# AIR QUALITY FORECASTS ON A KILOMETER-SCALE GRID OVER COMPLEX SPANISH TERRAINS

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#### 10 Abstract

11 CALIOPE-AQFS represents the current state-of-the-art in air quality forecasting 12 systems of high-resolution running on high-performance computing platforms. It 13 provides a 48-h forecast of NO<sub>2</sub>, O<sub>3</sub>, SO<sub>2</sub>, PM<sub>10</sub>, PM<sub>2.5</sub>, CO, and C<sub>6</sub>H<sub>6</sub> at 4-km 14 horizontal resolution over all of Spain, and at 1-km horizontal resolution over the most 15 populated areas in Spain with complex terrains (Barcelona, Madrid and Andalusia 16 domains). Increased horizontal resolution from 4 km to 1 km over the aforementioned 17 domains leads to finer texture and more realistic concentration maps, which is justified 18 by the increase in  $NO_2/O_3$  spatial correlation coefficients from 0.79/0.69 (4 km) to 19 0.81/0.73 (1 km). High resolution emissions using the bottom-up HERMESv2.0 model 20 are essential for improving model performance when increasing resolution on an urban 21 scale, but it is still insufficient. Decreasing grid spacing does not reveal the expected improvement in hourly statistics, i.e., decreasing NO<sub>2</sub> bias only by  $\sim 2 \mu gm^{-3}$  and 22 increasing  $O_3$  bias by ~1 µgm<sup>-3</sup>. The grid effect is less pronounced for PM<sub>10</sub> because 23 24 part of its mass consists of secondary aerosols, which are less affected than the locally 25 emitted primary components by a decreasing grid size. The resolution increase has the 26 highest impact over Barcelona, where air flow is controlled mainly by mesoscale 27 phenomena and a lower PBL. Despite the merits and potential uses of the 1-km 28 simulation, the limitations of current model formulations do not allow confirmation of 29 their expected superiority close to highly urbanized areas and large emission sources. 30 Future work should combine high grid resolution with techniques that decrease subgrid variability (e.g., stochastic fields methods), and also include models that consider urban
 morphology and thermal parameters.

### 3 1 Introduction

The World Health Organization (WHO) has recently shown that there is sufficient 4 5 evidence that particulate matter (PM), ozone  $(O_3)$  and nitrogen dioxide  $(NO_2)$  affect human health (WHO, 2013). Although  $NO_2$  and PM concentrations improved from 6 7 2002 to 2011 in Europe, the situation is still far from matching the WHO air quality 8 guidelines (AQG). The European annual limit values for NO<sub>2</sub> (annual) and PM<sub>10</sub> (daily) 9 had been exceeded at 42-43% of the traffic stations in 2011. For the same year, about 10 33% of the European urban population was exposed to  $PM_{10}$  concentrations above the 11 daily limit value, and nearly 88% was exposed to the respective WHO AQG (EEA, 12 2013). Air pollution legislation for the protection of the increasing city population has 13 recently increased the demand for forecasting systems that can assess and understand air 14 pollution dynamics, alert the population when health-related issues occur, and develop 15 emission abatement plans (EEA, 2011).

16 When applying an air quality modeling system, defining the grid resolution is an 17 important consideration. The potential benefits of higher-resolution modeling should be 18 weighed against the increased complexity of the inputs, increased CPU time, and disk 19 space requirements. In theory, higher resolution modeling is expected to yield more 20 accurate forecasts because of better resolved model input fields (topography, land cover 21 and emissions). Furthermore, high resolutions (ranging from 1 to 5 km) are essential to 22 reproduce mesoscale phenomena, e.g., those controlling  $O_3$  transport along the 23 mountainous northeastern Mediterranean coast (Fay and Neunhäuserer, 2006; Jiménez 24 et al., 2006). Even at fine scales, the modeled concentrations are not necessarily the best 25 (Mass et al., 2002; Gego et al., 2005; Valari and Menut, 2008) because increasing 26 emission and meteorology spatial resolution can also increase uncertainties, at the risk 27 of reduced model performance. Nowadays, fine horizontal resolution is a persistent 28 challenge when assessing health impacts and population exposure studies (Thompson et 29 al., 2013).

Several studies have evaluated the impact of increasing horizontal resolution on
different scales over the eastern and southeastern USA using the Community Multiscale
Air Quality (CMAQ) model and the Comprehensive Air Quality Model with Extensions

(CAMx), which range from 32 km - 12 km - 4 km (Cohan et al., 2006; Tesche et al., 1 2 2006; Queen and Zhang, 2008). They found no significant changes for  $O_3$  and PM (<5% 3 on average), and those changes were even smaller at resolutions of between 12 km and 4 4 km (<3%). Concerning PM components, Fountoukis et al. (2013) found that increased 5 resolution provides differences mostly for primary PM rather than secondary PM. 6 Recently, a model intercomparison exercise, named ScaleDep, was performed to 7 determine the effect of grid resolution on air quality modeling performance over Europe at a regional and urban scale (Cuvelier et al., 2013). The exercise, involving five 8 9 chemical transport models (CTMs) (EMEP, CHIMERE, CMAQ, LOTOS-EUROS and 10 RCGC) running under the same conditions over the full year 2009 and at four 11 resolutions (56, 28, 14, and 7 km), showed that it is difficult to define a grid size that is 12 adequate for resolving the urban signal under all conditions affecting Europe. Still, a 14-13 km resolution seems to be a good compromise between background applications and 14 those reproducing most of the urban signals (7 km resolution). However, the ScaleDep 15 exercise did not distinguish between the different topographies or complex 16 meteorological patterns which are characteristic of the Iberian Peninsula.

17 Few studies have been performed over selected areas in Spain; and of those, the focus 18 has been mainly on  $O_3$  and  $NO_2$ . Vivanco et al. (2008) evaluated the annual impact by increasing model resolution (to 36, 19, and 7 km) over Madrid for NO2 and O3. They 19 20 used the WRF-CHIMERE model, disaggregating the EMEP emission inventory based 21 on land use information. Their evaluation showed that the model improved more for 22 NO<sub>2</sub> than O<sub>3</sub>, with the most significant improvement achieved when resolution 23 increases from 36 to 19 km rather than to 7 km, which they linked to increased 24 uncertainty in the emission data introduced with the disaggregation techniques. Jiménez 25 et al. (2006) used the MM5-CMAQ model along with a bottom-up emission model 26 (EMICAT2000) to assess the influence of grid resolution on O<sub>3</sub> (at 8, 4, and 2 km) over 27 the complex terrain of the northeastern Iberian Peninsula (Catalonia) during a summer 28 pollution episode. They found that both an improved performance of the mesoscale 29 phenomena and a better allocation of emissions for the 2-km resolution improve the 30 capability of the model to simulate exceedances of European limit values. An important 31 issue in both studies is the emission modeling approach (top-down vs. bottom-up) when 32 applying high resolutions at the local scale (<10 km). As Fountoukis et al. (2013) and 33 Timmermans et al. (2013) demonstrate, in the range of local scale (e.g., the greater Paris area), the grid resolution is not currently the major source of discrepancies in model performance, but instead the predicted concentrations and corresponding gradients are more consistent with observed concentrations when provided by bottom-up emission inventories rather than down-scaled inventories. If local variation in input data (e.g., emission patterns or land use) cannot be properly characterized, modeling with a finer grid resolution may not provide any great advantages.

7 Increasing resolution is a technical challenge, since computational cost markedly 8 increases in inverse proportion to grid spacing. The current progress in computation 9 allows increased model resolution and for multiple spatial scales to be investigated with 10 the aim of establishing adequate grid size for forecasting air quality at the local scale. 11 Recently, Colette et al. (2014) evaluated the impact of increasing resolution up to 2 km 12 over the European continent by using the CHIMERE model for an episode of air 13 pollution in 2009. They simulated 2 million grid cells using over 2000 CPUs of a high 14 performance computing system, which was hosted by the French Computing Centre for 15 Research and Technology (CCRT/CEA).

16 In terms of computational resources, horizontal resolution is critical to an operational air 17 quality forecast. In Europe, operational air quality systems use resolutions between 12-18 25 km, while applications to a single country can reach resolutions between 4-10 km 19 (Zhang et al., 2012). Over Spain, there are three systems providing air quality forecasts 20 running at different horizontal resolutions. The lowest resolution system is the 21 Technical University of Madrid's OPANA (OPerational Atmospheric Numerical model 22 for urban and regional Areas), running at 27 km x 27 km and based on the 23 MM5/CMAQ/EMIMO models (San José et al., 2009). It is followed by the Spanish 24 meteorological office's (AEMET, system 25 http://www.aemet.es/es/eltiempo/prediccion/calidad\_del\_aire), which forecasts at 10 km 26 x 10 km using the HIRLAM-HRN/MOCAGE/GEMS-TNO models. The CALIOPE Air 27 Quality Forecast System (CALIOPE-AQFS; Baldasano et al., 2011; Pay et al., 2012a; 28 and references therein), of the Barcelona Supercomputing Center-Centro Nacional de 29 Supercomputación (BSC-CNS), runs at the highest resolution, 4 km x 4 km, and it is 30 based on the WRFv3.5/CMAQv5.0.1/HERMESv2.0/BSC-DREAM8bv2 models. 31 Moreover, CALIOPE-AQFS provides 1-km x 1-km resolution forecasts for the Madrid 32 and Barcelona metropolitan areas (since 2009) and the Andalusian region (since 2013).

Such resolution has been possible thanks to both the high performance computing
 resources at the BSC-CNS and the availability of detailed emission data covering Spain.

3 The present work aims to assess the impact of increasing the horizontal resolution from 4 4 km to 1 km, specifically over areas affected by heterogeneous emission patterns and 5 complex terrains such as the Barcelona and Madrid metropolitan areas (BCN and 6 MAD) together with the Andalusian region (AND). For that purpose, the CALIOPE-7 AQFS forecasts pollutant concentrations (O<sub>3</sub>, NO<sub>2</sub>, and PM<sub>10</sub>) at two horizontal 8 resolutions: first at a 4-km resolution covering Spain (IP4), and second at a 1-km 9 resolution covering the AND, BCN, and MAD domains. Here, a study is performed for 10 the period April 2013, which presents a seven day air pollution episode. We use 11 observations from routine air quality monitoring networks to evaluate both resolutions.

