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Spatio-temporal variability of concentrations and speciation of particulate matter across Spain in the CALIOPE modeling system

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ABSTRACT

The CALIOPE high-resolution air quality modeling system (4 km \times 4 km, 1 h) estimates particulate matter from two aerosol models, CMAQv4.5 (AERO4) and BSC-DREAM8b. While CMAQv4.5 calculates biogenic, anthropogenic and sea-salt aerosols; BSC-DREAM8b provides hourly estimates of the natural mineral dust contribution from North Africa deserts. This paper presents an evaluation of the CALIOPE system to reproduce the spatial and temporal variability levels of PM2.5, PM10 and chemical composition (nitrate, non-marine sulfate, ammonium, organic and elemental carbon, sea-salt, and desert dust) across Spain. The evaluation is performed against ground-based observations for the year 2004, when a number of time series of chemically speciated compounds were available. A new data set of Saharan dust PM10 concentration is used to evaluate the PM10 contribution modeled by BSC-DREAM8b. The results indicate that both natural aerosol sea-salt and desert dust accomplish the model performance criteria (MFE < 75% and MFB \pm 60%). Modeled PM10 sea-salt is highly dependent on wind speed and presents high correlation with experimental data in coastal areas (r = 0.67). The BSC-DREAM8b is able to reproduce the daily variability of the observed levels of desert dust and most of the outbreaks affecting southern Spain. Species in the equilibrium (e.g. sulfate/nitrate/ammonium) are highly correlated each other and show high dependency on ammonia emissions. Non-marine sulfate and ammonium are underestimated by a factor of 3. An underestimation of nitrate was also seen (factor of 2). Fine carbonaceous aerosols present the highest underestimations (factor of 4) in part related to the state-of-the-science concerning secondary organic aerosol formation pathways. Spatial and seasonal variability of PM2.5, PM10 and their chemical compounds increase the correlation with observations when multiplicative bias-correction factors for the aforementioned underestimated species are taking into account. Furthermore, simulated spatial and seasonal patterns of aerosol agree with those described in related studies based on experimental values.

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1. Introduction

Particulate matter (PM) is of major scientific interest due to its demonstrated impact on human heath (WHO, 2006), ecosystems (Lovett et al., 2000), Earth's climate (Jacobson, 2001; Ramanathan et al., 2001; IPCC, 2007) and visibility (Hyslop and White, 2008). Although air quality in Europe has improved substantially over the past decades, air pollution by PM is a major threat to human health (EEA, 2010). Not only do concentrations determine the effects of particles on human health, but also (1) their size, conditioning the degree of penetration into the respiratory system, (2) their composition, since the toxicity of the particles depends on the occurrence of specific hazardous PM compounds (Brauer et al., 1995) and (3) time of exposure. Although the effects of long-term exposure are much more uncertain than the effects of short-term exposure, they are believed to have a greater effect on health (Dockery et al., 1993; Pope et al., 2002). Recent studies carried out in Spain demonstrate that the levels of the finest PM fractions are important risk factors for daily cardiovascular mortality (Pérez et al., 2009; Maté et al., 2010) increasing the number of hospital admissions due to pollution caused by fine-particles (Linares and Díaz, 2010a,b). Current European standards on air quality, driven by the directive 2008/50/EC (European Commission, 2008), set an annual limit value for PM10 (particles below 10 μ m in diameter) of 40 μ g m⁻³; and a daily mean value of 50 μ g m⁻³ should not be exceeded 35 times per year. For the

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first time, a European directive introduces a limit value for PM2.5 (particles below 2.5 μ m in diameter) in the air quality standards to be accomplished in two stages: Stage I, a mean maximum annual limit value of 25 μ g m⁻³ (deadline: 2015); and Stage II, a mean maximum annual limit value of 20 μ g m⁻³ (deadline: 2020).

A number of studies (e.g. Sanz et al., 2000; Dueñas et al., 2002; Millán et al., 2002; Querol et al., 2004, 2008; Ribas and Peñuelas, 2004; Escudero et al., 2007a) show that the Iberian Peninsula (IP) exceeds some of the thresholds of air quality established in the legislation. Exceedances of the PM10 limit values were not only caused by anthropogenic emissions but also the long-range transport of atmospheric mineral dust from deserts. The Sahara desert is one of the most important dust sources on the Planet, with northern Africa being responsible for half of the global emissions of mineral dust (Middleton and Goudie, 2001; Prospero et al., 2002). In the IP, the mineral fraction of PM comes mainly from local resuspension and external contributions such as the Saharan/Sahelian dust (Querol et al., 1998, 2001; Artíñano et al., 2001; Rodríguez et al., 2001, 2002).

Urban to regional air quality models play a key roles in assessing and understanding the dynamics of air pollution, forecasting the air quality and developing emissions abatement plans and alerting the population when health-related issues occur (European Commission, 2008). Examples of air quality forecasting systems currently operating in Europe are compiled in Menut and Bessagnet (2010) and in Balk et al. (2010), the latter in the framework of the COST ES0602 action (www.chemicalweather.eu), which also complement other air quality European initiatives such as the Global Monitoring for Environment and Security (GEMS, http://www.gmes.info/), PROtocol MOniToring for the GMES Service Element (PROMOTE, http://www. gse-promote.org/) and Monitoring Atmospheric Composition and Climate (MACC, htpp://www.gmes-atmosphere.eu/). Regional air quality models are useful tools to describe the continental transport of air pollutants. However, a higher spatial resolution is required to evaluate the exposure of the population to PM pollution in urban areas, especially over complex topography such as over the IP (Jiménez et al., 2006). Up to now, there are several evaluation studies demonstrating the difficulty for models to simulate PM10 and PM2.5 concentration over Europe (Pay et al., 2010a and references therein). Most of them deal with regional background stations, mainly from EMEP network, and quantify the model performance to reproduce regional background aerosols. Overall these show a large gap between simulated and observed PM10 concentration (broadly a factor of 2) due to deficiencies in the physical and chemical understanding of the processes which lead to the formation of secondary aerosol and to the lack of accurate particulate matter emission inventories. It is worth noting that only a few published works (Hodzic et al., 2005; Flaounas et al., 2009; Chemel et al., 2010) report on performance characteristics of long-term aerosol simulations over urban domains hampered by the lack of measured PM compounds.

The CALIOPE modeling system is a state-of-the-art modeling system, specially developed with high spatial ($4 \text{ km} \times 4 \text{ km}$) and temporal resolution (1 h) to forecast air quality across Spain. The modeling system was first evaluated in Pay et al. (2010a) for Europe, and Baldasano et al. (2011) for Spain, including gas and particulate pollutants over the full year 2004. The purpose of the present paper is to complement these results by (1) providing a detailed quantitative analysis of the capabilities of the CALIOPE modeling system to simulate daily aerosol distribution over Spain in terms of concentration, size and chemical composition, and (2) estimating and assessing the spatial and seasonal distribution of the different aerosols over Spain based on comparing model outputs and observations. The study is carried out for the full year 2004 for two main reasons: (1) the HERMES emission database was compiled for this year and (2) a number of data sets on levels of PM

compounds are available for this year. An important novel approach of this work is the evaluation of the BSC-DREAM8b model using an experimental data set of African PM10-dust obtained with the methodology described in Escudero et al. (2007b). In addition, measurements of chemically speciated aerosol are available in a long time series for 2004 for different environments, which allow evaluating the skill of CALIOPE to reproduce PM in urban/suburban/ rural areas.

The outline of the paper is as follows. Section 2 describes the CALIOPE system and the observational data sets used for the evaluation. In Section 3, measurements of aerosol size and composition from IDAEA-CSIC/EMEP are used to identify the origin of the discrepancies in modeled aerosol and to quantify the deviation from observations. It also estimates and assesses the spatial and seasonal distribution of aerosols over Spain, for both size and composition, taking into account the discrepancies outlined in the previous section. Finally, conclusions and suggestions for further work are given in Section 4.

2. Methods

2.1. Modeling system formulation

The CALIOPE project has established an air quality forecasting system for Spain (Baldasano et al., 2008a). Current forecasts and near real-time evaluation are available through http://www.bsc.es/caliope. CALIOPE encompasses a high-resolution air quality forecasting system being applied to Europe as a mother domain (12 km \times 12 km, 1 h) (Pay et al., 2010a) and Spain as the nested domain (4 km \times 4 km, 1 h) (Baldasano et al., 2011).

Meteorological input data for the chemistry transport model are calculated using the Advanced Research Weather Research and Forecasting (WRF-ARW) model v3.0.1.1 (Michalakes et al., 2004; Skamarock and Klemp, 2008). For the Spanish domain WRF-ARW is configured with a grid of 397×397 points corresponding to a 4 km \times 4 km horizontal resolution and 38σ vertical levels with 11 characterizing the planetary boundary layer (PBL). The model top is defined at 50 hPa to resolve the troposphere–stratosphere exchanges properly.

The Models-3 Community Multiscale Air Quality Modeling System (Models-3/CMAQ, Byun and Schere, 2006), is a threedimensional Eulerian photochemical transport model that uses state-of-the-science routines to model the behavior of air pollutants from regional to local scales, due to its generalized coordinate system and its advanced nesting grid capability. CMAQ version 4.5, used in this study, has been extensively evaluated under various conditions and locations in the US (Appel et al., 2007, 2008; Roy et al., 2007), European continent (Matthias, 2008; Pay et al., 2010a), and Spain (Jiménez et al., 2006; Baldasano et al., 2011). It includes gas, aerosol and heterogeneous chemistry. According to the work by Jiménez et al. (2003) the photochemical mechanism used in this study is the Carbon Bond IV mechanism (CBM-IV, Gery et al., 1989).

The aerosols are modeled using the AERO4 module (Binkowski and Roselle, 2003). This module comprises the following aerosol compounds: nitrate, sulfate, ammonium, elemental and organic carbon (with three subcomponents: primary, secondary anthropogenic and secondary biogenic), soil, sodium, and chlorine. Unspecified anthropogenic aerosols and aerosol water are additionally kept as separate components. Aerosols are represented by three size modes (Aitken, accumulation- and coarse-mode) each of them assumed to have a lognormal distribution. The coarse particles mode is treated as dry and chemically inert with a fixed geometric standard deviation of 2.2 (Kelly et al., 2010), thus there is no dynamic interaction between the fine and coarse-mode. Secondary inorganic aerosols (SIA) are generated by nucleation processes from their precursors to form nitrate ammonium and non-marine sulfate aerosols. The thermodynamic equilibrium between gas and inorganic fine aerosols is determined by the ISORROPIA model (Nenes et al., 1998). Secondary organic aerosol (SOA) can be formed from aromatics (anthropogenic organic aerosols) and terpenes (biogenic organic aerosols) (Schell et al., 2001). The production of sea-salt aerosol (SS) is implemented as a function of wind speed and relative humidity (Gong, 2003; Zhang et al., 2005) through the AERO4 aerosol module. Aerosol deposition is treated by a secondgeneration deposition velocity scheme (Binkowski and Shankar, 1995; Venkatram and Pleim, 1999). In this study simulated PM concentrations do not include the water mass and Stokes diameter are used for characterizing aerosol geometry.

