

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

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Abstract

CALIOPE-AQFS represents the current state-of-the-art in air quality forecasting systems of high-resolution running on high-performance computing platforms. It provides a 48-h forecast of ~~the main pollutants~~ (NO₂, O₃, SO₂, PM₁₀, PM_{2.5}, CO, and C₆H₆) ~~found at a~~ 4-km horizontal resolution over all of Spain, and at 1 km over the most populated areas in Spain with complex terrains (Barcelona, Madrid and Andalucia domains). Increased horizontal resolution from 4 km to 1 km over the aforementioned domains leads to finer texture and more realistic concentration maps, which is justified by the increase in NO₂/O₃ spatial correlation coefficients from 0.79/0.69 (4 km) to 0.81/0.73 (1 km). High resolution emissions using the bottom-up HERMESv2.0 model are essential for improving model performance when increasing resolution on an urban scale, but it is still insufficient. Decreasing grid spacing does not reveal the expected improvement in hourly statistics, i.e., decreasing NO₂ bias only by ~2 μgm⁻³ and increasing O₃ bias by ~1 μgm⁻³. The grid effect is less pronounced for PM₁₀ because part of its mass consists of secondary aerosols, which are less affected than the locally emitted primary components by a decreasing grid size. The resolution increase has the highest impact over Barcelona, where air flow is controlled mainly by mesoscale phenomena and a lower PBL. Despite the merits and potential uses of the 1-km simulation, the limitations of current model formulations do not allow ~~confirming~~ their expected superiority close to highly urbanized areas and large sources. Future work should combine high grid resolution with techniques that decrease subgrid variability  and also include models that consider urban morphology and thermal parameters.

1 **1 Introduction**

2 The World Health Organization (WHO) has recently shown that there is sufficient
3 evidence ~~supporting the belief~~ that particulate matter (PM), ozone (O₃) and nitrogen
4 dioxide (NO₂) affect human health (WHO, 2013). Although NO₂ and PM
5 concentrations improved from 2002 to 2011 in Europe, the situation is still far from
6 matching the WHO air quality guidelines (AQG). The European annual limit values for
7 NO₂ (annual) and PM₁₀ (daily) had been exceeded at 42-43% of the traffic stations in
8 2011. For the same year, about 33% of the European urban population was exposed to
9 PM₁₀ concentration₁ above the daily limit value, and nearly 88% ~~was~~₂ exposed to the
10 respective WHO AQG (EEA, 2013). Air pollution legislation for the protection of the
11 increasing city population has recently increased the demand for urban air pollution
12 forecasting systems that can assess and understand ~~its~~₃ dynamics, alert the population
13 when health-related issues occur, and develop emission abatement plans (EEA, 2011).

14 When applying an air quality modeling system, defining the grid resolution is an
15 important consideration; ~~the~~₄ potential benefits of higher-resolution modeling should be
16 weighed against: ~~the~~₅ increased complexity of the inputs, CPU time, and disk space
17 requirements. In theory, higher resolution modeling is expected to yield more accurate
18 forecasts because of better resolved model input fields (topography, land cover and
19 emissions); and **better mathematical characterization of physical and chemical**
20 **processes**. Furthermore, high resolutions (ranging from 1 to 5 km) are essential to
21 reproduce mesoscale phenomena, e.g., those controlling O₃ transport along the
22 mountainous northeastern Mediterranean coast (Fay and Neunhäuserer, 2006; Jiménez
23 et al., 2006). Even at ~~the~~₆ finest scale, the modeled concentrations are not necessarily the
24 best (Mass et al., 2002; Gego et al., 2005; Valari and Menut, 2008); because increasing
25 emission and meteorology spatial resolution can also increase uncertainties, at the risk
26 of reduced model performance. Nowadays, fine horizontal resolution is a persistent
27 challenge when assessing health impact₇ and population exposure studies (Thompson et
28 al., 2013).

29 Several studies have evaluated the impact of increasing horizontal resolution on
30 different scales over the eastern and southeastern USA using the Community Multiscale
31 Air Quality (CMAQ) model and the Comprehensive Air Quality Model with Extensions
32 (CAMx), which range from 32 km – 12 km – 4 km (Cohan et al., 2006; Tesche et al.,
33 2006; Queen and Zhang, 2008). They found no significant changes for O₃ and PM (<5%

1 on average), and those changes were even lower at resolutions of between 12 km and 4
2 km (<3%). Concerning PM components, Fountoukis et al. (2013) found that increased
3 resolution provides differences mostly for primary PM rather than secondary PM.
4 Recently, a model intercomparison exercise, named ScaleDep, was performed to
5 determine the effect of grid resolution on air quality modeling performance over Europe
6 at a regional and urban scale (Cuvelier et al., 2013). The exercise, involving five
7 Chemical Transport Models (CTMs) (EMEP, CHIMERE, CMAQ, LOTOS-EUROS
8 and RCGC) running under the same conditions over the full year 2009 and at four
9 resolutions (56, 28, 14, and 7 km), showed that it is difficult to define a grid size that is
10 adequate for resolving the urban signal under all conditions affecting Europe. Still, a 14-
11 km resolution seems to be a good compromise between background applications and
12 those reproducing most of the urban signals (7 km resolution). However, the ScaleDep
13 exercise did not distinguish between the different topographies or complex
14 meteorological patterns which are characteristic of the Iberian Peninsula.

15 Few studies have been performed over selected areas in Spain; and of those, the focus
16 has been mainly on O₃ and NO₂. Vivanco et al. (2008) evaluated the annual impact by
17 increasing the resolution (to 36, 19, and 7 km) over Madrid for NO₂ and O₃. They used
18 the WRF-CHIMERE model, disaggregating the EMEP emission inventory based on
19 land use information. Their evaluation showed that the model improves greatly for NO₂
20 than O₃, and the most significant improvement is achieved when resolution increases
21 from 36 to 19 km rather than to 7 km, which is linked to increased uncertainty in the
22 emission data introduced with the disaggregation techniques. Jiménez et al. (2006) used
23 the MM5-CMAQ model along with a bottom-up emission model (EMICAT2000) to
24 assess the influence of grid resolution on O₃ (at 8, 4, and 2 km) over the complex terrain
25 of the northeastern Iberian Peninsula (Catalonia) during a summer pollution episode.
26 They indicate that, due to improved performance of the mesoscale phenomena and a
27 better allocation of emissions, a 2-km resolution improves the capability of the model to
28 simulate exceedances of European limit values. An important issue in both studies is the
29 emission modeling approach (top-down vs. bottom-up) when applying high resolution
30 at the local scale (<10 km). As Fountoukis et al. (2013) and Timmermans et al. (2013)
31 demonstrate, in the range of local scale (e.g., the greater Paris area), the grid resolution
32 is not currently the major source of discrepancies in model performance, but in fact, the
33 predicted concentrations and corresponding gradients are more consistent with observed

1 concentrations when provided by bottom-up emission inventories rather than down-
2 scaled inventories. If local variation in input data (e.g., emission patterns or land use)
3 cannot be properly characterized, modeling with a finer grid resolution may not provide
4 any great advantages.

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

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

1 The previous works demonstrate there is not a single **answer** which explains the merits
2 of high-resolution modeling for all applications. The present work aims to assess the
3 impact of increasing the horizontal resolution from 4 km to 1 km, specifically over areas
4 affected by heterogeneous emission patterns and complex terrains such as the Barcelona
5 and Madrid metropolitan areas (BCN and MAD) together with the Andalusian region
6 (AND). For that purpose, CALIOPE-AQFS forecasts pollutant concentrations (O_3 , NO_2 ,
7 and PM_{10}) at two horizontal resolutions: first at a 4-km resolution covering Spain (IP4),
8 and second at a 1-km resolution covering the AND, BCN, and MAD domains. The
9 study is performed for the period April 2013, which presented seven days of an air
10 pollution episode. We use observations from routine air quality monitoring networks to
11 evaluate both resolutions.

~~12 Section 2 describes the configuration and computational setup of CALIOPE-AQFS; it
13 analyses the domains and the period under study and it defines the methodology used to
14 evaluate the resolution increase.~~ Section 3 quantifies the impact of resolution increase
15 on forecasting hourly concentrations (and exceedances) in terms of: pollutant, domain,
16 building density and major emission sources. Section 4 concludes with the main results
17 and some recommendations.

18 **2 Methodology**

19 **2.1 Domain and period under study**

20 **Figure 1** shows the ~~main~~ NO_2 emission patterns and topographic characteristics of the
21 ~~domains: the Barcelona and Madrid metropolitan areas (BCN and MAD) and the~~
22 ~~Andalusian region (AND).~~ BCN is a coastal area characterized by several valleys
23 perpendicular to the coastline and two main mountain ranges, one coastal (500 m
24 height) and one **pre-coastal** (1000-1700 m height). These features induce mesoscale
25 phenomena such as sea-breeze and mountain-valley winds. On the other hand, MAD is
26 a continental region with a much simpler topography that includes the Tajo valley in the
27 ~~southern of MAD~~ and the mountain range of the Central System ~~located in the~~
28 ~~northwestern MAD~~, with summits reaching 2500 m ~~height.~~ ~~These features bring~~
29 ~~different locally-driven flows.~~

30 The urban contribution in BCN (3.1 million inhabitants) is accompanied by industrial
31 and power generation emissions, the road network and the harbor; ~~meanwhile~~ the

1 Spanish capital of MAD is mainly influenced by emission from the urban area (5.8
2 million inhabitants) and the road network that connects MAD with the surrounding
3 commercial and industrial zones as well as the urban areas.

4 AND is the southern-most region in Spain, with complex topography characterized by
5 the large depression of the Guadalquivir Basin (delimited by the Iberian Massif and the
6 Betic Range), which crosses the region from NE to SW over a 60–km stretch. About
7 three quarters of AND has a mountainous orography, including the Sierra Nevada (3481
8 m). AND includes one of the five biggest cities in Spain, Seville (~700 000 inhabitants),
9 which hosts industrial and electric generation activities around the Algeciras bay, and it
10 is affected by dense maritime traffic through the Strait of Gibraltar.

11 The study is performed over April 2013. At the beginning and end of the month, the
12 synoptic circulation was controlled by a low pressure system displaced over the south of
13 the British Isles and which affected Western Europe by leading to atmospheric
14 instability over the IP. This pattern is typical of transitional months such as April and
15 November (García-Valero et al., 2012; Valverde et al., 2014), which produce
16 precipitation and decreased temperatures because of cold and humid winds entering
17 from the Atlantic Ocean. In contrast, from 12-18 April there was a high pressure system
18 crossing the Iberian Peninsula in a SW-NE direction, transporting dust from the Sahara
19 Desert and increased temperatures of up to 25-28°C. During the latter episode,
20 available air quality stations at the study domains displayed several exceedances of the
21 European limit values (8 exceedances of the NO₂ hourly limit value, 25 exceedances of
22 the O₃ information threshold, and 31 exceedances of the PM₁₀ daily limit value).

