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Frontiers in air quality modelling

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Abstract. The first pan-European kilometre-scale atmospheric chemistry simulation is introduced. The continental-scale air pollution episode of January 2009 is modelled with the CHIMERE offline chemistry transport model with a massive grid of 2 million horizontal points, performed on 2000 CPU of a high-performance computing system hosted by the Research and Technology Computing Center at the French Alternative Energies and Atomic Energy Commission (CCRT/CEA). Besides the technical challenge, we find that model biases are significantly reduced, especially over urban areas. The high-resolution grid also allows revisiting of the contribution of individual city plumes to the European burden of pollution, providing new insights to target the appropriate geographical level of action when designing air pollution mitigation strategies.

1 Introduction

Beside the challenges in understanding and representing physical and chemical processes, the most common limitation mentioned in geophysical modelling studies regards spatial resolution, often restricted because of computational cost. With the ever growing computing facilities, such limitations are systematically pushed back and we are now able to introduce a continental scale atmospheric chemistry simulation with the CHIMERE model at 2-km resolution, whereas such high spatial resolutions are limited to local air quality models (for a comprehensive review see Zhang et al., 2012; and references therein). Capturing small-scale air pollution features over such a large domain at a glance will allow (1) comparing of local air quality throughout the continent based on homogenous input data, (2) improving of model performance in terms of scores, exceedances, and exposure over a large domain, (3) better represent pollution transport from sources to remote areas and therefore revisit the assessment of the respective role of local sources versus regional background of air pollution. This paper puts into perspective this achievement with the current state of the art in regional atmospheric chemistry modelling. We highlight the strength and weaknesses of the simulation and discuss the relevance of very high performance computing in the field of environmental modelling.

One of the main air pollution outbreaks in Europe in 2009 occurred between 9 and 16 January with particulate matter (fine, less than 2.5 µm and coarse, less than 10 µm, i.e. PM$_{2.5}$ and PM$_{10}$) and nitrogen dioxide (NO$_2$) levels reaching alert levels (MEEDDM, 2010; Chiappini, 2009; Favez et al., 2010) in various places. The main driving factor for this event was an intense cold spell (DWD, 2010) that led to high anthropogenic emissions of pollutants from domestic heating and a stable meteorological situation enhancing the accumulation of air pollution in the thin planetary boundary layer. While Western Europe was exposed to high levels of pollution at the beginning of the period, the pollution plume moved eastwards towards the end of the period as a result of the displacement of the cold spell and the advection of Western pollution. The conjunction of long-range transport and contribution of local sources supports the relevance of this event to investigate the added value of high-resolution continental-scale atmospheric chemistry simulations. The choice of a wintertime event to assess the added value of spatial resolution is further supported by the findings of Fountoukis et al. (2013), who, by comparing 36 and 4 km simulations over the Paris area, demonstrated the lower sensitivity in summer, when the formation of secondary organic aerosols smear out the spatial variability.

The horizontal resolution in the simulation used here is about 2 km, the lower limit being set by the availability of proxy data in building the emission inventories rather
than computational capacities or cost. With such a resolution, the European continent is covered with slightly more than 2 million horizontal grid cells. These numbers should be put into perspective with current practices in air pollution forecasting, where a resolution of about 10 km is achieved by the Modelling Atmospheric Composition Change (MACC) project of the Global Monitoring for Environment and Security programme of the European Commission (www.gmes-atmosphere.eu, Zyranyov et al., 2011) and the US air quality forecasting system of the National Oceanic and Atmospheric Administration and Environmental Protection Agency (Eder et al., 2009). The factor 5 between MACC forecasts (~10 km) and the present work (~2 km) leads to a computational demand two orders of magnitude (5 x 5 for the number of horizontal points, and another factor 5 for the increment in the time step) above current practices. Such a change in the order of magnitude of the spatial refinement of the European air quality model can however only be achieved over a limited period of time and we will focus here on a single air pollution event rather than an annual simulation, as would be common practice in regular air quality model evaluation studies.