Emissions are processed using the HERMES model (Baldasano et al., 2008b) specifically developed for Spain. Anthropogenic emissions are calculated by sector-specific sources or by individual installations and stacks following a bottom-up approach. Emissions are derived from the aggregation in space from 1 km \times 1 km to $4 \text{ km} \times 4 \text{ km}$. According to natural biogenic emissions, HERMES calculates volatile organic compounds from vegetation as a function of temperature and photosynthetically active radiation (Guenther et al., 1995; Parra et al., 2004, 2006). Land-use categories for each grid cell are obtained from CORINE Land Cover 2000 map starting with a resolution of 100 m. In this study HERMES provides a comprehensive description of the gas-phase primary emissions speciated to the CBM-IV chemical speciation profile. The CBM-IV mechanism in CMAOv4.5 is modified from the original in Gerv et al. (1989) in order to include gaseous species that are necessary to link gas-phase chemistry to aerosol formation (sulfuric acid, toluene, xylene, cresol, and monoterpenes). On the other hand, HERMES estimates PM10 and PM2.5 primary emission from anthropogenic origin, and after that it especiates fine-particles in (1) organic aerosol, (2) elemental carbon, (3) sulfate, (4) nitrate; and (5) other fine PM based on profiles in SPECIATE 3.2 (http://www.epa.gov/ttn/chief/ emch/speciation/cbiv-profiles_mar_4_2002.xls), which is sector dependent. Note that HERMES quantifies PM emissions resulting from paved road resuspension (Pay et al., 2010b) which is an important feature for PM10 studies in urban air pollution (Querol et al., 2008; Amato et al., 2010).

Due to the importance of trans-boundary air pollution, HERMES also estimates emissions from Portugal and southern France using a top-down approach from the EMEP emission inventory (50 km \times 50 km horizontal resolution) spatially disaggregated to 4 km \times 4 km horizontal resolution.

The CMAQ horizontal grid resolution corresponds to that of WRF. Its vertical structure was obtained by a collapse from the 38 WRF layers to a total of 15 layers steadily increasing from the surface up to 50 hPa with a stronger density within the PBL. The mean altitude of the lowest layer of CMAQ in CALIOPE system is of 19.5 \pm 0.5 m above ground level. Boundary and initial conditions are derived from CALIOPE-EU simulation (12 km \times 12 km grid for Europe and 1 h in time) (Pay et al., 2010a).

The long-range transport of mineral dust from Sahara desert is modeled by BSC-DREAM8b (Pérez et al., 2006a,b) over the domains of study. In this framework, the model results for PM in both domains are achieved by adding the Saharan dust contribution from BSC-DREAM8bin to the CMAQ anthropogenic species over the accumulation- and coarse-modes following the work by Jiménez-Guerrero et al. (2008).

As CALIOPE is a fundamental model system, the authors wish to stress that apart from the discussion of Figs. 4, 9 and 10 and 11, neither correction factors nor any adjusting model parameterization were applied to the model output or the original model codes.

2.2. Particulate matter observations in Spain: levels and chemical speciation

The levels and chemical speciated aerosol simulated by CALIOPE are compared against observations provided by the IDAEA-CSIC (Institute of Environmental Assessment and Water Research, Spanish Research Council). They provided measurements from selected Spanish stations for the whole year 2004. Fig. 1 shows the location of the 25 stations which accomplish the objectives of spatial cover for representing most of the geographical areas in Spain and also anthropogenic activities. Thus, the selected stations are representative of the rural background, suburban background, urban background and urban-industrial background. Table 1 provides detailed characteristics of the aforementioned stations.

Besides PM10 and PM2.5 concentrations, six aerosol components are considered: total carbon (TC), sea-salt (as sum of marine sulfate, chloride and sodium), non-marine sulfate (nm- SO_4^{2-}), nitrate (NO₃), ammonium (NH₄⁺) and desert dust.

Measured PM10 and PM2.5 concentrations are estimated by means of the standard gravimetric method according with the Directive 1999/30/EC. Bulk acid digestion and water extraction were applied to sub-samples, as described by Querol et al. (2001). Non-mineral carbon concentrations are quantified indirectly by a CHNS elemental analyzer after subtraction of Carbon from CaCO₃, as deduced stoichiometrically from experimental Ca levels. The soluble anions (SO₄²⁻, NO₃, Cl⁻) in the water extracts were quantified by ion chromatography. NH₄⁺ analyses were performed by FIA colorimetry or ion specific electrode. Marine aerosol is determined as the sum of the Na⁺ (ICP-AES) and Cl⁻ (ion chromatography).

Desert dust from Sahara is estimated by a new methodology developed by the IDAEA-CSIC (Escudero et al., 2007b) which estimates daily African PM load during dust outbreaks in southern Europe. The daily net dust load in PM10 attributable to an African episode in a given region can be obtained by subtracting the daily regional background level (RB) from the PM10 concentration value for the day with African dust outbreak at the rural station. The RB is obtained from the monthly moving 40th percentile from the PM10 time series after a prior extraction of the data of the days with African dust transport.

2.3. Evaluation methods

Modeled raw outputs are post-processed in order to be compared with quantified measured aerosols. A list of CMAQ aerosol module variables can be found in Binkowski and Roselle (2003) and the sea-salt species have been updated in CMAQv4.5 as described by Shankar et al. (2005). From these variables, fine-particles [nm-SO₄^{2–}], [NO₃], [NH₄⁺], [TC] are approximated by summing the appropriate Aitken- and accumulation-mode concentrations. TC is estimated by summing the modeled concentrations of primary organic aerosol, anthropogenic and biogenic secondary organic aerosol (SOA) and elemental carbon. To compare with measured PM10 sea-salt aerosol (SS), modeled sodium and chlorine (both in accumulation- and coarse-mode) and sulfate (coarse-mode) are counted in the variable SS (SS = [SS]_{fine} + [SS]_{coarse}). Modeled PM10 desert dust (DD) is estimated from BSC-DREAM8b corresponding bins \leq 10 µm (DD = [DD]_{fine} + [DD]_{coarse}).

The BSC-DREAM8b is run offline and the outputs are then added to the CMAQ-calculated PM (Jiménez-Guerrero et al., 2008). BSC-DREAM8b used a 50 km \times 50 km horizontal resolution for integrations; its outputs are interpolated to the CMAQ's Lambert conformal conic grid with a 4 km \times 4 km horizontal resolution in order to add the corresponding aerosol compounds. After the interpolation, fine-particle mass is approximated by summing the modeled concentrations of all Aitken- and accumulation-mode



Fig. 1. Location and type of the IDAEA-CSIC and EMEP stations over Spain with data available on a daily basis in 2004. Different types of stations are represented by symbols and color codes. The various symbols represent the major emissions types 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). The map is split in eight zones in accordance with the geographical area. Table 1 lists the main characteristics and numbers of each station. Color chart represents terrain elevation in meters. Black circles indicate place names. Names in italic show geographical systems.

species from CMAQ (PM_{i+j}) and the corresponding BSC-DREAM8b bins with diameter $\leq 2.5 \ \mu m$ ([DD]_{fine}) (Eq. (1)):

$$PM2.5 = \left[nm - SO_4^{2-}\right] + \left[NO_3^{-}\right] + \left[NH_4^{+}\right] + [TC] + [SS]_{fine} + [DD]_{fine} + Unspeciated Mass_{fine}$$
(1)

Analogously, modeled PM10 is the sum of the Aitken, accumulation- and coarse-mode (2.5–10 μm) species from CMAQ (PM_{i+j+k}) and the corresponding BSC-DREAM8b bins with diameter \leq 10 μm (Eq. (2)):

$$PM10 = [PM2.5] + [SS]_{coarse} + [DD]_{coarse} + Unspecified Mass_{coarse}$$
(2)

Note that PM_{i+j} and PM_{i+j+k} are used as an approximation of PM2.5 and PM10, respectively, since measurements (PM2.5 and PM10) are based on aerodynamic diameters and model only considers the Stokes diameter to characterize the aerosol geometry. For more detail on this issue, see Jiang et al. (2006).

It is worth mentioning that CMAQv4.5(AERO4) only estimates nm-SO₄²⁻, NO₃⁻, and NH₄⁺ in the fine fraction. Meanwhile, calcium and sodium salts with SO_4^{2-} and NO_3^{-} are not taken into account, since coarse-mode is modeled as chemically inert.

Representativeness challenges continue to be present whenever gridded simulations are compared to observed data at a point in time and space as chemistry transport models simulations represent a volumetric average over an entire grid cell, meanwhile the stochastic compound embedded in the observations is not accounted for. Measurements have their own uncertainty due to biases and artifacts related to sampling and laboratory analysis methods. The European legislation (2008/50/EC) requires that the uncertainty of PM10 and PM2.5 measurements meet the quality objective of 25%. However, legislation does not establish uncertainty criteria for chemical species yet. According to Putaud et al. (2004), inorganic species may be accurately measured with an uncertainty of about $\pm 10\%$ for major species. Consideration of uncertainties and spatial scale should be given when comparing model simulation and ambient measurements for model performance evaluation because neither one represents the absolute truth.

Hence, a variety of statistical metrics and graphic methods are used to compare the observed and modeled concentrations (Dennis et al., 2010). Besides mean and standard deviation of modeled and measured values we calculate mean bias (MB), root mean square error (RMSE), correlation coefficient (r) and the fraction of modeled concentrations that are within a factor of two and five of the measured values. We also calculate the mean fractional bias (MFB) and the mean fractional error (MFE) following the recommendation of Boylan and Russell (2006). They proposed that the model performance goal is met when both the MFE and MFB are less than or equal to 50% and \pm 30%, respectively. The model performance criterion is met when both MFE \leq 75% and MFB \pm 60%. These criteria and goals are selected to provide metrics for the CALIOPE performances. Since measurements are on a daily basis, aerosols are compared in terms of daily averages from the modeling system.

3. Results and discussions

The annual and seasonal statistics averaged over all studied stations based on daily values are indicated in Table 2. Scatter plots of the observed and modeled annual average concentration of each aerosol compound by station are shown in Fig. 2, where points represent the annual mean and lines indicate the 75th and 25th

Table 1

Location and characteristics of the stations for 2004.

