# Air-quality in the mid-21st century for the city of Paris under two climate scenarios; from regional to local scale.

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# 16 Abstract

17

18 Ozone and PM2.5 concentrations over the city of Paris are modeled with the CHIMERE air-19 quality model at 4km x 4km horizontal resolution for two future emission scenarios. A high 20 resolution (1x1 km) emission projection until 2020 for the greater Paris region is developed by 21 local experts (AIRPARIF) and is further extended to year 2050 based on regional scale emission projections developed by the Global Energy Assessment. Model evaluation is performed based 22 23 on a 10-year control simulation. Ozone is in very good agreement with measurements while PM<sub>2.5</sub> is underestimated by 20% over the urban area mainly due to a large wet bias in wintertime 24 25 precipitation. A significant increase of maximum ozone relative to present time levels over Paris 26 is modeled under the "business as usual" scenario (+7 ppb) while a more optimistic mitigation 27 scenario leads to moderate ozone decrease (-3.5 ppb) in year 2050. These results are substantially 28 different to previous regional scale projections where 2050 ozone is found to decrease under both 29 future scenarios. A sensitivity analysis showed that this difference is due to the fact that ozone 30 formation over Paris at the current, urban scale study, is driven by VOC-limited chemistry, 31 whereas at the regional scale ozone formation occurs under NO<sub>x</sub>-sensitive conditions. This

1 explains why the sharp  $NO_x$  reductions implemented in the future scenarios have a different 2 effect on ozone projections at different scales. In rural areas projections at both scales yield 3 similar results showing that the longer time-scale processes of emission transport and ozone 4 formation are less sensitive to model resolution. PM<sub>2.5</sub> concentrations decrease by 78% and 89% 5 under "business as usual" and "mitigation" scenarios respectively compared to present time period. The reduction is much more prominent over the urban part of the domain due to the 6 7 effective reductions of road transport and residential emissions resulting in the smoothing of the 8 large urban increment modelled in the control simulation.

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#### 10 **1** Introduction

11 Climate change can affect air quality through a number of mechanisms related to meteorological 12 variables such as temperature, humidity, precipitation, solar radiation, wind speed and the 13 planetary boundary layer height. If the effects of atmospheric pollutants (greenhouse gasses, 14 other gaseous species and aerosols) on climate change have been extensively investigated over 15 the last decades (IPCC, 2001, 2007) the impact of these emissions and changed climate on air 16 quality has only raised interest during the last few years and many issues remain open. Most of 17 these studies use global chemistry-transport models (CTMs) to study the impact of changing 18 climate on tropospheric ozone at either global (Brasseur et al., 1998; Liao et al., 2006; Prather et 19 al., 2003; Szopa and Hauglustaine, 2007) or regional scale (Murazaki and Hess, 2006; Szopa et 20 al., 2006). The resolution of these studies, typically a few hundreds of kilometers, is insufficient 21 to capture the high spatial variability of air pollution due to the short lifetime of tropospheric 22 ozone and particulate matter (in the range of some hours up to a few days) (Seinfeld and Pandis, 23 2006) as well as to the sharp horizontal gradients of anthropogenic emissions over urban areas. 24 Moreover, the chemistry-climate models used in such large-scale studies suffer from a simplistic 25 representation of regional scale chemistry.

Only few, recent, applications study the impact of changing climate on air quality using regional CTMs. These models include more sophisticated chemistry at typical resolutions of tenths of kilometers and therefore provide a better understanding of the underlying mechanisms at regional scale. The first documented modeling study on the impact of future conditions at regional scale is that of Hogrefe et al. (2004) while at the same time Knowlton et al. (2004) took a step forward and performed a health impact assessment study for the greater New York region. Since then several researchers performed similar regional applications in order to derive future
 predictions of ozone and/or aerosols using offline climate-chemistry models (Colette et al., 2013;
 Kelly et al., 2012; Lam et al., 2011; Langner et al., 2005; Nolte et al., 2008; Steiner et al., 2006;
 Szopa and Hauglustaine, 2007; Tagaris et al., 2009), on-line climate-chemistry models (Forkel
 and Knoche, 2006, 2007) or multi-model studies (Colette et al., 2012; Langner et al., 2012).

6 An open research challenge today is the impact of climate change on air-quality at urban scale. 7 Enhanced health effects due to increasing urbanization (UNFPA, 2007) should be addressed 8 primarily at such scale especially given that the efforts to mitigate air-pollution are more intense 9 in areas where the largest health benefits are observed (Riahi et al., 2011). The common 10 approach to study the effects of climate change and emissions at city level is to use regional scale 11 CTMs. However the resolution of current model studies remains too coarse to represent the high 12 spatiotemporal variability of pollutants at urban scale. Regional scale emission inventories fail to 13 represent the plethora of emission activities in terms of e.g. the underlying technologies used 14 within a specific source sector or to take into account the habitual patterns of the population at 15 large cities (Markakis et al., 2010, 2012). The substantial part of the published work uses 16 emission projections based on scenarios developed to represent changes at global scale and are 17 rarely well suited for local scale assessment e.g. they lack the detailed representation of all 18 emission activities at a finer scale. For example, the more recent Representative Concentration 19 Pathways (RCPs) (van Vuuren et al., 2011) developed to support modeling activities for the new 20 assessment of the Intergovernmental Panel on Climate Change (IPCC) incorporate a diversity of 21 radiative forcing scenarios that may be suitable for large scale chemistry-climate models but not 22 appropriate to represent emission trends for specific cities and therefore regional scale air-quality 23 assessment (Butler et al., 2012). Grid resolution is another key issue for urban scale modeling as 24 shown by previous research (Arunachalam et al., 2006; Cohan et al., 2006; Valari and Menut, 25 2008). The assumption of instantaneous mixing of emissions within the volume of the regional 26 scale CTMs' grid cells is unsuitable to represent local scale chemistry (such as ozone titration) 27 when the grid-cell size is too large compared to the actual size of the emitting areas. It has been 28 shown previously that use of coarse resolution leads to underprediction of daily ozone maxima 29 and to overprediction of daily average ozone levels (Arunachalam et al., 2006; Tie et al., 2010; 30 Vautard et al., 2007). Valin et al. (2011) show that grid resolution finer than ~10 km is often 31 necessary to capture ozone chemistry near areas with large  $NO_x$  emissions. Due to the non-

1 linearity of ozone chemistry the inability of large grid sizes to differentiate urban and rural areas 2 can have a profound effect on the simulated ozone concentrations (Silman et al., 1990). Forkel 3 and Knoche (2007) found that the increase of future maximum ozone concentrations in Europe 4 occurred in areas where high mixing ratios of NO<sub>x</sub> coincided with increased isoprene emissions, 5 indicating that the failure of coarser resolutions to efficiently distinguish between urban and rural 6 areas would result in overestimation of future ozone in cities near to high isoprene emitting 7 sources. The current challenge is to develop realistic long-term emission projections at a city-8 scale and bridge the gap between local, regional and global scale modeling.

9 The aim of our study is to develop mid-21<sup>st</sup> century horizon air quality projections over the 10 greater Paris area under two consistent emission and climate scenarios (10-year continuous 11 simulation). To the authors knowledge this is the first time where a 10-year air-quality projection 12 under climate and city level emission changes has been conducted at urban scale and resolution 13 as fine as 4 km. Local emission projections used in this study are compiled by merging several 14 pieces of information: i) a high resolution (i.e. 1 km) city specific emission inventory developed 15 by AIRPARIF (present time emissions); ii) a city-scale short-term projection (until 2020) 16 considering air-quality policies already in place and planned for the city for the next years iii) two mid-21<sup>st</sup> century regional scale emission scenarios including both climate and regional air 17 18 quality policies developed by the Global Energy Assessment (GEA) (Riahi et al., 2012). The 19 "reference" (REF) scenario is consistent with long-term climate outcomes of the RCP-8.5 20 adopting all current and planned air quality legislation until 2030 and assumes that no climate policies are implemented thereafter. The "mitigation" (MIT) scenario is consistent with long-21 22 term climate outcomes of the RCP-2.6 and additionally assumes stringent climate mitigation 23 policies. The final fine scale, mid-21<sup>st</sup> horizon emission inventory for the greater Paris area, the 24 Ile-de-France region (IdF), was obtained by scaling the 2020 high resolution emission inventory 25 to the year 2050 using the regional scale GEA scenarios.