23 2.2 CALIOPE-AQFS

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

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

14 Vertically, WRF-ARW is configured with 38 sigma layers up to 50 hPa, with 11
15 characterizing the planetary boundary layer (PBL); meanwhile CMAQ vertical levels
16 are obtained by collapsing from the 38 WRF levels to a total of 15 layers that steadily
17 increase from the surface up to 50 hPa. Six layers are within the PBL, and the first layer
18 depth is 39 m.

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

25 Currently, a new CMAQ version is being tested in the CMAS community, namely
26 CMAQv5.0 ([CMAQ, 2012](#)). It includes substantial scientific improvements over
27 Version 4.5 and is especially devoted to improving SOA formation as well as the
28 dynamic interactions of fine and coarse aerosols. Based on the evaluation results from
29 the previous CMAQ version within CALIOPE-AQFS (4.5 vs. 5.0) ([Pay et al. 2012b](#)),
30 CMAQ has been updated to Version 5.0.1 using the CB05 chemical mechanism
31 ([Yarwood et al., 2005](#)), the AEROS for aerosol modeling, and the in-line photolysis
32 calculation.

1 CALIOPE-AQFS considers desert dust contribution by means of the BSC-
2 DREAM8bv2, which runs off-line at a 0.5° x 0.5° 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 and at the required resolution and domain.
5 After interpolating, the modeled PM₁₀ concentration is the sum of Aitken,
6 accumulation and coarse-mode modes from CMAQ, and the corresponding BSC-
7 DREAM8bv2 bins with a diameter of ≤ 10 μm (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 in one CPU
19 (9.86 years). The storage for the April 2013 output files was 6.13 TB (~200 GB/day).

20 **2.4 Evaluating the increase in resolution**

21 Comparing CALIOPE-AQFS ~~grid resolutions~~ and measurements was done ~~in terms of~~
22 gas-phase and aerosol concentrations (O₃, NO₂, and PM₁₀). Representativeness continue
23 to be a challenge when comparing gridded simulations to observational data at a point
24 in time and space, as modeled concentrations represent a volumetric average over an
25 entire grid cell. Furthermore, the stochastic compound embedded in the observations is
26 not accounted for. Concerning temporal representativeness in the present comparison,
27 both modeled and measured concentrations are averaged hourly. CALIOPE-AQFS
28 operationally receives air quality measurements from Spanish administrative networks
29 in near real time (NRT) without any ~~quality data or~~ quality control. For the present
30 study, NRT measurements are filtered by removing data before and after measurement
31 interruptions or calibrations. Also, a minimum cut-off threshold of 1 μgm⁻³ is applied to

1 the observed concentrations in order to avoid unrealistic observations. After filtering,
2 the number of stations is 48/30/36 for O₃, 51/42/42 for NO₂, and 52/15/33 for PM₁₀ at
3 ~~AND/BCN/MAD.~~

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

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

14 The evaluation is based on discrete statistics performed on an hourly basis. We consider
15 the correlation coefficient (r), mean (absolute, relative, and fractional) biases (MB,
16 MNBE, and MFB), and error (MAE, MNGE, and MFE). Root Mean Square Error
17 (RMSE) is also calculated because it intensifies large differences between measured and
18 observed concentrations (Table A1).

19 In order to evaluate the effect of resolution increase on forecasted exceedances and non-
20 exceedances of limit values established by the European legislation, we calculate
21 categorical statistics based on comparisons with fixed concentration thresholds (T). The
22 calculated statistics are accuracy (A), bias (B), probability of detection (POD), critical
23 success index (CSI), and false alarm ratio (FAR), whose formulas and descriptions are
24 explained in Table A2 and elsewhere in Kang et al. (2005) and Eder et al. (2006). The
25 2008/50/EC directive sets an information threshold of 180 µg m⁻³ for maximum daily O₃
26 concentrations (Max 1h O₃) and a target value of 120 µg m⁻³ for the maximum daily 8-h
27 running O₃ mean (Max 8h O₃), which should not be exceeded more than 25 days per
28 year. It establishes a limit value of 200 µg m⁻³ for maximum daily NO₂ concentrations
29 (Max 1h NO₂), and 50 µg m⁻³ for the daily PM₁₀ mean (Mean 24h PM₁₀), which should
30 not be exceeded more than 35 times per year. Therefore, categorical evaluation will be
31 performed for Max 1h NO₂, Max 1h and Max 8h O₃, and Mean 24h PM₁₀. Note that
32 mean and maximum concentrations are calculated by considering at least 75% of the

1 data in the corresponding time base, i.e., values of at least 18 hours per day for Mean 24
2 h, Max 1h, and Max 8h; and 6 hours for 8 h values, as established by 2008/50/EC.

3 **3 Concentration maps and spatial representativeness**

4 To analyze the spatial differences between resolutions, [Figs. 3, 4, and 5](#) show the
5 monthly mean concentration maps for April 2013 over MAD, BCN and AND domains
6 at 4 km (left panels) and 1 km (right panels) for NO₂, O₃, and PM₁₀, respectively.

7 The maps of NO₂ and PM₁₀ at both resolutions display similar distribution along the
8 MAD and BCN urban plumes. On-road traffic constitutes the main source of primary
9 pollutants in MAD and BCN. HERMESv2.0 estimates that 75% and 59% of NO_x
10 emissions are produced by on-road traffic in both domains, respectively. Consequently,
11 when the resolution increases, the monthly mean O₃ concentration maps are almost
12 identical, although the NO_x titration effect on O₃ is significant along highways and
13 major point sources. In AND, NO₂ and O₃ concentrations are also conserved between
14 resolutions along the shipping route crossing the Strait of Gibraltar towards the
15 Mediterranean Sea.

16 However, the definition of NO₂ concentrations along highways connecting the biggest
17 cities with the rest of the country and industrial sectors are more easily identified at 1-
18 km simulations than at 4 km, especially along those roads from/to Barcelona (e.g., the
19 AP7 Mediterranean highway and C32, which connects the harbor and the airport) and
20 Madrid (the A-2 and A-6 in the north, and A-3, A-4 and A-5 in the south). In the same
21 way, 1-km O₃ maps are more textured than those at 4 km along highways, because the
22 titration effect is more significant at 1 km, due to less dilution within grid cells. The
23 titration effect of NO_x on O₃ over the main sources is more forceful in BCN than in
24 MAD, given that BCN has a larger concentration gradient resulting from complex
25 topography and recirculation flows that accumulate pollutants.

26 The improvement of the definition along roads in AND is lower than that observed in
27 the MAD and BCN domains, due to the fact that the AND domain is bigger and
28 displays lower traffic emission sources than the MAD or BCN domains. Regarding
29 PM₁₀, the main component in AND is the desert dust (~40% in both resolutions) from
30 North Africa. This is because there were two episodes on 14-19th and 25-26th April that
31 affected the IP, as shown by the S-N PM₁₀ gradient ([Fig.5e and f](#)). The desert dust is
32 transported from long-range simulation with BSC-DREAM8bv2.

1 Over complex terrains, the 1-km simulation ~~allows reproducing~~ more realistic NO₂
2 concentration maps because of its more detailed topographic information ~~input~~. For
3 instance, the BCN 1-km simulation displays the lowest NO₂ concentrations (< 10 µgm⁻³)
4 along the coastal chain (500 m height) and pre-coastal chain (1000-1700 m height),
5 except for the city's urban hill, where concentrations reach 20-40 µgm⁻³. In contrast, the
6 4-km simulation provides smoother NO₂ concentrations without any concentration
7 gradient. Thus, the 1-km simulation generates slightly higher O₃ background
8 concentrations than ~~does~~ the 4-km simulation along the BCN pre-coastal chain (66-70
9 vs. 70-74 µgm⁻³), as well as across the Iberian Massif (AND), where the O₃ ~~map~~
10 displays significant structure due to the higher resolution topography that ~~shapes~~
11 basin and the ~~on-road traffic~~.

12 [Figures 3, 4, and 5](#) include dots corresponding to mean concentrations at air quality
13 stations that help to qualitatively evaluate the modeled spatial representativeness at both
14 resolutions. Note the ~~high concordance~~ between NO₂ observations and the 1-km
15 simulation near the primary suburban traffic roads in BCN (e.g., Vilafranca, Igualada,
16 Manresa, and Mataró). Regarding O₃, although observed concentrations depict an
17 overall tendency of the model to underestimate concentrations at both resolutions, the 1-
18 km simulation ~~displays a higher accord~~ with measurements at rural background stations
19 (e.g., El Atazar, San Martín, Villa de Prado, Villarejo and Orosco stations in MAD), and
20 at suburban traffic stations (e.g., Manresa, Igualada and Vilafranca in BCN, with
21 modeled O₃ concentrations of around 54-58 µgm⁻³ at 1 km, and 60-66 µgm⁻³ at 4 km).
22 For PM₁₀, comparisons with measurements show that modeled concentrations are
23 underestimated over background areas, mainly outside the urban/suburban area, as
24 already discussed in [Pay et al. \(2012a\)](#). However, PM₁₀ measurements at the
25 urban/suburban stations of Vilafranca, Sant Celoni, and Mataró in BCN (14-16 µgm⁻³)
26 show a higher ~~concordance~~ at 1 km than at 4 km (12-14 µgm⁻³ vs. 8-10 µgm⁻³).

27 The spatial variability of the increased resolution is quantitatively analyzed by means of
28 concentration maps, shown in [Fig. 6](#) for NO₂, O₃ and PM₁₀ over AND, BCN and MAD.
29 Over all domains, the ~~explained spatial variability~~ improves as a function of the
30 resolution increase for NO₂ and O₃, ~~which is sustained~~ by the increase in monthly r
31 from 0.79 (4 km) to 0.81 (1 km) for NO₂, and from 0.69 to 0.73 for O₃. The slopes
32 improve with the resolution increase, from 0.72 (4 km) to 0.77 (1 km) for NO₂, and
33 from 0.50 (4 km) to 0.54 (1 km) for O₃. This results from the improved model

1 performance at urban stations, indicating that CALIOPE-AQFS ~~explains better~~ the
2 magnitude of the variability between urban regions at 1 km than at 4 km. In contrast, for
3 PM₁₀, the monthly r decreases from 0.67 to 0.58 when the resolution increases.
4 ~~Although~~ the PM₁₀ spatial variability over BCN and MAD improves when the
5 resolution increases (r increases by 0.01 and 0.04, respectively), the global correlation
6 coefficient ~~deviates mainly~~ at the AND stations (52/100), ~~where r decreases~~ by 0.1 from
7 0.36 (4 km) to 0.26 (1 km). Despite the unfavorable effect of the resolution increase in
8 PM₁₀ over AND, the NO₂ and O₃ concentrations show the highest absolute increase ~~in~~
9 the spatial r over this domain, from 0.62 (4 km) to 0.71 (1 km) for NO₂ and from 0.58 (4
10 km) to 0.64 (1 km) for O₃ (increasing r by 0.09 and 0.06, respectively).