12 Section 2 describes the configuration and computational setup of CALIOPE-AQFS. 13 Section 3 quantifies the impact of resolution increase on forecasting hourly 14 concentrations (and exceedances) in terms of: pollutant, domain, building density and 15 major emission sources. Section 4 concludes with the main results and some 16 recommendations.

# 17 2 Methodology

# 18 **2.1** Domain and period under study

19 Figure 1 shows the primary NO<sub>2</sub> emission patterns and topographic characteristics of 20 the BCN, MAD and AND domains. BCN is a coastal area characterized by several 21 valleys perpendicular to the coastline and two main mountain ranges, one coastal (500 22 m height) and one pre-coastal (1000-1700 m height). These features induce mesoscale 23 phenomena such as sea-breeze and mountain-valley winds. On the other hand, MAD is 24 a continental region with a much simpler topography that includes the Tajo valley in the 25 south and the mountain range of the Central System in the north, with summits reaching 26 2500 m.

The urban contribution in BCN (3.1 million inhabitants) is accompanied by industrial and power generation emissions, the road network and the harbor: Meanwhile, the Spanish capital of MAD is mainly influenced by emissions from the urban area (5.8 million inhabitants) and the road network that connects MAD with the surrounding commercial and industrial zones, as well as the urban areas. AND is the southern-most region in Spain, with complex topography characterized by the large depression of the Guadalquivir Basin (delimited by the Iberian Massif and the Betic Range), which crosses the region from NE to SW over a 60–km stretch. About three quarters of AND has a mountainous orography, including the Sierra Nevada (3481 m). AND includes one of the five biggest cities in Spain, Seville (~700 000 inhabitants), which hosts industrial and electric generation activities around the Algeciras bay, and it is affected by dense maritime traffic through the Strait of Gibraltar.

8 The study is performed for April 2013. At the beginning and end of the month, the 9 synoptic circulation was controlled by a low pressure system located over the south of 10 the British Isles and affected Western Europe by leading to atmospheric instability over 11 the IP. This pattern is typical of transitional months such as April and November 12 (García-Valero et al., 2012; Valverde et al., 2014), which produce precipitation and low 13 temperatures because of cold and humid winds from the Atlantic Ocean. In contrast, 14 from 12-18 April there was a high pressure system crossing the Iberian Peninsula in a 15 SW-NE direction, transporting dust from the Sahara Desert and increased temperatures 16 up to 25-28°C confirmed by the mineral dust forecasts of the BSC-DREAM8b model 17 (http://www.bsc.es/earth-sciences/mineral-dust-forecast-system/bsc-dream8b-forecast).

During the latter episode, available air quality stations in the study domains observed several exceedances of the European limit values (8 exceedances of the NO<sub>2</sub> hourly limit value, 25 exceedances of the O<sub>3</sub> information threshold, and 31 exceedances of the PM<sub>10</sub> daily limit value).

#### 22 2.2 CALIOPE-AQFS

23 CALIOPE-AQFS has provided 48-h air quality forecasts for Europe and Spain since 24 October 2006 (www.bsc.es/caliope) and has been described and evaluated in detail 25 elsewhere (Baldasano et al., 2008, 2011; Pay et al., 2011, 2012a). Briefly, it integrates a 26 meteorological model (WRF-ARW v3.5; Skamarock and Klemp, 2008), an emission 27 model (HERMESv2; Guevara et al., 2013), a chemical transport model (CMAQv5.0.1; 28 Byun and Schere, 2006; Appel et al., 2013), and a mineral dust atmospheric model 29 (BSC-DREAM8bv2; Pérez et al., 2006; Basart et al., 2012); together, all of these 30 comprise an air quality forecast system.

Figure 1 shows the working domains of the CALIOPE-AQFS. First, CALIOPE-AQFS
was run over Europe at 12-km x 12-km horizontal resolution using initial/boundary

1 conditions from the Final Analyses of the National Centers of Environmental Prediction 2 (FNL/NCEP). The analyses began at 12 UTC, at intervals of 6 h (0.5°x0.5°) for 3 meteorology. The global model LMDz-INCA2 (3.75°x2.5°, Szopa et al., 2009) was used 4 for chemistry. Then, CALIOPE-AQFS was run at 4-km x 4-km horizontal resolution 5 (IP4) over the IP using one-way nesting. In the present work CALIOPE-AQFS runs at 1 6 km x 1 km over the AND, BCN and MAD domains, with a nesting of over IP4 domain. 7 HERMESv2.0 forecasts anthropogenic emissions for the year 2009 by following a 8 bottom-up methodology (point, linear and area), and biogenic emissions using the 9 MEGANv2.0.4 model (Guenther et al., 2006). Emissions are aggregated into 1-km grids 10 for AND, BCN and MAD 1-km simulations, and into 4 km grid for IP4.

Vertically, WRF-ARW is configured with 38 sigma layers up to 50 hPa, with 11 characterizing the planetary boundary layer (PBL). Meanwhile, the CMAQ vertical levels are obtained by collapsing from the 38 WRF levels to a total of 15 layers whose depth increases with height from the surface up to 50 hPa. Six layers are within the PBL, and the first layer depth is 39 m.

The present WRF setup uses the Rapid Radiation Transfer Model (RRTM) and Dudhia for long- and short-wave radiation, respectively, the Kain-Fritsch cumulus parameterization (Kain and Fritsch, 1990), the single-moment 3-class (WSM3) microphysics scheme; and the Yonsei University PBL scheme (YSU). The Noah landsurface model (NoahLSM), based on the U.S. Geological Survey's (USGS) land-use data from the year 1993, is used by default in the present WRF configuration.

Currently, a new CMAQ version is being tested in the CMAS community, namely 22 23 CMAQv5.0 (CMAQ, 2012). It includes substantial scientific improvements over 24 Version 4.5 and is especially devoted to improving SOA formation as well as the 25 dynamic interactions of fine and coarse aerosols. Based on the evaluation results from 26 the previous CMAQ version within CALIOPE-AQFS (4.5 vs. 5.0) (Pay et al. 2012b), 27 CMAQ has been updated to Version 5.0.1 using the CB05 chemical mechanism 28 (Yarwood et al., 2005), the AERO5 for aerosol modeling, and the in-line photolysis 29 calculation. Although not used here, future simulations will use the updated version of 30 the aerosol module (AERO6) which includes significant improvement about the science 31 on the primary organic aerosol aging and the sulfur chemistry.

1 CALIOPE-AQFS considers desert dust contribution by means of the BSC-2 DREAM8bv2, which runs off-line at a  $0.5^{\circ} \times 0.5^{\circ}$  resolution covering Europe, North 3 Africa and the Middle East. Its outputs are mass conservative interpolated to the 4 CMAQ's Lambert conformal conic grids at the required resolution and domain. After 5 interpolating, the modeled PM<sub>10</sub> concentration is calculated as the sum of Aitken, 6 accumulation and coarse-mode modes from CMAQ, and the corresponding BSC-7 DREAM8bv2 bins with a diameter of  $\leq 10 \ \mu m$  (Pay et al., 2012a).

#### 8 2.3 Computational strategy

9 Running CALIOPE-AQFS at 4 and 1 km is a technical challenge. The simulations are 10 run on MareNostrum supercomputer (Intel Xeon E5-2670, 16 CPUs and 64 GB RAM 11 memory per node) at BSC-CNS. Table 1 depicts the computational requirements for 12 forecasting air quality at 48 h for each domain. The number of CPUs was chosen to 13 maximize CPU efficiency. Thanks to the parallelization of meteorological and air 14 quality models, MareNostrum uses up to 256 CPUs. Due to the variable nature and 15 complex dependencies, the computational time for forecasting 48 h of air quality fields 16 for the 4 domains is 8-9 hours. The most computationally demanding domain is the 17 AND, at 1-km resolution (366x358 cells, 256 CPU max., and 300 min). For the April 18 2013 simulation, times add up to 2880 CPU hours/day, or 86400 CPU hours on a single 19 processor (9.86 years). The storage for the April 2013 output files was 6.13 TB (~200 20 GB/day).

#### 21 **2.4** Evaluating the increase in resolution

22 Comparing CALIOPE-AQFS model outputs and measurements was done for gas-phase 23 and aerosol concentrations ( $O_3$ ,  $NO_2$ , and  $PM_{10}$ ). Representativeness continues to be a 24 challenge when comparing gridded simulations to observational data at a point in time 25 and space, as modeled concentrations represent a volumetric average over an entire grid 26 cell. Furthermore, the stochastic compound embedded in the observations is not 27 accounted for. Concerning temporal representativeness in the present comparison, both 28 modeled and measured concentrations are averaged hourly. CALIOPE-AOFS 29 operationally receives air quality measurements from Spanish administrative networks 30 in near real time (NRT) without any quality control. For the present study, NRT 31 measurements are filtered by removing data before and after measurement interruptions or calibrations. Also, a minimum cut-off threshold of 1  $\mu$ gm<sup>-3</sup> is applied to the observed concentrations in order to avoid unrealistic observations. After filtering, the number of stations is 48/30/36 for O<sub>3</sub>, 51/42/42 for NO<sub>2</sub>, and 52/15/33 for PM<sub>10</sub> in the AND/BCN/MAD domains, respectively.

5 The meteorological fields are evaluated for wind speed (U10) and wind direction 6 (WD10) at 10 m, and temperature at 2 m (T2M). The 10 METAR stations are all 7 located at airports (6/2/2 stations in AND/BCN/MAD) and are discussed in Sect. S1.

Figure 2 shows the location of the air quality and METAR (METeorological Aerodrome Report) stations over the respective domains. The spatial representativeness of the air quality network is highly variable. The influence of the station type is based on two classifications of air quality monitoring stations, the environment type (rural, R; suburban, S; and urban, U), and the dominant emission source (traffic, T; industrial, I; and background, B). These were derived from the Council decision 97/100/EC (Garber et al. 2002).