After describing the methodology in Section 2, an evaluation of the modelling system is described in Section 3 and scenario results are presented in Section 4. Our conclusions are discussed in Section 5.

29

#### 30 2 Materials and methods

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

An "off-line" modeling approach is used to assess air-quality in the study region as a response to climate and emission changes. For both climate and air-quality simulations we use a dynamical downscaling approach consisting of two one-way nesting steps: from global to regional scale over Europe and from regional to local scale over the IdF region. IPSL-CM5A-MR global circulation model (Dufresne et al., 2013) is used to derive future projections of the main climate drivers (temperature, solar radiation etc.) using a set of available global greenhouse gas emission scenarios (RCP-2.6 and RCP-8.5).

12 Global climate output is downscaled with the Weather Research and Forecasting (WRF) mesoscale climate model (Skamarock and Klemp, 2008) at two steps: first over a 50km 13 14 resolution grid over Europe and then on a 10 km resolution grid over France using one-way 15 nesting. We drive the CTM over IdF using the 10km meteorology acknowledging both the 16 computation overhead of a refined meteorology and the results of previous work over the same 17 region using the CHIMERE CTM (Menut et al., 2005a; Valari and Menut., 2008). More 18 specifically Menut et al. (2005a) showed that apart from the coastal areas where a refined 19 meteorology improved air-quality modelling results, in the rest of France ozone peaks were 20 better captured with lower resolution meteorological input. Valari and Menut (2008) showed that 21 a refined meteorological input gives similar results for ozone and that model performance is 22 much more sensitive to the resolution of emissions than to meteorology. These results suggest 23 that in areas having the geographical characteristics of the greater Paris area (flat topography at 24 great distance from any mountains or the ocean) increasing the resolution of the meteorological 25 input does not necessarily improve the results of the chemistry-transport modelling. On the other 26 hand Flagg and Taylor (2011) investigated the impact of the resolution of the surface layer input 27 data on modelling results of high resolution simulations in an urban coastal environment and 28 found that these results were quite dependent on that resolution. There are indications that one 29 needs to go to very high resolution in order to capture the urban heat island circulation (LeRoyer 30 et al. 2014) and the effects of the surface layer changes (Flagg and Taylor, 2008) in an urban 31 meteorological simulation but this analysis is beyond the scope of the present work.

1 Air-quality data are downscaled with the CHIMERE model (Menut et al., 2013) a detailed 2 description of which can be found at: http://www.lmd.polytechnique.fr/chimere. CHIMERE is an 3 off-line chemistry-transport model (CTM), which models atmospheric chemistry and transport, 4 forced by anthropogenic emissions, biogenic emissions, a meteorological simulation and 5 boundary conditions. It is a cartesian-mesh grid model including gas-phase, solid-phase and 6 aqueous chemistry, biogenic emissions modeling depending on meteorology with the MEGAN 7 model (Guenther et al., 2006), dust emissions (Menut et al., 2005b) and resuspension (Vautard et 8 al., 2005). Gas-phase chemistry is based on the MELCHIOR mechanism (Lattuati, 1997) which 9 includes more than 300 reactions of 80 gaseous species. The aerosols model species are sulfates, 10 nitrates, ammonium, organic and black carbon and sea-salt (Bessagnet et al., 2010) and the gas-11 particle partitioning of the ensemble Sulfate/Nitrate/Ammonium is treated by the ISORROPIA 12 code (Nenes et al., 1998) implemented on-line in CHIMERE.

13 Initial and boundary conditions are taken from global scale concentrations modelled with the 14 coupled LMDz-INCA global chemistry model (Hauglustaine et al., 2004; Szopa et al., 2012) and are then downscaled with CHIMERE using two-level one-way nesting first at a 0.5° resolution 15 16 grid (~50 km) over Europe (Colette et al., 2013) and then at the 4km resolution grid over the IdF 17 region (local scale runs). For more details on the global and regional modeling exercises the 18 reader is referred to Hauglustaine et al. (2014) and Colette et al. (2013) respectively. We only 19 note that the global scale simulations done with the LMDz-INCA model has a slightly different 20 chemical speciation from that of CHIMERE. Finally Table 1 presents the basic configuration of 21 the global, regional and local simulations used in the present study.

We performed three sets of 10-year runs at local scale: a continuous control run (CTL) from 1995 to 2004 representing present time air-quality and two continuous runs over the 2045-2054 decade representing air-quality projections to the mid-21<sup>st</sup> century horizon under two different pathways of climate and emissions. To minimize the effect of inter-annual variability and increase the statistical significance of model output longer term simulations should have been performed. However, given the high resolution of the modeling exercise the computational cost of longer simulations was considered too high for the present study.

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#### 30 2.1 Models setup

1 The modeling domain has a horizontal resolution of 4 km and consists of 39 grid cells in the 2 west-east direction and 32 grid cells in the north-south direction, thus covering a total region of 3 156 x 128 km<sup>2</sup>. Twelve  $\sigma$ -p hybrid vertical layers were used to represent the atmospheric column 4 from the surface up to approximately 500hPa (~5.5 km) with the thickness of the first layer being 5 8 m. Simulations were performed with a time step of 5 minutes. Hourly meteorological fields 6 necessary for CHIMERE were modeled with WRF model v3.4. WRF simulations were carried 7 out on a 10 km resolution grid of 90x85 cells. Vertically the domain is divided into 31  $\sigma$ -layers 8 extending up to 55hPa ( $\sim 20$  km). The time integration step was set to 1 minute. The physical 9 options used for these simulations are the WRF Single Moment 6-class microphysics scheme 10 (Hong and Lim, 2006), RRTM (rapid radiative transfer model) long-wave radiation scheme (Mlawer et al., 1997), Dudhia short-wave radiation scheme (Dudhia, 1989), NOAH land surface 11 12 model (Chen and Dudhia, 2001), Asymmetric Convective Model version 2 Planetary Boundary 13 Layer scheme (Pleim, 2007) and Kain-Fritsch cumulus parameterization scheme (Kain, 2004).

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#### 15 **2.2 Data and metrics for model evaluation**

16 CHIMERE model evaluation has been performed in numerous studies at both regional (see e.g. 17 Solazzo et al., 2013a, 2013b and references therein) and urban scales (Hodzic et al., 2005; 18 Vautard et al., 2007). However, due to the climate component of the study our modeling setup 19 requires a different evaluation framework. We use 29 surface monitors of the local air-quality 20 network AIRPARIF classified as urban (17 sites), suburban (4 sites) and rural (8 sites). Only 21 monitoring sites with more than 70% of available data (hourly values) through the 10-year period 22 were considered for the evaluation process. Observations are conducted at a height ranging from 23 3 to 8m from the surface, which makes the first model layer concentrations directly comparable 24 with the observations.

Given that adverse health impact is mainly correlated to daytime ozone levels in the present framework we focus on the evaluation of daytime and maximum ozone only. A variety of metrics are used to evaluate model performance. For ozone the widely used mean normalized bias (MNB) and mean normalized gross error (MNGE) are estimated. US EPA suggests that MNB for modeled ozone concentrations should lie within the +-15% range and MNGE not exceeding 35%. Regarding PM<sub>2.5</sub> the Mean Fractional Bias (MFB) and the Mean Fractional Error (MFE) are used. A literature review of targeted range of values for fine particles can be found in 1 EPA (2007). The narrowest of the reported ranges suggests that for modeled  $PM_{2.5}$  MFB should 2 fall within the +-30% range and MFE not exceeding 50%.