Section 2 presents the details of the methods in summarizing the setup of the CHIMERE chemistry transport model and highlighting the main features of input data (especially high-resolution emissions). Section 3 is devoted to the analysis of the results in terms of model validation, exceedances, exposure modelling as well as pollution budgets and fluxes.

2 Methods

2.1 Chemistry transport model

The CHIMERE offline chemistry transport model (CTM) (Menut et al., 2013) was used to model the transport and transformation of trace species in the lower troposphere. The model is being used by several international teams for research and environmental policy underpinning. It has been involved in a number of model inter-comparison initiatives (Colette et al., 2011; Cuvelier et al., 2007; van Loon et al., 2007; Vautard et al., 2009; Galmarini et al., 2012) and is now part of the MACC European pre-operational air composition forecasting system (Zyranyov et al., 2012) within the Copernicus programme. The model and further documentation can be found at www.lmd.polytechnique.fr/chimere. The version used here is CHIMERE-2008 (Bessagnet et al., 2008).

2.2 Model setup

Three spatial resolutions will be used in the present study. Apart from geometrical difference, and the related increment in time step (in order to comply with the Courant–Friedrich–Levy criterion), the model setup is identical for all simulations.

The coarser domain has a resolution of 0.875 and 0.4375 degrees of longitude and latitude (i.e. approximately 57 km). Global chemical fields from a 6-year reanalysis with the LMDz4-INCA3 model (Folberth et al., 2006; Hauglustaine et al., 2004) are used as boundary conditions for the coarse domain. The coarse simulation is then used as boundary conditions for two finer resolution simulations: 0.125 and 0.0625 degrees (i.e. approximately 8 km at the centre of the domain), and ultimately 0.03125 and 0.015625 degrees (i.e. approximately 2 km). The geographical domain is identical for the two fine resolutions; it extends from 10.4375° W to 30.4375° E and from 35.906250° S to 61.84375° N, while the coarse domain is 13.43750° W to 32.93750° E and 34.40625° S to 62.84375° N. (see full model domains in Fig. 2). The vertical discretisation is identical for all three simulations; it has 9 layers extending from approximately 997 hPa to 500 hPa.

The 57 km and 8 km resolution simulations include a one month spin up before 8 January 2009. At that date, the 2 km simulation starts using initial conditions of the 8 km simulation so that only 8 January 2009 is considered as a spin up for the very high-resolution simulation and discarded from the quantitative analysis below. The temporal increment of the 2 km simulation is 140 s for the physical integration and 20 s for the chemical solver. The simulation covers 9 days (8 January 2009 to 16 January 2009).

The main user-dependent physical and chemical parameterisation is summarized here, but the reader is referred to Menut et al. (2013) for details on the corresponding model components and references as well as non user-specific model characteristics. The gas-phase chemical mechanism is MELCHIOR2, which consists of a simplified version (40 species and 120 reactions) of the full chemical mechanism based on the concept of chemical operators. Modelled particulate matter includes primary particulate matter, and secondary inorganic (nitrate, sulphate, ammonium based on the ISORROPIA thermodynamic equilibrium model) and organic aerosol (whose formation is represented according to a single-step oxidation of the relevant anthropogenic and biogenic precursors and gas–particle partitioning of the condensable oxidation products). Biogenic emissions are computed with MEGAN version 2.04, sea-salt and desert dust are taken into account, but non-African mineral dust is excluded. Horizontal transport is solved with the second-order Van Leer scheme. Subgridscale convective fluxes are accounted for. The present setup also benefits from an improved representation of turbulent mixing in urban areas that yields lower horizontal wind and vertical mixing in order to better capture the difference between the urban canopy (where the first CHIMERE model level lies) and the top of the urban sub-layer (of which the lowermost meteorological model level is representative) as described in Terrenoire et al. (2013).
2.3 Input data: meteorology