- The evaluation is based on discrete statistics performed on an hourly basis. We consider the correlation coefficient (r, Eq. A1), mean (absolute, relative, and fractional) biases (MB, MNBE, and MFB, Eq. A2-A4), and error (MAE, MNGE, and MFE, Eq. A5-A7). Root Mean Square Error (RMSE, Eq. A8) is also calculated because it intensifies large differences between measured and observed concentrations (Table A1).
- 20 In order to evaluate the effect of increased resolution on forecast exceedances and non-21 exceedances of limit values established by the European legislation, we calculate 22 categorical statistics based on comparisons with fixed concentration thresholds (T). The 23 calculated statistics are accuracy (A, Eq. A9), bias (B, Eq. A10), probability of detection 24 (POD, Eq. A12), critical success index (CSI, Eq. A11), and false alarm ratio (FAR, Eq. 25 A13), whose formulas and descriptions are provided in Table A2 and also in Kang et al. (2005) and Eder et al. (2006). The 2008/50/EC directive sets an information threshold 26 of 180 µgm<sup>-3</sup> for maximum daily O<sub>3</sub> concentrations (Max 1h O<sub>3</sub>) and a target value of 27 120  $\mu$ gm<sup>-3</sup> for the maximum daily 8-h running O<sub>3</sub> mean (Max 8h O<sub>3</sub>), which should not 28 be exceeded more than 25 days per year. It establishes a limit value of 200 µgm<sup>-3</sup> for 29 maximum daily NO<sub>2</sub> concentrations (Max 1h NO<sub>2</sub>), and 50  $\mu$ gm<sup>-3</sup> for the daily PM<sub>10</sub> 30 mean (Mean 24h  $PM_{10}$ ), which should not be exceeded more than 35 times per year. 31 32 Therefore, categorical evaluation will be performed for Max 1h NO<sub>2</sub>, Max 1h and Max

8h O<sub>3</sub>, and Mean 24h PM<sub>10</sub>. Note that mean and maximum concentrations are calculated
by considering at least 75% of the data in the corresponding time base, i.e., values of at
least 18 hours per day for Mean 24 h, Max 1h, and Max 8h; and 6 hours for 8 h values,
as established by 2008/50/EC.

# 5 **3** Concentration maps and spatial representativeness

6 To analyze the spatial differences between resolutions, Figs. 3, 4, and 5 show the 7 monthly mean concentration maps for April 2013 over MAD, BCN and AND domains 8 at 4 km (left panels) and 1 km (right panels) for NO<sub>2</sub>, O<sub>3</sub>, and PM<sub>10</sub>, respectively.

9 The maps of  $NO_2$  and  $PM_{10}$  at both resolutions display similar distribution along the 10 MAD and BCN urban plumes. On-road traffic constitutes the main source of primary 11 pollutants in MAD and BCN. HERMESv2.0 estimates that 75% and 59% of NOx 12 emissions are produced by on-road traffic in both domains, respectively. Consequently, 13 when the resolution increases, the monthly mean  $O_3$  concentration maps are almost 14 identical, although the NO<sub>x</sub> titration effect on O<sub>3</sub> is significant along highways and 15 major point sources. In AND, NO<sub>2</sub> and O<sub>3</sub> concentrations are also conserved between 16 resolutions along the shipping route crossing the Strait of Gibraltar towards the 17 Mediterranean Sea.

18 However, the definition of NO<sub>2</sub> concentrations along highways connecting the biggest 19 cities with the rest of the country and industrial sectors are more easily identified at 1-20 km simulations than at 4 km, especially along those roads from/to Barcelona (e.g., the 21 AP7 Mediterranean highway and C32, which connects the harbor and the airport) and 22 Madrid (the A-2 and A-6 in the north, and A-3, A-4 and A-5 in the south). In the same 23 way, 1-km  $O_3$  maps are more textured than those at 4 km along highways, because the 24 titration effect is more significant at 1 km, due to less dilution within grid cells. The 25 titration effect of NO<sub>x</sub> on O<sub>3</sub> over the main sources is more forceful in BCN than in 26 MAD, given that BCN has a larger concentration gradient resulting from complex 27 topography and recirculation flows that accumulate pollutants.

The improvement of the definition along roads in AND is lower than that observed in the MAD and BCN domains, due to the fact that the AND domain is bigger and displays lower traffic emission sources than the MAD or BCN domains. Regarding  $PM_{10}$ , the main component in AND is the desert dust (~40% in both resolutions) from North Africa. This is because there were two episodes on 14-19th and 25-26th April that affected the IP, as shown by the S-N PM<sub>10</sub> gradient (Fig.5e and f). The desert dust is
 transported from long-range simulation with BSC-DREAM8bv2.

3 Over complex terrains, the 1-km simulation produces more realistic looking NO<sub>2</sub> 4 concentration maps because of its more detailed topographic information. For instance, the BCN 1-km simulation displays the lowest NO<sub>2</sub> concentrations ( $< 10 \ \mu gm^{-3}$ ) along 5 6 the coastal chain (500 m height) and pre-coastal chain (1000-1700 m height), except for the city's urban hill, where concentrations reach 20-40  $\mu$ gm<sup>-3</sup>. In contrast, the 4-km 7 8 simulation provides smoother NO<sub>2</sub> concentrations without any concentration gradient. 9 Thus, the 1-km simulation generates slightly higher O<sub>3</sub> background concentrations than the 4-km simulation along the BCN pre-coastal chain (66-70 vs. 70-74 µgm<sup>-3</sup>), as well 10 11 as across the Iberian Massif (AND), where the O<sub>3</sub> displays significant structure due to 12 the higher resolution topography that follows the basin and the areas of high-density on-13 road traffic.

14 Figures 3, 4, and 5 include dots corresponding to mean concentrations at air quality 15 stations that help to qualitatively evaluate the modeled spatial representativeness at both 16 resolutions. Note the strong correlation between NO<sub>2</sub> observations and the 1-km 17 simulation near the primary suburban traffic roads in BCN (e.g., Vilafranca, Igualada, 18 Manresa, and Mataró). Regarding O<sub>3</sub>, although observed concentrations depict an 19 overall tendency of the model to underestimate concentrations at both resolutions, the 1-20 km simulation shows better agreement with measurements at rural background stations 21 (e.g., El Atazar, San Martín, Villa de Prado, Villarejo and Orosco stations in MAD), and 22 at suburban traffic stations (e.g., Manresa, Igualada and Vilafranca in BCN, with modeled  $O_3$  concentrations of around 54-58 µgm<sup>-3</sup> at 1 km, and 60-66 µgm<sup>-3</sup> at 4 km). 23 24 For PM<sub>10</sub>, comparisons with measurements show that modeled concentrations are underestimated over background areas, mainly outside the urban/suburban areas, as 25 26 already discussed in Pay et al. (2012a). However,  $PM_{10}$  measurements at the urban/suburban stations of Vilafranca, Sant Celoni, and Mataró in BCN (14-16 µgm<sup>-3</sup>) 27 show a higher correlation at 1 km than at 4 km (12-14  $\mu$ gm<sup>-3</sup> vs. 8-10  $\mu$ gm<sup>-3</sup>). 28

The spatial variability of the increased resolution is quantitatively analyzed by means of concentration maps, shown in Fig. 6 for NO<sub>2</sub>, O<sub>3</sub> and PM<sub>10</sub> over AND, BCN and MAD. Over all domains, the correlation improves as a function of the resolution increase for NO<sub>2</sub> and O<sub>3</sub>, indicated by the increase in monthly r from 0.79 (4 km) to 0.81 (1 km) for

1  $NO_2$ , and from 0.69 to 0.73 for  $O_3$ . The slopes improve with the resolution increase, 2 from 0.72 (4 km) to 0.77 (1 km) for NO<sub>2</sub>, and from 0.50 (4 km) to 0.54 (1 km) for O<sub>3</sub>. 3 This results from the improved model performance at urban stations, indicating that 4 CALIOPE-AQFS better captures the magnitude of the variability between urban regions 5 at 1 km than at 4 km. In contrast, for  $PM_{10}$ , the monthly r decreases from 0.67 to 0.58 6 when the resolution increases. While the PM<sub>10</sub> spatial variability over BCN and MAD 7 improves when the resolution increases (r increases by 0.01 and 0.04, respectively), the 8 global correlation coefficient decreases at the AND stations (52/100) by 0.1 from 0.36 9 (4 km) to 0.26 (1 km). Despite the unfavorable effect of the resolution increase in  $PM_{10}$ 10 over AND, the NO<sub>2</sub> and O<sub>3</sub> concentrations show the highest absolute increase for the 11 spatial r over this domain, from 0.62 (4 km) to 0.71 (1 km) for NO<sub>2</sub> and from 0.58 (4 km) to 0.64 (1 km) for  $O_3$  (increasing r by 0.09 and 0.06, respectively). 12

13 4 Temporal evaluation

#### 14 4.1 Pollutant

15 Table 2 presents the statistical evaluation by pollutant, with a focus on the reproduction 16 of high concentrations established by the European directive (2008/50/EC). Depending 17 on the pollutant's lifetime and variability, as well as its dependency on precursors, 18 increased resolution shows different impacts. The resolution increase has a positive effect on NO<sub>2</sub>, decreasing its bias by 2.0  $\mu$ gm<sup>-3</sup> (from -4.5 to -2.5  $\mu$ gm<sup>-3</sup>) but it also 19 increases the absolute (squared) error by 0.3  $\mu$ gm<sup>-3</sup> to 0.9  $\mu$ gm<sup>-3</sup>. This positive effect is 20 21 sustained by the relative variability, where the MB (MFB) is reduced by 42% (19%) and 22 MAE (MFE) only increases by 2% (1%). The r does not between resolution (r=0.54) 23 partially due to the fact that the NO<sub>2</sub> is a primary pollutant and emissions at both 24 resolutions are modeled using the same approach. The bias improvement at 1-km 25 resolution is justified, because in theory the higher resolution leads to better emission 26 allocation from point, linear or area sources, decreases the artificial dilution of 27 emissions compared to the larger grid area and, due to the decrease of artificial dilution, 28 it treats chemistry more properly near large emission sources.