3 We extract these metrics from the daytime concentrations values and not the decade average 4 bearing in mind that this is not typical for runs forced by climate simulationsbut for operational 5 forecast evaluation. Here we do not use re-analyses to force the regional CTM but global and 6 regional climate simulations. Consequently one should expect lower scores than those yielded by 7 an air-quality forecast simulation, especially in the presence of climate biases (Colette et al., 8 2013; Menut et al., 2013). Since this work is original in its concept we aim to evaluate whether 9 the urban scale setup is sufficiently realistic but utilizing metrics that are timely averaged on a 10 scale finer than a climatological one.

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## 12 **2.3 Climate and emission scenarios**

13 Two long-term scenarios are used in the global-scale simulations of future climate conditions: 14 RCP-8.5 and RCP-2.6. RCP-8.5 (Riahi et al., 2011) is a "reference" type scenario with no mitigation for greenhouse gases leading to a global radiative forcing of 8.5 W  $m^{-2}$  by the end of 15 the century. RCP-2.6 (van Vuuren et al., 2011a) leads to a global radiative forcing of 2.6 W m<sup>-2</sup> 16 17 by the end of the century (2100), consistent with the goal of limiting the increase of global 18 average temperature due to human activity to 2°C. Both RCP scenarios include century-long 19 estimates of air pollutant emissions, including aerosols, and were used as input in the LMDz-20 INCA global chemistry model.

21 Present time emission estimates for the IdF region are available in hourly intervals, with a spatial 22 resolution of 1x1 km. The emission dataset was compiled by local experts using a variety of city-23 specific information integrating a number of anthropogenic activities in the IdF region 24 (AIRPARIF, 2012). The spatial allocation of emissions is completed with proxies such as high-25 resolution population maps, road network and the location of industrial units. It includes 26 emissions of CO, NO<sub>x</sub>, Non-methane Volatile Organic Compounds (NMVOCs), SO<sub>2</sub>, PM<sub>10</sub> and 27 PM<sub>2.5</sub> with a monthly, weekly and diurnal (source specific) temporal resolution. The distribution 28 of emissions among different activity sectors reveals that in the IdF region the principal emitter 29 of NO<sub>x</sub>, on annual basis, is the road transport sector (50%), for NMVOCs the use of solvents 30 (46%) and for fine particles the residential sector (40%). Annual emissions rates within the 31 simulated decade were kept constant; only the vertical distribution of point source emissions across model layers varies in time since it depends on several meteorological variables such as
 temperature and wind (plume-rise algorithm) (Scire et al., 1990). Finally the 1 km resolution
 emissions were aggregated to the 4 km resolution grid.

4 Mid-21<sup>st</sup> century (year 2050) emission projections for IdF incorporate local scale information, 5 including the implementation of planed policies. The short-term component consists of the local 6 scale 2020 emission projection compiled by AIRPARIF in the framework of the "Plan de 7 Protection de l'Atmosphere d'Ile de France" (PPA) and with the support of the "Direction 8 Regionale et Interdepartementale de l'Environnement et de l'Energie d'Ile de France" (DRIEE-9 IF). Emissions for 2020 correspond to a "business as usual" scenario assuming the 10 implementation of all regulatory measures planned by the PPA.

11 The long-term component of the projection is established by linking two primary sources of 12 information: (i) the aforementioned 2020 AIRPARIF inventory and (ii) a set of scenarios from 13 the recently published Global Energy Assessment (GEA) (Riahi et al., 2012) to extend 14 projections until 2050. The GEA scenarios, while consistent with similar long-term climate 15 outcomes as the RCPs, include a more detailed representation of short-term air quality 16 legislations from the GAINS model (Amann et al., 2011). Outputs are based on the MESSAGE model (Riahi et al., 2007) and include estimates of energy and greenhouse gas emissions for 11 17 global regions while air pollutants are further available at 0.5° x 0.5° resolution estimated based 18 19 on inventory data described in Granier et al. (2011) and Lamarque et al. (2010) and an exposure-20 driven algorithm for the downscaling of the regional air-pollutant emission projections. The 21 GEA scenarios have been used to estimate global health impacts of outdoor air pollution (Rao et 22 al., 2012, 2013) as well as for regional impacts analysis (Colette et al., 2012, 2013). Two GEA 23 scenarios are used in our analysis: i) a "reference" (REF) scenario (consistent with long-term 24 climate outcomes of the RCP-8.5) which adopts all current and planned air quality legislation 25 until 2030 and assumes that no climate policies are implemented thereafter ii) a "mitigation" 26 (MIT) scenario (consistent with long-term climate outcomes of the RCP-2.6) which additionally 27 assumes stringent climate mitigation policies consistent with a target of 2°C global warming by 28 the end of the century (2100). Thus the MIT scenario can be used to assess the co-benefits of 29 climate induced strategies. We derived national-wide annual totals (for France) in 2020 and 2050 30 for a number of pollutants from the GEA scenarios. 2050 over 2020 scaling ratios were 31 calculated for each pollutant for each source sector accounted for in the GEA dataset. The

resulting ratios are used as coefficients to scale the local 2020 AIRPARIF inventory without
 modifying the spatial and temporal patterns.

3 We decided to derive scaling ratios based on national totals and not from the GEA 0.5° resolution 4 grid to avoid adopting the relatively simplistic approaches of the GEA downscaling algorithm for 5 the distribution of emissions at the finer scale. Thus, our approach benefits from the highly 6 resolved emission variability over Paris from the AIRPARIF inventory but it must be noted that 7 at the same time we inherently assume the energy transition over the 2020 to 2050 period 8 according to the REF and MIT scenarios. This is particularly relevant as the GEA scenarios 9 implement significant improvements in energy efficiency and a complete implementation and 10 continuation of air-quality policies in Europe. Thus we observe significant reductions in 11 emissions for the IdF region in 2050 (Table 2). While using different scenarios would lead to 12 alternative estimates of 2050 emissions, it is important to note that significant technological 13 change in the energy systems would likely mean that air pollutant emissions in Europe would 14 decline over the long-term in any case, although the exact distribution of such reductions is 15 uncertain.

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### 18 **3** Model evaluation based on the control simulation (present time)

### 19 3.1 Meteorology

It is beyond the purpose of this study to provide an in-depth meteorology evaluation, however to build confidence in the meteorological fields used as input for our air-quality simulations we present a short model evaluation based on the CTL run.

23 WRF meteorology for the CTL simulation ran over a 10 km resolution grid. Model results are 24 compared to surface observations from 7 meteorological stations found inside the IdF region. 25 Only one of these monitors (MONT) is actually inside the city of Paris. The mean wintertime 26 (DJF) and summertime (JJA) modeled and observed daily average values were calculated for 4 27 different variables relevant to air quality: 2m temperature, wind speed at 10m, relative humidity 28 and total precipitation (Table 3). A warm bias during summer reaches +0.8°C in the city center 29 (MONT station). This is bound to lead to enhanced ozone formation due to the thermal 30 decomposition of PAN releasing NO<sub>x</sub> (Sillman and Samson, 1995). Modeled wind speeds are 31 higher than the observed ones both during winter and summer, a bias consistent with previous 32 studies (see e.g. Jimenez et al. (2012) for WRF or Vautard et al. (2012) for other models). This is

expected to enhance pollutants' dispersion and lead to less frequent stagnation episodes. The
model underestimates summertime relative humidity (bias=-13.3%). A systematic wet bias in
wintertime precipitation is observed in the city reaching +26.5%. This is expected to lead to
underestimation of PM levels through rain scavenging (Fiore et al., 2012).

