

Emission inventories and modeling requirements for the development of air quality plans. Application to Madrid (Spain)

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HIGHLIGHTS

- A comprehensive and flexible urban emission inventory was developed for Madrid.
- Options for multi-scale consistency are discussed (from European to street level).
- Urban background NO₂ concentration levels well described by CMAQ (MB – 2.2 µg/m³).
- NO₂ concentration levels in Madrid are dominated by local traffic (up to 90%).
- A 31% reduction of NO_x emissions may allow Madrid meeting the NO₂ European standards.

ABSTRACT

Modeling is an essential tool for the development of atmospheric emission abatement measures and air quality plans. Most often these plans are related to urban environments with high emission density and population exposure. However, air quality modeling in urban areas is a rather challenging task. As environmental standards become more stringent (e.g. European Directive 2008/50/EC), more reliable and sophisticated modeling tools are needed to simulate measures and plans that may effectively tackle air quality exceedances, common in large urban areas across Europe, particularly for NO₂. This also implies that emission inventories must satisfy a number of conditions such as consistency across the spatial scales involved in the analysis, consistency with the emission inventories used for regulatory purposes and versatility to match the requirements of different air quality and emission projection models. This study reports the modeling activities carried out in Madrid (Spain) highlighting the atmospheric emission inventory development and preparation as an illustrative example of the combination of models and data needed to develop a consistent air quality plan at urban level. These included a series of source apportionment studies to define contributions from the international, national, regional and local sources in order to understand to what extent local authorities can enforce meaningful abatement measures. Moreover, source apportionment studies were conducted in order to define contributions from different sectors and to understand the maximum feasible air quality improvement that can be achieved by reducing emissions from those sectors, thus targeting emission reduction policies to the most relevant activities. Finally, an emission scenario reflecting the effect of such policies was developed and the associated air quality was modeled.

1. Introduction

Modeling is an essential tool for the development of atmospheric emission abatement measures and Air Quality Plans (AQP). Frequently, these plans are related to urban environments where the emission sources as well as the exposed population are concentrated (Vlachokostas et al., 2009; EEA, 2011). Developing a set of reliable tools for air quality modeling at urban scale is a very challenging task due to the fact that urban environments are particularly complex. The

environments are characterized by the presence of several pollutants emitted from multiple sources. Moreover, a series of different spatial and temporal scales are involved in the chemical transformation and transport processes of such pollutants. The inherent complexity of urban environments requires simulation tools to assess air quality levels to be able to support the analysis and evaluation of a variety of policies and emission abatement measures (Denby et al., 2011).

As environmental standards increase in strictness, more reliable modeling tools are needed to simulate any measure or plan intended to effectively tackle air quality exceedances. This fact implies the need to count on reliable and flexible inventories that integrally describe the emissions of urban sources, in accordance with the requirements of

the applied air quality models (FAIRMODE, 2010). There are a number of global and regional inventories that have been found useful to support air quality modeling studies (Pouliot et al., 2012; European Commission, 2009; Vestreng, 2003). Even at the urban scale, numerous inventories have been developed all over the world (Sturm et al., 1999; Sowden et al., 2008; Venegas and Mazzeo, 2006; Ho and Clappier, 2011 among others). However, there is a lack of harmonized and scientifically sound methodologies to address the compilation of urban scale inventories and to secure their consistency with existing regional and national inventories (Vedrenne et al., 2012).

Nitrogen dioxide is a clear example of a legislated air pollutant (EU Directive 2008/50/EC) with important implications for human health (Latza et al., 2009) that still poses an important challenge. Despite recent efforts made in Europe, ambient air concentrations of NO_x lag clearly behind the decreasing trend of NO_x emissions (Guerreiro et al., 2010). This is relevant for the compliance of NO_2 limit values, especially in urban environments. In 2010, 22 of the 27 EU Member States recorded exceedances of the limit value (EEA, 2012). Madrid (Spain) is one of the European cities where NO_2 is the main air quality issue and is legally bounded to develop an AQP to meet the required limit values (further details are provided in Section 2).

Specifically from the perspective of emission inventories, NO_x are also especially interesting since actual emission rates and chemical speciation depend to a large extent on how engines and combustion devices are operated and maintained as well as technological changes. This is particularly true for the road transport sector (Lee et al., 2013; Simmons and Seakins, 2012; Liu et al., 2009; Grice et al., 2009) which often constitutes the single most important source of NO_x emissions in urban environments. In addition, urban scale inventories usually need a fine spatial and temporal resolution, which cannot be achieved by downscaling methods or top-down inventories. This implies that methods to relate emissions with transport patterns and relevant activity data are need in the compilation of local inventories (Aritzegui et al., 2004).

The present study describes the modeling activities carried out for Madrid (Spain). The developed work is an illustrative example of the combination of models and emission data that are needed to provide a comprehensive picture of air quality at the urban scale and thus, provide the basis for the formulation of AQP.

2. Case Study

Madrid is the capital and largest city in Spain, located in the center of the Iberian Peninsula with a total population of 5 million people in its metropolitan area. Despite the experienced population and traffic increase, air quality levels have improved in the city over the last decade. However, some pollutants like nitrogen dioxide (NO_2) still exceed the limit values (LV) according to the European legislation. The NO_2 annual average recorded in most of the city's traffic air quality monitoring stations is usually above the LV ($40 \mu\text{g}/\text{m}^3$). This phenomenon is basically attributed to heavy traffic levels and to a strong dieselization of the fleet in recent years (Kassomenos et al., 2006).

In 2007, 80% of the monitoring stations exceeded the ambient air quality standards. As a consequence, important modeling efforts are being made to improve the knowledge about air quality dynamics in Madrid and to identify the most effective abatement options to meet the NO_2 LV in the near future. This work constitutes an extension of the integrated assessment modeling activities in Spain, which intends to provide useful tools for local policymakers to this respect (Borge et al., 2007). In particular it reports on the methods and results of the development and assessment of a local AQP enacted by the Madrid municipality (Madrid City Council, 2012) with a temporal horizon up to 2015. According to the latest data available (year 2012), the situation has improved substantially although 10 (out of 24) monitoring stations still report NO_2 annual means above the LV.

3. Methods

3.1. Mesoscale modeling

Urban concentration levels depend on the atmospheric phenomena that occur at different spatial scales, namely from international scales of thousands of km to street levels of a few meters (Monteiro et al., 2007). Additionally, these levels present complex interactions with a large variety of chemicals in the atmosphere. Up to now, no single model can describe these processes consistently so a combination of models is needed to address such description. Moreover, the choice of the model is basically dependent on the purpose of the simulation. In this context, last-generation, 3D Eulerian models equipped with full photochemical schemes can consistently describe transport and transformation processes of NO_x and tropospheric O_3 (the main species involved in the complex dynamics of photochemical chemistry) from continental to urban scale.