	Latitude ^a	Longitude ^a	Altitude (m)	Source	Station Name	Code	Costal ^b	Geograp.	Type ^d	Number of data for daily concentration				
						Name		location ^c		Total PM10	Total PM2.5	Speciated PM2.5 ^e	Sea-salt ^f	Desert dust ^g
1	+38.908	-6.359	188	IDAEA-CSIC	Badajoz	BAD	-	SW	UB	363	363	58	52	_
2	+38.476	-6.923	393	EMEP	Barcarrota	BAR	-	SW	RB	343	343	_	-	106
3	+41.385	+2.119	68	IDAEA-CSIC	Barcelona	BARC	N-MED	NE	UB	344	344	84	53	-
4	+37.604	-0.978	20	IDAEA-CSIC	Basterreche	BAS	S-MED	SE	UT	366	-	_	78	-
5	+42.335	-3.638	889	IDAEA-CSIC	Burgos	BUR	-	N	SB	350	366	85	87	-
6	+41.281	-3.143	1360	EMEP	Campisábalos	CAM	-	С	RB	323	304	_	-	70
7	+42.319	+3.316	23	EMEP	Cap de Creus	CAP	N-MED	NE	RB	326	317	_	-	57
8	+41.395	+0.721	470	EMEP	Els Torms	ELS	-	NE	RB	341	336	_	-	65
9	+37.270	-6.924	10	IDAEA-CSIC	Huelva	HUL	S-ATL	SW	UB	362	312	43	44	-
10	+42.653	-2.501	810	RedESP	Izki	IZK	-	N	RB	159	-	_	-	39
11	+40.074	-0.213	175	IDAEA-CSIC	L'Acora	LAC	C-MED	E	UB	283	31	_	88	-
12	+41.780	+2.378	730	IDAEA-CSIC	Montseny	MON	-	NE	RB	320	320	37	113	59
13	+43.406	-2.704	116	RedESP	Mundaka	MUN	N-ATL	Ν	RI	283	-	_	-	35
14	+43.439	-4.850	134	EMEP	Niembro	NIE	N-ATL	N	RB	228	228	-	_	20
15	+42.634	-7.703	506	EMEP	O Saviñao	OSA	W-ATL	NW	RB	333	306	_	-	28
16	+39.963	-0.250	163	IDAEA-CSIC	Onda	OND	C-MED	E	SI	191	26	_	58	-
17	+43.250	-2.155	215	RedESP	Pagoeta	PAG	N-ATL	N	RB	283	-	_	-	38
18	+39.564	+2.623	117	IDAEA-CSIC	Palma	PAL	C-MED	В	SB	315	-	75	71	81
19	+41.288	-5.867	985	EMEP	Peñausende	PEÑ	-	С	RB	348	339	_	-	65
20	+38.683	-4.089	670	IDAEA-CSIC	Puertollano	PUE	-	С	SI	302	-	98	97	-
21	+39.522	-4.353	1241	EMEP	Risco Llano	RIS	-	С	RB	327	326	_	-	69
22	+37.652	-1.018	15	IDAEA-CSIC	Santa Ana	SAN	S-MED	SE	SI	_	323	83	-	-
23	+37.996	-5.666	569	RedESP	Sierra Norte	SIE	-	SW	RB	292	-	_	-	96
24	+37.237	-3.534	1230	EMEP	Viznar	VIZ	-	SE	RB	351	344	_	-	118
25	+39.086	-1.102	885	EMEP	Zarra	ZAR	-	E	RB	353	355	-	-	84

^a "+" is used to indicate North latitude and East longitude; "-" is used to indicate South latitude and West longitude.

^b Classification of the coastal stations according to their location: northern Mediterranean (N-MED), central Mediterranean (C-MED), southern Mediterranean (S-MED), southern Atlantic (N-ATL), western Atlantic (W-ATL).

^c Classification of the stations according to the geographical area: southeastern Spain (SE), southwestern Spain (SW), Balearic Islands (B), eastern Spain (E), central Spain (C), northeaster Spain (NE), northern Spain (N), northwestern Spain (NW).

^d Classification of the stations according to the area type where they are located (first code: urban (U), suburban (S), and rural (R)), and according to what type of sources dominates the air pollution levels at the station (second code: traffic (T), industrial (I), and background (B)) following the specifications described in the Council Decision97/101/ECdated January 27, 1997 (OJ L 35, 5.2.1997, p.14) and amended by Commission Decision 2001/752/EC (OJ L 282, 26.10.2001, p.69).

^e Speciated aerosol measurements corresponding to nitrate (NO₃), ammonium (NH₄), nm-sulfate (nm-SO₄²⁻), total carbon (TC).

^f Sea-salt aerosol (SS) indicates the speciated aerosol corresponding to the sum of sodium (Na⁺), chlorine (Cl⁻) and marine sulfate in the PM10 fraction.

^g Desert dust (DD) indicates the dust load from the Sahara desert in the PM10 fraction.

percentile distribution of observed (horizontal lines) and modeled (vertical lines) concentrations.

3.1. Evaluation of PM2.5 and PM10

A total of 18 and 24 stations are used to evaluate the simulated PM2.5 and PM10 concentrations, respectively (Table 1 and Fig. 1). For PM2.5 (PM10), 11 (15) stations are located in rural areas, 3 (4) in suburban areas, and 4 (5) in urban areas. In both cases rural background stations account for more than 50% of the number of sites (61% for PM2.5 and 58% for PM10).

Fig. 2a and b show that modeled PM2.5 and PM10 daily values are in good agreement with observations (r is 0.46 for PM2.5 and 0.59 for PM10) although modeled concentrations present a systematic negative bias. Differences in model performance for PM2.5 and PM10 are related to the major compounds in both fractions. CALIOPE's results indicate that over Spanish rural background areas carbonaceous and secondary inorganic aerosol are the major compounds in PM2.5, these account for 50–90% (30–70%) to PM2.5 (PM10), meanwhile mineral dust is the second contributor in PM10, it contributes 5-30% (10-50%) to PM2.5 (PM10). Based on correlation coefficients (Table 2), the model presents more accurate skills to reproduce the temporal evolution for PM10 than for PM2.5, which gives an indication that the main transport pattern are correctly represented by CALIOPE. In terms of normalized standard deviations, the amplitude of the daily variation is better reproduced for PM2.5 than for PM10 (0.64 vs. 0.45, respectively), especially in summer, which suggests that modeled coarse particles daily variation seem to be too low. One source of underestimation in coarse particles is related to the fact that CMAQv4.5(AERO4) does not consider the formation of Ca(NO₃)₂ and NaNO₃ salts in the coarse fraction which are significant in Spain from mid-spring to mid-autumn (Rodríguez et al., 2002; Querol et al., 2008).

Fig. 2a and b show that the model skills are different at each site. Since all stations of the same category display similar features, the results are discussed in terms of the mean values, averaged over all rural (Fig. 3a and b), suburban (Fig. 3c and d) and urban (Fig. 3e and f) stations. For more details, time series for representative stations are shown in supplementary material Fig. S1.

Time series for PM2.5 and PM10 at rural sites show that model reproduces observations with correlation coefficients reaching 0.47 (PM2.5) and 0.65 (PM10), and with mean bias of $-5.4 \,\mu\text{g}$ PM2.5 m⁻³ and $-10.1 \,\mu\text{g}$ PM10 m⁻³. Modeled PM at urban and suburban stations presents similar performance, the correlation coefficients reach *r* of 0.49–0.52 (PM2.5) and 0.47–0.51 (PM10) and annual MB are around $-16 \,\mu\text{g}$ PM2.5 m⁻³ and $-25 \,\mu\text{g}$ PM10 m⁻³, in both environments. Therefore, it can be deduced that 35% (40%) of the PM2.5 (PM10) bias at urban and suburban stations is due to the underestimation of the regional background levels.

Overall, CALIOPE reproduces the observed PM daily variability along the year, with spring—summer maxima and the minimum levels in April, November and December. Modeled PM presents the best skills in summer (Table 2), when high PM levels are related to the low wet deposition (related to low humidity), low air mass

Table 2

Annual and seasonal statistics for modeled aerosol calculated at IDAEA-CSIC stations in 2004. The calculated statistics are: observed mean (OM, $\mu g m^{-3}$), modeled mean (MM, $\mu g m^{-3}$), mean bias (MB, in $\mu g m^{-3}$), correlation coefficient (*r*), root mean square error (RMSE, in $\mu g m^{-3}$), normalized ratio of standard deviations of the modeled mean (σ (Mod./Obs.)). Data indicates the number of pairs model-observation used to calculate the statistics.