11 4 Temporal evaluation

12 ~~The present section discusses the temporal evaluation of the resolution increase by~~
13 ~~pollutant, environment, predominant emission sources, and study domain. Fig. 7~~
14 ~~summarizes the statistical evaluation.~~

15 4.1 Pollutant

16 Table 2 depicts the statistical evaluation by pollutant, with a focus on the reproduction
17 of high concentrations established by the European directive (2008/50/EC). Depending
18 on the pollutant's lifetime and variability, as well as its dependency on precursors,
19 increased resolution shows different impacts. The resolution increase has a positive
20 effect on NO₂, decreasing its bias by 2.0 $\mu\text{g m}^{-3}$ (from -4.5 to -2.5 $\mu\text{g m}^{-3}$); but it also
21 increases absolute (squared) errors by 0.3 $\mu\text{g m}^{-3}$ (0.9 $\mu\text{g m}^{-3}$). This positive effect is
22 sustained by the perceptual variability, where the MB (MFB) is reduced by 42% (19%);
23 whereas MAE (MFE) only increases by 2% (1%). The correlation coefficient does not
24 significantly change, which is obvious because emissions at both resolutions are
25 modeled using the same approach. The bias improvement at 1-km resolution is justified,
26 because the higher resolution leads to better emission allocation from point, linear or
27 area sources; it decreases the artificial dilution of emission compared to the larger grid
28 area; and, due to the decrease of artificial dilution, it treats chemistry more properly near
29 large emission sources.

30 In contrast, the resolution increase has a negative effect on hourly and Max 8h O₃
31 concentrations because it increases biases and errors by 0.1-0.8 $\mu\text{g m}^{-3}$. Relative
32 (fractional) biases and errors increase by 8% (15%) and 1% (1%), respectively, for

1 hourly O₃; and 6% and 4% for Max 1h O₃. However, the statistical evaluation alone is
2 not enough to explain the impact of the resolution increase on O₃ concentrations.

3 According to the categorical evaluation, only a few exceedances of the European target
4 and limit values were detected for Max 1h NO₂ (9), Max 1h O₃ (25), and Mean 24h PM₁₀
5 (31) in April 2013. Thus, categorical evaluation is performed on the temporal basis
6 established by the European legislation, but it uses a T based on the 75p of the observed
7 concentrations in each case. T corresponds to 71 μgm⁻³ for Max 1h NO₂, 108 (101)
8 μgm⁻³ for Max 1h (Max 8h) O₃, and 27 μgm⁻³ for Mean 24h PM₁₀.

9 Overall, CALIOPE-AQFS underestimates exceedances at both resolutions, indicating
10 that errors of missing observed exceedances are not totally resolved by a resolution
11 increase (a<d). The best performance is found for Max 1h NO₂, where bias (B)
12 improves from 37% (4 km) to 40% (1 km).

13 For NO₂ Max 1h, there are 953 observed exceedances (b+d) of the threshold (T=47
14 μgm⁻³). Increasing the resolution increases the POD from 49% (4 km) to 56% (1 km).
15 As POD, CSI studies the exceedances, but in a more coherent way by considering both
16 false alarms and missing events. Both POD and CSI increase by 14% and 20% when
17 resolution is increased. ~~The opposite effect appears for O₃. Aside from the fact that the~~
18 ~~O₃ POD is relatively low, the POD decreases as the resolution increases. Of the 1306~~
19 ~~observed exceedances of the 108 μgm⁻³ Max 1h, CALIOPE-AQFS detected 112~~
20 ~~exceedances at 4 km and only 96 at 1 km. Increasing resolution decreases POD and CSI~~
21 ~~by 22% and 25% for O₃ Max 1h, whereas they do not significantly change for Max 8h~~
22 ~~O₃ and Mean 24h PM₁₀.~~

23 FAR increases for Max 1h NO₂ (from 40% to 42%) and decreases for Max 1h O₃ (from
24 27% to 17%) when the resolution increases. In relative terms, this variability is more
25 significant for Max 1h O₃ (37 %) than for Max 1h NO₂ (5%), indicating that, in terms of
26 failures, the resolution has a positive global effect by reducing false exceedances.

27 For various reasons, accuracy (A) remains almost constant when the resolution
28 increases. Regarding NO₂ and O₃, it is due to a stable sum of b and c, increasing the b at
29 the cost of c, and vice versa. For NO₂, the number of hits (b) to forecast Max 1h at 1
30 km is higher than 4 km (537 vs. 466), but the number of correct negatives at 1 km is

1 lower than at 4 km (2439 vs. 2517). The resolution increase has the opposite effect on
2 O₃ over b and c for both Max 1h and Max 8h.

3 4.2 PM₁₀ components

4 The resolution increase has the lowest effect on PM₁₀ hourly concentrations and its
5 exceedances (<1%). PM₁₀ components are secondary inorganic aerosols (SIA), which
6 include sulfate (SO₄), nitrate (NO₃) ammonium (NH₄), secondary organic aerosol
7 (SOA), elemental carbon (EC), sea salt (SS), desert dust (DD), and primary PM (PPM).

8 Pay et al. (2012a) already evaluated the PM components at some Spanish urban and
9 rural background stations using the CALIOPE-AQFS based on CMAQv4.5, and they
10 showed that the model underestimated the secondary inorganic aerosols by a factor of 2-
11 3. The highest underestimation was found for fine carbonaceous aerosols (a factor of 4),
12 in part related to the state-of-the-science concerning secondary organic aerosol
13 formation pathways. The updated version of CMAQ, v5.0.1 includes scientific
14 improvements concerning SOA formation and aerosol dynamics, which could improve
15 the modeled PM₁₀ performance for its components.

16 Figure 8a shows that the resolution increase does not significantly change the PM₁₀
17 composition. DD remains the main contributor (~40-41%), followed by PPM (22-24%),
18 SIA (~21-22%), SS (9-11%), EC (~4%) and SOA (~0.6%). However, the effect of the
19 increased resolution on PM₁₀ component concentrations is different (Fig. 8b),
20 depending on their origin, atmospheric cycle and the way they are modeled. DD
21 concentrations do not change between resolutions, because they are mass conservative
22 when interpolated from 0.5°x0.5° till 1 km x 1 km.

23 Regarding SIA, increasing the resolution increases NO₃ and NH₄ concentrations by 4
24 and ~2%, respectively, and it decreases SO₄ by ~2%. The NH₄ increase means there are
25 more primary precursors (H₂SO₄ or HNO₃/NO₂) available to neutralize NH₃ (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
28 controlled to some extent by the ISORROPIA thermodynamic equilibrium.
29 Furthermore, the absence of aerosol measurements for April 2013 does not allow us to
30 explain this situation.

1 ~~The resolution increase displays the highest decrease for SS (~16%),~~ CMAQv5.0.1
2 simulates SS emission as a function of the wind speed and ~~the~~ relative humidity (Gong,
3 2003; Zhang et al., 2005). **Although not shown here, when the resolution increases, the**
4 **wind speed increases at the available PM₁₀ stations by ~1.4/0.4/0.2 ms⁻¹ over**
5 **AND/BCN/ MAD, and also over the open ocean.**

6 For primary PM components (EC and PPM), increasing resolution ~~presents the highest~~
7 ~~increase in~~ concentration (by 10 and ~12%, respectively). **As for NO₂, the 1-km**
8 **simulation leads to a reduced effect of artificial dilution of emissions in a grid cell, so**
9 **concentration gradients are stronger than in the 4-km simulation.**

10 **4.3 Domain**

11 ~~Due~~ to differences in geographical location and emission patterns over the domains
12 ~~under study, the resolution increase has different impacts~~ (Fig. 7). BCN shows the
13 highest NO₂ bias decrease (73%) when the resolution increases, ~~yet with~~ no effect on
14 the correlation (<7%). However, O₃ shows significant variability over BCN, increasing
15 r (by 4%) and MB (by 23%). To a lesser extent, MB also increases over AND (by 8%).
16 **Meanwhile, the variability over MAD is reduced (<4%). MB decreases for PM₁₀ (< 1**
17 **μgm⁻³) over the urban domains of MAD (3%) and BCN (16%), and increases over AND**
18 **(7%).**

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

25 As shown in **Sect. S1**, the pollutant transport at the BCN coastal domain is controlled by
26 mesoscale phenomena such as sea-breezes (day) and land-breezes (night), which are a
27 result of its complex topography and location (Baldasano et al., 1994; Millán et al.,
28 1997; Gonçalves et al., 2009). The NO₂ daily cycle is highly influenced by traffic
29 emissions (Fig. 9). Both resolutions show the highest underestimations ~~at~~ the morning
30 peak **(5-9 am)** (~20 μgm⁻³). Although the afternoon peak is well reproduced, there is
31 excessive variability at both resolutions, ~~which results from~~ problems with wind

1 direction. During the sea breeze period, the mean simulated wind was more easterly
2 than westerly, as registered by measurements (Sect. S1). Several works indicate that
3 WRF ~~does not faithfully reproduce~~ the morning and evening transition over the urban
4 environment, possibly because it does not model the heat retention in cities (Makar et
5 al., 2006; Appel et al., 2013). Increasing the resolution increases ~~the~~ NO₂ concentrations
6 from 14 μgm⁻³ (4 km) to 17 μgm⁻³ (1 km) during the morning hours after sunrise (5-9
7 am) and in the evening hours after sunset (5-9 pm). This behavior could be explained by
8 PBL variability when increasing the resolution, which decreases PBL height by ~33 m
9 for these hours.

10 NO₂ performance impacts the O₃ daily cycles over BCN, showing that 4- and 1-km
11 simulations underestimate maximum O₃ concentrations by ~20 μgm⁻³ at midday (1-4
12 pm), and ~~it overestimates~~ minimum O₃ concentration by ~20 μgm⁻³ in the morning
13 hours after sunrise (5-9 am). The resolution increase ~~allows slightly decreasing~~ O₃
14 concentrations at night, which is perhaps ~~controlled by the PBL decreasing at 1 km~~
15 during the early morning and late afternoon, when PBL ~~reaches the minimum height~~.
16 During these hours, the titration effect of NO₂ on O₃ is ~~more effective~~, improving the O₃
17 overestimation of the daily minimum, which allows a slightly ~~increasing~~ hourly r (2%).
18 However, O₃ underestimation increases in the late afternoon, contributing to an increase
19 in the hourly mean bias from ~9 μgm⁻³ (4 km) to ~11 μgm⁻³ (1 km).