5 6

#### 3.2 Ozone concentrations

7 Scatter plots in Fig. 2 compare modeled ozone with surface measurements. Mean daytime 8 surface ozone and  $O_x$  (NO<sub>2+</sub>O<sub>3</sub>) (averaged over the April-August period) as well as the daily 9 maximum of 8-hour running means (MD8hr) from the urban stations are shown. Modeled 10 daytime ozone is in very good agreement with measurements over the urban cluster with a bias 11 of only +0.25 ppb (0.8% overestimation). Note that the spread of ozone bias among individual 12 stations is also small. O<sub>x</sub> is often used as an indicator of ozone photochemical build-up because it 13 rules out titration process. Therefore, the low O<sub>x</sub> bias indicates that local scale modeling 14 reproduces sufficiently both urban titration (daytime) and photochemical formation.  $NO_2$  bias 15 (not shown) is also low (-0.32 ppb). Model results for ozone are considered satisfactory given the 16 nature of these runs (forced by climate simulations and not forecast meteorology as discussed in 17 Sect. 2.2). The MNB (+16.5%) and MNGE (+40.6%) are outside but near the US EPA targeted 18 ranges (+-15% and +35% respectively). For downtown Paris sites (not shown) MNB is +10.9% 19 and MNGE is equal to +39.8%. However, the model underestimates MD8hr by -7.8% (bias = -3.2ppb). The 95<sup>th</sup> percentile (not shown) of observed and modeled ozone daily maxima differ by 20 21 13.8 ppb (-20.1%) indicating that the model fails to reproduce ozone levels under extreme 22 photochemical episodes which in any case are produced in timely short periods of very specific 23 meteorological conditions characterized by stagnated air masses and low vertical mixing 24 favoring ozone build-up. The meteorological input used in our local simulation represents poorly 25 the observed wind speeds which are overestimated significantly (Sect. 3.1). This affects 26 stagnation (Jacob and Winner, 2009; Vautard et al., 2007) but also vertical diffusivity through an 27 increased boundary layer height. At the same time, the climate model suffers from a cold and 28 overcast bias (Colette et al., 2013) that inhibits the emission of biogenic precursors and ozone 29 buildup.

30 Overestimation of mean daytime ozone is observed in all suburban and rural stations (+10.9% 31 and +11.3% respectively) even though  $O_x$  is in relatively good agreement with measurements. This points to an underestimation of  $NO_2$  and titration is represented less faithfully. Ozone overestimation is due to some extent to the boundary conditions. This is supported by the fact that both upwind and downwind rural stations show the same level of overestimation portraying a spatially symmetric influence of boundary conditions (not shown). In the urban areas the strong titration probably depletes the excess of ozone from the boundaries pointing to a possible exaggeration of urban titration in the model. MNB and MNGE scores for suburban and rural sites (Fig. 2) in general lie close but outside the US EPA targeted ranges.

8

#### 9 3.3 Fine particulate matter

10 Modeled  $PM_{2.5}$  surface concentrations are compared to all available measurements (i.e four 11 urban sites only one of which located in Paris). Results for wintertime (DJF), summertime (JJA) 12 and on annual basis are shown in Fig. 3. The model underestimates  $PM_{2.5}$  by 20% in wintertime 13 (-14% at the Paris site). This underestimation could be attributed to the significant positive bias 14 in modeled precipitation and wind speeds (Sect. 3.1). Another explanation could be that the 15 monthly temporal profiles used to distribute the annual emissions of residential heating are based 16 on mean temperature variations. This could lead to strong underestimation of emissions in 17 wintertime. The analysis of the annual model results yields even higher underestimation for both 18 Paris site and domain-wide values (-18.9% and -26.6% respectively). The poorer model 19 performance during summertime suggests a possible underestimation of summertime emissions 20 in the local inventory. Vehicle-induced resuspension of particles might be a significant missing 21 source, especially during the drier summertime months (Schap et al., 2009 and references 22 therein).  $PM_{2.5}$  underestimation, to some extent, is due to a poor representation of secondary 23 organic aerosol (SOA) in the model (modeled SOA concentrations are <0.5% in wintertime and 24 ~1.5% in summertime). This is a well-known shortcoming of regional CTMs (Simpson et al., 25 2007; Solazzo et al., 2012; Stern et al., 2007). In particular Hodzic et al. (2009; 2010) found 26 significant underprediction of observed SOA concentrations modeled with CHIMERE in the 27 framework of the MILAGRO campaign. Modeled meteorology (wind and precipitation 28 overestimation) also contributes to the observed PM underestimation. Nevertheless, wintertime 29 and annual metrics are within the US EPA targeted range (MFB within +-30% and MFE <50%).

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#### 31 **3.4** Conclusions on model evaluation

1 The model is able to represent the main features of ozone photochemical cycle (ozone formation 2 and titration) over the Ile-de-France region but fails to reproduce high ozone episodes. Urban 3 features of ozone chemistry, such as ozone titration are resolved showing that the emission 4 inventory is realistic. Therefore, this local scale modeling makes it possible to discern between 5 urban, suburban and rural areas and provides a good representation of both ozone build-up and 6 titration processes. The underestimation of  $PM_{2.5}$  is attributed to errors in modeled meteorology 7 (high precipitation and winds) and possibly to missing emission sources and chemical processes 8 (secondary organic aerosol).

9

#### 10 4 Future scenario analysis

#### 11 **4.1** Climate projections for the mid-21<sup>st</sup> century in Paris

Table 4 summarizes the projections of key meteorological variables. It is beyond the scope of this study to disentangle the effects of climate and emission changes on air-quality and therefore only a brief qualitative discussion is provided in this section.

Under the REF scenario, which assumes that greenhouse gases increase monotonically until the end of the century, the annual mean surface temperatures in the mid-21<sup>st</sup> horizon exhibits an overall domain-wide increase of +1.1 °C (+1.0°C over the city) compare to the present time period. Consequently, enhanced ozone formation is expected especially in the rural part of the domain due to increase of biogenic organic compounds (BVOCs). Monoterpenes are especially sensitive to temperature while isoprene to both temperature and sunlight. We do not observe any significant changes to short-wave radiation, RH or precipitation under the REF scenario.

22 A -0.5°C decrease in the annual surface temperature compared to present time period is modeled 23 under the MIT trajectory, with two particularly cold years (2051 and 2052) lowering the decade 24 average. The inter-annual variability (standard deviation) is 2.54°C around a decade average of 25 11.3°C while under the REF scenario the variability is 0.8°C around a decade average of 13°C. 26 The internal natural variability of temperature can be very large at regional scale (Deser et al., 27 2012), which stresses the need for longer simulation periods over which the inter-annual 28 variability would be smoothed. Lower temperatures are expected to inhibit ozone formation 29 while the drop of shortwave downward radiation by 16.6% relative to present will lead to less 30 BVOCs in the rural areas. We observe a significant increase in total precipitation compared to 31 the control simulation by a factor of 3.2 under the MIT pathway where, similarly to temperature,

two particularly wet years (2050 and 2051) are simulated. Modeled precipitation is high only on a limited amount of days within those two years (standard deviation=0.73 around a decade average of 0.32). In the regional simulation from which boundaries were utilized to force the local runs, climate is also shown significantly wetter under MIT by a factor of 3.6.

5

#### 6 4.2 Ozone projections

Mid-21<sup>st</sup> century ozone projections averaged over the April through August period are presented in Fig. 4 for the REF and the MIT scenarios. Panel a shows present time MD8hr levels (CTL simulation) whereas panels b and c provide differences between each of the REF and MIT scenarios and the CTL simulation. The CTL simulation is able to dissociate the fast ozone titration process from the longer scale photochemical build-up: low ozone levels are modeled over the city of Paris (36-38 ppb) due to titration by NO (road transport mainly) and higher levels (47-50 ppb) are modeled at the surrounding rural area due to photochemical formation.