PM10 Iactor 2 PM10 7889 23.5 8.0 -15.5 0.59 22.3 0.45 21 Winter 1949 20.4 5.4 -15.0 0.52 23.7 0.24 18 Spring 1964 21.2 6.7 -14.5 0.60 19.4 0.46 16 Summer 1953 24.2 8.6 -15.6 0.64 21.1 0.50 24 PM25 8.6 -16.4 0.55 15.2 0.29 16 Spring 1246 13.4 4.6 -8.8 0.48 12.0 0.58 18 Summer 1387 15.7 7.9 -7.8 0.48 11.6 0.75 33 Fall 1350 13.8 6.0 -7.7 0.51 11.3 0.71 29 Annual 563 3.5 1.4 -2.1 0.62 2.9 0.46 30 Winter		Data	OM	MM	MB	R	RMSE	σ (Mod./Obs.)	% within
PM10 FM10 Annual 7889 23.5 8.0 -15.5 0.59 22.3 0.45 21 Winter 1949 20.4 5.4 -15.0 0.52 23.7 0.24 18 Spring 1964 21.2 6.7 -14.5 0.60 19.4 0.46 16 Summer 2002 27.8 11.1 -16.7 0.59 24.6 0.51 27 Fall 1953 24.2 5.6 -8.7 0.46 12.6 0.64 24 PM2.5 - - 8.6 4.8 12.0 0.58 18 Symme 1387 15.7 7.9 -7.8 0.48 12.0 0.58 18 Summer 1387 15.7 7.9 -7.8 0.48 11.6 0.75 33 Fall 1350 1.3 4.0 -7.7 0.51 13.0 0.71 29 Summer 149 </td <td></td> <td>_</td> <td></td> <td></td> <td>_</td> <td></td> <td>_</td> <td></td> <td></td>		_			_		_		
Annual 7889 25.5 8.0 -15.5 0.59 22.3 0.45 21 Winter 1949 20.4 5.4 -15.0 0.52 23.7 0.24 18 Spring 1964 21.2 6.7 -14.5 0.60 19.4 0.46 16 Summer 2002 27.8 11.1 -16.7 0.59 24.6 0.51 27 Fall 1953 24.2 8.6 -15.6 0.64 21.1 0.50 24 Minter 1255 14.0 3.6 -10.4 0.55 15.2 0.29 16 Spring 1246 13.4 4.6 -8.8 0.48 11.6 0.75 33 Fall 1350 13.8 6.0 -7.7 0.51 11.3 0.71 29 mmsulfate 127 2.9 0.7 -2.2 0.46 3.4 0.18 10 Spring 147 3.3 1.2 -2.0 0.65 2.5 0.47 28 Summer 140	PM10				45.5	0.50		0.45	24
Winter 1949 20.4 5.4 -15.0 0.52 2.5.7 0.24 18 Spring 1964 21.2 6.7 -14.5 0.60 19.4 0.46 16 Summer 2002 27.8 11.1 -16.7 0.59 24.6 0.51 27 Fall 1953 24.2 8.6 -15.6 0.64 21.1 0.50 24 PM2.5 0.64 12.6 0.64 24 Winter 1255 14.0 3.6 -10.4 0.55 15.2 0.29 16 Spring 1246 13.4 4.6 -8.8 0.48 11.6 0.75 33 Fall 1350 13.8 6.0 -7.7 0.51 11.3 0.71 29 Inm-sulfate 2.2 0.65 2.5 0.47 28 Summer 149 3.3 2.2 -2.1 0.68 2.3 </td <td>Annual</td> <td>/889</td> <td>23.5</td> <td>8.0</td> <td>-15.5</td> <td>0.59</td> <td>22.3</td> <td>0.45</td> <td>21</td>	Annual	/889	23.5	8.0	-15.5	0.59	22.3	0.45	21
Spring 1964 21.2 6.7 -14.5 0.60 19.4 0.46 16 Summer 2002 27.8 11.1 -16.7 0.59 24.6 0.51 27 Fall 1953 24.2 8.6 -15.6 0.64 21.1 0.50 24 PM2.5 3.6 -10.4 0.55 15.2 0.29 16 Spring 1246 13.4 4.6 -8.8 0.48 11.6 0.75 33 Fall 1350 13.8 6.0 -7.7 0.51 11.3 0.71 29 nm-suffate Annual 563 3.5 1.4 -2.1 0.62 2.9 0.46 30 Winter 127 2.9 0.7 -2.2 0.46 3.4 0.18 10 Spring 147 3.3 1.2 -2.0 0.65 2.5 0.47 28 Summer 149 3.5	Winter	1949	20.4	5.4	-15.0	0.52	23.7	0.24	18
Summer 2002 27.8 11.1 -16.7 0.59 24.6 0.51 27 Fall 1953 24.2 8.6 -15.6 0.64 21.1 0.50 24 PM2.5 - - 0.66 12.6 0.64 24 Winter 1255 14.0 3.6 -10.4 0.55 15.2 0.29 16 Spring 1246 13.4 4.6 -8.8 0.48 11.6 0.75 33 Fall 1350 13.8 6.0 -7.7 0.51 11.3 0.71 29 nm-sulfate - - 0.65 2.5 0.47 28 30 Winter 127 2.9 0.7 -2.2 0.46 3.4 0.18 10 Spring 147 3.3 1.2 -2.0 0.65 2.5 0.47 28 Summer 149 4.3 2.2 -2.1 0.68 2.3 0.39 31 Winter 127 2.6 0.7 -1.9 <	Spring	1964	21.2	6.7	-14.5	0.60	19.4	0.46	16
Fail 1953 24.2 8.6 -15.6 0.64 21.1 0.50 24 PM2.5 Annual 5253 14.2 5.6 -8.7 0.46 12.6 0.64 24 Winter 1255 14.0 3.6 -10.4 0.55 15.2 0.29 16 Spring 1246 13.4 4.6 -8.8 0.48 11.6 0.75 33 Fall 1350 13.8 6.0 -7.7 0.51 11.3 0.71 29 mm-sulfate - - 0.62 2.9 0.46 30 30 Vinter 127 2.9 0.7 -2.2 0.46 3.4 0.18 10 Spring 147 3.3 1.2 -2.0 0.65 2.5 0.47 28 Summer 149 3.5 1.4 -2.1 0.73 2.8 0.38 34 Summer 149 0.7 0.5 -0.2 0.22 0.9 0.75 40 Fall 140 1.6	Summer	2002	27.8	11.1	-16.7	0.59	24.6	0.51	27
PM2.5 Annual 5253 14.2 5.6 -8.7 0.46 12.6 0.64 24 Winter 1255 14.0 3.6 -10.4 0.55 15.2 0.29 16 Spring 1246 13.4 4.6 -8.8 0.48 11.6 0.75 33 Fall 1350 13.8 6.0 -7.7 0.51 11.3 0.71 29 mm-suffatt - Annual 563 3.5 1.4 -2.1 0.62 2.9 0.46 30 Winter 127 2.9 0.7 -2.2 0.46 3.4 0.18 10 Spring 147 3.3 1.2 -2.0 0.65 2.5 0.47 28 Summer 149 4.3 2.2 -2.1 0.68 2.7 0.54 44 Signing 147 1.9 0.8 -1.1 0.28 2.4 0.43 28 Summe	Fall	1953	24.2	8.6	-15.6	0.64	21.1	0.50	24
Annual 5253 14.2 5.6 -8.7 0.46 12.6 0.64 24 Winter 1255 14.0 3.6 -10.4 0.55 15.2 0.29 16 Spring 1266 13.4 4.6 -8.8 0.48 12.0 0.58 18 Summer 1387 15.7 7.9 -7.8 0.48 11.6 0.75 33 Fall 1350 13.8 6.0 -7.7 0.51 11.3 0.71 29 nm-sulfate - - 0.62 2.9 0.46 30 00 Winter 127 2.9 0.7 -2.2 0.46 3.4 0.18 10 Spring 147 3.3 1.2 -2.0 0.65 2.5 0.47 28 Summer 149 4.3 2.2 -2.1 0.73 2.8 0.38 34 Nitrate 127 2.6 0.7 -1.9 0.34 3.5 0.23 25 Spring 147 1.9 0.8 -1.1	PM2.5								
Winter 1255 14.0 3.6 -10.4 0.55 15.2 0.29 16 Spring 1246 13.4 4.6 -8.8 0.48 12.0 0.58 18 Summer 1387 15.7 7.9 -7.8 0.48 11.6 0.75 33 Fall 1350 13.8 6.0 -7.7 0.51 11.3 0.71 29 nm-sulfate - - 0.62 2.9 0.46 30 Winter 127 2.9 0.7 -2.2 0.66 2.5 0.47 28 Summer 149 4.3 2.2 -2.1 0.68 2.7 0.54 44 Fall 140 3.5 1.4 -2.1 0.73 2.8 0.38 34 Nitret 127 2.6 0.7 -1.9 0.34 3.5 0.23 25 Spring 147 1.9 0.8 -1.1 0.28 2.4 0.43 28 Summer 149 0.7 0.5 -0.2	Annual	5253	14.2	5.6	-8.7	0.46	12.6	0.64	24
Spring 1246 13.4 4.6 -8.8 0.48 12.0 0.58 18 Summer 1387 15.7 7.9 -7.8 0.48 11.6 0.75 33 Fall 1350 13.8 6.0 -7.7 0.51 11.3 0.71 29 nm-sulfate - - 0.62 2.9 0.46 30 Winter 127 2.9 0.7 -2.2 0.46 3.4 0.18 10 Spring 147 3.3 1.2 -2.0 0.65 2.5 0.47 28 Summer 149 4.3 2.2 -2.1 0.68 2.7 0.54 44 Fall 140 3.5 1.4 -2.1 0.73 2.8 0.38 34 Nitrate - - 1.7 0.7 -1.0 0.36 2.3 0.39 31 Winter 127 2.6 0.7 -1.9 0.34 3.5 0.23 25 Spring 147 1.9 0.8 -0	Winter	1255	14.0	3.6	-10.4	0.55	15.2	0.29	16
Summer 1387 15.7 7.9 -7.8 0.48 11.6 0.75 33 Fall 1350 13.8 6.0 -7.7 0.51 11.3 0.71 29 nm-sulfate Annual 563 3.5 1.4 -2.1 0.62 2.9 0.46 30 Winter 127 2.9 0.7 -2.2 0.46 3.4 0.18 10 Spring 147 3.3 1.2 -2.0 0.65 2.5 0.47 28 Summer 149 4.3 2.2 -2.1 0.68 2.7 0.54 44 Fall 140 3.5 1.4 -2.1 0.73 2.8 0.38 34 Nitrate N Annual 563 1.7 0.7 -1.0 0.36 2.3 0.39 31 Winter 127 2.6 0.7 -1.9 0.34 3.5 0.23 25 Spring 147 1.9 0.6 -1.3 0.42 2.4 0.27 21	Spring	1246	13.4	4.6	-8.8	0.48	12.0	0.58	18
Fail 13.8 6.0 -7.7 0.51 11.3 0.71 29 nm-sulfate Annual 563 3.5 1.4 -2.1 0.62 2.9 0.46 30 Winter 127 2.9 0.7 -2.2 0.46 3.4 0.18 10 Spring 147 3.3 1.2 -2.0 0.65 2.5 0.47 28 Summer 149 4.3 2.2 -2.1 0.68 2.7 0.54 44 Fall 140 3.5 1.4 -2.1 0.73 2.8 0.38 34 Nitrate	Summer	1387	15.7	7.9	-7.8	0.48	11.6	0.75	33
nn-sulfate Annual 563 3.5 1.4 -2.1 0.62 2.9 0.46 3.0 Winter 127 2.9 0.7 -2.2 0.46 3.4 0.18 10 Spring 147 3.3 1.2 -2.0 0.65 2.5 0.47 28 Summer 149 4.3 2.2 -2.1 0.68 2.7 0.54 44 Fall 140 3.5 1.4 -2.1 0.73 2.8 0.38 34 Nitrate Nitrate N Nitrate N <	Fall	1350	13.8	6.0	-7.7	0.51	11.3	0.71	29
Annual 563 3.5 1.4 -2.1 0.62 2.9 0.46 30 Winter 127 2.9 0.7 -2.2 0.46 3.4 0.18 10 Spring 147 3.3 1.2 -2.0 0.65 2.5 0.47 28 Summer 149 4.3 2.2 -2.1 0.73 2.8 0.38 34 Nitrate - - 0.73 2.8 0.38 34 Nitrate - - 0.36 2.3 0.39 31 Winter 127 2.6 0.7 -1.9 0.34 3.5 0.23 25 Spring 147 1.9 0.8 -1.1 0.28 2.4 0.43 28 Summer 149 0.7 0.5 -0.2 0.22 0.9 0.75 40 Fall 140 1.6 0.8 -0.8 0.49 2.0 0.05 32 Ammonium - - 0.34 2.1 0.47 21 <	nm-sulfa	te							