The mesoscale modeling system is based on the Weather Research and Forecasting (WRF) (Skamarock and Klemp, 2008), the Sparse Matrix Operator Kernel Emissions (SMOKE) modeling system (Institute for the Environment, 2009), and the Community Multiscale Air Quality (CMAQ) (Byun and Ching, 1999; Byun and Schere, 2006). Details about specific configuration and adaptation to the Spanish conditions can be found respectively in Borge et al. (2008a, 2008b; 2010a).

Four nested domains (Fig. 1) were used in order to capture international, national, regional and local contributions to NO_2 to ambient concentration in Madrid with a maximum resolution of 1 km^2 (Table 1). Each domain used dynamic boundary conditions from its immediate mother domain, except D1 that was run with fixed lateral chemical boundary conditions (details regarding CMAQ boundary conditions can be found in Borge et al., 2010a). A similar nesting approach was applied for the simulation of meteorology. The mother domain for the WRF model (slightly larger than D1 shown in Fig. 1) was run with initial and boundary conditions from the National Centers for Environmental Prediction (NCEP) Global Tropospheric Analyses with $1^\circ \times 1^\circ$ spatial resolution and temporal resolution of 6 h. This mesoscale configuration was found useful to describe urban background pollution levels, meeting the EU benchmarks for regulatory NO_2 modeling. The model uncertainty according to the Relative Directive Error (RDE) for this application reaches 23.7% (hourly LV) and 22.4% (annual LV), well below the maximum RDE criteria of 50% and 30%, respectively (Fig. 2). This corresponds to a global mean bias (MB) of $-2.2 \mu\text{g}/\text{m}^3$, a mean fractional bias of -14.1% and a global correlation factor (r) of 0.63.

3.2. Hotspot modeling

Despite a satisfactory performance of the mesoscale system, NO_2 presents strong concentration gradients that cannot be reproduced by mesoscale Eulerian models since large concentration variations typically exist within the extension of a grid cell. Such gradients have been observed in many urban environments (e.g. Vardoulakis et al., 2011) including specific measurement campaigns with passive samplers performed in Madrid (Karanasiou et al., 2011). In order to depict street level concentration gradients, specific and local-scale tools are needed; either high-resolution flow models that can resolve the buildings or semi-empirical street canyon models able to capture this variability (Vardoulakis et al., 2003).

To this respect, CFD (Computational Fluid Dynamic) models are very expensive computationally and can only be applied to spatially and temporally restricted domains. For this reason, simpler, parameterized operational street canyon models are preferred for planning and regulatory purposes (Vardoulakis et al., 2007). Street-scale systems, such as the Operational Street Pollution Model (OSPM) (Berkowicz et al., 2008) used in this study, are based on a combined plume and box model that can simulate in-street emissions and dispersion (including traffic-induced turbulence) according to local building

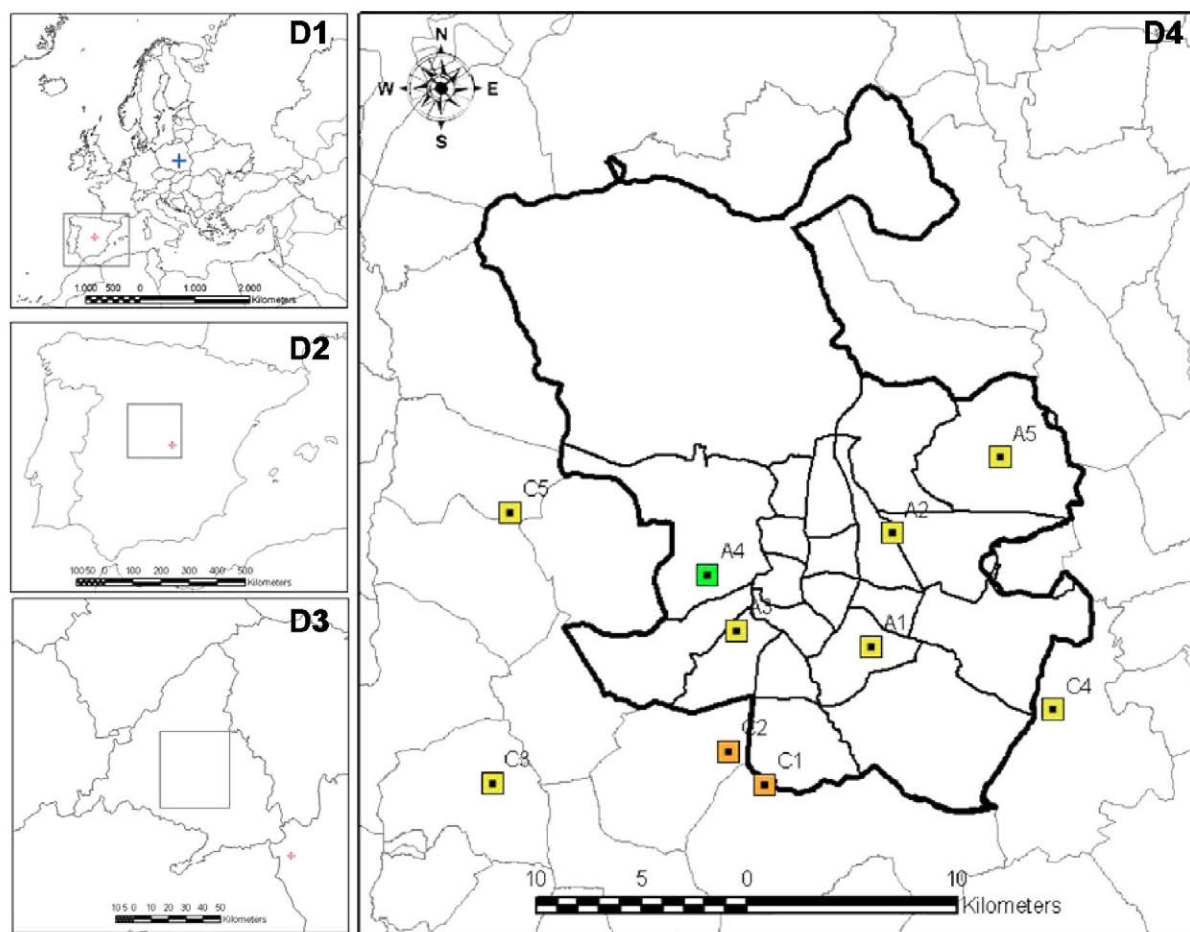


Fig. 1. CMAQ modeling domains. Note: the colored squares represent the location of air quality monitoring stations used for evaluation purposes in the innermost modeling domain (1 km resolution). Squares in green, yellow and orange indicate the station type (suburban, urban background and traffic respectively) according to the air quality monitoring network (A – Madrid City Council, C – Madrid Greater Region). (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)

geometry and street configuration. Besides the short computational time requirements, these models provide a rough representation of very fast chemistry (i.e. primary NO oxidation depending on O₃ background levels), which dominates NO₂ levels at traffic locations. Street canyon models however, need to be carefully coupled to the mesoscale model system (meteorology, background concentration) in order to obtain a consistent representation of air quality.