As an offline CTM, CHIMERE requires prescribed meteorological fields which were provided here by ECMWF with the IFS model at 16 km resolution with data assimilation. The same fields are interpolated linearly at 57, 8, and 2 km to drive the relevant resolution CTM simulations. Unfortunately, performing a continental-scale high-resolution meteorological simulation including urban island effects is a challenge in itself and was ruled out of the present initiative, so that the resolution of the meteorological model does not match that of the CTM. It should be noted however that the added value of resolution in meteorological forcing is expected to be negligible compared with that of emission and chemistry transport (Valari and Menut, 2008).

2.4 Input data: emissions

The efforts of the CHIMERE development team for the present initiative were focused on the improvement of the anthropogenic emissions module (Menut et al., 2012). A few years ago, frontiers in atmospheric chemistry simulation were set by the resolution of emission inventories that were delivered at approximately 50 km at continental scale, whereas higher resolution (up to 1 km) could only be reached over a restricted area with bottom-up inventories. However, because of heterogeneities in the input data used in bottom-up emission data sets and lack of required information at the local scale, such an approach cannot yet be contemplated at continental scale. Instead, we used a top-down approach, i.e. downsampling techniques to disaggregate coarse inventories using proxies such as point source location and high-resolution population density map. Here the resolution limitation is thus prescribed by the proxy data: in our case the population density map that was made available at a resolution of about 1 km by the Joint Research Centre of the European Commission.

More specifically, the anthropogenic emission used here is the inventory developed by INERIS in the framework of the EC4MACS project (European Consortium for Modelling of Air Pollution and Climate Strategies, http://www.ec4macs.eu/). It is described in further detail in Bessagnet et al. (2012) and Terrenoire et al. (2013) and consists of an optimal combination of various sources.

National annual (2009) emission totals for each activity sector are the same as used in EMEP models provided at 0.5 × 0.5 degree resolution (Vestreng et al., 2007). For primary PM$_{2.5}$, however, GAINS (Amann et al., 2011) national emissions totals were considered more reliable (Z. Klimont, personal communication, 2011) for the following countries: CZ, BA, BE, BY, ES, FR, HR, IE, LT, LU, MD, MK, NL, CS, and TR.

This hybrid of EMEP and GAINS national and annual totals is then spatialised following various proxies. The basic spatial distribution follows the TNO-MACC inventory at 0.125 × 0.0625 degree resolution (Kuenen et al., 2011) for SNAP activity sectors 3, 7, 8, 9, 10. Residential emissions (SNAP 2) were regridded according to population density maps and correlation coefficient between population and emissions derived from a French bottom-up inventory (Bessagnet et al., 2012). For SNAP sectors 1, 4, 5, 6 emissions are distributed over the artificial land use corresponding to EPER industries (eea.europa.eu/data-and-maps/data/eper-the-european-pollutant-emission-register-4).

Annual maps of surface fluxes of emissions are then ultimately distributed along the year and in the vertical according to the Eurodelta protocol (Thunis et al., 2008). An additional correction is applied to take into account the impact of day-to-day temperature variations on the emission due to residential heating (degree-days concept) (Terrenoire et al., 2013).

![Fig. 1. Map of NO$_2$ (µg m$^{-3}$) on 10 January 2009 at 16:00 UTC in the 57 km (top) and 2 km (bottom) resolution CHIMERE simulations.](image-url)
The above methodology is used to produce emission at the relevant resolution (57, 8, or 2 km) using the same coarse-level total fluxes (on the EMEP 0.5 degrees grid). A comparison of primary PM$_{2.5}$ emission maps at 57 and 2 km is provided in Fig. 1. The areas where the coarse emission data set can be refined using appropriate proxies (population density, point sources) stand out clearly from the locations where such a downscaling is not possible (e.g. over sea surfaces). Note that the total emitted flux is identical in both cases.