In contrast, the resolution increase has a negative effect on hourly and Max 8h  $O_3$ concentrations increasing the bias and error by 0.1-0.8 µgm<sup>-3</sup>. Relative (fractional) bias and error increase by 8% (15%) and 1% (1%), respectively, for hourly  $O_3$ ; and 6% and 4% for Max 1h  $O_3$ . According to the categorical evaluation, only a few exceedances of the European target and limit values were detected for Max 1h NO<sub>2</sub> (9), Max 1h O<sub>3</sub> (25), and Mean 24 PM<sub>10</sub> (31) in April 2013. Thus, categorical evaluation is performed on the temporal basis established by the European legislation, but it uses a T based on the 75<sup>th</sup> percentile (75p) of the observed concentrations in each case, where T is 71  $\mu$ gm<sup>-3</sup> for Max 1h NO<sub>2</sub>, 108 (101)  $\mu$ gm<sup>-3</sup> for Max 1h (Max 8h) O<sub>3</sub>, and 27  $\mu$ gm<sup>-3</sup> for Mean 24h PM<sub>10</sub>.

7 Overall, CALIOPE-AQFS underestimates exceedances at both resolutions, indicating 8 that errors of missing observed exceedances are not completely resolved by an 9 increasing the horizontal resolution (a<d). The best performance is found for Max 1h 10 NO<sub>2</sub>, where categorical bias (B, Eq. A10) improves from 37% (4 km) to 40% (1 km), 11 although the tendency to underestimating exceedances remain with the resolution 12 increase (B < 100%).

13 For NO<sub>2</sub> Max 1h, there are 953 observed exceedances (b+d) of the threshold (T=47 14  $\mu$ gm<sup>-3</sup>). Increasing the resolution increases the POD from 49% (4 km) to 56% (1 km). 15 As with POD, CSI examines the exceedances, but in a more comprehensive way by 16 considering both false alarms and missing events. POD and CSI increase by 14% and 17 20%, respectively, when resolution is increased form 4 km to 1 km. For  $O_3$ , POD has a relatively low POD value, and it decreases as the resolution increases. Of the 1306 18 observed exceedances of the 108 µgm<sup>-3</sup> Max 1h, CALIOPE-AQFS detected 112 19 exceedances at 4 km and only 96 at 1 km. Increasing resolution decreases POD and CSI 20 21 by 22% and 25% for O<sub>3</sub> Max 1h, whereas they do not significantly change for Max 8h 22 O<sub>3</sub> and Mean 24h PM<sub>10</sub>.

FAR increases for Max 1h NO<sub>2</sub> (from 40% to 42%) and decreases for Max 1h O<sub>3</sub> (from 27% to 17%) when the resolution increases. In relative terms, this variability is more 25 significant for Max 1h O<sub>3</sub> (37%) than for Max 1h NO<sub>2</sub> (5%), indicating that, in terms of 26 failures, the resolution has a positive global effect by reducing false exceedances.

For various reasons, accuracy (A) remains almost constant when the resolution increases. For NO<sub>2</sub> and O<sub>3</sub>, this is due to a stable sum of b and c, increasing the b at the cost of c, and vice versa. For NO<sub>2</sub>, the number of hits (b) to forecast Max 1h at 1 km is higher than 4 km (537 vs. 466), but the number of correct negatives at 1 km is lower than at 4 km (2439 vs. 2517). The resolution increase has the opposite effect on O<sub>3</sub>, decreasing the number of hits by 14% for Max 1h and 33% for Max 8h. The increase of
the number of correct negatives is less than 2% in both Max 1h and Max 8h O<sub>3</sub>.

#### 3 4.2 PM<sub>10</sub> components

The resolution increase has the lowest effect on  $PM_{10}$  hourly concentrations and its exceedances (<1%).  $PM_{10}$  components are secondary inorganic aerosols (SIA), which include sulfate (SO4 =  $SO_4^{2^-}$ ), nitrate (NO3 =  $NO_3^-$ ), ammonium (NH4 =  $NH_4^+$ ), secondary organic aerosol (SOA), elemental carbon (EC), sea salt (SS), desert dust (DD), and primary PM (PPM).

9 Pay et al. (2012a) evaluated the PM components at some urban and rural background 10 stations in Spain using the CALIOPE-AQFS based on CMAQv4.5, where they showed 11 that the model underestimated SIA by a factor of 2-3. The highest underestimation was 12 found for fine carbonaceous aerosols (a factor of 4), in part related to the state-of-the-13 science concerning SOA formation pathways. The updated version of CMAQ, v5.0.1 14 includes scientific improvements to SOA formation and aerosol dynamics, which could 15 improve the modeled  $PM_{10}$  performance and its components.

Figure 8a shows that the resolution increase does not significantly change the  $PM_{10}$ composition. DD remains the main contributor (~40-41%), followed by PPM (22-24%), SIA (~21-22%), SS (9-11%), EC (~4%) and SOA (~0.6%). However, the effect of the increased resolution on  $PM_{10}$  component concentrations is different (Fig. 8b), depending on their origin, atmospheric cycle and the way they are modeled. DD concentrations do not change between resolutions, because they are mass conservative when interpolated from  $0.5^{\circ}x0.5^{\circ}$  till 1 km x 1 km.

23 For SIA, increasing the resolution increases NO3 and NH4 concentrations by 4 and 24 ~2%, respectively, and it decreases SO4 by ~2%. The NH4 increase means there are 25 more primary precursors ( $H_2SO_4$  or  $HNO_3/NO_2$ ) available to neutralize  $NH_3$  (gas) to 26 NH4 (aerosol). However, the variability between SO4 and NO3 is more difficult to 27 explain, due to the nonlinearity of photochemistry and aerosol formation, which is controlled to some extent by the ISORROPIA thermodynamic equilibrium. 28 29 Furthermore, the absence of aerosol measurements for April 2013 does not allow us to 30 investigate this further.

The resolution increase results in a decrease in SS of ~16%, the largest change of all the PM<sub>10</sub> components. CMAQv5.0.1 simulates SS emission as a function of the wind speed and the relative humidity (Gong, 2003; Zhang et al., 2005). Although not shown here, when the resolution increases, the wind speed decreases at the available PM<sub>10</sub> stations by ~1.4/0.4/0.2 ms<sup>-1</sup> over AND/BCN/ MAD, and also over the open ocean.

For primary PM components EC and PPM, increasing resolution results in higher
concentrations (by 10 and ~12%, respectively). As for NO<sub>2</sub>, the 1-km simulation leads
to a reduced effect of artificial dilution of emissions in a grid cell, so concentration
gradients are stronger than in the 4-km simulation.

#### 10 4.3 Domain

11 The resolution increase has varying impacts due to differences in geographical location 12 and emission patterns over the domains (Fig. 7). BCN shows the highest  $NO_2$  bias 13 decrease (73%) when the resolution increases, but no effect on the correlation (<7%). 14 However, O<sub>3</sub> shows significant variability over BCN, increasing r by 4% and MB by 15 23%. To a lesser extent, MB also increases over AND (by 8%). Meanwhile, the variability over MAD is reduced (bias differences <4%). MB decreases for PM<sub>10</sub> (bias 16 differences  $<1 \ \mu gm^{-3}$ ) over the urban domains of MAD (3%) and BCN (16%), and 17 18 increases over AND (7%).

Figure 9 analyzes the impact of the resolution increase on daily cycles. Although PBL measurements are not available, PBL daily cycles are displayed together in order to find some correlations with the daily pollutant variability. Due to the lamination of PBL growth by the Mediterranean sea breezes, the PBL reaches its maximum height at midday, being the highest in MAD (1600 m AGL) followed by AND (1000 m AGL) and BCN (900 m AGL).

As shown in Sect. S1, the pollutant transport at the BCN coastal domain is controlled by mesoscale phenomena such as sea-breezes (day) and land-breezes (night), which are a result of its complex topography and location (Baldasano et al., 1994; Millán et al., 1997; Gonçalves et al., 2009). The NO<sub>2</sub> daily cycle is highly influenced by traffic emissions (Fig. 9). Both resolutions show the highest underestimations for the morning peak (6 am) (~20  $\mu$ gm<sup>-3</sup>). Although the afternoon peak is well reproduced, there is excessive variability at both resolutions, a result of problems with wind direction. 1 During the sea breeze period, the mean simulated wind was more easterly than westerly, 2 as registered by measurements (Sect. S1). Several works indicate that WRF has 3 difficulty reproducing the morning and evening transition over the urban environment, possibly because it does not model the heat retention in cities (Makar et al., 2006; Appel 4 5 et al., 2013). Increasing the resolution increases  $NO_2$  concentrations by 21% from 14 6  $\mu$ gm<sup>-3</sup> (4 km) to 17  $\mu$ gm<sup>-3</sup> (1 km) during the morning hours after sunrise (5-9 am) and in the evening hours after sunset (5-9 pm). This behavior could be explained by PBL 7 8 variability when increasing the resolution, which decreases PBL height by ~33 m for 9 these hours.