Winter1272.90.7 -2.2 0.463.40.1810Spring1473.31.2 -2.0 0.652.50.4728Summer1494.32.2 -2.1 0.732.80.3834NitrateAnnual5631.70.7 -1.0 0.362.30.3931Winter1272.60.7 -1.9 0.343.50.2325Spring1471.90.8 -1.1 0.282.40.4328Summer1490.70.5 -0.2 0.220.90.7540Fall1401.60.8 -0.8 0.492.00.0532Annual5631.90.6 -1.3 0.342.10.4721Winter1271.90.6 -1.3 0.342.10.4721Syring1472.10.6 -1.5 0.322.20.4421Summer1491.80.3 -1.4 0.152.10.4012Fall1401.70.7 -1.0 0.481.60.7129OC+EC $OC+EC$ $OC+EC$ $O.53$ 6.00.3111Spring1474.31.2 -3.0 0.513.90.5017Summer1494.61.4 -3.2 0.314.20.5025Fall1405.71.7 </td <td>Annual</td> <td>563</td> <td>3.5</td> <td>1.4</td> <td>-2.1</td> <td>0.62</td> <td>2.9</td> <td>0.46</td> <td>30</td>	Annual	563	3.5	1.4	-2.1	0.62	2.9	0.46	30
Spring1473.31.2 -2.0 0.65 2.5 0.47 28 Summer1494.3 2.2 -2.1 0.68 2.7 0.54 44Fall140 3.5 1.4 -2.1 0.73 2.8 0.38 34Nitrate 0.73 2.8 0.39 31 Winter127 2.6 0.7 -1.9 0.34 3.5 0.23 25 Spring147 1.9 0.8 -1.1 0.28 2.4 0.43 28 Summer149 0.7 0.5 -0.2 0.22 0.9 0.75 40 Fall140 1.6 0.8 -0.8 0.49 2.0 0.05 32 Annual 563 1.9 0.6 -1.3 0.34 2.1 0.47 21 Winter 127 1.9 0.6 -1.3 0.34 2.1 0.47 21 Winter 127 1.9 0.6 -1.3 0.42 2.4 0.27 21 Spring 147 2.1 0.6 -1.5 0.32 2.2 0.44 21 Summer 149 1.8 0.3 -1.4 0.15 2.1 0.40 12 Fall 140 1.7 0.7 0.53 6.0 0.31 11 Spring 147 4.3 1.2 -3.0 0.51 3.9 0.50 17 Summer	Winter	127	2.9	0.7	-2.2	0.46	3.4	0.18	10
Summer1494.32.2 -2.1 0.68 2.7 0.54 44Fall1403.5 1.4 -2.1 0.73 2.8 0.38 34 NitrateNitrateNitrateNitrateNitrateNitrateNitrateAnnual563 1.7 0.7 -1.0 0.36 2.3 0.39 31 Winter 127 2.6 0.7 -1.9 0.34 3.5 0.23 25 Spring 147 1.9 0.8 -1.1 0.28 2.4 0.43 28 Summer 149 0.7 0.5 -0.2 0.22 0.9 0.75 40 Fall 140 1.6 0.8 -0.8 0.49 2.0 0.05 32 AmmoniumAnnual 563 1.9 0.6 -1.3 0.34 2.1 0.47 21 Winter 127 1.9 0.6 -1.3 0.34 2.1 0.47 21 Spring 147 2.1 0.6 -1.5 0.32 2.2 0.44 21 Summer 149 1.8 0.3 -1.4 0.15 2.1 0.40 12 Fall 140 1.7 0.7 -1.0 0.48 1.6 0.71 29 OC+ECNoNo -1.4 0.53 6.0 0.31 11 Spring 147 4.3 1.2 -3.0 0.51 3.9 0.50 17	Spring	147	3.3	1.2	-2.0	0.65	2.5	0.47	28
Fail 140 3.5 1.4 -2.1 0.73 2.8 0.38 34 Nitrate Annual 563 1.7 0.7 -1.0 0.36 2.3 0.39 31 Winter 127 2.6 0.7 -1.9 0.34 3.5 0.23 25 Spring 147 1.9 0.8 -1.1 0.28 2.4 0.43 28 Summer 149 0.7 0.5 -0.2 0.22 0.9 0.75 40 Fall 140 1.6 0.8 -0.8 0.49 2.0 0.05 32 Ammonium Annual 563 1.9 0.6 -1.3 0.34 2.1 0.47 21 Winter 127 1.9 0.6 -1.5 0.32 2.2 0.44 21 Summer 149 1.8 0.3 -1.4 0.15 2.1 0.40 12 Fall 140 1.7 0.7 -1.0 0.48 1.6 0.71 29 OC+EC Mnu	Summer	149	4.3	2.2	-2.1	0.68	2.7	0.54	44
NirrateAnnual5631.70.7 -1.0 0.362.30.3931Winter1272.60.7 -1.9 0.343.50.2325Spring1471.90.8 -1.1 0.282.40.4328Summer1490.70.5 -0.2 0.220.90.7540Fall1401.60.8 -0.8 0.492.00.0532AmmoniumMonomiumMinterNineMonomiumAnnual5631.90.6 -1.3 0.422.40.2721Spring1472.10.6 -1.5 0.322.20.4421Summer1491.80.3 -1.4 0.152.10.4012Fall1401.70.7 -1.0 0.481.60.7129OC + ECMinuer1275.81.3 -4.5 0.536.00.3111Spring1474.31.2 -3.0 0.513.90.5017Summer1494.61.4 -3.2 0.314.20.5025Fall1405.71.7 -4.0 0.435.40.4126See SaltMinuer133<	Fall	140	3.5	1.4	-2.1	0.73	2.8	0.38	34
Annual5631.70.7 -1.0 0.362.30.3931Winter1272.60.7 -1.9 0.343.50.2325Spring1471.90.8 -1.1 0.282.40.4328Summer1490.70.5 -0.2 0.220.90.7540Fall1401.60.8 -0.8 0.492.00.0532AmmoniumNNNNNNAnnual5631.90.6 -1.3 0.422.40.2721Winter1271.90.6 -1.5 0.322.20.4421Spring1472.10.6 -1.5 0.322.20.4421Summer1491.80.3 -1.4 0.152.10.4012Fall1401.70.7 -1.0 0.481.60.7129OC+ECAnnual5635.11.4 -3.7 0.454.90.4220Winter1275.81.3 -4.5 0.536.00.3111Spring1474.31.2 -3.0 0.513.90.5017Summer1494.61.4 -3.2 0.314.20.5025Fall1405.71.7 -4.0 0.701.21.1434Spring2101.31.0 -0	Nitrate								
Winter1272.60.7 -1.9 0.343.50.2325Spring1471.90.8 -1.1 0.282.40.4328Summer1490.70.5 -0.2 0.220.90.7540Fall1401.60.8 -0.8 0.492.00.0532Ammonium -1.3 0.342.10.4721Winter1271.90.6 -1.3 0.342.10.4721Spring1472.10.6 -1.5 0.322.20.4421Summer1491.80.3 -1.4 0.152.10.4012Fall1401.70.7 -1.0 0.481.60.7129OC+EC -1.4 0.152.10.4012Annual5635.11.4 -3.7 0.454.90.4220Winter1275.81.3 -4.5 0.536.00.3111Spring1474.31.2 -3.0 0.513.90.5017Summer1494.61.4 -3.2 0.314.20.4126Sea Salt -1.7 -4.0 0.701.21.1434Spring2101.31.0 -0.3 0.651.21.1741Winter1531.30.9 -0.4 0.701.21.1434Spring	Annual	563	1.7	0.7	-1.0	0.36	2.3	0.39	31
Spring1471.90.8 -1.1 0.282.40.4328Summer1490.70.5 -0.2 0.220.90.7540Fall1401.60.8 -0.8 0.492.00.0532Ammonium -1.3 0.342.10.4721Winter1271.90.6 -1.3 0.422.40.2721Spring1472.10.6 -1.5 0.322.20.4421Summer1491.80.3 -1.4 0.152.10.4012Fall1401.70.7 -1.0 0.481.60.7129OC+ECAnnual5635.11.4 -3.7 0.454.90.4220Winter1275.81.3 -4.5 0.536.00.3111Spring1474.31.2 -3.0 0.513.90.5017Summer1494.61.4 -3.2 0.314.20.5025Fall1405.71.7 -4.0 0.435.40.4126SeasitAnnual7381.20.9 -0.3 0.651.21.1741Winter1531.30.9 -0.4 0.701.21.1434Spring2101.31.0 -0.3 0.711.21.0737 <td< td=""><td>Winter</td><td>127</td><td>2.6</td><td>0.7</td><td>-1.9</td><td>0.34</td><td>3.5</td><td>0.23</td><td>25</td></td<>	Winter	127	2.6	0.7	-1.9	0.34	3.5	0.23	25
Summer1490.70.5 -0.2 0.220.90.7540Fall1401.60.8 -0.8 0.492.00.0532Ammonium -0.8 0.492.00.0532Annual5631.90.6 -1.3 0.342.10.4721Winter1271.90.6 -1.3 0.422.40.2721Spring1472.10.6 -1.5 0.322.20.4421Summer1491.80.3 -1.4 0.152.10.4012Fall1401.70.7 -1.0 0.481.60.7129OC+EC -3.7 0.454.90.4220Winter1275.81.3 -4.5 0.536.00.3111Spring1474.31.2 -3.0 0.513.90.5017Summer1494.61.4 -3.2 0.314.20.5025Fall1405.71.7 -4.0 0.435.40.4126Sea Salt -3.1 -0.3 0.651.21.1741Winter1531.30.9 -0.4 0.701.21.1434Spring2101.31.0 -0.3 0.711.21.0737Summer1931.20.7 -0.7 0.621.11.4447Fall182<	Spring	147	1.9	0.8	-1.1	0.28	2.4	0.43	28
Fall1401.60.8 -0.8 0.492.00.0532AmmoniumAnnual5631.90.6 -1.3 0.342.10.4721Winter1271.90.6 -1.3 0.422.40.2721Spring1472.10.6 -1.5 0.322.20.4421Summer1491.80.3 -1.4 0.152.10.4012Fall1401.70.7 -1.0 0.481.60.7129OC+ECVAnnual5635.11.4 -3.7 0.454.90.4220Winter1275.81.3 -4.5 0.536.00.3111Spring1474.31.2 -3.0 0.513.90.5017Summer1494.61.4 -3.2 0.314.20.5025Fall1405.71.7 -4.0 0.435.40.4126Sea SaltMinuel7381.20.9 -0.3 0.651.21.1741Winter1531.30.9 -0.4 0.701.21.1434Spring2101.31.0 -0.3 0.711.21.0737Summer1931.21.1 -0.1 0.521.11.4447Fall1821.20.7 -0.4 0.621	Summer	149	0.7	0.5	-0.2	0.22	0.9	0.75	40
Ammonium Annual 563 1.9 0.6 -1.3 0.34 2.1 0.47 21 Winter 127 1.9 0.6 -1.3 0.42 2.4 0.27 21 Spring 147 2.1 0.6 -1.5 0.32 2.2 0.44 21 Summer 149 1.8 0.3 -1.4 0.15 2.1 0.40 12 Fall 140 1.7 0.7 -1.0 0.48 1.6 0.71 29 OC+EC Annual 563 5.1 1.4 -3.7 0.45 4.9 0.42 20 Winter 127 5.8 1.3 -4.5 0.53 6.0 0.31 11 Spring 147 4.3 1.2 -3.0 0.51 3.9 0.50 17 Summer 149 4.6 1.4 -3.2 0.31 4.2 0.50 25 Fall 140 5.7 1.7 -4.0 0.43 5.4 0.41 <td< td=""><td>Fall</td><td>140</td><td>1.6</td><td>0.8</td><td>-0.8</td><td>0.49</td><td>2.0</td><td>0.05</td><td>32</td></td<>	Fall	140	1.6	0.8	-0.8	0.49	2.0	0.05	32