20 In BCN, the PM₁₀ underestimation is not systematic throughout the daily cycle (Fig. 9),
21 which shows a bias of ~20/10 μgm⁻³ at day/night time. The higher daytime
22 underestimation as compared to the nighttime cannot be explained by the current
23 results, but it could be a result of **missing sources** and problems with PBL
24 overestimation and ~~emission dilutions~~. The resolution increase ~~allows reducing~~
25 by ~1 μgm⁻³ (16%), especially during early morning and late afternoon, when the
26 highest PBL variability between resolutions is ~~detected~~. Although the evaluation of
27 T2M, U10 and WD10 indicates that the resolution increase has a ~~low~~ effect over BCN
28 (Sect. S1), the reduction of the artificial dilution of NO₂ emissions –together with a
29 lower PBL height at 1 km than at 4 km during the night and early morning– **allows**
30 **improving** NO₂, O₃ and PM₁₀ concentrations, which in turn decreases their biases.

31 In AND, the model ~~at both resolutions~~ underestimates observed NO₂ concentrations
32 throughout the daily cycle (~5 μgm⁻³), with the highest underestimation at the morning

1 peak ($\sim 25 \mu\text{gm}^{-3}$) and the lowest at the afternoon peak ($\sim 10 \mu\text{gm}^{-3}$). The resolution
2 increase reduces the bias from **-3.5 to 2** μgm^{-3} (by 43%) and increases r by 7% (from
3 0.39 to 0.41). As in BCN, the NO_2 underestimation directly impacts the O_3 daily cycle
4 (which the resolution increase cannot resolve), increasing the bias by $\sim 1 \mu\text{gm}^{-3}$, a
5 phenomenon that is predominant in the morning hours. In the case of PM_{10} , the daily
6 cycle indicates that the biases are almost systematic throughout the ~~daily cycle~~ (~ 22
7 μgm^{-3}). Increasing the resolution increases the bias by less than 4% in the late
8 afternoon, which is perhaps dominated by the ~~PBL decrease~~. When the resolution is
9 increased, NO_2 performs better because of the improved model performance for the
10 **temperature and wind speed** (Sect. S1), as well as the lower ~~nocturnal~~ and higher ~~diurnal~~
11 PBL. Meanwhile the O_3 and PM_{10} performance do not significantly change.

12 During April 2013, the main flow over MAD was controlled by S-SW synoptic winds
13 channeled by orographic barriers in the NW domain and the Tajo valley (Valverde et
14 al., 2014). The NO_2 daily cycle depicts a high influence of traffic emissions (Fig. 9),
15 showing significant model underestimation at both resolutions for the
16 morning/afternoon peaks ($\sim 15/10 \mu\text{gm}^{-3}$). Note that, in terms of **mean and variability**
17 resulting from southeastern winds, the model performs well at the afternoon peak. NO_2
18 performance leads to more accurate O_3 daily cycles than in AND and BCN, especially
19 in the early morning, when the titration effect of NO_2 is ~~more efficient~~ because the NO_2
20 morning peak underestimation is ~~lower~~ when compared to the other domains.
21 Meanwhile, the modeled PM_{10} at both resolutions presents a profile controlled by traffic
22 emissions. Observed concentrations display a flatter daily cycle, in which the model
23 underestimation reaches $40 \mu\text{gm}^{-3}$ in the morning. Increasing resolution shows a
24 **positive effect** for NO_2 , and PM_{10} increases r by 0.01 and reduces MB and RMSE by
25 $0.1\text{-}0.2 \mu\text{gm}^{-3}$. However, it ~~depicts~~ the lowest variability when compared to the other
26 domains ($<5\%$ for bias, error and r), ~~which is~~ the result of a relatively simpler
27 topography and meteorological patterns.

28 **4.4 Environment and major sources**

29 **Figure 7** shows that the resolution impact also depends on the type of area and the
30 dominant emission source. Theoretically, the meteorological fields of urban areas differ
31 from those of surrounding rural areas because of their different morphology (radiation

1 trapping and wind profiles), surface materials (heat storage) and variable energy
2 consumption (heat release).

3 Increasing resolution reduces the NO₂ bias at suburban and urban stations by 1.8-2 μgm⁻³
4 ³, and, to a lesser extent, by 1.2 μgm⁻³ at rural stations. The correlation coefficients also
5 improve at suburban stations (from 0.48 to 0.52) and rural stations (from 0.34 to 0.35).
6 ~~That is not surprising, because the 1-km grid allows~~ better allocation of land-use
7 categories (urban vs. rural) and of their fraction in a grid cell ~~than does a 4-km grid~~. The
8 NO₂ biases ~~exhibit a relative~~ 39% (65%) ~~decrease~~ at urban (background) stations, but
9 O₃ biases increase by 9% (5%). For PM₁₀, the resolution increase does not significantly
10 change ~~as a function of area type, and it depicts variation in biases and errors~~ of less
11 than ±4% (<0.5 μgm⁻³).

12 The low improvement at urban stations is **obviously** because the NoahLSM land-surface
13 model does not consider the effect of urban morphology or thermal parameters in order
14 to accurately model meteorological fields. Modeling air quality on an urban scale over
15 cities requires a description of the heat/momentum exchange between buildings and the
16 lower atmospheric layers. For instance, the impact of using an urban model on
17 meteorological fields over the greater Paris area was studied by [Kim et al. \(2013\)](#) using
18 WRF with the Urban Canopy Model, demonstrating that, ~~below a 1000-m height~~,
19 overestimations of wind speed were significantly reduced.

20 ~~The r-effect-of~~ increasing resolution is **positive** for primary pollutants near important
21 emission sources. For example, it reduces NO₂ biases at traffic (industrial) stations by
22 ~3 μgm⁻³ (2 μgm⁻³), but it increases O₃ biases by ~2 μgm⁻³ (1 μgm⁻³). However, the
23 resolution increase in the range of 4-1 km does not exhibit the expected improvement
24 on the hourly statistics that are based on the constraints of the current model
25 formulation. In other words, it cannot resolve the subgrid air quality variability merely
26 by increasing resolution. For instance, although on-road traffic emissions are estimated
27 by following a bottom-up approach along highways and routes, heterogeneity is lost in
28 the CTM volume averaging process, which artificially dilutes emission rates over the
29 grid cells. The resolution effect is ~~the~~ lowest at background stations, which ~~are~~ not
30 influenced by any single source, but rather by the integrated contribution from all
31 sources upwind of the stations where variations are less than 1% for O₃ and PM₁₀ (< 1

1 μgm^{-3}). However, background NO_2 levels increase by $\sim 1 \mu\text{gm}^{-3}$ (48%) from 4 km to 1
2 km.

3 **Figure 10** shows the temporal series and daily cycles for NO_2 and O_3 at traffic and
4 background stations throughout the episode of 12-18th April, 2013. At traffic stations,
5 the temporal series show a remarkable O_3 daily cycle (observed 25p = $23.2 \mu\text{gm}^{-3}$ and
6 75p = $77.5 \mu\text{gm}^{-3}$), due to O_3 destruction caused by high NO_x levels (observed 50p =
7 $34.5 \mu\text{gm}^{-3}$). In contrast, the NO_2 -limited regime at background sites (observed 50p =
8 $19 \mu\text{gm}^{-3}$) allows higher O_3 concentrations (observed 25p = $38 \mu\text{gm}^{-3}$ and 75p = 89
9 μgm^{-3}) than in high NO_2 environments.

10 During the episode mentioned above, the resolution increase at traffic stations had a
11 positive effect by increasing the correlation coefficient for NO_2 (from 0.73 to 0.76) and
12 O_3 (from 0.83 to 0.86), and also by decreasing the NO_2 mean bias by $\sim 5 \mu\text{gm}^{-3}$ (from 6
13 to $1 \mu\text{gm}^{-3}$). The NO_2 daily cycle improves in the morning hours after sunrise, reducing
14 bias by $5\text{-}10 \mu\text{gm}^{-3}$ and contributing to a reduction in O_3 overestimations ($\sim 5 \mu\text{gm}^{-3}$). In
15 contrast, at background stations, where the NO_x/O_3 chemistry is less dominant, the
16 resolution effect is not significant. Such behavior indicates that finer resolution
17 improves the performance, because horizontal resolution affects the representation of
18 chemical processes near large emission sources, such as the efficient formation of O_3
19 and nighttime O_3 titration (Mathur et al., 2005). However, the loss of subgrid variability
20 and improved meteorological fields (transport and temperature) are required.

21 **5 Conclusions**

22 The present work shows the effects of increasing the horizontal resolution from 4 km to
23 1 km using the CALIOPE-AQFS on pollutant concentrations (NO_2 , O_3 , and PM_{10}) over
24 three Spanish domains (AND, BCN and MAD) in April 2013.

25 The global features of concentration maps at both resolutions are quite similar, with
26 zones of high/low concentration identically located, which is obviously because both
27 simulations are based on the same emission dataset. Further comparisons demonstrate
28 that increasing the resolution provides better-defined and more realistic concentration
29 structures over large sources (roads and industries) and complex terrains (more sharply
30 defined orographic hills). The titration effect on O_3 concentrations along highways and
31 major point sources is more evident in 1-km simulations than at 4 km, because the latter

1 ~~is affected by higher dilution within the grid cells.~~ This improvement is quantified by an
2 increase in spatial correlation coefficients of 3% (6%) for NO₂ (O₃).

3 However, the resolution increase in the range of 4-1 km does not exhibit the expected
4 improvement in hourly statistics for any pollutant. Hourly correlation coefficients do
5 not significantly change, and absolute (relative) errors and biases vary < 2 μgm⁻³ (9%).
6 The merit of the resolution increase may be underrated when classical statistics are
7 applied at measurement stations (Mass et al., 2002; Gego et al., 2005). For instance,
8 although the structure of important NO₂ urban plume features (> 40 μgm⁻³) often
9 become more realistic (stronger and more defined plumes) as resolution increases,
10 statistics are deeply degraded by even small timing and spatial errors.

11 The resolution increase has a significant impact on reducing NO₂ hourly bias (by 42%,
12 2 μgm⁻³), without any significant change in the error and the r (<2%), ~~but it increases~~ O₃
13 hourly biases (<1 μgm⁻³). The main differences between resolutions appear at daytime
14 and nighttime traffic peaks, when the mixing height experiences rapid changes,
15 allowing the 1-km simulation to slightly reduce NO₂ underestimation in the morning by
16 ~5-10 μgm⁻³. The O₃ daily cycles **at large sources** depict a high influence of hourly NO₂
17 concentrations, increasing the hourly O₃ bias by ~3 μgm⁻³. ~~That~~ behavior is controlled
18 by the daytime O₃ underestimation and, to a lesser extent, by the nighttime
19 overestimation. The resolution increase ~~allows reducing~~ the O₃ overestimations at night
20 (by ~5 μgm⁻³), partly because of higher nocturnal NO₂ concentrations.