10 NO<sub>2</sub> performance impacts the O<sub>3</sub> daily cycles over BCN, showing that 4- and 1-km simulations underestimate maximum  $O_3$  concentrations by ~20 µgm<sup>-3</sup> at midday (1-4 11 pm) and overestimate minimum  $O_3$  concentrations by ~20  $\mu$ gm<sup>-3</sup> in the morning hours 12 after sunrise (5-9 am). The resolution increase results in slightly lower  $O_3$ 13 14 concentrations at night, which is perhaps the result of lower PBL heights in the 1-km 15 simulation during the early morning and late afternoon, when PBL height tends to be the lowest. During these hours, the titration effect of NO<sub>2</sub> on O<sub>3</sub> is greater, improving 16 the O<sub>3</sub> overestimation of the daily minimum, which allows a slightly higher hourly r 17 (2%). However, O<sub>3</sub> underestimation increases in the late afternoon, contributing to an 18 increase in the hourly mean bias from ~9  $\mu$ gm<sup>-3</sup> (4 km) to ~11  $\mu$ gm<sup>-3</sup> (1 km). 19

In BCN, the  $PM_{10}$  underestimation is not systematic throughout the daily cycle (Fig. 9), 20 which shows a bias of  $\sim 20$  (10)  $\mu gm^{-3}$  at day (night) time. The higher daytime 21 22 underestimation as compared to the nighttime cannot be explained by the current 23 results, but it could be a result of missing sources (e.g., fugitive agricultural emissions 24 and wind-blown dust) and problems with PBL height overestimation and an excess dilution of emissions. The resolution increase reduces the bias by  $\sim 1 \ \mu gm^{-3}$  (16%), 25 26 especially during early morning and late afternoon, when the highest PBL height 27 variability between resolutions is observed. Although the evaluation of T2M, U10 and 28 WD10 indicates that the resolution increase has a small effect over BCN (Sect. S1), the 29 reduction of the artificial dilution of NO<sub>2</sub> emissions -together with a lower PBL height 30 at 1 km than at 4 km during the night and early morning– improves NO<sub>2</sub>, O<sub>3</sub> and PM<sub>10</sub> 31 concentrations, which in turn decreases their biases.

1 For AND, the model underestimates observed NO<sub>2</sub> concentrations throughout the daily cycle by  $\sim 5 \ \mu \text{gm}^{-3}$  for both the 4-km and 1-km simulations, with the highest 2 underestimation during the morning peak (~25  $\mu$ gm<sup>-3</sup>) and the lowest during the 3 afternoon peak (~10  $\mu$ gm<sup>-3</sup>). The resolution increase reduces the bias from -3.5 to -2 4  $\mu$ gm<sup>-3</sup> (by 43%) and increases r by 7% (from 0.39 to 0.41). As in BCN, the NO<sub>2</sub> 5 underestimation directly impacts the O<sub>3</sub> daily cycle (which the resolution increase 6 cannot resolve), increasing the bias by  $\sim 1 \ \mu \text{gm}^{-3}$ , a phenomenon that is predominant in 7 the morning hours. In the case of PM<sub>10</sub>, the daily cycle indicates that the biases are 8 almost systematic throughout the day ( $\sim 22 \ \mu gm^{-3}$ ). Increasing the resolution increases 9 10 the bias by less than 4% in the late afternoon, which is perhaps dominated by the 11 decrease in PBL height. When the resolution is increased, NO<sub>2</sub> performs better partially due to the improved model performance for the temperature and wind speed (Sect. S1), 12 13 as well as the lower nighttime and higher daytime PBL. Meanwhile, the  $O_3$  and  $PM_{10}$ 14 performance do not significantly change.

15 During April 2013, the main flow over MAD was controlled by S-SW synoptic winds 16 channeled by orographic barriers in the NW domain and the Tajo valley (Valverde et 17 al., 2014). The NO<sub>2</sub> daily cycle depicts a high influence of traffic emissions (Fig. 9), 18 significant model underestimation showing at both resolutions for the morning/afternoon peaks ( $\sim 15/10 \ \mu gm^{-3}$ ). Note that, in terms of temporal variability of 19 20 the NO<sub>2</sub> concentrations, the model performs well at the afternoon peak when the air 21 flow is controlled by southeastern winds. The NO<sub>2</sub> performance leads to more accurate 22 O<sub>3</sub> daily cycles than in AND and BCN, especially in the early morning, when the 23 titration effect of NO<sub>2</sub> is greater because the NO<sub>2</sub> morning peak underestimation is smaller when compared to the other domains. Meanwhile, the modeled PM<sub>10</sub> at both 24 25 resolutions presents a profile controlled by traffic emissions. Observed concentrations display a flatter daily cycle, in which the model underestimation reaches 40 µgm<sup>-3</sup> in the 26 morning. Increasing resolution shows a positive effect for NO<sub>2</sub> and PM<sub>10</sub> because it 27 increases r by 0.01 and reduces MB and RMSE by 0.1-0.2 µgm<sup>-3</sup>. However, it has the 28 29 lowest variability when compared to the other domains (<5 % for bias, error and r), 30 likely the result of a relatively simpler topography and meteorological patterns.

#### 1 4.4 Environment and major sources

Figure 7 shows that the resolution impact also depends on the type of area and the dominant emission source. Theoretically, the meteorological fields of urban areas differ from those of surrounding rural areas because of their different morphology (radiation trapping and wind profiles), surface materials (heat storage) and variable energy consumption (heat release).

7 Increasing resolution reduces the NO<sub>2</sub> bias at suburban and urban stations by  $1.8-2 \,\mu \text{gm}^{-1}$  $^{3}$ , and to a lesser extent by 1.2  $\mu$ gm<sup>-3</sup> at rural stations. The correlation coefficients also 8 9 improve at suburban stations (from 0.48 to 0.52) and rural stations (from 0.34 to 0.35). 10 This may be due to better allocation of land-use categories (urban vs. rural) and of their 11 fraction in a grid cell in the 1-km simulation than in 4-km simulation. The NO<sub>2</sub> bias 12 decreases by 39% (65%) at urban (background) stations, but O<sub>3</sub> biases increase by 9% 13 (5%). For  $PM_{10}$ , the resolution increase does not significantly change as a function of area type, with differences in bias and error of less than  $\pm 4\%$  (<0.5 µgm<sup>-3</sup>) between the 14 15 two resolutions.

16 The low improvement at urban stations may be because the NoahLSM land-surface 17 model does not consider the effect of urban morphology or thermal parameters in order 18 to accurately model meteorological fields. Modeling air quality on an urban scale over 19 cities requires a description of the heat/momentum exchange between buildings and the 20 lower atmospheric layers. For instance, the impact of using an urban model on 21 meteorological fields over the greater Paris area was studied by Kim et al. (2013) using 22 WRF with the Urban Canopy Model, demonstrating that below 1 km height 23 overestimations of wind speed were significantly reduced.

24 The impact on r of increasing resolution is less than 2% for primary pollutants near 25 important emission sources. For example, it reduces NO<sub>2</sub> biases at traffic (industrial) stations by ~3  $\mu$ gm<sup>-3</sup> (2  $\mu$ gm<sup>-3</sup>), but it increases O<sub>3</sub> biases by ~2  $\mu$ gm<sup>-3</sup> (1  $\mu$ gm<sup>-3</sup>). 26 27 However, the resolution increase in the range of 4-1 km does not exhibit the expected 28 improvement on the hourly statistics that are based on the constraints of the current 29 model formulation. In other words, it cannot resolve the subgrid air quality variability 30 merely by increasing resolution. For instance, although on-road traffic emissions are 31 estimated by following a bottom-up approach along highways and routes, heterogeneity 32 is lost in the CTM volume averaging process, which artificially dilutes emission rates over the grid cells. The resolution effect is low at background stations, which genearally are not influenced by any single source, but rather by the integrated contribution from all sources upwind of the stations where variations are less than 1% for  $O_3$  and  $PM_{10}$  (<  $1 \mu gm^{-3}$ ). However, background NO<sub>2</sub> levels increase by ~1  $\mu gm^{-3}$  (48%) from 4 km to 1 km.

Figure 10 shows the temporal series and daily cycles for NO<sub>2</sub> and O<sub>3</sub> at traffic and background stations throughout the episode of 12-18<sup>th</sup> April, 2013. At traffic stations, the temporal series show a remarkable O<sub>3</sub> daily cycle (observed  $25p = 23.2 \ \mu gm^{-3}$  and  $75p = 77.5 \ \mu gm^{-3}$ ), due to O<sub>3</sub> destruction caused by high NO<sub>x</sub> levels (observed 50p = $34.5 \ \mu gm^{-3}$ ). In contrast, the NO<sub>2</sub>–limited regime at background sites (observed 50p = $19 \ \mu gm^{-3}$ ) allows for higher O<sub>3</sub> concentrations (observed  $25p = 38 \ \mu gm^{-3}$  and 75p = 89 $\mu gm^{-3}$ ) than in high NO<sub>2</sub> environments.

13 During the episode mentioned above, the resolution increase at traffic stations had a positive effect by increasing the r for NO<sub>2</sub> (from 0.73 to 0.76) and O<sub>3</sub> (from 0.83 to 14 0.86), and also decreasing the NO<sub>2</sub> mean bias by ~5  $\mu$ gm<sup>-3</sup> (from 6 to 1  $\mu$ gm<sup>-3</sup>). The 15 NO<sub>2</sub> daily cycle improves in the morning hours after sunrise, reducing bias by 5-10 16  $\mu$ gm<sup>-3</sup> and contributing to a reduction in O<sub>3</sub> overestimations (~5  $\mu$ gm<sup>-3</sup>). In contrast, at 17 18 background stations, where the  $NO_x/O_3$  chemistry is less dominant, the resolution effect 19 is not significant. Such behavior indicates that finer resolution improves the model 20 performance, because horizontal resolution affects the representation of chemical 21 processes near large emission sources, such as the formation of O<sub>3</sub> and nighttime O<sub>3</sub> 22 titration (Mathur et al., 2005).

# 23 **5 Conclusions**

The present work shows the effects of increasing the horizontal resolution from 4 km to 1 km using the CALIOPE-AQFS on pollutant concentrations (NO<sub>2</sub>, O<sub>3</sub>, and PM<sub>10</sub>) over three Spanish domains (AND, BCN and MAD) in April 2013.

The global features of concentration maps at both resolutions are quite similar, with zones of high/low concentration identically located, which is expected since both simulations are based on the same emission dataset. Further comparisons demonstrate that increasing the resolution provides better-defined and more realistic concentration structures over large emission sources (roads and industries) and complex terrains (more sharply defined orographic hills). The titration effect on  $O_3$  by  $NO_x$  along highways and major point sources is more evident in 1-km simulations than at 4 km, since NOx
concentrations tend to be higher in the 1-km simulation due to less dilution. This
improvement is quantified by an increase in spatial correlation coefficients of 3% (6%)
for NO<sub>2</sub> (O<sub>3</sub>).