2.5 Observation data used for the validation

The model results will be interpolated at the location of air quality monitoring stations reported in the AIRBASE repository (eea.europa.eu/themes/air/airbase) that includes Europe-wide hourly NO$_2$ and daily PM$_{10}$ and PM$_{2.5}$ measurements from urban, suburban, and rural sites. The number of station per type and compound is given in Table 1.

2.6 Computational setup and cost

The simulations were parallelised on 2000 Intel Xeon processors at 2.7 GHz with 64 G memory by node. Tests were performed with up to 4000 processors with less satisfactory scaling properties. The total cost of the project was about 110 000 h core, or 13 yr on a single processor, but the wall-clock time for 24 h of simulation was only 5 h 40 (excluding the pre-processing). The storage of input/output files also constitutes an achievement with more than 2 T of data.

3 Results

3.1 Qualitative presentation

The maps in Fig. 2 illustrate the added value of increased resolution by showing snapshots of the simulations of nitrogen dioxide at about 57 and 2 km of resolution. A more dynamic view is accessible in the movie provided in the Supplement. While the coarse run captures well the main emission hotspots, local maxima over individual roads and isolated point sources (e.g. the oil rigs in the North Sea) are smeared out because of the numerical diffusion of the Eulerian model.

A more quantitative analysis of the added value of resolution will be provided in the following paragraphs, but the benefit of such a high-resolution simulation in raising awareness of the contribution of local sources in the accumulation of air pollution at the continental scale is worth mentioning. While discussions about the design of efficient air pollution mitigation policy measures often fall back to the choice of the most relevant geographical scale (local, national, continental), these high-resolution maps constitute a relevant reminder that in the field of atmospheric pollution as well, little brooks make great rivers.

A comparison with the observations allows discussing of the improvement in the performance of the model. An average of time series of NO$_2$ at all urban, suburban and rural monitoring stations in the Paris area (geographical box 47° N, 50° N, 1.5° E, 3.5° E) is given in Fig. 3. The upper panel (urban stations) illustrates well the improvement brought about by the increased resolution to capture high pollution levels in urban areas. Whereas the coarse simulation is only able to reproduce the diurnal cycle and the daily variability with an important average bias, moving to a 8 km resolution improves the average level and daytime values, while night-time and morning peaks are further improved when moving to a 2 km resolution. The differences between the 8 km and 2 km setups almost vanish when focusing on suburban stations (second panel of Fig. 3), and looking further at rural stations (low

Fig. 2. Map of emissions of fine primary particulate matter (PPM) emissions (in molecule cm$^{-2}$ s$^{-1}$) over the day of 10 January 2009 in the 57 km (top) and 2 km (bottom) resolution domains.

Fig. 3. Average time series of NO$_2$ at urban, suburban and rural monitoring stations in the Paris area (geographical box 47° N, 50° N, 1.5° E, 3.5° E).
Table 1. Model performances (root mean square error, mean fractional bias and average spatial correlation) of each configuration over the whole modelling domain for three types of background stations (urban, suburban, rural) between 9 and 16 January 2009. The average of observations over the whole period is also given as well as the number of exceedances (Exceed.) of regulatory thresholds: 200 µg m\(^{-3}\) for hourly NO\(_2\) and 50 µg m\(^{-3}\) for daily PM\(_{10}\).