21 ~~Concerning the capability of forecasting 75p exceedances in the observed maximum 1h~~
22 ~~concentrations, the increased resolution has a positive effect: it increases the~~ number of
23 hits that forecast 75p exceedances in the observed Max 1h NO₂ (537 vs. 466 over 953
24 exceedances), ~~and it reduces the false alarms for Max 1h O₃ exceedances (FAR~~
25 ~~improves by 37%).~~

26 The grid effect is less pronounced for PM₁₀ than for NO₂ and O₃. When the resolution
27 increases, the **low increment** of PM₁₀ mean concentrations (<0.1 μgm⁻³) is the result of
28 compensating biases of PM₁₀ components, which is controlled mainly by the PPM and
29 EC increase as well as the SS decrease.

30 BCN is the domain where the resolution increase has the **highest** effect, with changes in
31 bias (error) of 16-73% (< 5%), followed by AND with 4-43% (< 5%) and MAD < 3-5%

1 (< 1%). In BCN, as in the western Mediterranean Basin, the transport of O₃ and its
2 precursors is governed by mesoscale circulation. In that sense, the resolution increase
3 has a ~~great~~ impact over BCN, where induced mesoscale phenomena control the air flow;
4 ~~meanwhile~~ synoptic transport is more prominent in MAD and AND. The benefits of
5 increasing the resolution to 1 km over rural areas (Mass et al., 2002) are that it **increases**
6 **the accurate representation of mesoscale meteorological structures such as orographic**
7 **wind and circulation**. Over urban areas along the western Mediterranean coast (Toll and
8 Baldasano, 2002; Jiménez et al., 2006; Fay and Neunhäuserer, 2006), further
9 improvements and **urbanization steps** are required before seeing any benefits in
10 increasing the resolution to 1 km.

11 In urban areas ~~or~~ near large emission sources (industrial and traffic stations), NO₂ and
12 O₃ concentrations are more sensitive to changes in the grid resolution. The
13 concentration increase in primary anthropogenic pollutants (NO₂, PPM and EC) is
14 obvious because **the high resolution allows better allocation of emissions at point,**
15 **linear and area sources.** ~~What is more, it decreases~~ the artificial dilution of emissions
16 when compared to the larger grid area. **However, the 1-km simulation attempt to more**
17 **accurately describe the chemical formation of O₃ and dilution of NO₂ concentrations**
18 **over those areas was not generally successful.**

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

1 until mixing conditions allow the pollutants to disperse above these urban canopy levels
2 (Kim et al., 2013; Ching, 2013). Third, the USGS land-use data used in the WRF model
3 is based on 1993 data, and urban changes in MAD and BCN over the last 20 years are
4 significant.

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

1 **Appendix A**

2 Table A1. Definition of the discrete statistics used in the evaluation. Where $C_m(x, t)$ and
 3 $C_o(x,t)$ are the modeled and observed concentrations at a location (x) and time (t); N is
 4 the number of pairs of data. \bar{C}_m and \bar{C}_o are the modeled and observed mean
 5 concentrations over the whole period, respectively.

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

6

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

Statistic	Formula	
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

1

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4

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 (399x399 cells)	AND-1km (366x358 cells)	BCN-1km (146x146 cells)	MAD-1km (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 modeling	256 CPU/210 min	256 CPU/220 min	128 CPU/150 min	128 CPU/110 min

5

6

1 Table 2. Discrete and categorical statistics for NO₂, O₃, O₃-8h, and PM₁₀ for April 2013
 2 as a function of horizontal resolution (4 km and 1 km). n indicates the number of pairs
 3 of data used in the discrete evaluation on an hourly basis. OM and MM depict the
 4 measured and modeled mean concentrations, respectively. T is the threshold applied in
 5 the categorical evaluation. Max 1h and mean 24h concentrations are calculated by
 6 considering ≥75% of the hours in a day, as established by Directive 2008/50/EC.

	NO ₂	O ₃	O ₃ -8h	PM ₁₀
n (stations)	90761 (135)	76471 (114)	3248 (114)	66642 (100)
OM (μgm ⁻³)	22.0	68.4	88.6	20.6
EU LV/TV (μgm ⁻³) (temp basis)	200 (Max 1h)	180 (Max 1h)	120 (Max 8h)	50 (Mean 24h)
EU LV/TV exceedances	0	25	0	31
T* (μgm ⁻³) (temp. basis)	71 (Max 1h)	108 (Max 1h)	101 (Max 8h)	27 (Mean 24h)

Discrete evaluation

	4 km	1 km						
MM (μgm ⁻³)	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 (μgm ⁻³)	-4.5	-2.6	-10.5	-11.3	-16.3	-17.2	-6.7	-6.6
MAE (μgm ⁻³)	12.9	13.2	19.7	19.8	18.4	19.2	12.6	12.7
RMSE (μgm ⁻³)	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

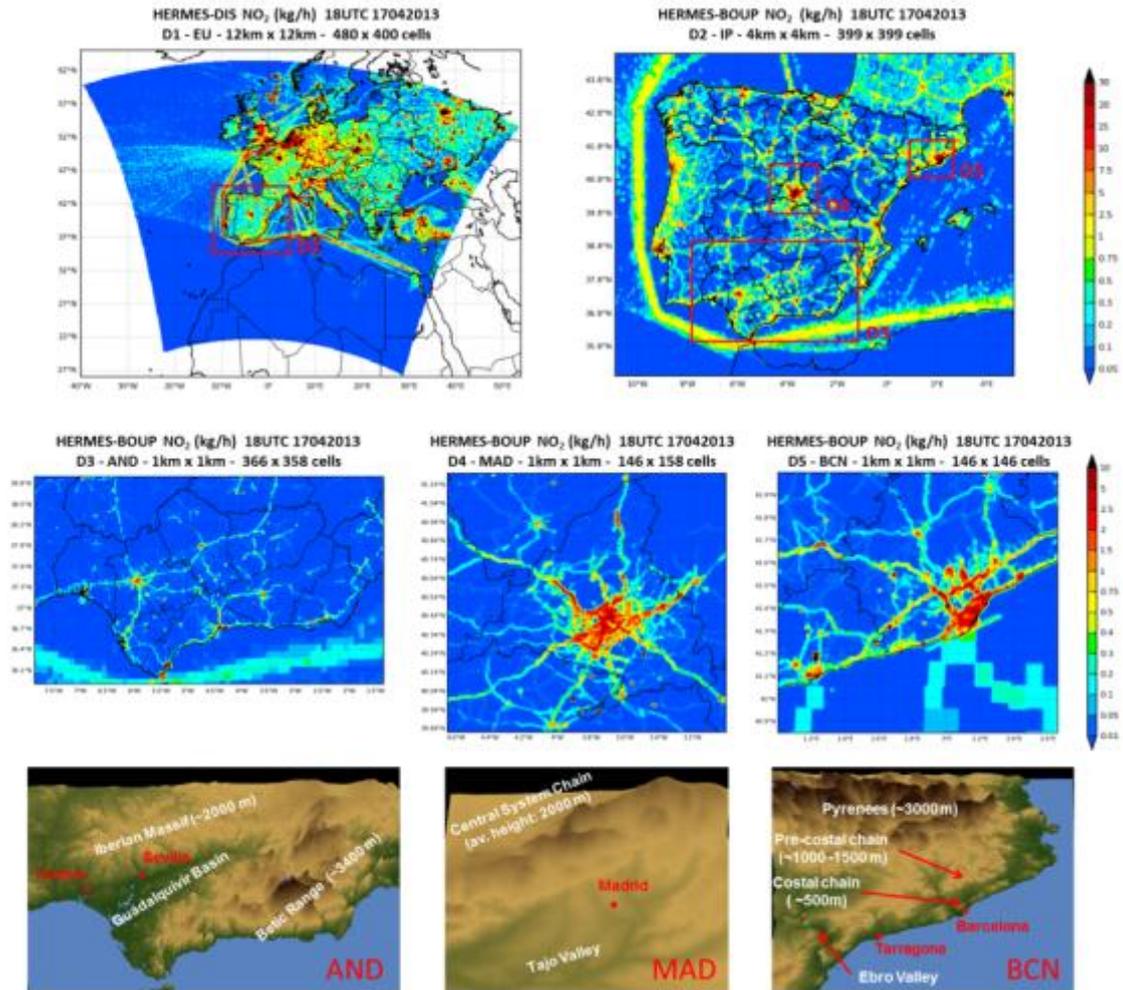
Categorical evaluation (Threshold = T*)

	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

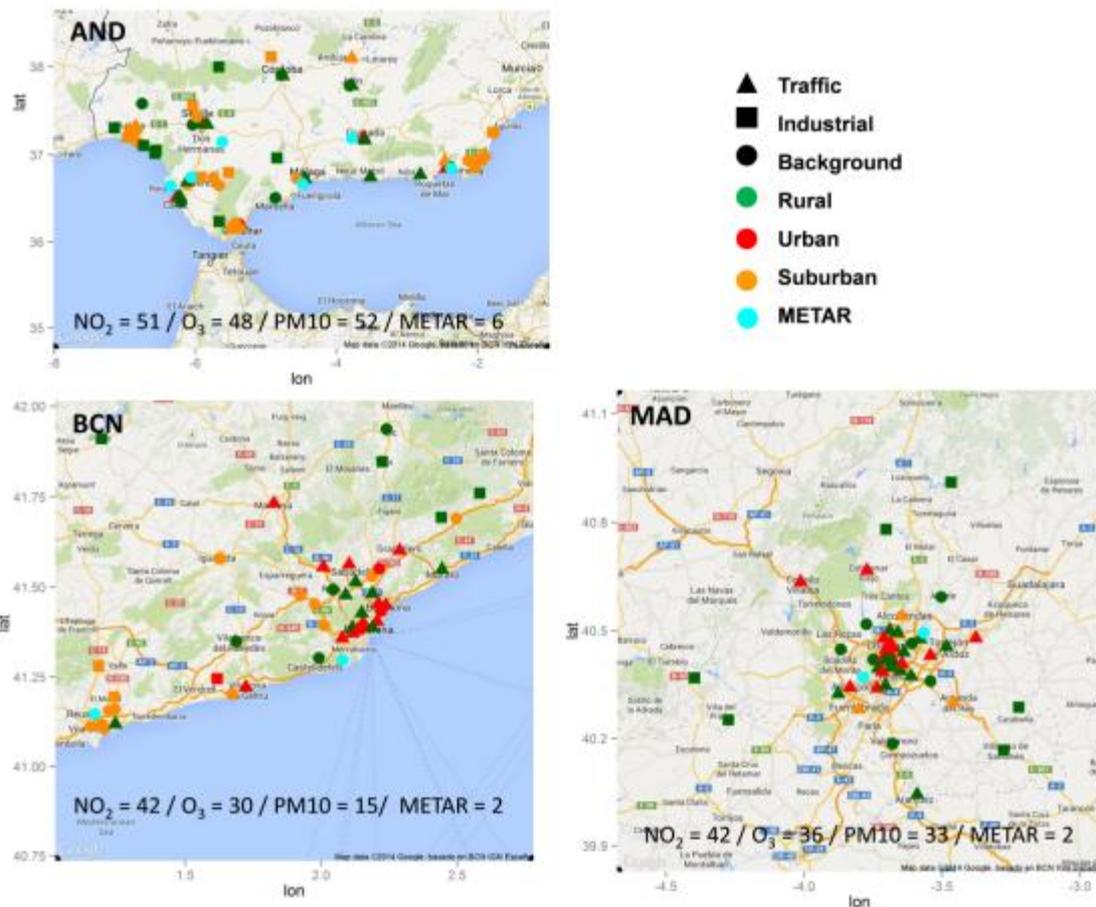
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2 Fig. 1. CALIOPE-AQFS nesting strategy (D-domains) and study domains (Andalucia,
 3 AND; Madrid, MAD; and Barcelona, BCN). Colour chart at D-domains shows NO₂
 4 emission rate (kg h⁻¹) for 17th April, 2013 at 18UTC. HERMES-DIS model generates
 5 emissions at 12 km x 12 km over Europe (the mother domain, D1) by performing
 6 disaggregation from the EMEP database. HERMES-BOUP model estimates emissions
 7 at 1 km x 1 km, following a bottom-up approach.