5 However, the resolution increase in the range of 4-1 km does not exhibit the expected improvement in hourly statistics for any pollutant. Hourly correlation coefficients do 6 not significantly change, and absolute (relative) errors and biases vary  $< 2 \mu \text{gm}^{-3}$  (9%). 7 8 The merit of the resolution increase may be underrated when classical statistics are 9 applied at measurement stations (Mass et al., 2002; Gego et al., 2005). For instance, although the structure of important NO<sub>2</sub> urban plume features (> 40  $\mu$ gm<sup>-3</sup>) often 10 11 become more realistic (stronger and more defined plumes) as resolution increases, 12 statistics are deeply degraded by even small timing and spatial errors.

13 The resolution increase has a significant impact on reducing NO<sub>2</sub> hourly bias (by 42%, 14  $2 \mu \text{gm}^{-3}$ ), without any significant change in the error and the r (<2%). However, O<sub>3</sub> hourly biases increased by  $<1 \ \mu gm^{-3}$ . The main differences between resolutions appear 15 at daytime and nighttime traffic peaks, when the mixing height experiences rapid 16 17 changes, allowing the 1-km simulation to slightly reduce NO<sub>2</sub> underestimation in the morning by ~5-10  $\mu$ gm<sup>-3</sup>. The O<sub>3</sub> daily cycles at large emission sources depict a high 18 influence of hourly NO<sub>2</sub> concentrations, increasing the hourly O<sub>3</sub> bias by  $\sim$ 3 µgm<sup>-3</sup>. This 19 20 behavior is controlled by the daytime  $O_3$  underestimation and, to a lesser extent, by the 21 nighttime overestimation. The resolution increase reduces the O<sub>3</sub> overestimations at night by ~5  $\mu$ gm<sup>-3</sup> partly because of higher nocturnal NO<sub>2</sub> concentrations. 22

Increasing the horizontal resolution improves the model's ability to forecast 75p exceedances in the observed maximum 1h concentrations. The number of hits that forecast 75p exceedances in the observed Max 1h NO<sub>2</sub> increases from 466 to 537 over 953 exceedances, while the FAR for Max 1h O3 exceedances is reduced by 37%.

The grid effect is less pronounced for  $PM_{10}$  than for NO<sub>2</sub> and O<sub>3</sub>. When the resolution increases, the low gradient of  $PM_{10}$  mean concentrations (<0.1 µgm<sup>-3</sup>) is the result of compensating biases of  $PM_{10}$  components, which is controlled mainly by the PPM and EC increase, as well as the SS decrease.

BCN is the domain where the resolution increase has the largest effect, with changes in 1 2 bias (error) of 16-73% (< 5%), followed by AND with 4-43% (< 5%) and MAD < 3-5%3 (< 1%). In BCN, as in the western Mediterranean Basin, the transport of  $O_3$  and its 4 precursors is governed by mesoscale circulations. In that sense, the resolution increase 5 has a large impact over BCN, where induced mesoscale phenomena control the air flow. 6 Conversely, synoptic transport is more prominent in MAD and AND. Increasing the 7 resolution to 1 km over rural areas (Mass et al., 2002) could contribute to improve the 8 representation of mesoscale meteorological structures such as orographic wind and 9 circulation. Over urban areas along the western Mediterranean coast, further 10 improvements (e.g., models that consider the urban morphology and thermal 11 parameters) are required before seeing any benefits in increasing the resolution to 1 km 12 (Toll and Baldasano, 2002; Jiménez et al., 2006; Fay and Neunhäuserer, 2006).

In urban areas and near large emission sources (industrial and traffic stations), NO<sub>2</sub> and O<sub>3</sub> concentrations are more sensitive to changes in the grid resolution. The concentration increases in primary anthropogenic pollutants (NO<sub>2</sub>, PPM and EC) because the high resolution may better allocate emissions at point, linear and area sources. Increase resolution also reduces the artificial dilution of emissions when compared to the larger grid area. Despite the reasons above, moving to 1-km horizontal resolution generally did not result in better performance for O<sub>3</sub> and NO<sub>2</sub>.

20 This analysis demonstrates weaknesses in the current model formulations that cannot be 21 resolved with only high-resolution modeling. The subgrid air quality variability at 1-km 22 resolution could be not reproduced over large emission sources or urban areas, because 23 a finer spatial structure is expected but unresolved. While the work presented here 24 demonstrates the feasibility of modeling at fine-scales, there remain many challenges. 25 First, there is a loss of subgrid emission heterogeneity. Emission inputs to CTM are an 26 average rate, which accounts for the volume averaged quantity of mass released per unit 27 of time. No other information regarding emission allocation (e.g., point, linear or per 28 area) is considered; for instance a large amount of mass can be emitted by a small 29 portion of the grid surface or by several sources scattered around it (Galmarini et al., 30 2008; Cassiani et al., 2010; Ching and Majeed, 2012). Despite the fact that emissions 31 are estimated by following a bottom-up approach emission model, emission 32 heterogeneity is lost in the volume averaging process performed within CTM. The loss 33 is even higher when resolution decreases (from 1 km to 4 km). Second, there is a low

degree of complexity in flow and dispersion details at urban scales, where most of the pollutants come from street canyons and/or tree canopies, where they are transported until mixing conditions allow the pollutants to disperse above these urban canopy levels (Kim et al., 2013; Ching, 2013). Third, the USGS land-use data used in the WRF model is based on 1993 data, and urban changes in MAD and BCN over the last 20 years are significant.

7 Since temperature and wind speed are very sensitive to the ratio of building width to 8 road width, future improvement for fine-scale modeling should focus on using an urban 9 canopy model that considers effects on the transfer of energy and momentum between 10 urban structures and the lower atmosphere. This is crucial for modeling meteorology 11 and air quality at fine-scales in urban environments. However, it requires an urban 12 canopy scheme and a canopy parameter database (urban fraction, building height and 13 area). Furthermore, in order to gain any benefits from increasing resolution, the 14 meteorological modeling should be updated to include a better description of land use, 15 instead of relying on USGS data from the year 1993. To this end, the Coordination of 16 Information on the Environment (CORINE) provides a high resolution (100 m) land use 17 database, which was developed by the European Environmental Agency and updated to the year 2006 (CLC2006) (EEA, 2007). This could be implemented in the WRF model 18 19 following the methodology described in Pineda et al. (2004).

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# 3 Appendix A

4 Table A1. Definition of the discrete statistics used in the evaluation. Where  $C_m(x, t)$  and 5  $C_o(x,t)$  are the modeled and observed concentrations at a location (x) and time (t); N is 6 the number of pairs of data.  $\overline{C_m}$  and  $\overline{C_o}$  are the modeled and observed mean 7 concentrations over the whole period, respectively.

Statistic	Formula				
Correlation coefficient	$r = \frac{\sum_{i=1}^{N} (C_m(x,t) - \bar{C_m})(C_o(x,t) - \bar{C_o})}{\sqrt{\sum_{i=1}^{N} (C_m(x,t) - \bar{C_m})^2} \sqrt{\sum_{i=1}^{N} (C_o(x,t) - \bar{C_o})^2}}$	A1			
Mean bias	$MB = \frac{1}{N} \sum_{i=1}^{N} \left( C_m(x,t) - C_0(x,t) \right)$	A2			
Mean normalized bias error	$MNBE = \frac{1}{N} \sum_{i=1}^{N} \frac{(C_m(x,t) - C_0(x,t))}{C_o(x,t)} \cdot 100$	A3			
Mean fractional bias	$MFB = \frac{1}{N} \sum_{i=1}^{N} \frac{(C_m(x,t) - C_0(x,t))}{(C_o(x,t) + C_m(x,t))/2} \cdot 100$	A4			
Mean absolute error	$MAE = \frac{1}{N} \sum_{i=1}^{N}  C_m(x,t) - C_0(x,t) $	A5			
Mean normalized gross error	$MNGE = \frac{1}{N} \sum_{i=1}^{N} \frac{ C_m(x,t) - C_o(x,t) }{C_o(x,t)} \cdot 100$	A6			
Mean fractional error	$MFE = \frac{1}{N} \sum_{i=1}^{N} \frac{ C_m(x,t) - C_o(x,t) }{(C_m(x,t) + C_o(x,t))/2} \cdot 100$	A7			
Root mean squared error	$RMSE = \sqrt{\frac{1}{N} \sum_{i=1}^{N} (C_m(x,t) - C_o(x,t))^2}$	A8			

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9 Table A2. Definition of the categorical statistics used in the evaluation. Exceedance 10 analysis is based on a comparison with a fixed threshold concentration (T), where a is 11 the number of false alarms, b is the number of hits, c is the number of correct negatives, 12 and d is the number of misses.

Statistic	Formula	
	Tormula	

Accuracy	$A = \left(\frac{b+c}{a+b+c+d}\right) \cdot 100$	A9
Bias	$B = \left(\frac{a+b}{b+d}\right) \cdot 100$	A10
Critical success index	$CSI = \left(\frac{b}{a+b+d}\right) \cdot 100$	A11
Probability of detection	$POD = \left(\frac{b}{b+d}\right) \cdot 100$	A12
False alarm ratio	$FAR = \left(\frac{a}{a+b}\right) \cdot 100$	A13

# 2 Acknowledgements

The Spanish administrations "Generalitat de Catalunya", "Junta de Andalucia", and "Comunidad de Madrid" are acknowledged for providing air quality measurements. The CALIOPE-AQFS team (G.Arévalo, K. Serradell, D. Carrió, M. Castrillo, A. Soret, S. Basart and S. Gassó) and F. Benincasa are also thanked for their technical support. This work is funded by the post-doctoral grant held by M.T. Pay in the Beatriu de Pinós programme (2011 BP-A 00427), Andalusian contract (NET838690), and the Severo Ochoa Program awarded by the Spanish Government (SEV-2011-00067).