Projections show an increase in maximum ozone (+7 ppb) over the city of Paris under the REF scenario relative to CTL and only a small decrease under the MIT scenario (-3.5 ppb). The average daytime ozone (not shown) also increases substantially in downtown Paris by approximately +10 ppb under REF. The results are consistent with the near future projection of Roustan et al. (2011) who also employed local scale emissions and modelled maximum ozone increase in Paris for the year 2020. Reductions in rural ozone are modelled under both scenarios: REF: -3.2 ppb and MIT: -13.5 ppb.

21 Previous studies have shown that anthropogenic emissions over urban areas are the driver of 22 ozone concentrations compared to climate, boundary conditions and other influencing factors 23 (Colette et al., 2013). Therefore, to gain further insight in the ozone response one should study 24 the differences between present and future time ozone precursors' emissions. However emission 25 changes under different chemical regimes may also lead to different ozone responses. For 26 example, air-quality projections conducted at a 0.5°x0.5° resolution grid over Europe forced with 27 GEA emissions and the same meteorology as in our study found significant decrease compared 28 to present time values in ozone concentrations over the Paris region at the 2050 horizon for both 29 REF (-5 ppb) and MIT (-16 ppb) scenarios (Colette et al., 2013), in contrast to our findings 30 showing increase in ozone under the REF scenario and only small decrease under the MIT 31 scenario. We are interested to see whether this difference in the sign between the local and

regional ozone projections could be explained by the photochemical regime under which ozone
 is produced in each simulation.

3

#### 4 4.2.1 Analysis of the chemical regimes

5 It is known that areas under  $NO_x$ -sensitivity or VOC-sensitivity are more effectively mitigated 6 when  $NO_x$  or NMVOCs emission reductions are implemented respectively (Beekman and 7 Vautard, 2010). Previous modeling exercises using the CHIMERE model have shown that the 8 ozone photochemistry in Paris is VOC-sensitive (Beekman and Derognat, 2003; Beekman and 9 Vautard, 2010; Deguillaume et al., 2008).

10 In this experiment we perform a model sensitivity analysis in which anthropogenic  $NO_x$  and 11 NMVOCs emissions let vary from a factor of one fourth up to two times their actual value with a 12 step of 0.25 (all combinations of 8 different scenarios for both  $NO_x$  and NMVOCs). In total 64 13 simulations were performed for a typical summertime ozone episode. Emission changes were 14 applied over the whole 3-dimensional domain. We conducted the same experiment twice: once 15 using as initial local-scale emission inventory developed by AIRPARIF (LOC case) and once 16 using the regional scale inventory (REG case). In order to isolate the effect of emissions only and 17 exclude the potential impact of other resolution-related effects the sensitivity analysis for the 18 REG realization was not performed in the native REG domain; instead we first downscaled the 19 regional emission inventory on the 4km grid and then ran the regional scale simulation. 20 Obviously, emission totals between the two simulations do not match and no direct comparison 21 should be made based on these results. However, it is interesting to see how different emission 22 starting points may lead to ozone production under different photochemical regimes and 23 therefore to different future ozone projections. We note that present time  $NO_x/NMVOCs$ 24 emission ratios within the ozone period over Paris are much higher in the LOC inventory than in 25 the REG dataset (0.5 vs. 0.2). This ratio is a key factor for the modeling of ozone production. We 26 chose to run the sensitivity analysis over a one-day, present time, summer ozone episode that 27 emission fluxes and meteorological conditions favoring the photochemical ozone built-up are 28 similar in most ozone events. We also note that while boundary conditions may impact local 29 scale model predictions, we focus here on the impact of local emissions through the use of a 30 common set of boundary conditions, to ensure that the differences arise from the sensitivity to 31 local emissions.

1 Iso-contours of day ozone maxima concentrations as a function of NO<sub>x</sub> and NMVOCs emission 2 rates are given for the two simulations (isopleth plots of Fig. 5). Two cases are discussed 3 separately: downtown chemistry (panels a and b) and in-plume chemistry (panels c and d). "In-4 plume" corresponds to the chemical reactions taking place when high concentrations of nitrous 5 oxides coming from the city and NMVOCs from the rural areas help to accumulate ozone on the 6 downwind site of the city. To extract "in-plume" results we identify at each hour of the simulated 7 episode the grid cell with the maximum concentrations inside the domain. Downtown chemistry 8 in the LOC simulation is mainly sensitive to  $NO_x$  changes with sharp ozone increases related to 9  $NO_x$  reductions. This is a typical ozone response near high  $NO_x$  emission sources where titration 10 and removal of radicals by nitrogen dioxide are the main drivers of ozone concentrations (VOC-11 sensitive regime). Under the REF scenario NO<sub>x</sub> reductions in 2050 are much more drastic than 12 NMVOC ones (Table 2) because the GEA-derived scaling coefficients used to project local scale 13 emissions from 2020 to 2050 mainly implement NO<sub>x</sub> reductions through domestic heating and 14 road transport regulation. Under VOC-sensitive conditions NO<sub>x</sub> reductions without additional 15 regulation of NMVOCs lead to inhibited ozone titration and increase in ozone concentrations. 16 Ozone mitigation would most likely be more effective if the domestic/industrial use of solvents 17 (the main source of NMVOCs) were targeted as well. Under the MIT scenario however, where 18 both  $NO_x$  and NMVOCs are mitigated more effectively than in REF, ozone concentrations 19 decrease in 2050 compared to present time levels. This feature stands-out in the NO<sub>x</sub>-limited 20 rural areas in which ozone follows the fate of its precursors showing a much more drastic 21 decrease of ozone under MIT (Fig. 4c) compared to REF (Fig. 4b). On the other hand chemistry 22 in the REG simulation is driven by a NO<sub>x</sub>-sensitive regime. The titration process is much less 23 pronounced compared to the LOC run. As a result ozone decreases in response to the significant 24 NO<sub>x</sub> reductions (see also discussion in Sect. 2.3) under both REF and MIT scenarios. 25 In the regional setup both urban and rural chemistry is characterized by NO<sub>x</sub>-sensitive conditions 26 (panel b vs. d). On the contrary, if urban chemistry is clearly VOC-sensitive at the local scale run 27 (panel a), in-plume ozone built-up is found on the ridge line separating the two regimes (panel c).

28 This is consistent with previous studies (Menut et al., 2000; Sillman et al., 2003). Urban ozone

29 levels modeled at the local scale are much lower than those modeled at the regional scale by

30 almost 50 ppb ( $\sim$ 54 ppb vs.  $\sim$ 100 ppb). This is due to the higher NO<sub>x</sub> emission ratios prescribed

31 by the local scale inventory (LOC) compared to the regional inventory (REG). As far as the in-

plume chemistry is concerned the difference between the two simulations is less pronounced (~20 ppb). This is consistent with the fact that the reductions in rural ozone modelled at regional scale (REF: -3.2 ppb and MIT: -13.5 ppb) and in our study for the two future projections relative to present time levels are similar. Photochemical ozone built-up occurs at longer time and space scale compared to titration and therefore the increase in the resolution of the emissions brought about at the local scale does not provide much new information to the modeling of rural ozone.