In this study, outputs from WRF and CMAQ were used to provide wind conditions and pollution background concentration at roof level, inputs to which street canyon models are very sensitive. In addition, consistent emission data have to be used across the scales and models. In this application, a common traffic model is used to provide the activity data (intensity, fleet composition) and relevant variables (average speed, etc.) needed for traffic emission computation (as discussed in the next section). The results indicated that when properly fed (meteorology, background pollution and traffic conditions), the street-canyon model is able to achieve a reasonable performance (RDE < 20%) even

at heavily trafficked hotspots with hourly peak values close to 400 µg NO₂/m³, such as the traffic station illustrated in Fig. 3.

The results obtained after identifying and analyzing this traffic hotspot within the city of Madrid strongly highlight the need of conducting such studies in order to estimate the effectiveness of specific policies and control measures as well as to evaluate the resulting technological improvements in vehicles (Beelen et al., 2009). To this respect, hotspot-modeling activities provide valuable information regarding the street level, which if overlooked, might originate ineffective legislative measures (Giannouli et al., 2011).

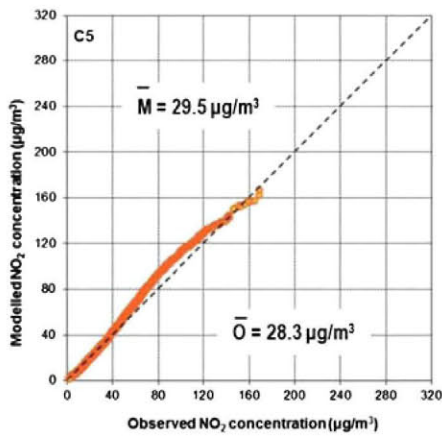
3.3. Emission inventories

For the completion of an efficient air quality management framework, the existence of a robust emission inventory is crucial (Moussiopoulos et al., 2009). Moreover, emissions constitute a key input to air quality models since they are deemed being one of the main sources of uncertainty (Russell and Dennis, 2010). This issue is also relevant for the analysis of alternatives to improve air quality in a given region in future years as a result of the implementation of emission abatement (FAIRMODE, 2010).

As for the implications related with multi-scale studies, emission estimates constitute one of the most challenging aspects. Emission-related inputs must be as detailed and specific as possible for the different domains involved in the simulation, and simultaneously they must be consistent across the scales (Borge et al., 2009). In addition, they have to be flexible and detailed enough to reflect the outcome of relevant measures and meet the modeling system requirements (Borge

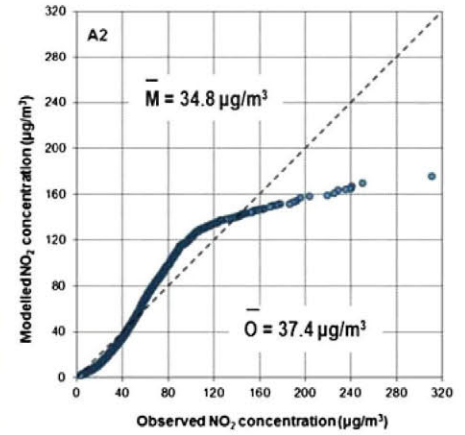
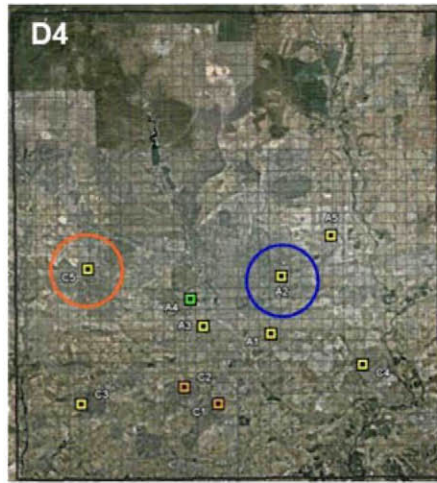
Table 1
Spatial domains for the mesoscale modeling system.

Domain	Geographic scope	X–Y dimensions (km)	Horizontal resolution (km)
D1	Europe	6144 × 5376	48
D2	Iberian Peninsula	1200 × 960	16
D3	Greater Madrid Region	192 × 192	4
D4	Madrid Metropolitan Area	40 × 44	1



RDE* hourly LV = 1 %
RDE* annual LV = 3 %

MB = 1.2 µg/m³ ME = 16.2 µg/m³ r = 0.608



RDE* hourly LV = 20 %
RDE* annual LV = 6 %

MB = -2.5 µg/m³ ME = 18.4 µg/m³ r = 0.608

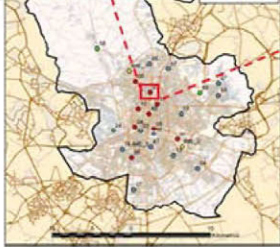
* Relative Directive Error (RDE)
Directive 2008/50/EC

$$RDE = \frac{|O_{LV} - M_{LV}|}{LV}$$

where: O_{LV} – closest observed concentration to the limit value concentration (LV)

M_{LV} – correspondingly ranked modelled concentration

Fig. 2. Computation and results of the Relative Directive Error (RDE) for the innermost domain and two examples for individual monitoring stations (C5 in the left and A2 in the right). C5 is an example of an urban background station. Although A2 is also labeled as an urban background site it is more influenced by direct traffic emissions. Mean error (MB) and error (ME) are larger but the correlation coefficient (r) is the same. According to the Directive 2008/50/EC, the uncertainty for NO₂ modeling should be assessed by the Maximum RDE (MRDE) found at 90% of the available stations. The results for this application are 23.7% (hourly limit value) and 22.4% (annual limit value).