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- 3 632-655, 2012.

- 1 Table 1. CALIOPE-AQFS computational requirements, in terms of Central Processor
- 2 Units (CPU) and computational time (in min), for simulating 48h air quality forecasts as
- 3 a function of the domain: IP-4km (D2), AND-1km (D3), BCN-1km (D5) and MAD-
- 4 1km (D4), all of which are described in Figure 1. D-domains are described in Figure 1.

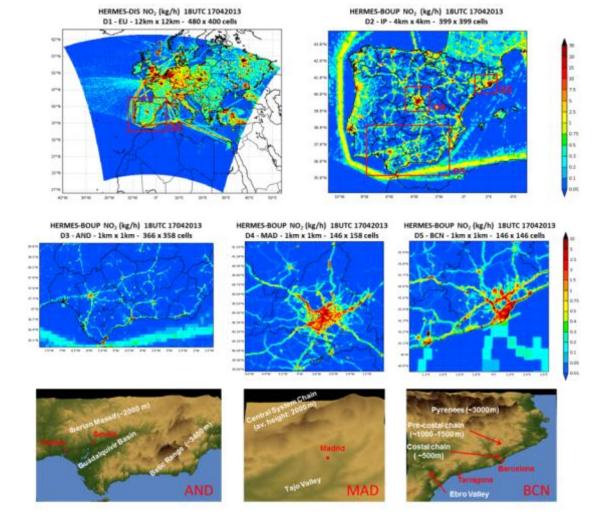
	IP-4km	AND-1km	BCN-1km	MAD-1km
	(399x399 cells)	(366x358 cells)	(146x146 cells)	(146x158 cells)
Meteorological Modeling	128 CPU/15 min	256 CPU/80 min	128 CPU/20 min	128 CPU/20 min
Emission Modeling	1 CPU/ 4 min	1 CPU/ 4 min	1 CPU/1 min	1 CPU/1 min
Air Quality	256 CPU/210	256 CPU/220	128 CPU/150	128 CPU/110
modeling	min	min	min	min

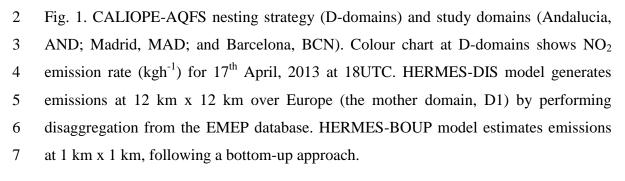
Table 2. Discrete and categorical statistics for NO<sub>2</sub>, O<sub>3</sub>, O<sub>3</sub>-8h, and PM<sub>10</sub> for April 2013 as a function of horizontal resolution (4 km and 1 km). n indicates the number of pairs of data used in the discrete evaluation on an hourly basis. OM and MM depict the measured and modeled mean concentrations, respectively. T is the threshold applied in the categorical evaluation. Max 1h and mean 24h concentrations are calculated by considering  $\geq$ 75% of the hours in a day, as established by Directive 2008/50/EC.

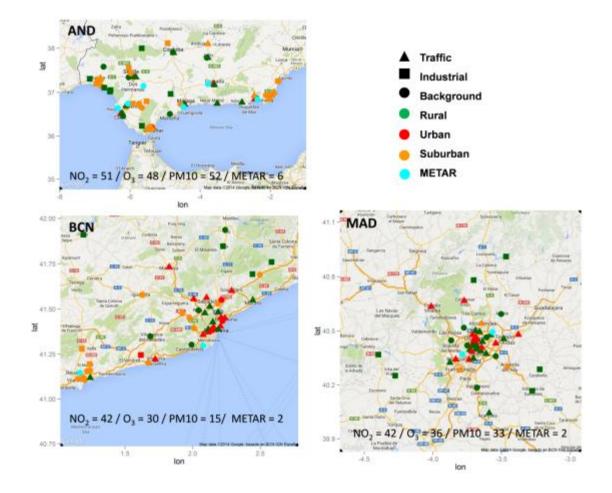
	N	O <sub>2</sub>	C	<b>)</b> <sub>3</sub>	O <sub>3</sub> .	-8h	PN	<b>I</b> <sub>10</sub>	
n (stations)	90761 (135)		76471 (114)		3248 (114)		66642 (100)		
OM ( $\mu$ gm <sup>-3</sup> )	22.0		68	68.4		88.6		20.6	
EU LV/TV ( $\mu gm^{-3}$ )	200 (Max 1h )		100 (1)	180 (Max 1h) 1		120 (Max 8h)		50 (Mean 24h)	
(temp basis)			160 (Max 111)		120 (Max 811)		50 (Mean 2411)		
EU LV/TV	0		n	25		0		31	
exceedances	0		2	20		0		51	
$T^*$ (µgm <sup>-3</sup> ) (temp.	71 (N	[av 1h)	108 (N	fax 1h)	101 (N	(av 8h)	27 (Me	an 24h)	
basis)	71 (Max 1h)		108 (Max 1h)		101 (Max 8h)		27 (Mean 24h)		
Discrete evaluation									
	4 km	1 km	4 km	1 km	4 km	1 km	4 km	1 km	
MM ( $\mu$ gm <sup>-3</sup> )	17.4	19.3	58.0	57.3	72.4	71.5	13.9	14.0	
r	0.54	0.54	0.61	0.61	0.54	0.51	0.45	0.44	
MB ( $\mu$ gm <sup>-3</sup> )	-4.5	-2.6	-10.5	-11.3	-16.3	-17.2	-6.7	-6.6	
MAE ( $\mu$ gm <sup>-3</sup> )	12.9	13.2	19.7	19.8	18.4	19.2	12.6	12.7	
<b>RMSE</b> ( $\mu$ gm <sup>-3</sup> )	19.8	20.4	24.6	24.7	21.8	22.8	17.2	17.4	
MNBE (%)	-20.4	-11.8	-15.3	-16.5	-18.4	-19.4	-32.5	-32.0	
MNGE (%)	58.5	59.9	28.8	28.9	20.8	21.7	61.1	61.6	
MFB (%)	-28.5	-23	-13.1	-15.1	-19.3	-20.5	-63.3	-64.1	
MFE (%)	69.2	68.7	37.1	37.4	22.5	23.5	85.7	87.1	
Categorical evaluation	Categorical evaluation (Threshold = $T^*$ )								
C	4 km	1 km	4 km	1 km	4 km	1 km	4 km	1 km	
a (false alarm)	306	384	41	19	17	6	131	133	
b (hits)	466	537	112	96	6	4	331	334	
c (correct negative)	2517	2439	1846	1868	2826	2837	1978	1976	
d (misses)	487	416	1194	1210	399	401	334	331	
B (%, 100)	37	40	8	7	1	1	42	42	
POD (%, 100)	49	56	9	7	1	1	50	50	
CSI (%, 100)	81	97	12	9	6	2	69	70	
FAR (%, 0)	40	42	27	17	74	60	28	28	
A (%, 100)	79	79	61	62	87	87	83	83	

T\* is defined as 75p of the observed concentrations estimated temporally, as established by EU Directive 2008/50/EC

7

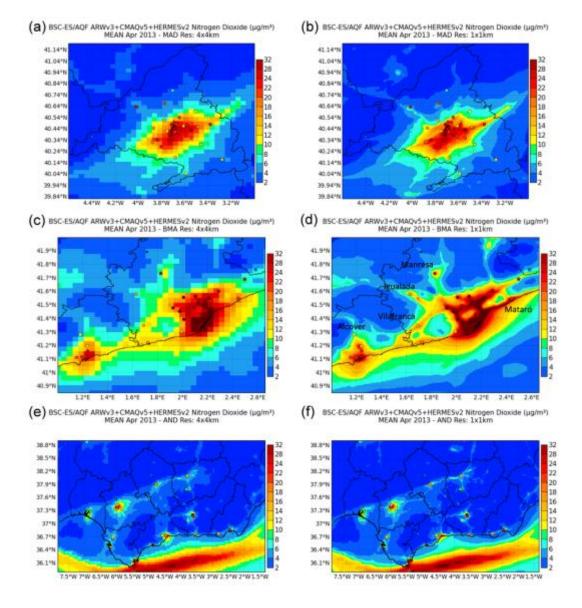






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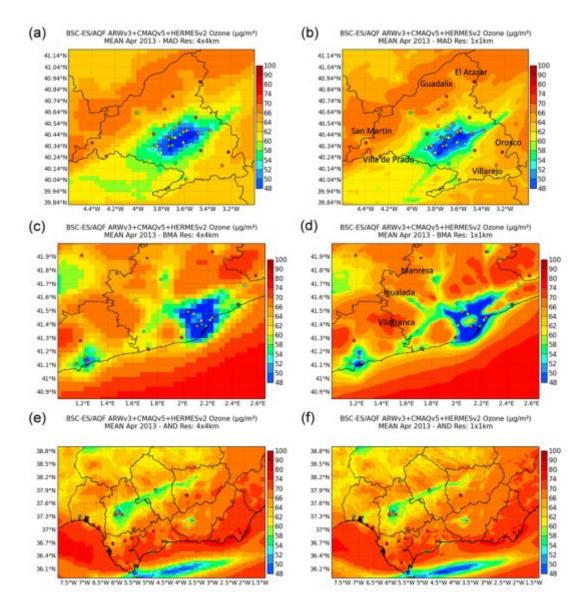
Figure 2. Air quality stations for  $NO_2$ ,  $O_3$  and  $PM_{10}$  in the three domains under study (AND, BCN and MAD) in April 2013. Different types of stations are shown by symbols and color codes. The various symbols represent the major emission type affecting each station (Traffic: triangle; Industrial: square; and Background: circle), while the colors reflect the environment of each station (Urban: red; Suburban: green; and Rural: orange). Cyan dots represent METAR stations used in Sect. S1.