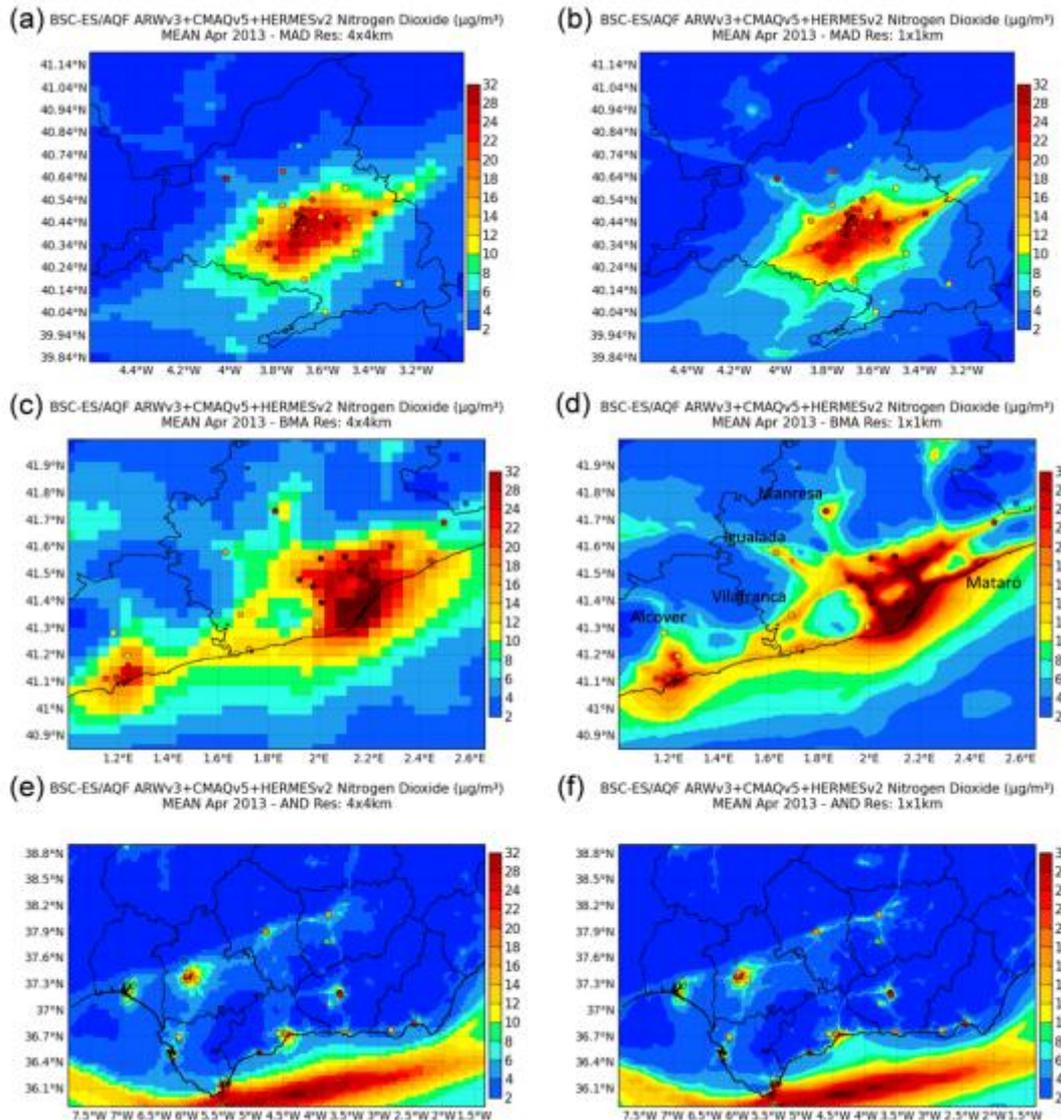
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2 Figure 2. Air quality stations for NO_2 , O_3 and PM_{10} in the three domains under study
 3 (AND, BCN and MAD) in April 2013. Different types of stations are shown by symbols
 4 and color codes. The various symbols represent the major emission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

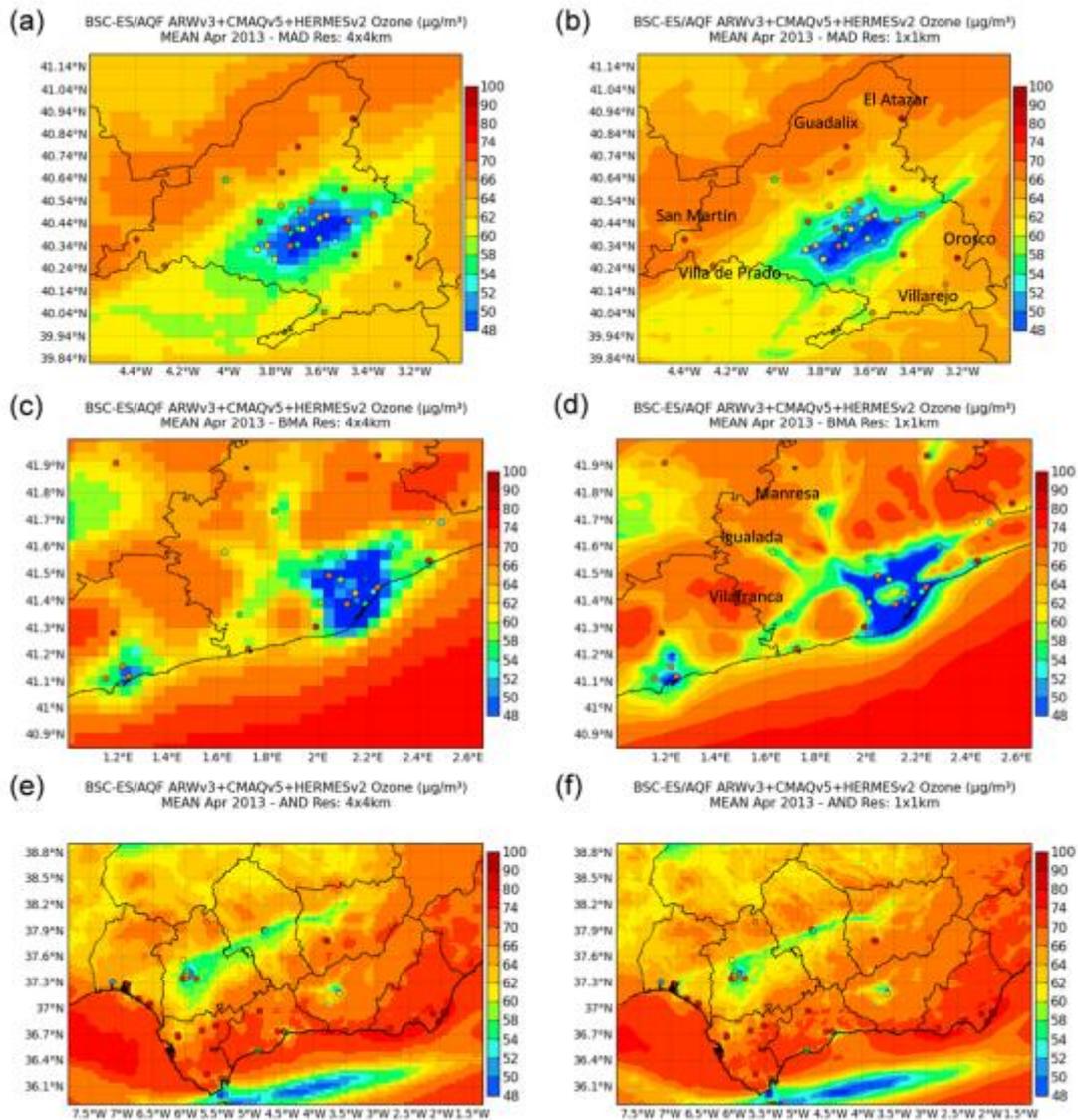
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2 Figure 3. CALIOPE-AQFS mean NO₂ concentration (µg⁻³) in April 2013 over (a,b)
 3 MAD, (c,d) BCN, and (e,f) AND, as a function of horizontal resolution: 4 km (left
 4 column) and 1 km (right column). Dots indicate mean concentration at air quality
 5 stations.