Annual 563 1.9 0.6 -1.3 0.34 2.1 0.47 21 Winter 127 1.9 0.6 -1.3 0.42 2.4 0.27 21 Spring 147 2.1 0.6 -1.5 0.32 2.2 0.44 21 Summer 149 1.8 0.3 -1.4 0.15 2.1 0.40 12 Fall 140 1.7 0.7 -1.0 0.48 1.6 0.71 29 OC+ECAnnual 563 5.1 1.4 -3.7 0.45 4.9 0.42 20 Winter 127 5.8 1.3 -4.5 0.53 6.0 0.31 11 Spring 147 4.3 1.2 -3.0 0.51 3.9 0.50 17 Summer 149 4.6 1.4 -3.2 0.31 4.2 0.50 25 Fall 140 5.7 1.7 -4.0 0.43 5.4 0.41 26 See SaltAnnual 738 1.2 0.9 -0.3 0.65 1.2 1.17 41 Winter 153 1.3 0.9 -0.4 0.70 1.2 1.14 34 Spring 210 1.3 1.0 -0.3 0.71 1.2 1.07 37 Summer 193 1.2 0.7 0.47 21.9 0.68 37 Winter 153 1.4 <td< td=""><td>Ammoniu</td><td>ım</td><td></td><td></td><td></td><td></td><td></td><td></td><td></td></td<>	Ammoniu	ım							
Winter1271.90.6 -1.3 0.422.40.2721Spring1472.10.6 -1.5 0.322.20.4421Summer1491.80.3 -1.4 0.152.10.4012Fall1401.70.7 -1.0 0.481.60.7129OC+ECOC+ECAnnual5635.11.4 -3.7 0.454.90.4220Winter1275.81.3 -4.5 0.536.00.3111Spring1474.31.2 -3.0 0.513.90.5017Summer1494.61.4 -3.2 0.314.20.5025Fall1405.71.7 -4.0 0.435.40.4126SeasattAnnual7381.20.9 -0.3 0.651.21.1741Winter1531.30.9 -0.4 0.701.21.1434Spring2101.31.0 -0.3 0.711.21.0737Summer1931.20.7 -0.4 0.621.11.0944Desert DustAnnual103019.413.5 -5.9 0.4721.90.6837Winter8814.45.3 -9.1 0.1830.54.4516Spring22	Annual	563	1.9	0.6	-1.3	0.34	2.1	0.47	21
	Winter	127	1.9	0.6	-1.3	0.42	2.4	0.27	21
Summer 149 1.8 0.3 -1.4 0.15 2.1 0.40 12 Fall 140 1.7 0.7 -1.0 0.48 1.6 0.71 29 OC+EC - - 0.48 1.6 0.71 29 Winter 127 5.8 1.3 -4.5 0.53 6.0 0.31 11 Spring 147 4.3 1.2 -3.0 0.51 3.9 0.50 17 Summer 149 4.6 1.4 -3.2 0.31 4.2 0.50 25 Fall 140 5.7 1.7 -4.0 0.43 5.4 0.41 26 Sea Salt - - - 0.40 0.70 1.2 1.17 41 Winter 153 1.3 0.9 -0.4 0.70 1.2 1.14 34 Spring 210 1.3 1.0 -0.3 0.71 1.2 1.07 37 Summer 193 1.2 1.1 -0.1 0.52 <t< td=""><td>Spring</td><td>147</td><td>2.1</td><td>0.6</td><td>-1.5</td><td>0.32</td><td>2.2</td><td>0.44</td><td>21</td></t<>	Spring	147	2.1	0.6	-1.5	0.32	2.2	0.44	21
Fall1401.70.7 -1.0 0.481.60.7129OC+ECAnnual5635.11.4 -3.7 0.454.90.4220Winter1275.81.3 -4.5 0.536.00.3111Spring1474.31.2 -3.0 0.513.90.5017Summer1494.61.4 -3.2 0.314.20.5025Fall1405.71.7 -4.0 0.435.40.4126See SaltAnnual7381.20.9 -0.3 0.651.21.1741Winter1531.30.9 -0.4 0.701.21.1434Spring2101.31.0 -0.3 0.711.21.0737Summer1931.21.1 -0.1 0.521.11.4447Fall1821.20.7 -0.4 0.621.11.0944Desert DustAnnual103019.413.5 -5.9 0.4721.90.6837Winter8814.45.3 -9.1 0.1830.54.4516Spring22817.08.3 -8.7 0.4417.90.7319	Summer	149	1.8	0.3	-1.4	0.15	2.1	0.40	12
OC+ECAnnual 563 5.1 1.4 -3.7 0.45 4.9 0.42 20 Winter 127 5.8 1.3 -4.5 0.53 6.0 0.31 11 Spring 147 4.3 1.2 -3.0 0.51 3.9 0.50 17 Summer 149 4.6 1.4 -3.2 0.31 4.2 0.50 25 Fall 140 5.7 1.7 -4.0 0.43 5.4 0.41 26 Sea SaltMnnual 738 1.2 0.9 -0.3 0.65 1.2 1.17 41 Winter 153 1.3 0.9 -0.4 0.70 1.2 1.14 34 Spring 210 1.3 1.0 -0.3 0.71 1.2 1.07 37 Summer 193 1.2 0.7 -0.4 0.62 1.1 1.09 44 Deser DuseAnnual 1030 19.4 13.5 -5.9 0.47 21.9 0.68 37 Winter 88 14.4 5.3 -9.1 0.18 30.5 4.45 16 Spring 228 17.0 8.3 -8.7 0.44 17.9 0.73 19	Fall	140	1.7	0.7	-1.0	0.48	1.6	0.71	29
Annual 563 5.1 1.4 -3.7 0.45 4.9 0.42 20 Winter 127 5.8 1.3 -4.5 0.53 6.0 0.31 11 Spring 147 4.3 1.2 -3.0 0.51 3.9 0.50 17 Summer 149 4.6 1.4 -3.2 0.31 4.2 0.50 25 Fall 140 5.7 1.7 -4.0 0.43 5.4 0.41 26 Sea Salt Annual 738 1.2 0.9 -0.3 0.65 1.2 1.17 41 Winter 153 1.3 0.9 -0.4 0.70 1.2 1.14 34 Spring 210 1.3 1.0 -0.3 0.71 1.2 1.07 37 Summer 193 1.2 0.7 -0.4 0.62 1.1 1.09 44 Desert Dust Ital 1.2 0.7 -0.4 0.62 1.1 1.09 44 Desert	OC+EC								
Winter1275.81.3 -4.5 0.536.00.3111Spring1474.31.2 -3.0 0.513.90.5017Summer1494.61.4 -3.2 0.314.20.5025Fall1405.71.7 -4.0 0.435.40.4126Sea SaltVAnnual7381.20.9 -0.3 0.651.21.1741Winter1531.30.9 -0.4 0.701.21.1434Spring2101.31.0 -0.3 0.711.21.0737Summer1931.21.1 -0.1 0.521.11.4447Fall1821.20.7 -0.4 0.621.11.0944Desert DustAnnual103019.413.5 -5.9 0.4721.90.6837Winter8814.45.3 -9.1 0.1830.54.4516Spring22817.08.3 -8.7 0.4417.90.7319	Annual	563	5.1	1.4	-3.7	0.45	4.9	0.42	20
Spring 147 4.3 1.2 -3.0 0.51 3.9 0.50 17 Summer 149 4.6 1.4 -3.2 0.31 4.2 0.50 25 Fall 140 5.7 1.7 -4.0 0.43 5.4 0.41 26 Sea Salt	Winter	127	5.8	1.3	-4.5	0.53	6.0	0.31	11
Summer 149 4.6 1.4 -3.2 0.31 4.2 0.50 25 Fall 140 5.7 1.7 -4.0 0.43 5.4 0.41 26 Sea Salt Namual 738 1.2 0.9 -0.3 0.65 1.2 1.17 41 Winter 153 1.3 0.9 -0.4 0.70 1.2 1.14 34 Spring 210 1.3 1.0 -0.3 0.71 1.2 1.07 37 Summer 193 1.2 1.1 -0.1 0.52 1.1 1.44 47 Fall 182 1.2 0.7 -0.4 0.62 1.1 1.09 44 Desert Dust Nanual 1030 19.4 13.5 -5.9 0.47 21.9 0.68 37 Winter 88 14.4 5.3 -9.1 0.18 30.5 4.45 16 Spring 228 17.0<	Spring	147	4.3	1.2	-3.0	0.51	3.9	0.50	17
Fall 140 5.7 1.7 -4.0 0.43 5.4 0.41 26 Sea Salt	Summer	149	4.6	1.4	-3.2	0.31	4.2	0.50	25
Sea Salt Annual 738 1.2 0.9 -0.3 0.65 1.2 1.17 41 Winter 153 1.3 0.9 -0.4 0.70 1.2 1.14 34 Spring 210 1.3 1.0 -0.3 0.71 1.2 1.07 37 Summer 193 1.2 1.1 -0.1 0.52 1.1 1.44 47 Fall 182 1.2 0.7 -0.4 0.62 1.1 1.09 44 Desert Dust V -0.4 0.62 1.1 1.09 44 Spring 1030 19.4 13.5 -5.9 0.47 21.9 0.68 37 Winter 88 14.4 5.3 -9.1 0.18 30.5 4.45 16 Spring 228 17.0 8.3 -8.7 0.44 17.9 0.73 19	Fall	140	5.7	1.7	-4.0	0.43	5.4	0.41	26
Annual 738 1.2 0.9 -0.3 0.65 1.2 1.17 41 Winter 153 1.3 0.9 -0.4 0.70 1.2 1.14 34 Spring 210 1.3 1.0 -0.3 0.71 1.2 1.14 34 Spring 210 1.3 1.0 -0.3 0.71 1.2 1.07 37 Summer 193 1.2 1.1 -0.1 0.52 1.1 1.44 47 Fall 182 1.2 0.7 -0.4 0.62 1.1 1.09 44 Desert Dust - - - 0.47 21.9 0.68 37 Winter 88 14.4 5.3 -9.1 0.18 30.5 4.45 16 Spring 228 17.0 8.3 -8.7 0.44 17.9 0.73 19	Sea Salt								
Winter 153 1.3 0.9 -0.4 0.70 1.2 1.14 34 Spring 210 1.3 1.0 -0.3 0.71 1.2 1.07 37 Summer 193 1.2 1.1 -0.1 0.52 1.1 1.44 47 Fall 182 1.2 0.7 -0.4 0.62 1.1 1.09 44 Desert Dust Annual 1030 19.4 13.5 -5.9 0.47 21.9 0.68 37 Winter 88 14.4 5.3 -9.1 0.18 30.5 4.45 16 Spring 228 17.0 8.3 -8.7 0.44 17.9 0.73 19	Annual	738	1.2	0.9	-0.3	0.65	1.2	1.17	41
Spring 210 1.3 1.0 -0.3 0.71 1.2 1.07 37 Summer 193 1.2 1.1 -0.1 0.52 1.1 1.44 47 Fall 182 1.2 0.7 -0.4 0.62 1.1 1.09 44 Desert Dust Annual 1030 19.4 13.5 -5.9 0.47 21.9 0.68 37 Winter 88 14.4 5.3 -9.1 0.18 30.5 4.45 16 Spring 228 17.0 8.3 -8.7 0.44 17.9 0.73 19	Winter	153	1.3	0.9	-0.4	0.70	1.2	1.14	34
Summer 193 1.2 1.1 -0.1 0.52 1.1 1.44 47 Fall 182 1.2 0.7 -0.4 0.62 1.1 1.09 44 Desert Dust	Spring	210	1.3	1.0	-0.3	0.71	1.2	1.07	37
Fall 182 1.2 0.7 -0.4 0.62 1.1 1.09 44 Desert Dust Annual 1030 19.4 13.5 -5.9 0.47 21.9 0.68 37 Winter 88 14.4 5.3 -9.1 0.18 30.5 4.45 16 Spring 228 17.0 8.3 -8.7 0.44 17.9 0.73 19	Summer	193	1.2	1.1	-0.1	0.52	1.1	1.44	47
Desert Dust Annual 1030 19.4 13.5 -5.9 0.47 21.9 0.68 37 Winter 88 14.4 5.3 -9.1 0.18 30.5 4.45 16 Spring 228 17.0 8.3 -8.7 0.44 17.9 0.73 19	Fall	182	1.2	0.7	-0.4	0.62	1.1	1.09	44
Annual 1030 19.4 13.5 -5.9 0.47 21.9 0.68 37 Winter 88 14.4 5.3 -9.1 0.18 30.5 4.45 16 Spring 228 17.0 8.3 -8.7 0.44 17.9 0.73 19	Desert Dust								
Winter 88 14.4 5.3 -9.1 0.18 30.5 4.45 16 Spring 228 17.0 8.3 -8.7 0.44 17.9 0.73 19	Annual	1030	19.4	13.5	-5.9	0.47	21.9	0.68	37
Spring 228 17.0 8.3 -8.7 0.44 17.9 0.73 19	Winter	88	14.4	5.3	-9.1	0.18	30.5	4.45	16
	Spring	228	17.0	8.3	-8.7	0.44	17.9	0.73	19
Summer 409 22.0 16.9 -5.0 0.45 26.5 0.61 43	Summer	409	22.0	16.9	-5.0	0.45	26.5	0.61	43
Fall 305 19.2 15.1 -4.1 0.68 12.9 1.08 48	Fall	305	19.2	15.1	-4.1	0.68	12.9	1.08	48