RDE hourly LV = 5 %
RDE annual LV = 13 %

MB = 5.0 µg/m³

ME = 22.8 µg/m³

r = 0,665

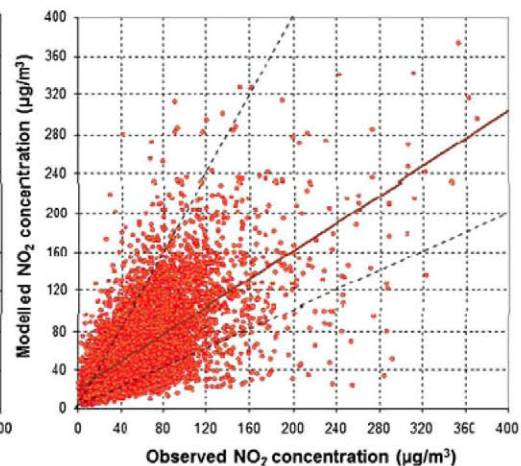
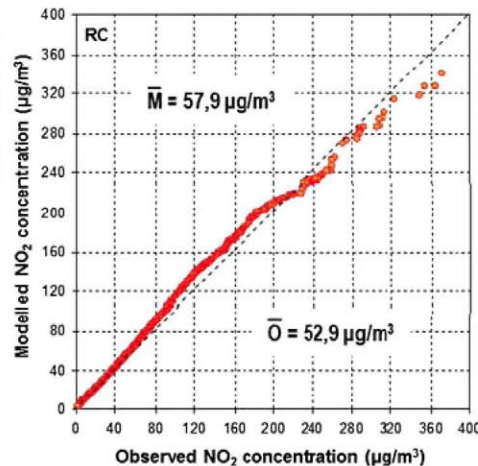
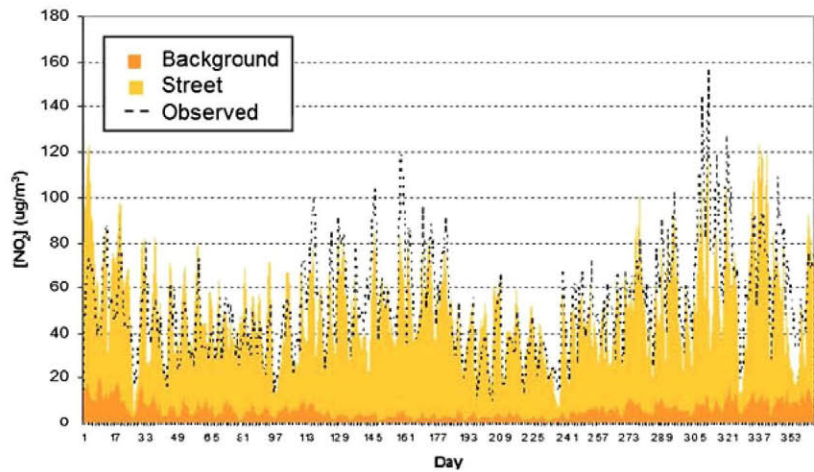


Fig. 3. Location and OSPM results (annual series, Q-Q and scatter plots) for a traffic station.

et al., 2010b). Consequently, a specific emission inventory has been developed and adapted for each of the four modeling domains in this application. Emission processing was performed with SMOKE in every case.

3.3.1. Domain 1 – Europe (D1)

Anthropogenic emissions were taken from the EMEP inventory, which consists of a gridded inventory ($50 \times 50 \text{ km}^2$) that covers Europe completely and which was compiled from national submissions to the Convention on Long-range Transboundary Air Pollution (LRTAP) (Vestreng, 2003). The temporal profiles and vertical distribution needed to resolve the emissions were those used in the EuroDelta experiment (van Loon et al., 2007). Biogenic VOCs (isoprene, monoterpenes and other biogenic volatile organic compounds) have been computed offline (the Global Emission Inventory Activity – GEIA) and processed into SMOKE by implementing the algorithms proposed by Guenther et al. (1996). Both inventories were consistent with the EMEP/CORINAIR methodology used to compute emissions in the Spain's National Emission Inventory (SNEI) (EEA, 2009).

3.3.2. Domain 2 – Iberian Peninsula (D2)

The emissions used for this domain were taken from the National Emission Inventories of Spain (SNEI) and Portugal (PNEI) and processed with SMOKE. Hourly, 16-km resolved emissions from 184 area-source categories were used along with detailed information regarding temporal patterns and release conditions of 1720 stacks belonging to 62 point-source categories. The inventory was chemically speciated according to the Carbon Bond CB05 mechanism, a lumped structure

chemical mechanism including 156 reactions and 69 species including aerosols (Yarwood et al., 2005). The chemical composition of VOCs, $\text{PM}_{2.5}$ and NO_x emissions in the inventory was defined through 221 chemical profiles built from the relevant information contained the EMEP/CORINAIR guidebook and the US EPA ESPECIATE database (EEA, 2009; Hsu et al., 2006).

3.3.3. Domain 3 – Greater Madrid region (D3)

The emission inventory compilation and implementation for D3 were the results of a thorough intercomparison exercise of two official inventories available for this area, the regional inventory and the regional disaggregation of the SNEI (Fig. 4) (Vedrenne et al., 2012). The analysis relied on the fundamental hypothesis that the accuracy of an emission estimate may be assessed by the degree of agreement between air quality observations and the results of an air quality model (CMAQ) feed with that emission information while keeping all other inputs (meteorology, boundary conditions, etc.) constant (Borge et al., 2010b). The analysis of the differential response of the model at representative points in the modeling domain (Fig. 4) along with the analysis of the differences on alternative emission estimates was used to find out which of the underlying methods and information used in both inventories may provide a more accurate estimate of the actual emissions (total amount, sector share and geographical and temporal distribution) in the Madrid region.

The results confirmed the lack of consistency between national, regional and local emission inventories, which has been a longstanding problem in multi-scale air quality modeling (FAIRMODE,