1

2 Figure 3. CALIOPE-AQFS mean NO<sub>2</sub> concentration ( $\mu$ gm<sup>-3</sup>) in April 2013 over (a,b)

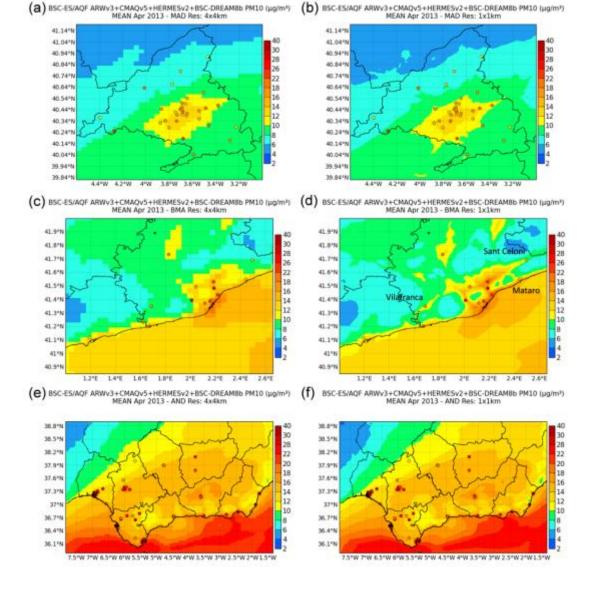
MAD, (c,d) BCN, and (e,f) AND, as a function of horizontal resolution: 4 km (left column) and 1 km (right column). Dots indicate mean concentration at air quality stations.



1

2 Figure 4. CALIOPE-AQFS mean  $O_3$  concentration ( $\mu$ gm<sup>-3</sup>) in April 2013 over (a,b)

MAD, (c,d) BCN, and (e,f) AND, as a function of horizontal resolution: 4 km (left
column) and 1 km (right column). Dots indicate mean concentration at air quality
stations.



1

2 Figure 5. CALIOPE-AQFS mean  $PM_{10}$  concentration ( $\mu$ gm<sup>-3</sup>) in April 2013 over (a,b)

MAD, (c,d) BCN, and (e,f) AND, as a function of horizontal resolution: 4 km (left column) and 1 km (right column). Dots indicate mean concentrations at air quality stations.

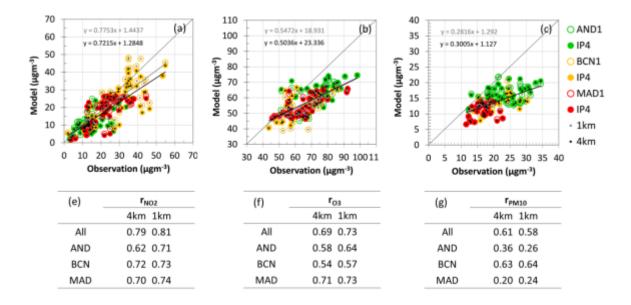




Figure 6. Monthly mean scatter plots for CALIOPE-AQFS (y-axis) and observed (xaxis) concentrations for the three study domains (AND in green, BCN in yellow, and MAD in red), as a function of horizontal resolution for (a) NO<sub>2</sub>, (b) O<sub>3</sub> and (c) PM<sub>10</sub>. Equations show the linear adjustment between models and observations at 1 km (light grey) and 4 km (dark grey). Spatial correlation coefficients as a function of resolution and domain are shown for (e) NO<sub>2</sub>, (f) O<sub>3</sub>, and (g) PM<sub>10</sub>.

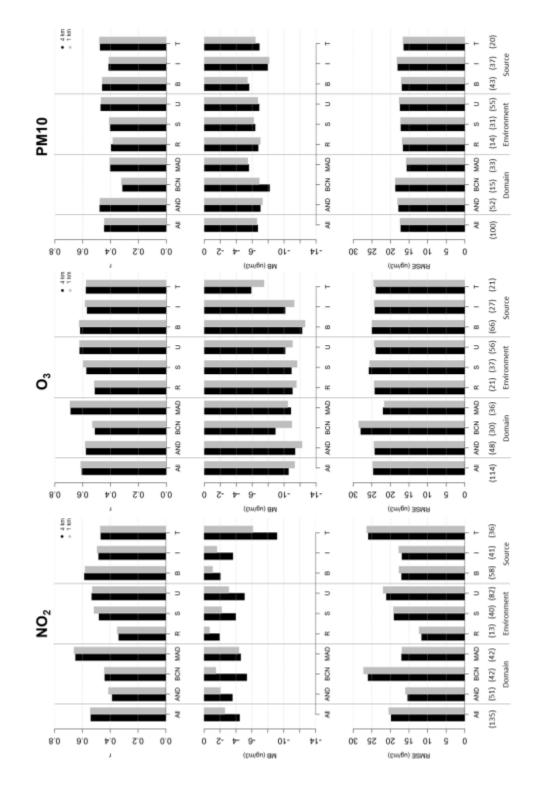


Figure 7. Statistics (r, MB, and RMSE in rows) for each pollutant (NO<sub>2</sub>, O<sub>3</sub>, and PM<sub>10</sub>
in columns) on an hourly basis as a function of horizontal resolution: 4 km (black) and 1
km (grey). Four categories are considered: all stations (all), domain (AND, BCN and
MAD), station environment (R, S, and U), and main sources (B, I, and T).

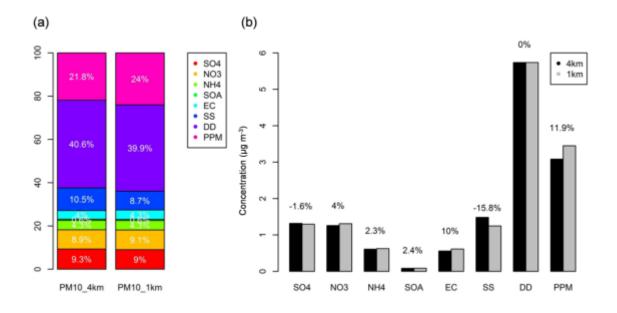




Figure 8. Resolution effect on  $PM_{10}$  components in April 2013. (a) Percentage of  $PM_{10}$ components: sulfate (SO4), nitrate (NO3), ammonium (NH4), secondary organic aerosol (SOA), elemental carbon (EC), sea salt (SS), desert dust (DD), and primary particulate matter (PPM). (b)  $PM_{10}$  component concentrations in the 1-km simulation (black) and 4-km simulation (grey). Numbers over bars indicate the % of increase when increasing resolution.

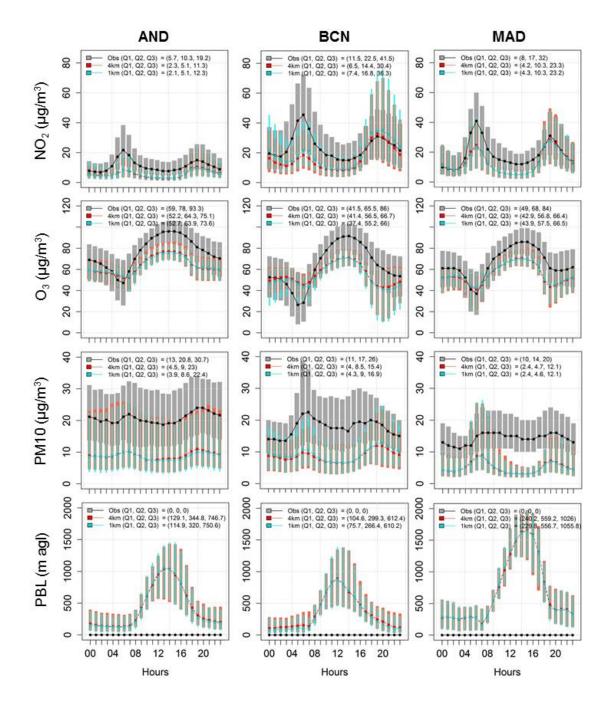


Figure 9. Daily cycles for NO<sub>2</sub>, O<sub>3</sub> and PM<sub>10</sub> for each study domain at available stations
as a function of resolution. No observations of PBL are available. Q1, Q2 and Q3
indicate quartiles for the daily cycle. Bars show Q1 and Q3 at each hour.

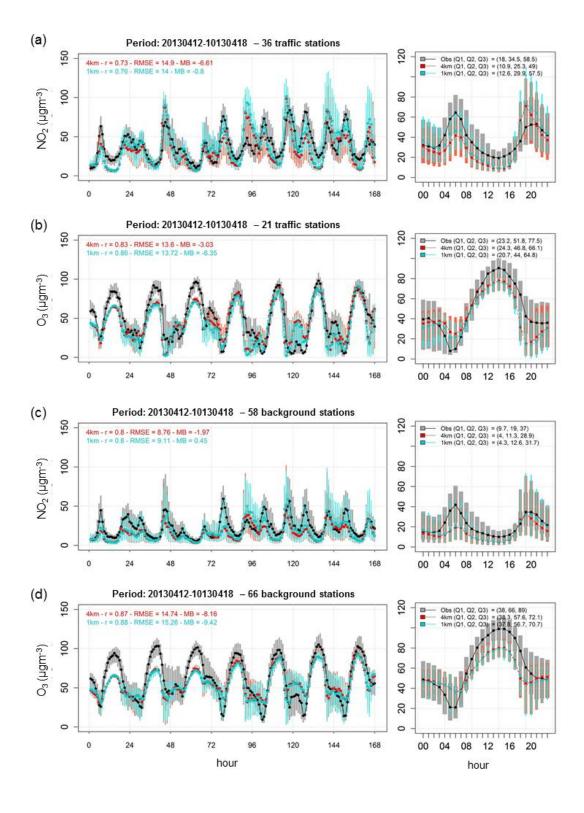


Figure 10. Temporal series and daily cycles for NO<sub>2</sub> and O<sub>3</sub> at background (a and b,
respectively) and traffic stations (c and d, respectively) for the episode of 12-18<sup>th</sup> April,
2013. Q1, Q2 and Q3 indicate quartiles for the daily cycle. Bars show Q1 and Q3 at
each hour.