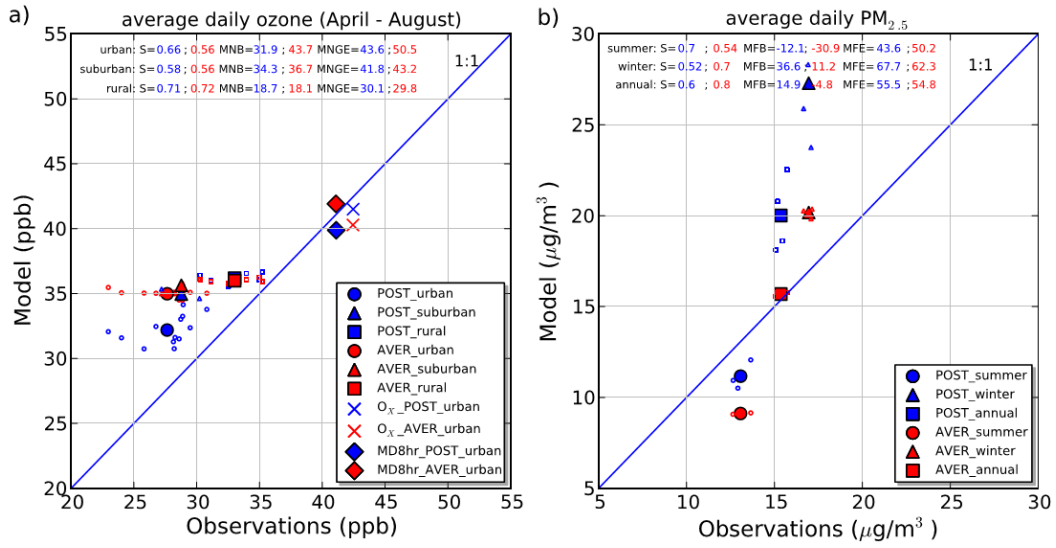


Figure 10. Average Mean diurnal variation of (a) ozone concentrations averaged over the from April- to August period (a) and (b) wintertime PM_{2.5} (b) concentrations in the urban area.

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Figure 11. Scatter plots and scores for the sensitivity test on the resolution of the emission inventory for ozone and PM_{2.5}.

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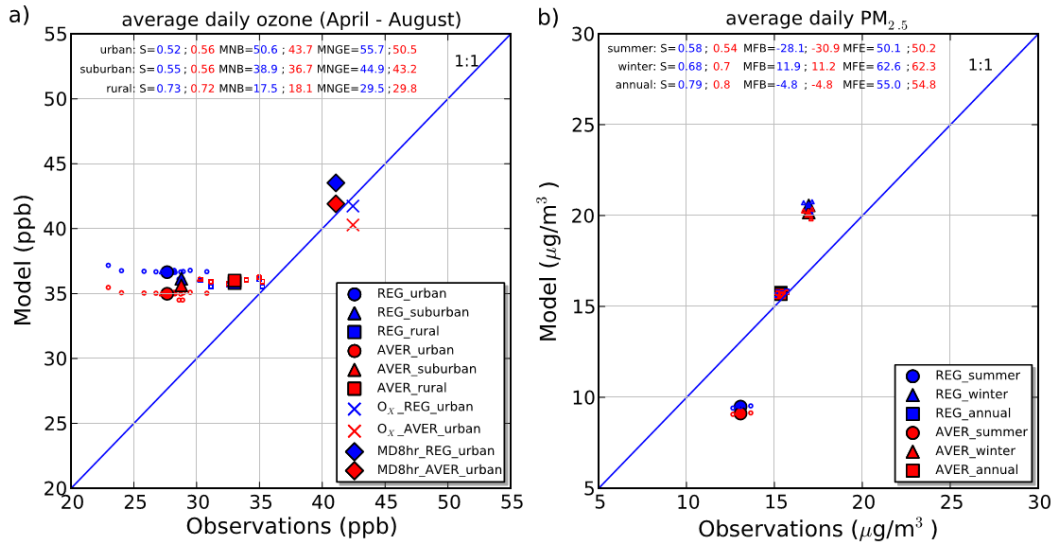
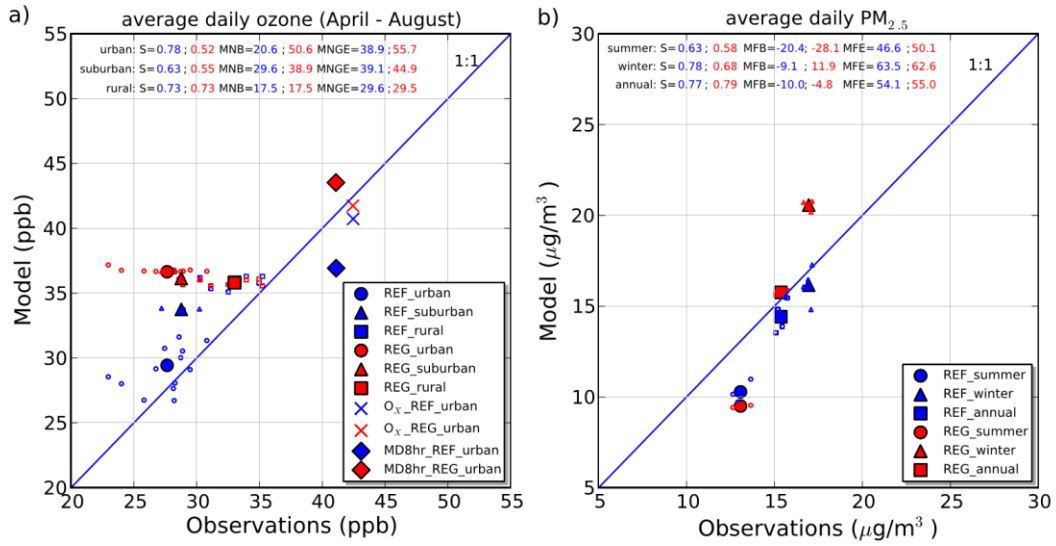


Figure 12. Scatter plots for the sensitivity test on model resolution for ozone and PM_{2.5}.

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Figure 13. Panel a: Scatter plots of daily average ozone concentrations at urban, suburban and rural stations from the REF and REG simulations. The odd oxygen (O_x) and daily maximum at urban locations is also shown. Panel b: daily average PM_{2.5} concentrations in wintertime (DJF), summertime (JJA) and on annual basis over urban stations.

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