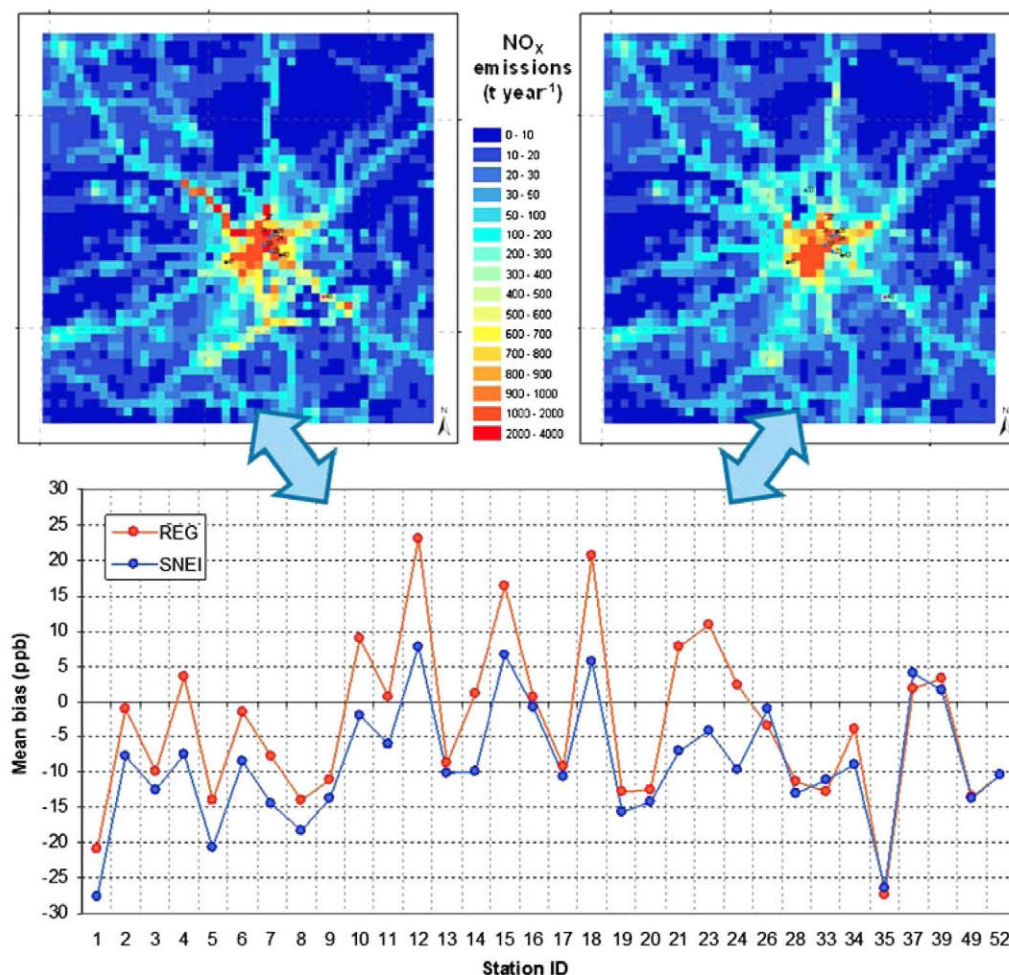


Fig. 4. Comparison of emission estimates in D3 based on the SNEI inventory (left) and the regional inventory (right) and response of the air quality model in particular locations (corresponding to air quality monitoring stations); national inventory in blue (SNEI) and regional inventory in orange (REG). (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)

Table 2
Summary of emissions (SNAP group level) in D4. Reported in metric tons per year.

SNAP group	CO	NH ₃	NO _x	PM ₁₀	PM _{2.5}	SO ₂	VOC
01 – Combustion in energy and transformation industries	225	0	243	50	29	1128	1
02 – Non-industrial combustion plants	10,004	0	3680	520	410	2731	1104
03 – Combustion in manufacturing industry	2238	0	10,689	265	210	2494	1217
04 – Production processes	1083	130	108	51	32	70	3782
05 – Extraction and distribution of fossil fuels	0	15	0	0	0	0	2056
06 – Solvent and other product use	0	212	0	0	0	0	48,828
07 – Road transport	22,070	250	27,961	1506	1205	157	4365
08 – Other mobile sources and machinery	2711	0	4171	360	360	287	769
09 – Waste treatment and disposal	441	2036	1769	26	26	6	5267
10 – Agriculture	357	1543	56	90	13	0	17
11 – Other sources and sinks (Nature)	32	605	125	0	0	0	4682
Total	39,161	4791	48,802	2868	2285	6873	72,088

2010). The resulting inventory for D3 was a combination of emission data from both official inventories based on the understanding of the reasons for discrepancies between them. Besides helping to understand which inventory provided a better estimate, the study was useful to identify preliminary ways to conciliate future editions of both inventories.

3.3.4. Domain 4 – Madrid Metropolitan Area (D4)

The criteria for the design and computation of the emission inventory for the innermost domain consisted in the combination of bottom-up and top-down methods paying special attention to keep the consistency across domains/inventories, the use of very detailed source-specific methods and the implementation of a flexible and detailed approach in order to reflect the outcome of relevant emission reduction measures. The computations carried out at this stage revealed that road traffic (SNAP group 07) is responsible for 57% of NO_x emissions in this modeling domain, as summarized in Table 2. The resulting inventory had the capability to simulate strategies aimed

at cutting down emissions from this sector such as implementing low emission zones (access restrictions by vehicle type, age or technology), changing speed limits or allowing penetration of new technologies (combustion engine standards, hybrid and electric vehicles, etc.). It also considered specific fleet turnover and limitations by segments (busses, taxis, light duty vehicles, passenger cars) as well as measures to alleviate urban congestion.

At this scale, the reference model for calculating emissions from road traffic was COPERT 4 (Ntziachristos et al., 2009), which was run for a great number of road links contained in 9 management areas as shown in Fig. 5. Alternatively, emissions from road traffic were computed with HBEFA 3.1, which is a model based on traffic situations (HBEFA, 2012). Further descriptions on the configuration of these models and on the estimation of traffic-related emissions can be found in Borge et al. (2012). The resulting emissions for each vehicle type are referred to link level and have a 1-hour temporal resolution. The subsequent spatial allocation of the emissions within the Eulerian grid was carried out through an overlapping process as

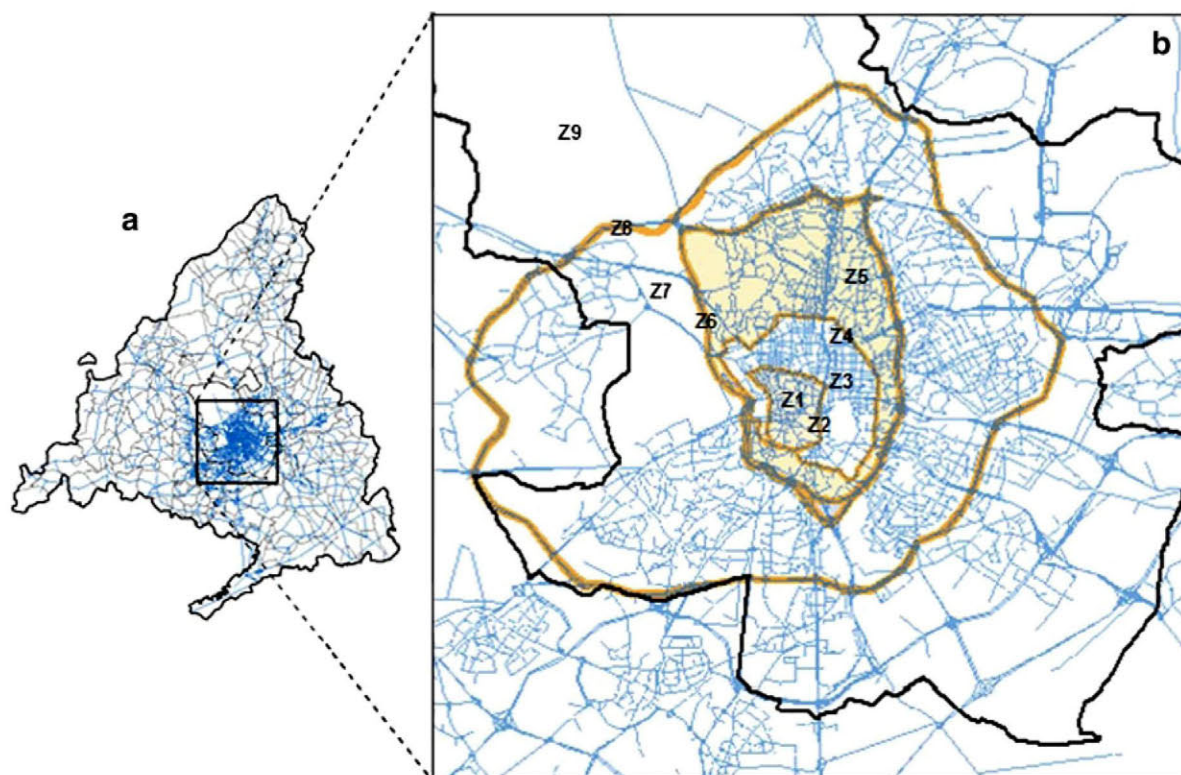


Fig. 5. Road network of the traffic model (a) and zoom to the city center with indication of the nine management areas (b) defined by the Madrid City Council, referred to as Z1 to Z9.