<table>
<thead>
<tr>
<th>Pollutant</th>
<th>Station type</th>
<th>Number of sites</th>
<th>Average observed</th>
<th>Exceed.</th>
<th>RMSE (µg m(^{-3}))</th>
<th>57 km Corr.</th>
<th>57 km MFB</th>
<th>Exceed.</th>
<th>RMSE (µg m(^{-3}))</th>
<th>8 km Corr.</th>
<th>8 km MFB</th>
<th>Exceed.</th>
<th>RMSE (µg m(^{-3}))</th>
<th>2 km Corr.</th>
<th>2 km MFB</th>
<th>Exceed.</th>
</tr>
</thead>
<tbody>
<tr>
<td>NO(_2)</td>
<td>Urban</td>
<td>721</td>
<td>47.3</td>
<td>223</td>
<td>33.00</td>
<td>0.68</td>
<td>1.04</td>
<td>0</td>
<td>24.37</td>
<td>0.66</td>
<td>0.59</td>
<td>130</td>
<td>21.51</td>
<td>0.64</td>
<td>0.38</td>
<td>317</td>
</tr>
<tr>
<td></td>
<td>Suburban</td>
<td>380</td>
<td>45.9</td>
<td>116</td>
<td>30.19</td>
<td>0.68</td>
<td>0.93</td>
<td>0</td>
<td>22.75</td>
<td>0.67</td>
<td>0.52</td>
<td>69</td>
<td>21.57</td>
<td>0.66</td>
<td>0.42</td>
<td>70</td>
</tr>
<tr>
<td></td>
<td>Rural</td>
<td>345</td>
<td>25.4</td>
<td>0</td>
<td>16.94</td>
<td>0.54</td>
<td>0.42</td>
<td>0</td>
<td>14.72</td>
<td>0.59</td>
<td>0.29</td>
<td>3</td>
<td>14.81</td>
<td>0.61</td>
<td>0.33</td>
<td>5</td>
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<tr>
<td>PM(_{10})</td>
<td>Urban</td>
<td>633</td>
<td>66.0</td>
<td>2648</td>
<td>48.70</td>
<td>0.32</td>
<td>0.81</td>
<td>209</td>
<td>43.62</td>
<td>0.31</td>
<td>0.59</td>
<td>751</td>
<td>42.29</td>
<td>0.29</td>
<td>0.44</td>
<td>1238</td>
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<tr>
<td></td>
<td>Suburban</td>
<td>293</td>
<td>58.5</td>
<td>1104</td>
<td>38.36</td>
<td>0.48</td>
<td>0.67</td>
<td>133</td>
<td>34.08</td>
<td>0.49</td>
<td>0.48</td>
<td>343</td>
<td>32.47</td>
<td>0.49</td>
<td>0.38</td>
<td>476</td>
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<tr>
<td></td>
<td>Rural</td>
<td>221</td>
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<td>462</td>
<td>23.47</td>
<td>0.59</td>
<td>0.28</td>
<td>46</td>
<td>21.59</td>
<td>0.65</td>
<td>0.23</td>
<td>84</td>
<td>20.37</td>
<td>0.64</td>
<td>0.16</td>
<td>131</td>
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<tr>
<td>PM(_{2.5})</td>
<td>Urban</td>
<td>161</td>
<td>44.6</td>
<td>27.51</td>
<td>0.62</td>
<td>0.58</td>
<td>0.39</td>
<td>24.69</td>
<td>0.57</td>
<td>0.39</td>
<td>0.22</td>
<td>22.78</td>
<td>0.58</td>
<td>0.22</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Suburban</td>
<td>68</td>
<td>43.7</td>
<td>26.68</td>
<td>0.67</td>
<td>0.59</td>
<td>0.37</td>
<td>23.53</td>
<td>0.58</td>
<td>0.37</td>
<td>0.27</td>
<td>22.32</td>
<td>0.56</td>
<td>0.27</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
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<td>27.2</td>
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<td>0.68</td>
<td>0.12</td>
<td>0.00</td>
<td>14.70</td>
<td>0.77</td>
<td>0.02</td>
<td></td>
<td>14.05</td>
<td>0.76</td>
<td>0.07</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

panel of Fig. 3), we find that the three model setups give very similar results.