7 This sensitivity analysis shows that the integration of high-resolution emission projections in the 8 modeling of ozone is critical. Clearly the regional and local simulation will respond differently to 9 emission changes in the future but unfortunately it is not possible to assess quantitatively the 10 impact of this change since besides the spatial gradients in emissions, projections also differ 11 between the LOC and REG simulations. The conclusion that can be drawn though from this 12 sensitivity analysis is that the simulation forced by the regional scale emissions fails to 13 distinguish the chemical regime between urban and rural chemistry (both shown as NO<sub>x</sub>-limited) 14 and that ozone titration over Paris is most likely underestimated in the regional setup which may 15 explain the ozone decrease modelled for 2050 over Paris under the REF scenario. Increase in 16 ozone due to sharp NO<sub>x</sub> emission reductions under a VOC-limited environment as modelled with 17 the local setup may be more realistic.

18

## 19 4.2.2 Future chemical regimes indicators

Based on the aforementioned previous studies as well as on the sensitivity exercise carried out here, ozone formation in the city of Paris occurs under VOC-sensitive conditions. However, it is plausible that the implementation of emission reductions in the long run causes a shift of the chemical regime towards a more  $NO_x$ -sensitive chemistry (Beekman and Vautard, 2010; Colette et al., 2012; Tarasson et al., 2003).

To investigate this shift, we use here a number of chemical regime indicators, which explain emission accumulation and radical production/loss processes:  $O_3/NO_y$ ,  $O_3/NO_z$  and  $H_2O_2/NO_z$ . Each indicator accounts for different aspects of ozone chemistry, for example the  $H_2O_2/NO_z$  ratio takes into account both the impact of emissions and radical production (Sillman, 1995) by comparing the  $HO_x + NO_x$  radical sink (yielding  $NO_z$ ) to the  $HO_2 + HO$  radical sink yielding  $H_2O_2$ . The latter sink corresponds to enhanced radical production, which favors  $NO_x$ -sensitive conditions (Kleinman, 1997). These indicators are used in order to avoid the large number of sensitivity runs such as the ones described in Sect. 4.2.1 that would be necessary to perform if
one used a more direct approach such as above.

3 Chemical regime indicators are estimated from domain-wide daytime O<sub>3</sub>, NO<sub>y</sub>, NO<sub>z</sub> and H<sub>2</sub>O<sub>2</sub> 4 concentrations for the CTL run and the two REF and MIT projections (Fig. 6). Higher  $O_3/NO_v$ , 5 O<sub>3</sub>/NO<sub>z</sub> and H<sub>2</sub>O<sub>2</sub>/NO<sub>z</sub> ratios point to more NO<sub>x</sub>-sensitive conditions while lower ratio values point to VOC-sensitive chemistry. A visual inspection of the figure reveals that all three 6 7 chemical indicator ratios increase under both REF and MIT scenarios compared to CTL due to 8 decreases in  $NO_{x}$  and  $NO_{z}$  concentrations induced by the implemented  $NO_{x}$  emission reductions. 9 For example the decade mean O<sub>3</sub>/NO<sub>y</sub> molar ratio shifts from the preset-time value of 0.64 to 10 2.75 and 10.6 in the REF and MIT scenarios respectively. Based on these indicators we deduce 11 that a shift towards more NO<sub>x</sub>-sensitive chemistry should be expected in 2050 under both REF 12 and MIT scenarios.

13

#### 14 **4.2.3 Ozone health indicators**

15 Here, we study how different ozone health indicators change under the two future scenarios 16 (Table 5). We focus on three indexes: i) the sum of the differences between maximum daily 8-17 hour running means and the 35 ppb threshold value (SOMO35); ii) the number of days where 18 maximum daily 8-hour running mean ozone concentration exceeds the 60 ppb threshold (Nd120) 19 and iii) the mean of the ten highest daily max concentrations during the ozone period (MTDM). 20 These metrics are typically used in health impact assessment studies and account for the non-21 linearity in the ozone dose-response function. For example SOMO35 was developed in the Joint 22 WHO/Convention Task Force in 2004 and represents the cumulative annual exposure to ozone. 23 The threshold value was established on the fact that a statistically significant increase in 24 mortality risk estimates was observed at ozone concentrations above 25-35ppb. .

Under the REF scenario SOMO35 modelled at local scale in the city is almost doubled compared to present-time (Table 5) while the reductions of the Nd120 index from 29 to 18 days points to less frequent ozone episodes in the future. The MTDM index remains constant suggesting that ozone episodes will not decrease in intensity.

All three health related proxies improve under the MIT scenario. SOMO35 decreases by 79%

30 and MTDM by 26.5% relative to CTL corresponding to future conditions with no significant

31 ozone episodes (Nd120 falls to zero). The modelled reductions in the health indicators represent

a much more efficient mitigation compared to maximum ozone concentrations (see discussion in Sect. 4.2) highlighting the high sensitivity of these indicators to their respective cut-off thresholds. It becomes clear, that from a human health perspective the MIT scenario is very effective and the co-benefits of climate and air quality strategies are significant. In the downwind rural areas south of Paris all health indicators under both scenarios are improved, as seen in Table 5, following the subsiding of ozone levels in the future (Fig. 4).

7

#### 8 4.3 PM<sub>2.5</sub>

9 Local scale PM<sub>2.5</sub> projections (wintertime) are presented in Fig. 7 as well as differences between 10 each of the REF and MIT scenarios and the CTL simulation. Major decrease in PM2.5 is 11 modelled under both REF and MIT projections compared to present time levels. This is mainly 12 due to significant reductions in primary particle emissions (Table 2). In downtown Paris, concentrations are reduced from 14.2  $\mu$ g/m<sup>3</sup> in CTL to 3.4  $\mu$ g/m<sup>3</sup> and 1.6  $\mu$ g/m<sup>3</sup> under the REF 13 14 and MIT scenarios respectively. The reductions are much stronger in the city than at rural areas 15 due to the effective mitigation of road transport emissions. Consequently the large wintertime urban increment (difference between urban and rural concentrations) of approximately 6.8  $\mu$ g/m<sup>3</sup> 16 modelled with CTL is significantly reduced in both future projections:  $\sim 1 \text{ µg/m}^3$  in REF and less 17 than 0.2  $\mu$ g/m<sup>3</sup> in MIT. 18

19

#### 20 5 Conclusions

21 Mid-21<sup>st</sup> century ozone and PM<sub>2.5</sub> projections over the city of Paris have been modelled with the 22 CHIMERE air-quality model at a 4 km horizontal resolution under two consistent climate and 23 emission scenarios: a reference and a mitigation scenario. To our knowledge, this is the first time 24 that a study of a 10-year air-quality projection under climate and city level emission changes is 25 conducted over a large European agglomeration at such fine scale. A key innovation of this work 26 is that we use local-scale emissions and their projections until 2020 developed by local experts 27 (AIRPARIF) and extend those until 2050 based on coefficients extracted from large-scale 28 emission projections. A 10-year control simulation served model evaluation purposes. 29 Furthermore, we investigate how ozone projections under the two future scenarios depend on the 30 photochemical regime.

Model evaluation showed a very good agreement between model and measurements for ozone and an underestimation of wintertime  $PM_{2.5}$  by 20% over the urban area, which is mainly attributed to a large wet bias in wintertime precipitation (+26.5%). The comparison between modeled and measured  $O_x$  showed that the model at 4 km resolution accurately resolves  $O_3$ titration by NO over the highly urbanized city of Paris. The decade average bias within the ozone period for  $O_x$  ( $O_3$ +NO<sub>2</sub>) and ozone is found to be less that 0.3 ppb.