renovation capacity due to the stable anticyclonic situation which generate recirculation episodes and accumulation in coastal areas (Baldasano et al., 1994; Millán et al., 1997), high natural resuspension of soil particles, enhanced frequency of African dust episodes and a high formation capacity of secondary aerosols by photochemical activity (Viana et al., 2002; Escudero et al., 2005). During this season the model is able to capture the amplitude of the daily variation for PM2.5 and PM10 (σ (Mod./Obs.) = 0.75 and σ (Mod./Obs.) = 0.51, respectively). Also high performances of CALIOPE are found during April, beginning November and December (Fig. 3a and b), where PM variability is attributed to the higher frequency of advection of Atlantic air masses, associated with a high rainfall rate.

However, in winter (February and ending November) the modeling system presents low skills to reproduce observation, underestimating the amplitude of the daily variation in PM2.5 and PM10 ($0.24 < \sigma$ (Mod./Obs.) < 0.29). Local anthropogenic emissions dominated high PM levels at urban and industrial sites in February 2004 (or winter 2004) when strong anticyclonic atmospheric condition (17 days) took place in the western Mediterranean Basin leading to the accumulation of PM (Pey et al., 2010). Under low dispersive conditions in February and ending November the modeling system presents the major problems.

3.2. Evaluation and assessment of PM speciation

The previous section analyzed the total concentration of PM2.5 and PM10. Although the modeling system reproduces quite well the temporal variability, systematic underestimations have been detected. It is important for these reasons to assess the modeling system ability to simulate the chemical composition of PM, not just the PM concentrations.

In the present section, we characterize the CALIOPE performance to reproduce the daily variability of the observed PM compounds $(nm-SO_4^{2-}, NO_3^{-}, NH_4^{+}, OC + EC, SS and DD)$ over Spain and we also estimate a multiplicative adjustment ratio for each compound. Several studies have demonstrated the benefit of adjusting sitespecific air quality models using observational data to reduce systematic model error (e.g. Hogrefe et al., 2006; Djalalova et al., 2010). Bias-adjustment strategies range from the relatively simple mean bias and multiplicative ratio adjustments used by McKeen et al. (2005) to the more complex Kalman filter techniques (Delle Monache et al., 2006; Kang et al., 2010). In this study we estimate a spatial homogenous correction factor based on the ratio of annual modeled mean over measured mean. Fig. 4 displays the annual time series of the raw and corrected daily modeled concentrations for PM chemical compounds for which the estimated adjusting ratio is different to 1.

A total of 8 stations are used to evaluate the simulated nm-SO₄⁻⁻, NO₃⁻, NH₄⁺ and TC concentrations in the PM2.5 fraction: 1 station is located in rural areas, 4 in suburban areas and 3 in urban areas (Table 1 and Fig. 1). 75% of these stations are background stations and the other 25% are industrial. SS is evaluated over 10 stations, 3 of them are inland stations (BAD, BUR and PUE), 3 are located over < 30 km away from the coastal line (LAC, MON and OND) areas, and 4 over the coastal line (BARC, BAS, HUL and PAL). DD is available over 16 rural background stations geographically distributed over Spain.

Furthermore, the present section provides a general description of the raw CALIOPE modeled seasonal mean spatial distribution of the aerosol compounds across Spain (Figs. 6 and 7).

3.2.1. Non-marine sulfate $(nm-SO_4^{2-})$

On average modeled and measured $nm-SO_4^{2-}$ concentrations present high correlation for daily concentrations (r = 0.62) (Fig. 2c). The modeling system reproduces 30% of the observed daily nm-SO₄⁻ concentrations indicating that modeled nm-SO $_4^{2-}$ is about 70% lower than observations along the trend line. At most stations, simulated annual means are underestimated by a factor of 3, with low absolute bias (MB = $-2.1 \ \mu g \ m^{-3}$). In order to examine the annual variability, Fig. 4a displays the annual time series of modeled and observed nm-SO₄²⁻ concentrations (black and red lines, respectively) on a daily basis averaged over all the stations. Modeled $nm-SO_4^{2-}$ concentrations agree with the daily observations, especially during summer and fall (r = 0.68 and r = 0.73, respectively) (Fig. 5). In this period nm-SO₄²⁻ exhibited a maximum (Observed Mean, OM = 4.3 μ g m⁻³) because of the higher SO₂ oxidation velocity under high insolation conditions (Querol et al., 2008). Nevertheless, the modeling system tends to underestimate the peaks during February, decreasing the correlation in winter (r = 0.46). SO₂ presents minimum



Fig. 2. Modeled vs. measured daily particulate matter (PM2.5 and PM10) and chemically speciated aerosols (nm-SO₄²⁻, NO₃⁻, NH₄⁺, TC, SS and DD) concentrations (in μ m⁻³) by station for the year 2004. Dots represent annual average concentration in the station. Horizontal and vertical lines indicate the 25 p and 75 p, respectively. PM10 (a) and PM2.5 (b) classifies stations as rural (green), suburban (orange) and urban (red). In c, d, e, and f each color represents a station (Fig. 1). SS (g) classifies coastal stations according to the ocean where they are located. DD (h) classifies stations according to geographic area. The expression located on the top indicates the adjust equation, the correlation coefficient (*r*), and the number of data points (*n*).

concentration in winter (OM = 2.9 μ g m⁻³) and the modeled nm-SO₄⁻ shows the largest underestimation (-2.2 μ g m⁻³)(Table 2). Winter underestimation of nm-SO₄⁻ is a common issue in most models in Europe, representing a direct couplet of sulfur chemistry with photochemistry, even detected with CMAQ over Europe (Matthias, 2008; Chemel et al., 2010). This feature can be explained

by the lack of model calculated oxidants or missing reactions (Kasibhatla et al., 1997; Tarrasón and Iversen, 1998; Schaap et al., 2004). After multiplying the model results by the estimated adjusting ratio (factor of 3) the corrected modeled nm- SO_4^{2-} concentrations (blue line in Fig. 4a) agree with the daily observations, especially during summer and fall.



Fig. 3. Time series of observed (red diamond) and simulated (blue continuous line) daily PM2.5 (left column) and PM10 (right column) for the year 2004 averaged over each type of site: (a, b) rural, (c, d) suburban, and (e, f) urban (Fig. 1 and Table 1 for the sites). The dashed lines represent the current limit values in Europe (European Commission, 2008) for PM10 (annual and daily limit values of 40 μ g m⁻³ and 50 μ g m⁻³ not to be exceeded 35 times per year, respectively) and for PM2.5 (annual limit value of 25 μ g m⁻³). Statistics for each type of site are showed in the figure (bottom right): Modeled Mean (MM), Observed Mean (OM), Mean Bias (MB) and correlation coefficient (*r*).

Fig. 6a and b shows the seasonal mean distribution of nm-SO₄⁻ concentration over Spain for winter and summer, respectively. nm-SO₄⁻ concentration has lower spatial variability than other PM2.5 species, with significant relationship with the distribution of emission sources, mainly downwind shipping and power plants, refineries and transformation industries that are the main responsible of high concentration of SO₂; nm-SO₄⁻ is mostly formed either from the gas-to-particle conversion of SO₂ in the atmosphere or from reactions in the aqueous phase. Because of the relatively slow rates of formation and removal, nm-SO₄⁻ becomes relatively well mixed (Park et al., 2006).

Over Spain, $(NH_4)_2SO_4$ presents minimum levels in winter (Fig. 6a) and reaches maximum concentrations in summer (Fig. 6b). This seasonal pattern is related to the stronger insolation favoring the higher oxidation of SO_2 to $nm-SO_4^{2-}$, low air renovation at regional scale, and the increment of the summer mixing layer depth favoring the regional mixing of polluted air masses. Overall, $nm-SO_4^{2-}$ concentrations decrease from the coast to the continent related to the fact that high-density of refineries and power plant are located near the coastal line, and also because shipping is also an important source of SO_2 .

During winter lowest concentrations ($<0.5 \ \mu g \ m^{-3}$) are detected over the main lberian mountain ranges meanwhile winter maximum concentrations around emission sources are $\sim 1 \ \mu g \ m^{-3}$. During summer the highest concentrations are found in the Atlantic regions (northwestern Spain) ($\sim 3.5 \ \mu g \ m^{-3}$) due to the presence of geographically-close power plants, and refineries in the area of Galicia, Ponferrada, Asturias and Bilbao where SO₂ modeled level is $\sim 10 \ \mu g \ m^{-3}$, or even 50 $\mu g \ m^{-3}$ around the La Coruña refinery. The dispersion pattern in northwestern Spain is significantly dominated by the northern and northeastern winds oxidizing SO₂ during the transport inland. High nm-SO₄^{2–} concentrations are present along the Mediterranean coast, ~2.5 μ g m⁻³. Northeastern Spain is mostly affected by the Andorra power plant from Teruel (Aragón region) where nm-SO₄^{2–} is secondarily formed during the canalization of the Ebro valley toward the Mediterranean Sea. Furthermore, shipping across the Mediterranean Sea may be a cause of high nm-SO₄^{2–} along the coast. Summer nm-SO₄^{2–} is low (~1–1.5 μ g m⁻³) in northern and southern plateaus corresponding with areas with low precursors emissions. The central Spanish plateau displays low levels of nm-SO₄^{2–} despite presenting one of the most important refinery and power plants around Puertollano (Castilla-La Mancha region).

Qualitative comparisons between the simulated pattern of seasonal nm- SO_4^{2-} and description based on measurements provided by Querol et al. (2008) reveals a good agreement. Such finding demonstrates the accuracy in the spatial description of the source regions and various industrial hot-spots.

Over the ocean high nm-SO₄²⁻ concentrations (~4 μ g m⁻³) are located in the shipping route from the Atlantic Ocean through the Strait of Gibraltar toward the Mediterranean Sea. As indicated in Baldasano et al. (2011), ship emissions are large contributors of the total SO₂ concentrations along the ship tracks, and the subsequent formation of nm-SO₄²⁻ downwind. Maximum nm-SO₄²⁻ concentrations are in the narrow Gibraltar regions with a dispersion pattern dominated by western winds. The high nm-SO₄²⁻ in this area is a combination of high emissions from shipping, industry, power generation and oil refinery processes in the Gibraltar bay and Algeciras industrial area (Pandolfi et al., 2011).



Fig. 4. Time series of observed (red lines), simulated (black line) and simulated corrected (blue line) concentration for daily nm-SO₄²⁻ (a), NO₃⁻ (b), NH₄⁺ (c) and TC (d) as an average of all the stations for the year 2004.

3.2.2. Nitrate (NO_3^-)

Fine ammonium nitrate (NH_4NO_3) is produced by the oxidation of NO_x to nitric acid (HNO_3) during night and day (e.g. Meng et al., 1997; Nguyen and Dabdub, 2002). HNO_3 dissolves in the aqueous particles along with the ammonia (NH_3) and produces NO_3^- determined by the composition of the particle. Therefore, the behavior of nitrate is one of the most intriguing issues of inorganic atmospheric aerosols because its concentration depends not only on the amount of gas-phase HNO_3 , but also on the availability of NH_3 and $nm-SO_4^2^-$, together with temperature and relative humidity (Mozurkewich, 1993; Nenes et al., 1998; Bassett and Seinfeld, 1983; Zhang et al., 2000; Schaap et al., 2004).