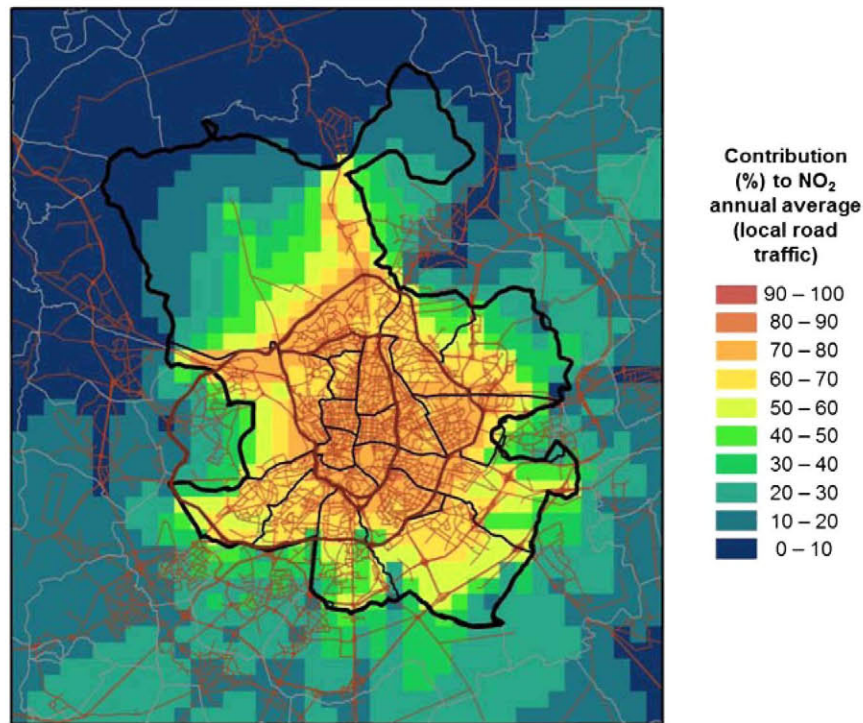


Fig. 6. Result of the source apportionment analysis for the road traffic sector (SNAP group 07).

described in Borge et al. (2008b). Additionally, this procedure allowed providing emissions directly to the street canyon model, which constituted a distinct advantage to keep consistency between the mesoscale and the street-scale models.

Besides road traffic, all the other relevant sectors were represented with a sufficient detail. Relatively important sources such as those of the domestic, residential and commercial sector (SNAP group 02) were inventoried under a bottom-up approach and classified according

to their consumed fuel. This issue allows any fuel change or boiler turn-over processes to be simple and quickly simulated.

4. Results and discussion

Once the emission calculations were accomplished, the modeling system was used to perform a series of analyses and experiments that resulted in the definition of a complete strategy aimed to meet

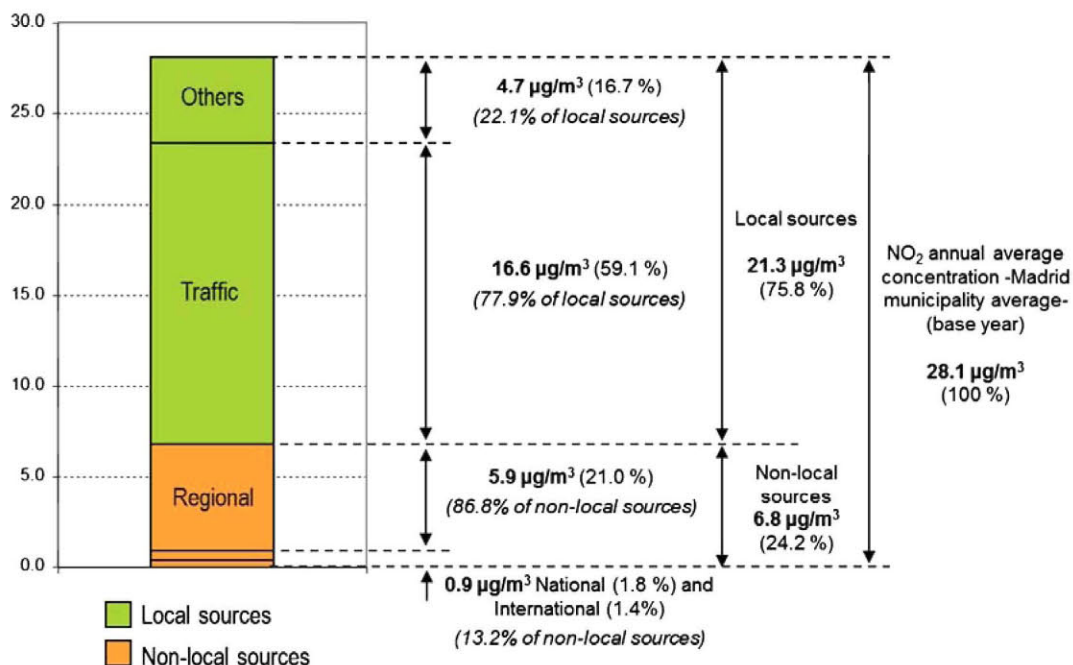


Fig. 7. Result of the source apportionment analysis (annual NO₂ mean for the whole Madrid municipality).

the NO₂ air quality standards required by 2015 in Madrid (Madrid City Council, 2012).

4.1. Source apportionment exercise

A source apportionment exercise of NO₂ levels is usually a necessary procedure to be carried out when developing an AQP intended to demonstrate future compliance under the European legislation as well as to define meaningful abatement options. The analysis for the relevant time period (e.g. annual basis for the NO₂ annual LV) provided essential information regarding the basic emission abatement strategy/course of action, a maximum feasible air quality improvement related to the main emitting sectors and a series of external constraints.

A zero-out methodology was followed in this application. The contribution of a particular emission source or region can be estimated through the brute force method (BFM), sometimes referred to as single-perturbation method (Samaali et al., 2011 and references within). This method relies on the analysis of the change in the pollutant concentration that would occur if a given emitting source is removed from the simulation (usually referred to as zero-out sensitivity runs). This approach has been used in the past to isolate the

response of complex, nonlinear systems to one particular sector in source apportionment and sensitivity analysis (Cohan et al., 2005). This method has inherent limitations in accurately describing sensitivities but it may be useful to approximate the effect of potential emission reductions in a particular source or origin area as pointed out before in several studies (Koo et al., 2009; Carmichael and Wild, 2010; Leung et al., 2007).