7 Under the reference scenario ozone increases over the city of Paris by +7 ppb relative to present 8 time values and only a small decrease is modelled under the mitigation scenario (-3.5 ppb). 9 Through a sensitivity analysis, we showed that ozone formation in Paris occurs under VOC-10 sensitive chemistry. Under such conditions and due to the stronger mitigation of  $NO_x$  compared 11 to NMVOCs in the REF scenario, titration is inhibited and ozone increases in the 2050 horizon. 12 Following the MIT trajectory both NO<sub>x</sub> and NMVOCs are more effectively regulated leading to 13 ozone reduction in 2050. The same sensitivity analysis applied on the regional emission 14 projections showed that ozone formation in Paris occurred under  $NO_x$ -sensitive conditions. The 15 discrepancy in the chemical regime is attributed to differences in ozone precursor emissions 16 prescribed by the two inventories: NO<sub>x</sub>/NMVOCs emission ratios are much lower in the regional 17 (0.2) compared to the local inventory (0.5). Modelling at regional scale most likely 18 underestimates ozone titration prescribing NO<sub>x</sub>-limited chemistry for Paris. Under this chemical 19 regime ozone precursor emission cutbacks prescribed even for the less optimistic REF scenario 20 benefits ozone air quality to such an extent that reductions are observed while the local setup 21 yields ozone increases instead. If the regional scale model is inadequate to resolve urban features 22 of ozone chemistry, mainly titration close to high NO<sub>x</sub> emission sources, it yields similar rural 23 ozone levels as the local scale run showing that the longer time-scale processes of emission 24 transport and ozone formation are less sensitive to model resolution in the high ozone 25 concentration plume. Our findings suggest that the estimates based on European scale 26 applications are likely to overestimate the downward ozone trend under VOC-sensitive 27 conditions while differences in rural areas are limited.

Finally in downtown Paris,  $PM_{2.5}$  concentrations are reduced by 78% and 89% under the REF and MIT scenarios respectively. The reduction is much more prominent over the urban part of the domain due to significant reductions in primary emissions as a result of the effective

mitigation of the road transport and residential sectors. Therefore the large urban increment
 modelled at the control run is significantly reduced in both future projections.

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1 Table 1. Configuration description of the global, regional and local simulations of the present

2 study.

				3
	Global <sup>a</sup>	Regional <sup>b</sup>	Local	-
GCM/CTM	LMDz-INCA	CHIMERE	CHIMERE	4
Horizontal	1.9° x 3.75°	$0.5^{\circ} \ge 0.5^{\circ}$	4km x 4km	5
Anthropogenic	DCD			5
emissions (present)	RCP	GEA	AIRPARIF	6
Anthropogenic		MIT / REF <sup>d</sup>	MIT/REF	7
emissions scenarios <sup>c</sup>	RCP-2.6 / RCP-4.5	MIII / KEF <sup>-</sup>	consistent <sup>e</sup>	1
Climate model	IPSL-CM5A-MR	WRF	WRF	8
Resolution	2.5° x 1.26°	$0.5^{\circ}$	10km	9
Climate scenarios	RCP-2.6 / RCP-4.5	RCP-2.6 / RCP-4.5	RCP-2.6 / RCP-4	-
				10

11

12 a For details see Hauglustaine et al. (2014).

13 b For details see Colette et al. (2013).

14 c Description given in Sect. 2.3.

15 d The mitigation (MIT) and the reference (REF) anthropogenic emission scenarios were

16 developed to be consistent with RCP-2.6 and RCP-4.5 respectively.

17 e The future emissions of the local scale simulations are developed based on a hybrid

18 methodology utilizing local and regional information (details in Sect. 2.3).

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1 Table 2. Current time and future emission estimates (kt) over a 50km<sup>2</sup> area around Paris

		NO <sub>x</sub> <sup>a</sup>	NMVOCs <sup>a</sup>	PM <sub>2.5</sub> <sup>b</sup>
	CTL	11.2	22.6	1.5
	2050 REF	3.1	10.6	0.34
	2050 MIT	1.0	8.1	0.18
3	a ozone period (April	l-August)		
4	b winter period			
5				
6				
7				
8				
9				
10				
11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26				

2 extracted from the  $1^{st}$  model layer.

- 1 Table 3. Observed and modelled meteorological variables over the Ile-de-France region.
- 2 Modelled data stem from the WRF simulation at 10km resolution. Absolute model bias is given
- 3 in parenthesis.

Variable	Summer (JJA)		Winter (DJF)	
variable	Obs	Model	Obs	Model
7 stations average				
T2 (°C)	19.2	19.8 (+0.6)	4.3	4.2 (-0.1)
WS10 (m/s)	2.9	3.7 (+0.8)	3.6	5.5 (+1.9)
RH (%)	69.1	55.8 (-13.3)	85.0	83.9 (-1.1)
PRECIP (mm/day)	0.076	0.073 (-0.003)	0.07	0.094 (+0.024)
MONT (Paris)				
T2 (°C)	20.1	20.9 (+0.8)	5.4	4.9 (-0.5)
WS10 (m/s)	2.8	3.3 (+0.5)	3.3	4.7 (+1.4)
RH (%)	64.9	50.8 (-14.1)	79.8	79.5 (-0.3)
PRECIP (mm/day)	0.073	0.085 (+0.012)	0.064	0.081 (+0.017)

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Table 4. Future changes in key meteorological variables<sup>a</sup> under the two simulated climate 

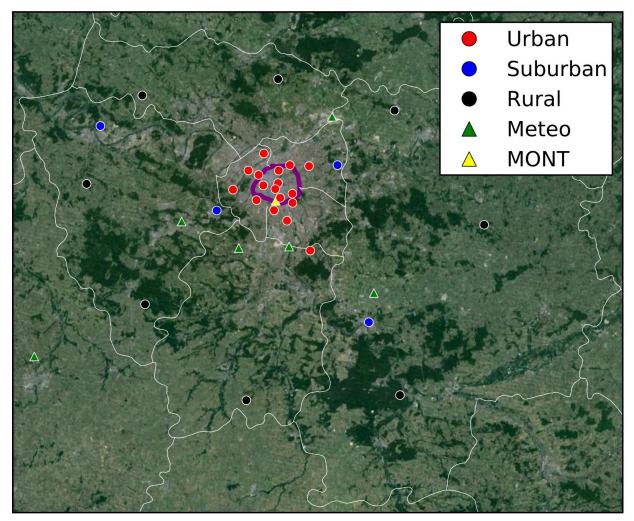
scenarios.

	Variable	CTL	$\Delta$ (REF-CTL)	$\Delta$ (MIT-CTL)
	Temperature (°C)	11.6	+1.1	-0.5
	RH (%)	72.3	+1.1	+1.7
	Precipitation (mm/day)	0.1	+0.007	+0.22
	Radiation (W/m <sup>2</sup> )	154.3	-5.4	-25.4
3	a Annual domain-wide values extracted from 10-yr mean of daily averages.			
4				
5				
6				
7				
8				

1 Table 5. Ozone exposure indicators for the control simulation and the relative differences

Variable	CTL	REF	MIT
City			
SOMO35 (ppb days)	13763	27470	2904
Nd120 (days)	29	18	-
MTDM (ppb)	53.9	54.7	39.6
Rural			
SOMO35 (ppb days)	39581	29101	2115
Nd120 (days)	134	4	-
MTDM (ppb)	63.8	52.9	38.9

2 between the latter and the two future projections for the city of Paris.