The performances of the model are further quantified in Table 1 that synthesizes a few key statistical metrics: the root mean square error (RMSE) and the mean fractional bias (MFB) between the model and observations (based on daily values) averaged across the Europe-wide AIRBASE network as well as daily spatial correlation coefficients throughout Europe averaged over the whole period. For the spatial correlation, we used Pearson’s definition: the covariance between X and Y divided by the product of the standard deviation of X times that of Y, where X is the set of observed values over Europe at a given time and Y the corresponding modelled set. The MFB is defined as the average of normalised differences between model and observation (i.e. model minus observation) divided by the average between model and observation. It ranges between −2 and 2 and has an ideal value of 0. The MFB provides additional information compared with RMSE by providing the sign of the difference on a normalised basis.

We find that increasing the resolution improves significantly the root mean square error and mean fractional bias, especially at urban sites. The difference between the RMSE and MFB at urban, suburban and rural stations decreases with the resolution, showing that we resolve better spatial gradients. This feature is more sensitive for NO\(_2\) than PM\(_{10}\) or PM\(_{2.5}\) because of the lower spatial variability of the latter, yielding this higher sensitivity of NO\(_2\) often reported at urban sites.

Table 1 somehow questions the relevance of very high-resolution simulations since the bulk of the improvement is achieved at 8 km resolution, for which the RMSE of daily mean NO\(_2\) averaged over the whole monitoring network is 24 % (10 % for PM\(_{10}\)) lower than the 57 km simulation, and the 2 km resolution adds only another 6 % (4 % for PM\(_{10}\)) reduction to achieve an average RMSE 30 % (14 % for PM\(_{10}\)) lower than the coarsest configuration.

In addition, the improvement in the spatial correlation is only seen at rural sites (and suburban sites for PM\(_{10}\)). Discussing spatial correlation clustered by station type is however slightly unfavourable given that it ignores the improvement in spatial gradient between urban/suburban/rural areas. The spatial correlation for NO\(_2\) over the whole network improves from 0.63 to 0.67 and 0.67 in the 57 km, 8 km, and 2 km setups, respectively.

Nevertheless one could have expected that the improvement would have been even larger. These figures illustrate the need to investigate further more sophisticated emission top–down downscaling algorithms, and to account for small-scale dynamical meteorological features (such as urban processes).

3.3 Exceedances

Table 1 only addresses average biases and similar limited added values of increased resolution were reported before as far as mean or fractional biases are concerned (Fountoukis et al., 2013; Tesche et al., 2006). On the contrary, the performances in terms of exceedances (which are notably challenging to capture with numerical models because of thresholding effects even if they carry a high regulatory relevance in the European air quality legislation; EC, 2008) are improved at very high resolution. For PM\(_{10}\), the daily limit value of 50 µg m\(^{-3}\) was exceeded 4214 times throughout the European monitoring network between 9 and 16 January 2009 (1147 monitoring stations and 8 days, i.e. almost 50 % of observations on average). Using a 57 km resolution, only 388 exceedances (or about 10 %) are captured by the CTM, but the increased resolution improves this number, with 28 % and 44 % of the exceedances captured by the model at 8 km and 2 km of resolution, respectively.

Similarly, no exceedances of the 200 µg m\(^{-3}\) threshold for hourly NO\(_2\) values were captured with the 57 km resolution setup, while this threshold had been exceeded 339 times over the period. Moving to higher resolution allows capturing of such events, to the extent of producing 15 % too many exceedances at 2 km resolution.
3.4 Budget and fluxes

The non-linearity of the atmospheric chemistry system yields a slight sensitivity to the resolution of the total mass of modelled particulate matter aggregated over Western Europe (10° W, 30° E, 36° N, 60° N). The average PM$_{10}$ load is 17.4, 19.5 and 20.7 µg m$^{-3}$, for the 57, 8, and 2 km resolution simulations respectively, while the corresponding figures for NO$_2$ are 10.2, 13.8, and 14.4 µg m$^{-3}$.