Reductions of 100% (zero-out) were simulated for the most relevant anthropogenic emissions, including road transport, industry, aviation and residential, commercial and institutional combustion (RCI). The total impact and therefore the maximum theoretical benefits that can be harvested by implementing abatement options in these sectors were derived from the comparison of the assessment of the individual runs with the base case (considering all emissions). This premise is illustrated in Fig. 6, from which it can be seen that reductions of up to 90% can be achieved theoretically after applying restrictions only to the road traffic sector. Moreover, reductions decrease with the distance to the city center, witnessing the preponderance of road traffic as the main polluting activity for an urban environment such as Madrid.

A similar approach was followed to estimate the contributions of different geographic areas. In this case outputs from CMAQ runs using

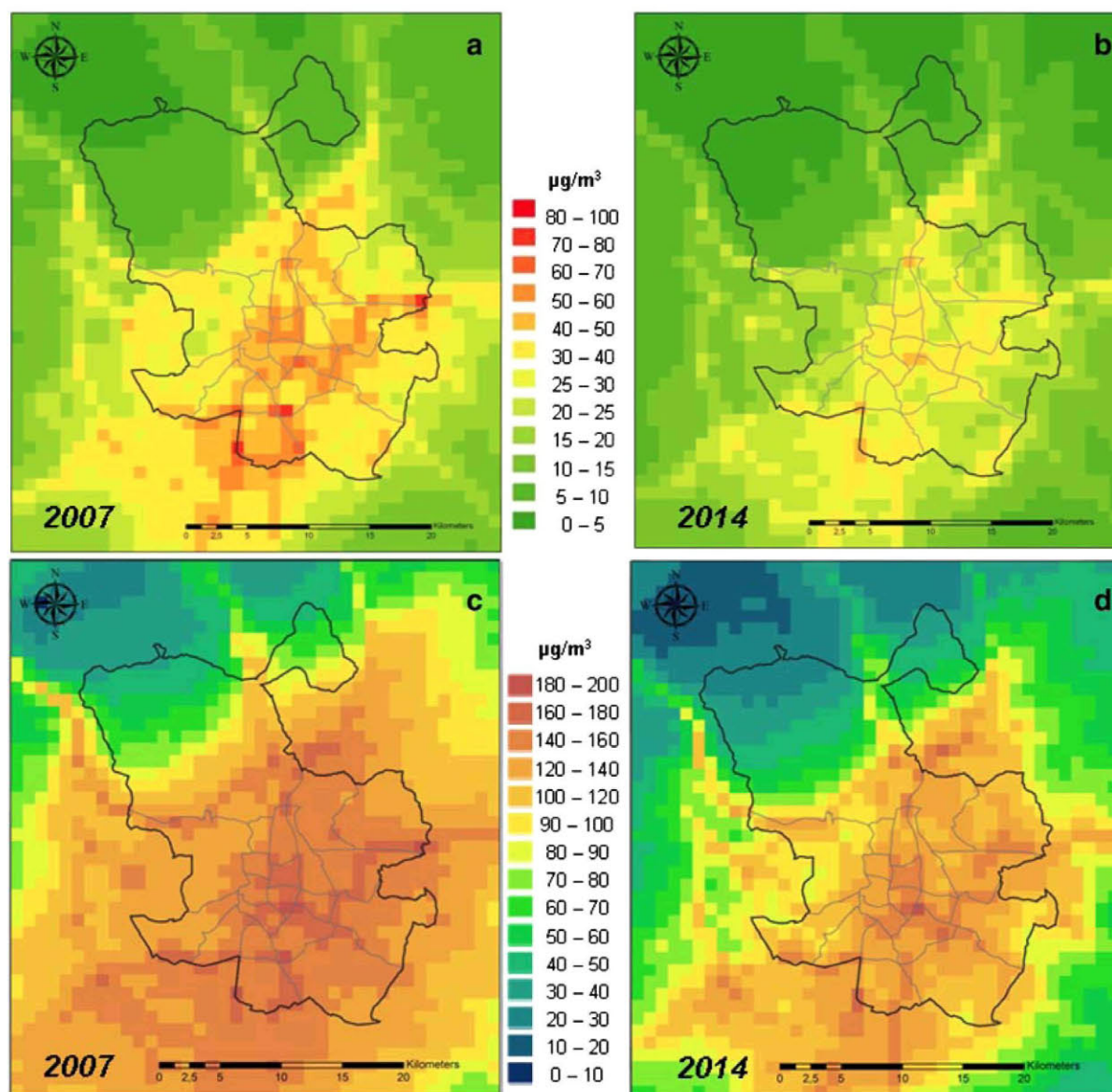


Fig. 8. Expected effect of the Madrid AQP in NO₂ concentration: annual limit value (a–b) and 1-hour limit value (c–d).

alternative boundary conditions and geographical masks (i.e. setting to zero the emissions in grid cells corresponding to a given administrative division, such as the municipalities of the Madrid Greater Region other than Madrid) were compared in order to derive the amount of NO_2 that can be related to different origin areas. Fig. 7 shows the average geographic apportionment structure for the entire Madrid municipality. This confirmed that Madrid is strongly dominated by local sources (76%), mainly road traffic, contributing up to $16.6 \mu\text{g}/\text{m}^3$ to the annual NO_2 value as an average over the Madrid municipality (nearly 80% of local contributions). Other local sources (mainly RCI and waste management) had a much lower influence, roughly $5 \mu\text{g}/\text{m}^3$ (around 20% of local contributions). The regional influence accounted for approximately 20% of the total mean concentration ($5.9 \mu\text{g}/\text{m}^3$), while the national and international influences were negligible (less than $1 \mu\text{g}/\text{m}^3$ combined). In this context, regional refers to other municipalities within the Madrid Greater Region and national refers to locations beyond the Madrid Greater Region.

It should be noted that non-local sources cannot be directly regulated or controlled by the Madrid City Council and therefore are outside the scope of any local plan or policy. However, most of the regional contribution originates in adjacent municipalities that are part of the Madrid metropolitan area where emissions are also strongly dominated by the road traffic sector. These results indicated that an efficient AQP for Madrid should include measures aimed at limiting local road traffic with an additional effect at the whole metropolitan area.

4.2. Emission scenario

Complementing the source apportionment exercise, the development and modeling of a future-year emission scenario is a crucial stage for the design of effective abatement options and assessment of the compliance with air quality standards (Boogaard et al., 2012). Since there are no universal solutions to improve air quality, the particular features of any reduction plan will depend on the causes of poor air quality levels. Future-year emission estimates should be consistent with the methods applied for the base year emission inventory compilation (Int Panis et al., 2004). Changes or updates of computation methods may lead to important deviations in future year estimates and therefore misleading information about the effectiveness of particular measures. For instance,

preliminary experiments revealed important differences (up to 20%) in NO_x emissions for the Madrid metropolitan area depending on the road traffic emission model used. Important differences were also found in critical parameters such as the NO_x emission speciation (NO/NO_x ratio) for future engine technologies. Further analysis on this issue and examples of consistent emission projection methods are provided in Lumbreras et al. (2008).