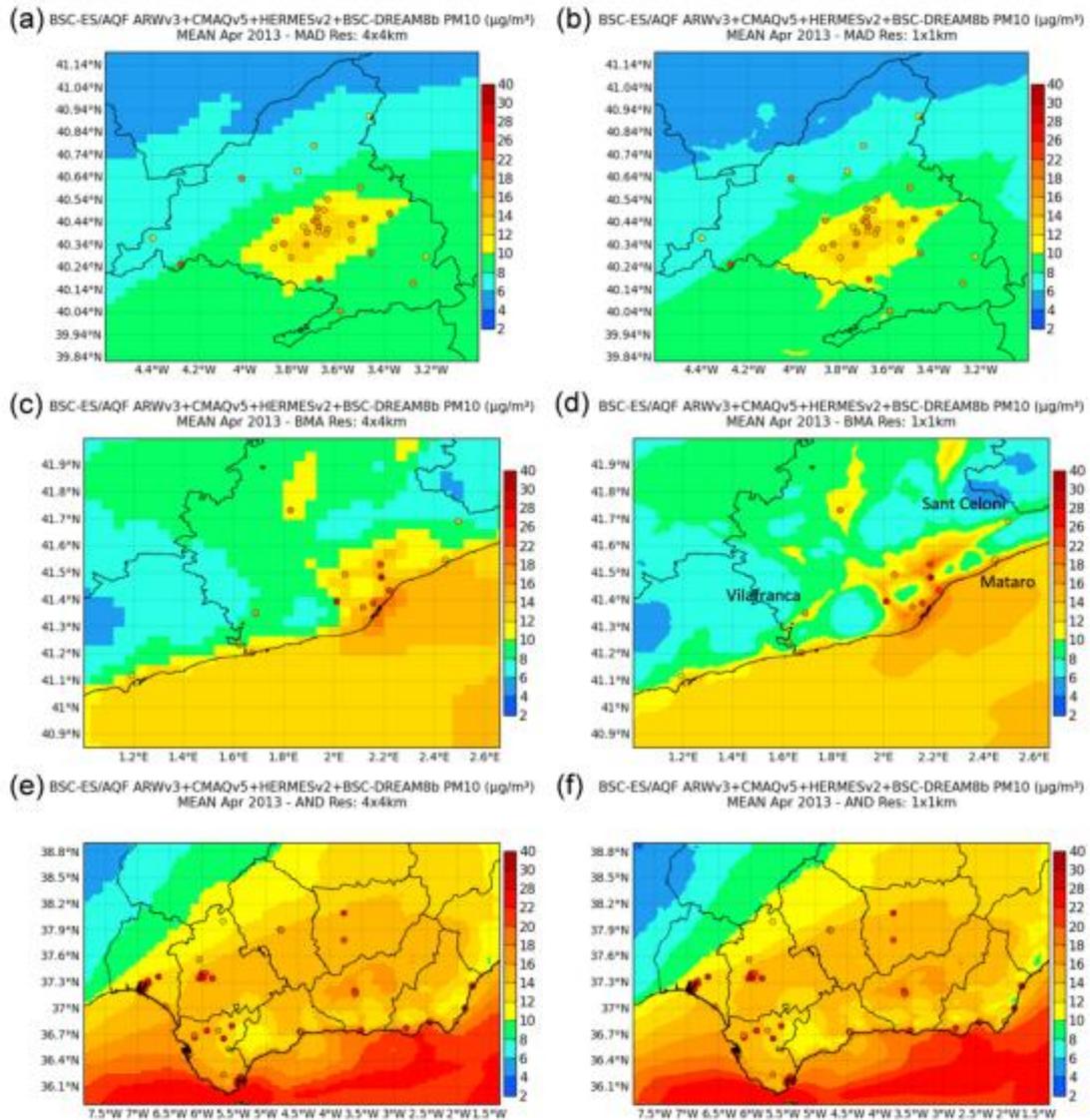
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2 Figure 4. CALIOPE-AQFS mean O₃ concentration (µg⁻³) in April 2013 over (a,b)
 3 MAD, (c,d) BCN, and (e,f) AND, as a function of horizontal resolution: 4 km (left
 4 column) and 1 km (right column). Dots indicate mean concentration at air quality
 5 stations.

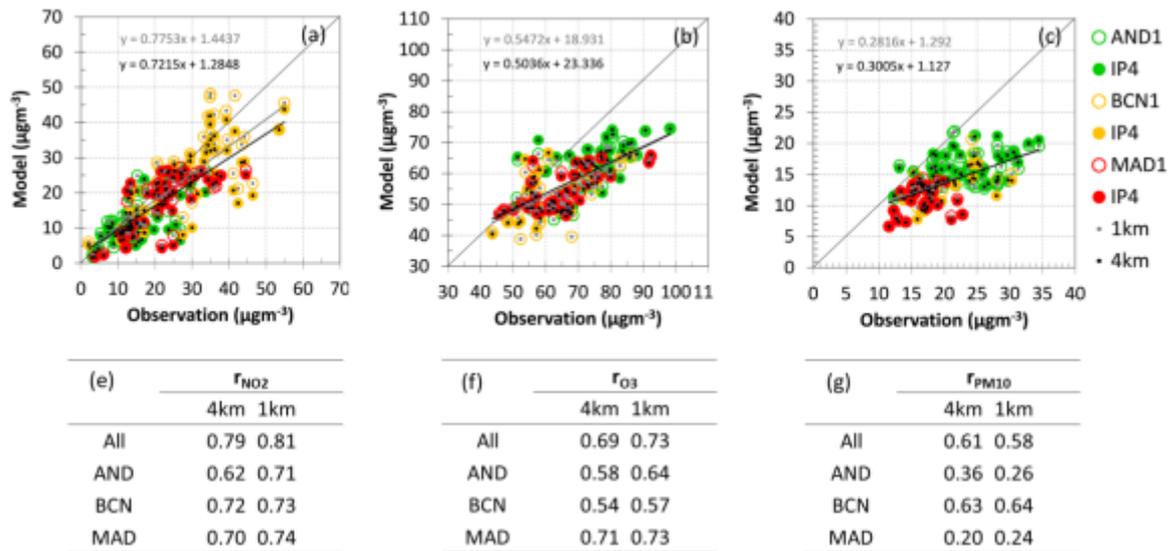
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2 Figure 5. CALIOPE-AQFS mean PM_{10} concentration ($\mu g m^{-3}$) in April 2013 over (a,b)
 3 MAD, (c,d) BCN, and (e,f) AND, as a function of horizontal resolution: 4 km (left
 4 column) and 1 km (right column). Dots indicate mean concentrations at air quality
 5 stations.

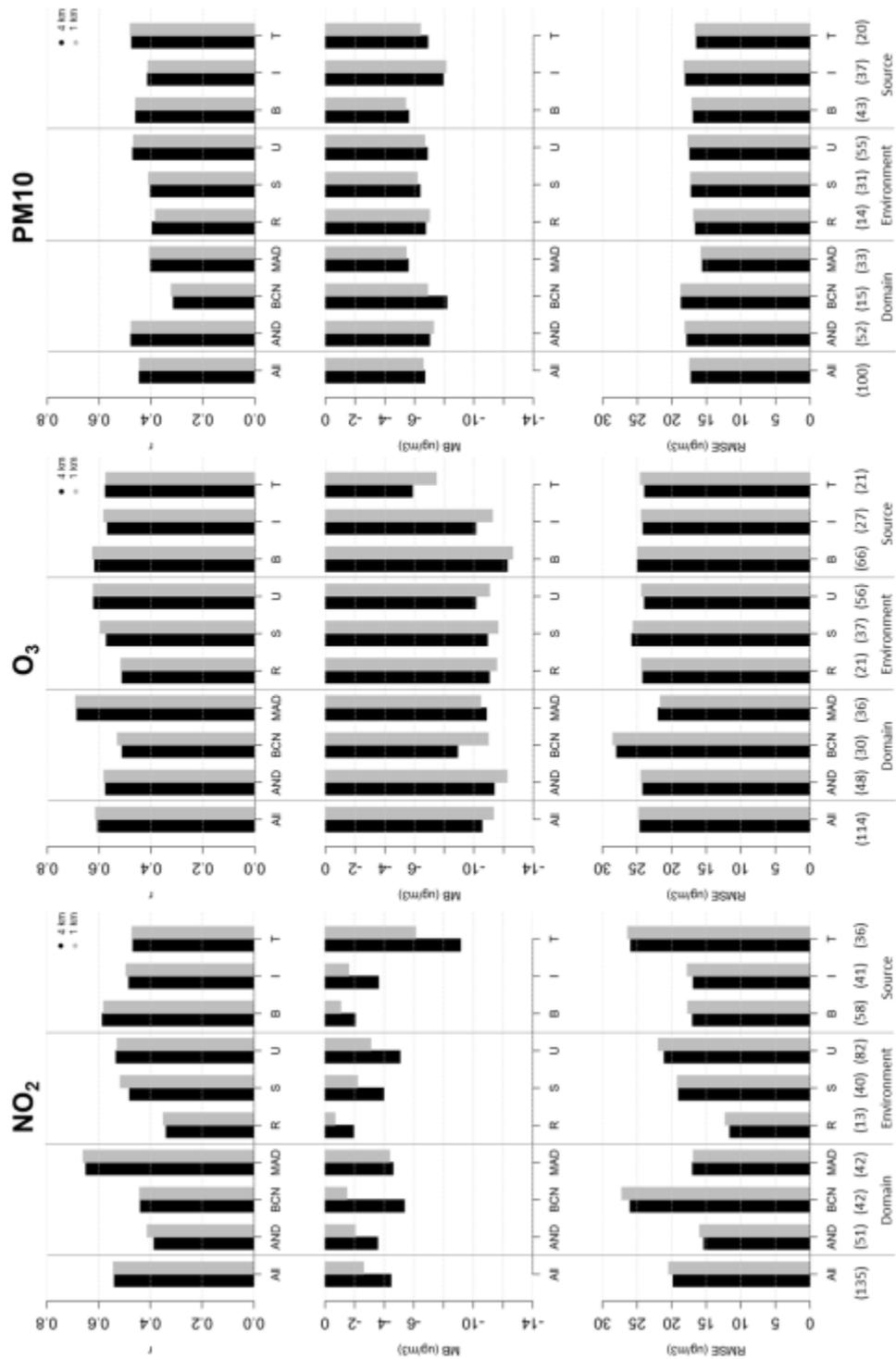
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2 Figure 6. Monthly mean scatter plots for CALIOPE-AQFS (y-axis) and observed (x-
 3 axis) concentrations for the three study domains (AND in green, BCN in yellow, and
 4 MAD in red), as a function of horizontal resolution for (a) NO₂, (b) O₃ and (c) PM₁₀.
 5 Equations show the linear adjustment between models and observations at 1 km (light
 6 grey) and 4 km (dark grey). Spatial correlation coefficients as a function of resolution
 7 and domain are shown for (e) NO₂, (f) O₃, and (g) PM₁₀.