Fig. 3. Average time series of hourly NO$_2$ (µg m$^{-3}$) observed at, from the top to the bottom, urban, suburban and rural background monitoring stations in the Paris area (black) and interpolated in the CHIMERE model simulations at the 57 km (solid), 8 km (dashed), and 2 km (dotted) resolutions.

3.5 Exposure

The sensitivity is even larger when it comes to exposure modelling. Because highly populated areas are correlated with emissions hotspots, the horizontal gradients matter when weighting pollution load by population density. Here we find that on average over the 9-day period, in the high-resolution simulation 87.9 million people (out of 588 million inhabitants included in the modelling domain) were exposed to daily concentrations exceeding the daily limit value of 50 µg m$^{-3}$ for PM$_{10}$ according to the EU Directive on ambient air quality (2008/50/EC). Using a 57 km grid, the same estimate would be 18.2 million people, i.e. almost a factor 5 underestimation if the additional increment of pollution in urban areas is ignored.

4 Conclusions

By producing the first continental-scale 2 km resolution air quality simulation performed on a 2000 CPU computer, we introduced the first steps of atmospheric chemistry modelling.
in the era of very high performance computing. This frontier simulation is two orders of magnitude above current practices in terms of computational demand.

Besides the technical challenge, which demonstrated the robustness of the CHIMERE air quality model, we discuss the added value in terms of air pollution modelling and decision support. The comparison with in situ observations shows that model biases are significantly improved despite some spurious added spatial variability attributed to the coarse resolution of the meteorological fields shortcomings in the emission downscaling process. The high-resolution emission inventory used here relies on a top–down approach that uses European-wide proxies for the spatial distribution. The quality of the emission spatialisation would be greatly improved by using national emission reported at finer resolution than current standards (e.g. 10 km instead of 50 km) instead of coarse high-level proxies.

The increased spatial resolution is clearly beneficial for the detection of exceedances, the quantification of transport fluxes and exposure modelling resulting from the non-linearity of the atmospheric system. We exhibit dynamic air pollution patterns that highlight the contribution of the city plumes to the background air pollution levels. Reciprocally, the contribution of long-range transport to exceedances of regulatory threshold values for PM$_{10}$ and NO$_2$ in urban areas is better assessed as a result of reduced diffusion. The underestimation in terms of net export flux of trace species is of the order of 25% for a city like Paris, which raises important concerns in studies focusing on impacts on remote areas with large-scale coarse models. Lastly, up to a factor 5 underestimation of the fraction of population exposed to detrimental levels of pollution can be derived with a simulation at coarse resolution if subgrid scale correction such as urban increments (Amann et al., 2011; Denby et al., 2011) are ignored to cope with the lack of an explicit high-resolution model.

We conclude that significant improvement can be sought after by refining the geographical mesh of atmospheric chemistry models. Better results can be achieved thanks to an improved representation of the contribution of local sources in the build-up of continental-scale events, in the net transport fluxes and in the aggregated population exposure to detrimental air pollution levels. This experiment opens new perspectives for decision making. In Europe, considering the achievements after two decades of efforts to reduce air pollutant emissions, the challenge is now to find the optimal trade-off between national and local air quality management strategies. While the first approach is based on sectoral strategies and energy policies, the latter builds upon new alternatives such as urban development. The strategies, the decision pathways and the involvement of individual citizens differ, and a compromise based on cost and efficiency must be found. We illustrated how high-performance computing in atmospheric science can contribute to this aim by improving modelling of exposure, exceedance, and export of pollution. Specifically, our finding on the underestimation of about 25% for the export fluxes of fine particulate matter of a city like Paris in 57 km simulation compared to the 2 km setup provides a new insight into the relative contribution of individual cities to the European burden, raising new questions with regards to the appropriate level of action for air pollution mitigation, although further developments are still needed to secure the results for routine policy use.

Supplementary material related to this article is available online at http://www.geosci-model-dev.net/7/203/2014/gmd-7-203-2014-supplement.zip.

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