Up to 70 abatement measures have been assessed and evaluated for the final definition of the Madrid AQP. Consistently with the results of the source apportionment exercise performed, most of the measures were targeted to the road traffic sector. Measures such as the definition of a Low Emission Zone (LEZ), reduction of road capacity and pedestrianized areas in the city center, and renovation of city bus fleet to incorporate clean technologies (electric, hybrid natural gas-fuelled busses) would achieve a 40% reduction of NO_x emissions from the road traffic sector in the modeling domain. Further details can be found in Madrid City Council (2012). As a result of all the measures included in the AQP, a global decrease of 31% in NO_x emissions is expected in the year 2014 within D4 (relative to the emissions of 2007 used as reference scenario or base year). Emissions, surrogate data and speciation profiles were updated to reflect the expected composition of fleet and other structural measures (expansion of the regulated parking system, regulation of the working schedule of taxis, etc.). This is also relevant for measures affecting other sectors such as fuel switch in domestic boilers or upgrades of engines and other equipment in waste water plants. The simulation of NO_2 ambient concentration values corresponding to this emission scenario pointed out that compliance with LV could be achieved in Madrid by 2015. Fig. 8 compares CMAQ outputs for 2007 (base year) and 2014 (implementation of the AQP). From this comparison it can be inferred that annual NO_2 levels may be reduced by 34% as an average; approximately $15 \mu\text{g}/\text{m}^3$ in the city center, also with an important impact in the metropolitan area ($-7 \mu\text{g}/\text{m}^3$ as an average in the modeling domain). 1-hour concentration peaks may also decline by 40% approximately in most of the city.

The modeling platform was also useful to estimate the effect of additional measures that may be applied under exceptional conditions or short-term exceedance situations (such as those produced under unfavorable meteorological conditions), as illustrated in Fig. 9. This can be accomplished by conveniently changing emission figures and

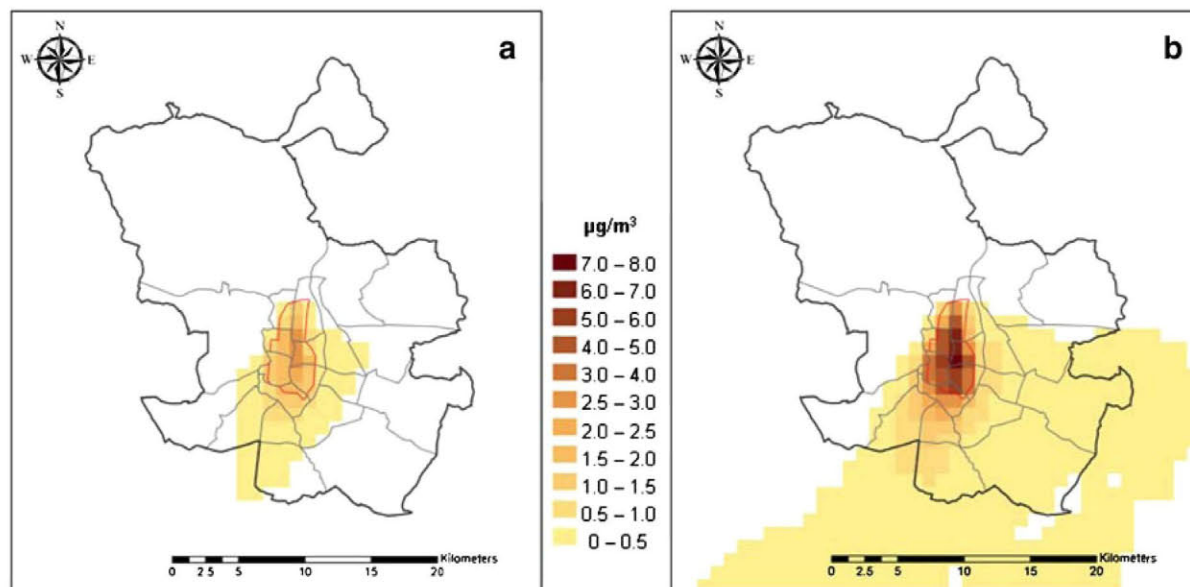


Fig. 9. Reduction of average NO_2 concentration due to passenger car access restriction to the Low Emission Zone (LEZ) (the area inside the inner red line) by 20% (a) and 50% (b) during a 24-h period under unfavorable meteorological conditions. (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)

surrogate data for specific SMOKE activities (linked to specific vehicle types and management areas). The same emissions can be used to feed the street-scale OSPM model to complement the analysis providing thus relevant information regarding compliance expectancy in hot spots.

5. Conclusions

The development and assessment of an AQP in an urban area constitute a very complex task from the air quality-modeling point of view. The definition of effective abatement measures implies the need of a previous analysis of source apportionment regarding both the geographic origin of pollutants and the identification of sources responsible for their emission. These analyses involve rather different temporal and spatial scales and require the combination and harmonization of models and data. Emission inventories play a crucial role in this context since the assessment of a given measure will entirely depend on how accurate is the representation of that measure in terms of emissions. Therefore, the emission processing system used in this kind of applications should be able to combine information from a variety of sources and it needs to be flexible and detailed enough to reflect the outcome of relevant emission reduction measures.

This paper summarizes the modeling activities carried out in Madrid (Spain) to develop an AQP to comply with the annual NO₂ European standards. The study shows that the SMOKE system is able to accommodate emissions from at least four emission inventories from the European scale EMEP inventory to a very detailed bottom-up emission inventory for the Madrid city.

The present study implemented a series of good practices that are recommendable when developing scenarios. This included focusing on the abatement measures on the emission sectors responsible for air pollution, according to the developed source apportionment study, the use of a projection model consistent with emission inventories, as well as transparent practices.

The source apportionment exercises made for Madrid AQP indicate that NO₂ ambient concentration values are strongly dominated by local sources with a remarkable contribution from road traffic. Therefore, a package of 70 measures, mostly targeted at this sector, was proposed and simulated. According to the results of this study, this scenario would cut down NO_x emissions by 31% and would allow the fulfillment of NO₂ limit values in Madrid by the end of 2014.

Conflict of interest

There is no conflict of interests.

Acknowledgments

The Madrid City Council provided the traffic model and supported this study. The CMAQ modeling system was made available by the US EPA and it is supported by the Community Modeling and Analysis System (CMAS) Center. The authors also acknowledge the use of emission datasets and monitoring data from the Spanish and Portuguese Ministries of Environment.

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