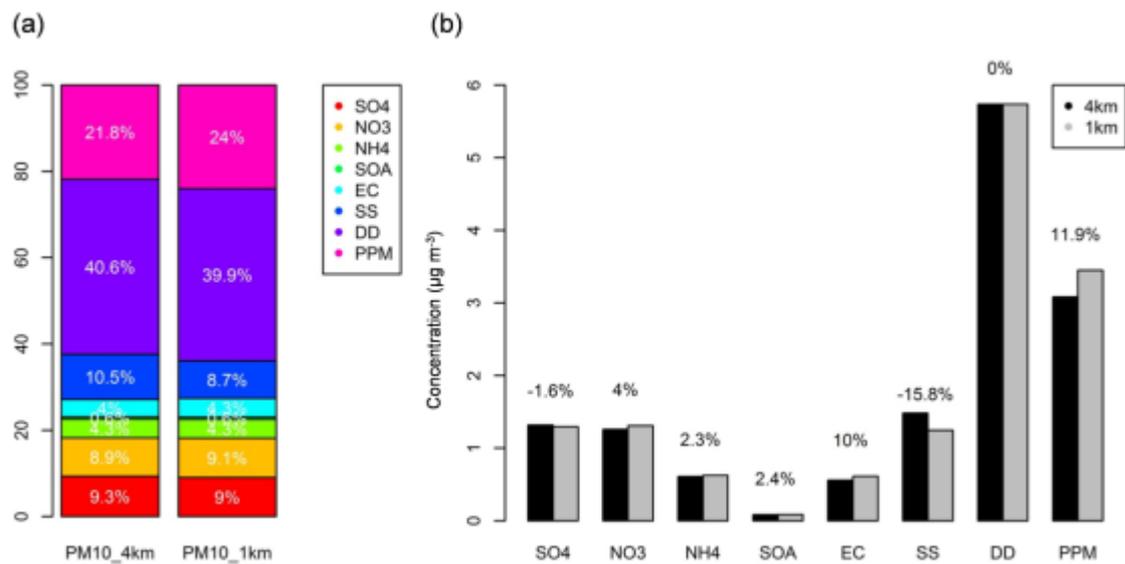
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2 Figure 7. Statistics (r , MB, and RMSE in rows) for each pollutant (NO_2 , O_3 , and PM_{10})
 3 in columns) on an hourly basis as a function of horizontal resolution: 4 km (black) and 1
 4 km (grey). Four categories are considered: all stations (all), domain (AND, BCN and
 5 MAD), station environment (R, S, and U), and main sources (B, I, and T).

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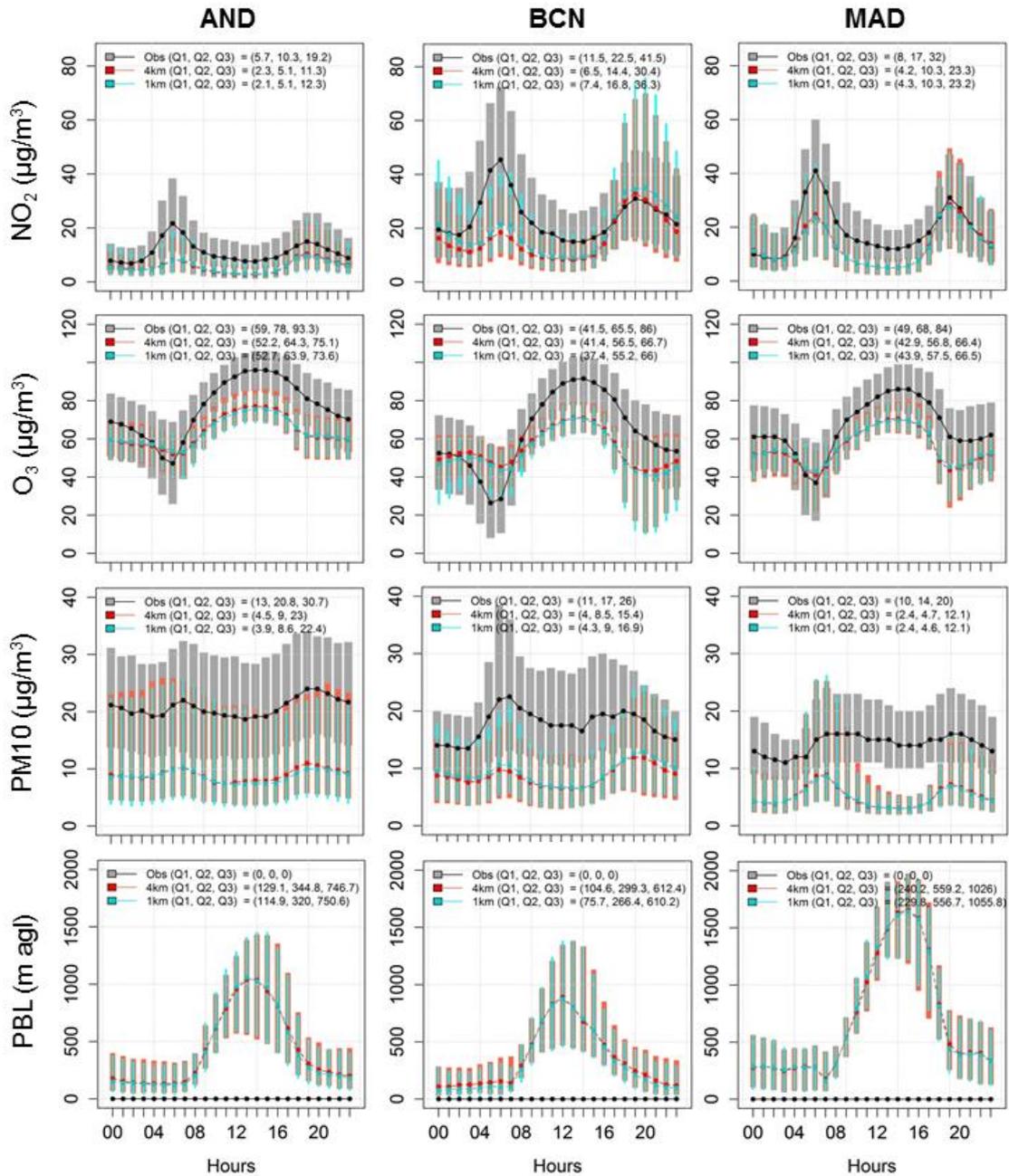


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2 Figure 8. Resolution effect on PM₁₀ components in April 2013. (a) Percentage of PM₁₀
 3 components: sulfate (SO₄), nitrate (NO₃), ammonium (NH₄), secondary organic
 4 aerosol (SOA), elemental carbon (EC), sea salt (SS), desert dust (DD), and primary
 5 particulate matter (PPM). (b) PM₁₀ component concentrations in the 1-km simulation
 6 (black) and 4-km simulation (grey). Numbers over bars indicate the % of increase when
 7 increasing resolution.

8

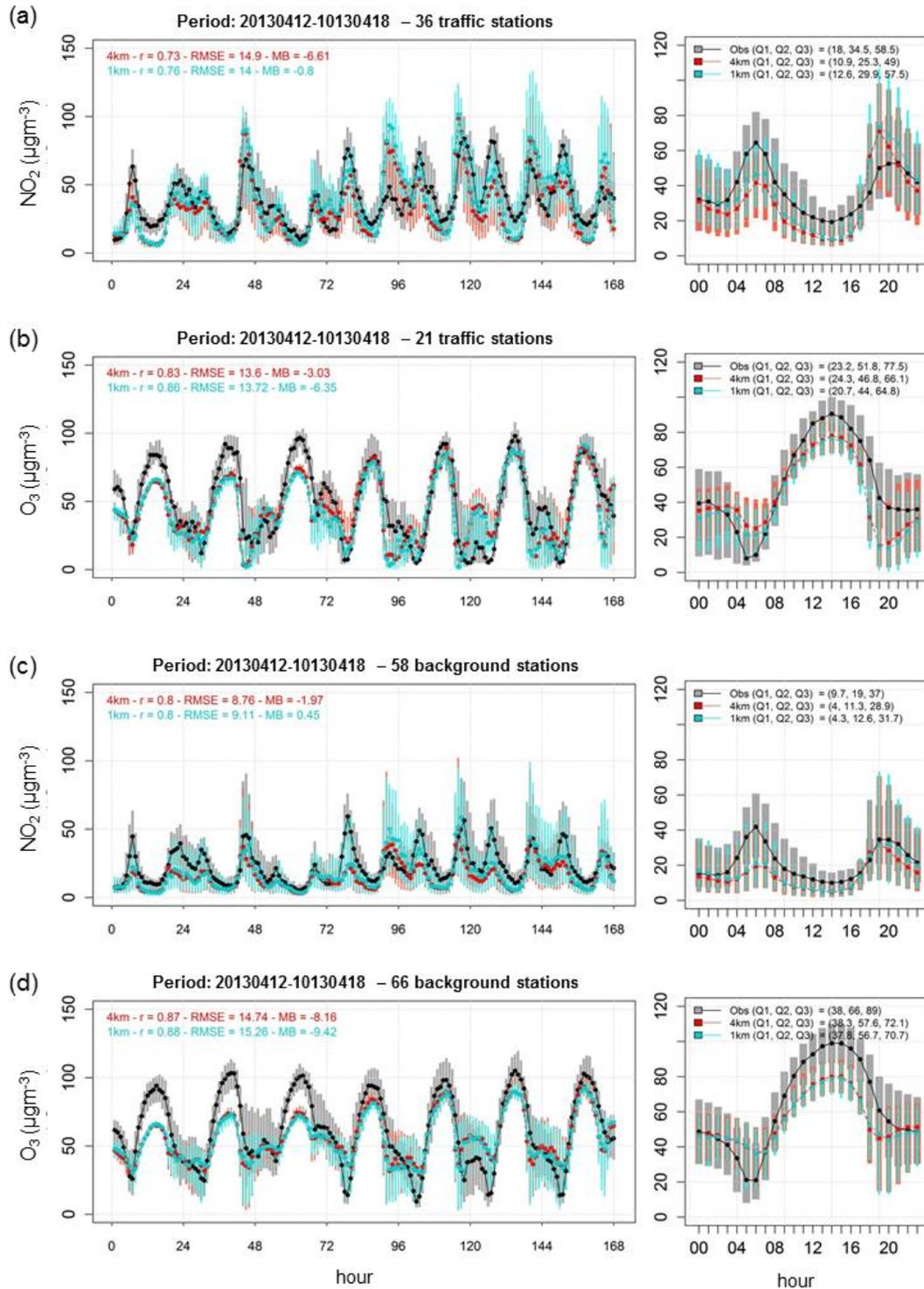
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2

3 Figure 9. Daily cycles for NO_2 , O_3 and PM_{10} for each study domain at available stations
 4 as a function of resolution. No observations of PBL are available. Q1, Q2 and Q3
 5 indicate quartiles for the daily cycle. Bars show Q1 and Q3 at each hour.

6



1

2 Figure 10. Temporal series and daily cycles for NO₂ and O₃ at background (a and b,
 3 respectively) and traffic stations (c and d, respectively) for the episode of 12-18th April,
 4 2013. Q1, Q2 and Q3 indicate quartiles for the daily cycle. Bars show Q1 and Q3 at
 5 each hour.