- Figure 1. Local scale (IdF) simulation domain, with the city of Paris in the center (area enclosed
   in the purple line). Circles correspond to sites of the local air-quality monitoring network
- 3 (AIRPARIF) with red for urban, blue for suburban and black for rural. Triangles correspond to
- 4 the meteorological stations (the yellow triangle correspond to the Montsouris meteorological
- 5 located in the city center).
- 6

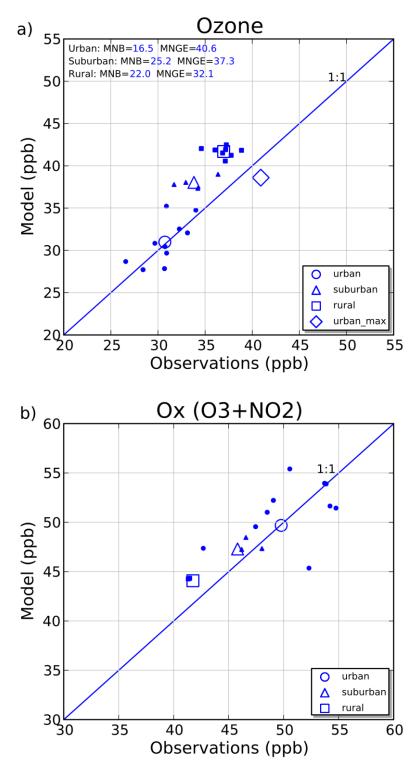
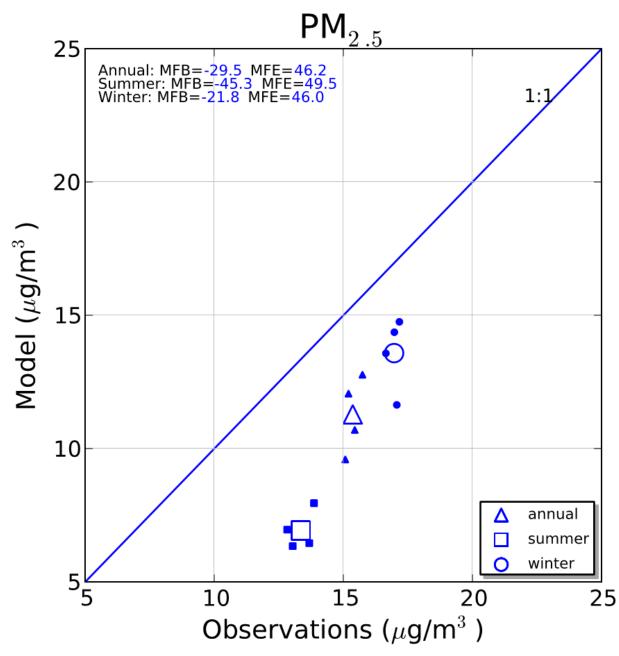
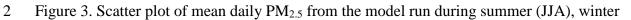


Figure 2. Scatter plots of mean daytime ozone (a) and O<sub>x</sub> (b) during the April to August period
from the model run against observations. Full small markers represent concentrations at
individual stations while large hollow markers stand for the aggregated value. "Urban\_max"
stands for the daily maximum of an 8-hr running average.

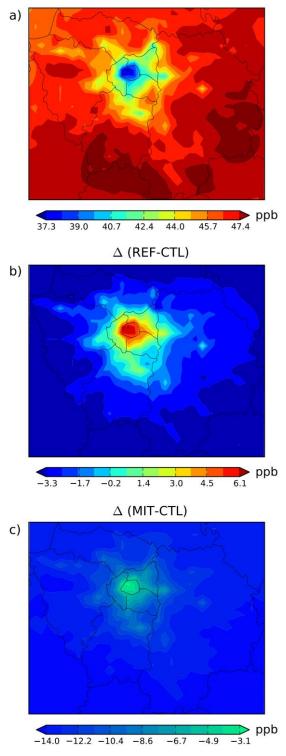




3 (DJF) and on annual basis against observations. Full small markers represent concentrations at

4 individual stations while large hollow markers stand for the aggregated value.

10yr mean of daily max O3 concentrations (APR-AUG



2 Figure 4. Present time ozone daily maximum (ppb) of 8-hr running means averaged over the

- 3 April-August period (a) and differences between the CTL run and the REF (b) and MIT
- 4 projections (c).
- 5

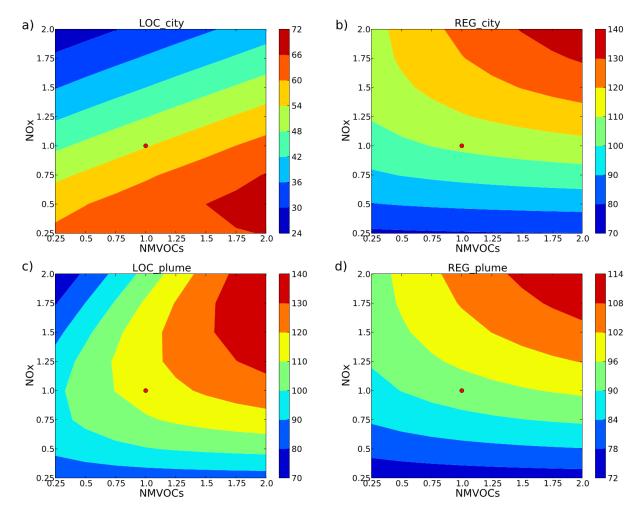




Figure 5. Isocontours of maximum ozone concentrations over downtown Paris (top panels) and
over the entire modelling domain (IdF) (bottom panels) as a function of ozone precursor
emissions. Axis units are the coefficients of emission changes (in molecules/cm<sup>2</sup>/s) with respect
to their baseline values (point 1.0, 1.0). Isopleths on the left are modelled with the LOC
simulation and on the right with the REG simulation.

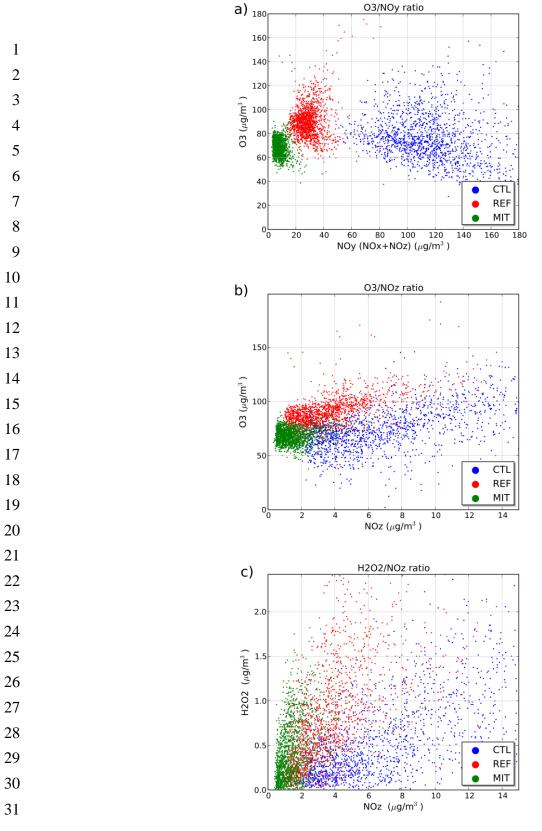
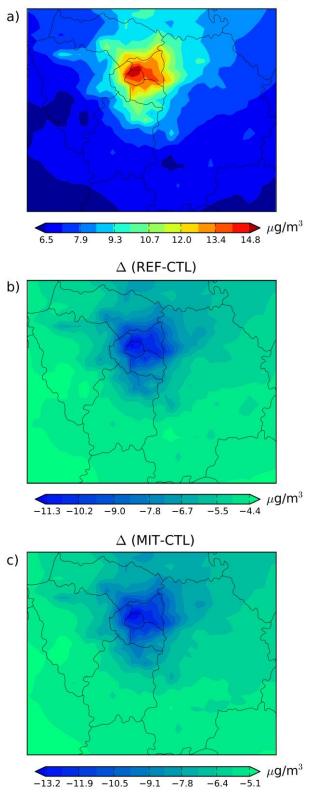


Figure 6. Ozone chemical regime indicators for the control run and the two future projections.
Dots represent daily means averaged during the ozone period over the grid-cells of the
downtown Paris area.

10yr mean of daily avg PM2.5 concentrations (DJF)



1

2 Figure 7. Wintertime (DJF) PM<sub>2.5</sub> daily average fields ( $\mu g/m^3$ ) for the control simulation (a) and

3 the differences between the latter and the REF (b) and MIT (c) future projections.