On average the CALIOPE system underestimates measured fine NO₃⁻ concentrations (Fig. 2d) by a factor 2 with very low bias $(MB = -1.0 \ \mu g \ m^{-3})$. The cause of this underestimation is related to errors in the simulation of NH_x (NH⁺₄ + NH₃), SO²₄⁻, and, to a lesser extent, TNO₃ (NO₃⁻ + HNO₃) (Yu et al., 2005). However, the temporal variability and therefore the transport patterns of NO₃⁻ are



Fig. 5. Mean Bias (a, in µg m⁻³) and Root Mean Squared Error (b, in µg m⁻³) for nm-SO₄²⁻ (green), NO₃⁻ (red), NH₄⁺ (dark blue), TC (yellow) and SS (cyan) averaged over the IDAEA-CSIC stations for each month of the year 2004.

captured (Fig. 4b) especially during fall (r = 0.49), although the modeling system hardly reproduces the amplitude of the events. Modeled NO₃ concentrations present a good agreement with the observation when a factor of 2 is applied to the results (blue line). As for nm-SO₄²⁻, the adapted results for NO₃ match consistently with the observations except for February.

The seasonal mean concentration pattern of fine NO₃⁻ over Spain (Fig. 6c and d) shows a marked seasonal evolution characterized by a summer minimum. The oxidation rate of NO_x ranges from 5% till 50% per hour (Spicer et al., 1981) faster than that of SO₂. Also, HNO₃ deposits rapidly and NH₄⁺ presents a low thermal stability in warm season (Harrison and Pio, 1983; Querol et al., 2008). Thus, NO₃⁻ is spatially less homogeneously distributed than nm-SO₄²⁻ in summer as shown seasonal distribution in Fig. 6d. Levels decrease from the coastal areas to inland, with the lowest regional background levels over central and southern plateaus (~0.3 µg m⁻³).

In contrast, high NO₃⁻ concentration and homogeneous distributions are found in winter (Fig. 6c) favored by the higher stability of NH₄NO₃ in cool conditions, which causes a higher portion of NO₃⁻ to partition to the aerosol, which has a longer lifetime than HNO₃ against deposition (Schaap et al., 2004) and favored by less dispersive conditions in the boundary layer. Overall, rural background NO₃⁻ concentrations remain ~ 1 µg m⁻³, reaching the lowest values in the higher mountains ranges (<0.5 µg m⁻³). High levels (~1.2 µg m⁻³) are found over the intensive agricultural areas of Plana de Lérida and Gerona and along the urban plume from Madrid metropolitan area which transports precursors from the agriculture and road traffic (see NH₃ and NO₂ patterns, Fig. S3, supplementary material) following a southwestern direction conditioned by the barrier of the Central System (2500 m height) and the canalization of Tajo Valley in the southern part (~ 1.2 µg m⁻³).

Furthermore, humidity plays a key role stabilizing NH₄NO₃ in the coasts, even at relatively high temperatures. In northeastern Spain (Galicia) there is no a significant seasonal trend of NO₃⁻. Levels remain ~0.9 µg m⁻³ due to the frequent Atlantic advection and the high-density industry located there. At Nervion valley, the seasonal trend of NO₃⁻ is smooth probably as a consequence of the significant summer breeze transport of atmospheric pollutants from the coastal industrial zones along the valley and the high humidity conditions.

Along the Mediterranean coast, the seasonal trend of NO_3^- presents low variability compared with central Spain as reported by Querol et al. (2008). The spatial difference is related to (1) the relative low disperse atmospheric condition prevailing in the Mediterranean Basin (Millán et al., 1997) compared with those in the Atlantic regions; (2) the high ammonia emission along the Mediterranean coast, above all in the northeast of Spain, caused by emission from intensive cultivation and farming (Fig. S3, supplementary material); and (3) the high humidity conditions which stabilize NH₄NO₃ even during the summer. Furthermore, traffic NH₃ emissions in urban

areas with high humidity (coastal areas) may also provoke high NH₄NO₃ levels (Athanasopoulou et al., 2008).

Thus, in contrast to nm-SO₄²⁻, NO₃⁻ levels are independent of the relevance of the regional/local emission of NO₃⁻ precursors. This spatial distribution shows no relationship either with road traffic flows (given that cities such as Madrid and Barcelona with similar inhabitants present different NO₃⁻ concentrations) or with industrial density so sites with similar industrial development (Huelva and Tarragona) have very different NO₃⁻ levels as shown by Querol et al. (2008).

3.2.3. Ammonium (NH⁺₄)

Along the year the overall variability of the modeled NH⁴ concentrations compares favorably with observations (Figs. 2e and 5) with r = 0.34. As for nm-SO₄²⁻, modeled annual mean of NH₄⁺ is underestimated by a factor of 3 with very low bias (MB = $-1.3 \,\mu g \, m^{-3}$). As for NO_3^- , NH_4^+ presents the worst performance during warmer seasons (0.15 < r < 0.32) while is better simulated during the colder seasons (0.42 < r < 0.48), especially during fall season (σ (Mod./Obs.) = 0.71, r = 0.48). When a factor of 3 is applied the magnitude of most of the single events is captured (Fig. 4c). Similar CMAQ model results were found in Matthias (2008) and Chemel et al. (2010) where NH₄⁺ concentration is underestimated in summer and the agreement with the observations is much better in winter and fall. The explanation is likely related to the uncertainties in the seasonal distribution of NH₃ emissions (Gilliland et al., 2006). There is not, to the author's knowledge, a large data set of NH₃ measurement available in Spain thus not so many conclusions can be taken over the equilibrium NO₃/NH₄. However, through the degree of $nm-SO_4^{2-}$ neutralization (DON) indicator, described in the supplementary material, it is possible to provide indirectly information about how well the modeling system estimates NH_3 concentrations. PM2.5 nm- SO_4^{2-} is almost always completely neutralized by NH₃ in the modeling system (Fig. S2, Supplementary material). Observations suggest that PM2.5 nm-SO $_{4}^{2-}$ is fully neutralized in warm seasons and exists in the form of NH⁺₄ bisulfate in the colder months. This finding suggests that there should be more free ammonia in the gas-phase or less NH⁺₄ in the photochemical model. The representation of NH₃ emissions can have a large effect on air quality model simulations of aerosol NO_3^- and NH_4^+ concentrations.

The bias performance of NH⁺₄ is strongly correlated with that of nm-SO²₄⁻ (r = 0.69) and NO³₃ (r = 0.62), because both anions are neutralized by NH⁺₄. Therefore, the performance of NH⁺₄ at single stations is mostly in between that of nm-SO²₄⁻ and NO³₃. At locations where nm-SO²₄⁻ dominates over NO³₃ the performance is close to that of nm-SO²₄⁻, whereas in regions with high modeled NH₄NO³ the performance is closer to NO³₃. CALIOPE performance for SIA is better for nm-SO²₄⁻ than for the other compounds based on high correlation coefficient (0.62 vs. 0.36–0.34) and normalized RMSE (0.82 vs. 1.3–1.1).



Fig. 6. Annual average concentration (in μ g m⁻³) for nm-SO₄²⁻ (a, b), NO₃⁻ (c, d), NH₄⁺ (e, f) and TC (g, h) in winter (right) and in summer (left) at lower-most level simulated by CALIOPE over Spain at a 4 km × 4 km spatial resolution. Winter corresponds to January, February and March. Summer corresponds to June, July and August. Red icons in a and b represent the Spanish refineries meanwhile white icons represent power plants.

NH[‡] is secondarily formed as NH₃ neutralizes H₂SO₄ and HNO₃. The amount of NH[‡] is highly dependent on the relative amounts of H₂SO₄ and NH₃, and indirectly of HNO₃. The seasonal distribution pattern of NH[‡] is shown in Fig. 6e and f for winter and summer, respectively. As for NO₃, the less dispersive conditions in the boundary layer favored a relative small spatial variability for NH[‡] during winter. Average levels are ~0.6 µg m⁻³, meanwhile over high mountain ranges the levels are low (<0.2 µg m⁻³). In summer NH[‡] concentrations present a more complex pattern. Concentrations over the coastal areas increase (~1.2 µg m⁻³) determined by its association with nm-SO[‡]. Over the Atlantic coast, northern winds determined the generation of NH[‡] inland.

3.2.4. Total carbon (TC)

The annual temporal evolution of the modeled TC concentrations compares reasonably well with the observations (r = 0.45), albeit the CALIOPE system underestimates the amplitude of daily concentrations, σ (Mod./Obs.) = 0.42 (Fig. 2f and Table 2). Most measured annual mean concentrations at individual stations are underestimated by a factor of 4. These results agree with a large number of studies showing that air quality models generally underestimate organic aerosol concentrations, especially in urban air masses (Carlton et al., 2008; Murphy and Pandis, 2009; Smyth et al., 2009 among others).

The urban background station at Barcelona presents the lowest underestimation (factor of 2). In cities, a large proportion of TC is emitted by road traffic. In large conurbations (>500,000 inhabitants), like Barcelona, the road traffic emission module is specifically described in HERMES, including the definition of the road network, divided in stretches (inside the 1 km \times 1 km cells), with specific temporary disaggregating profiles (distinguishing day-type: weekday-holiday, and month), average speed circulation, annual average daily traffic (number of vehicles per day), stretch length, route type (highway, road or urban) and circulation zones. Because of this complexity, the emissions at large cities are correctly described (Baldasano et al., 2008b) and hence the model tends to perform better with respect to areas with a larger uncertainty in emissions.

The TC temporal series averaged over all stations is shown in Fig. 5. The highest measured TC concentrations are found during winter (5.8 µg m⁻³). Although the modeled TC reproduces temporal variability with high correlation (r = 0.53), absolute MB and RMSE are relatively weak (MB = $-4.5 µg m^{-3}$, RMSE = $6 µg m^{-3}$). On the other hand, during summer TC is modeled with the lowest correlation (r = 0.31), partly due to lack of SOA yield paths. The SOA model adopted in CMAQv4.5 does not include biogenic SOA formation from isoprene and sesquiterpenes (Appel et al., 2008). The predominant vegetation types in the Spanish domain favor isoprene as the main biogenic volatile organic compound (bVOC) (Keenan et al., 2009). The absence of the isoprene-SOA route on SOA modeling in the domain of study may impact significantly air quality during summer, when elevated bVOC emissions combine with an enhanced photochemistry.

The model underestimations of TC are mainly related to (1) the state-of-the-science concerning to SOA formation pathways, as commented before (Eder and Yu, 2006; Edney et al., 2007; Appel et al., 2008) and (2) probable underestimation of primary carbonaceous emission (Cooke and Wilson, 1996; Bond et al., 2004; Tsyro et al., 2007).

Fig. 6 shows the winter (g) and summer (h) mean concentration of TC over Spain in 2004. Elemental carbon is a primary pollutant so its spatial variability is relatively high. Major sources of elemental carbon include diesel engines, particulate heavy-duty trucks, and combustion process (including biomass and fossil fuel), thus high levels are associated with urban areas. In contrast, organic carbon is emitted directly or, in a large proportion, formed from the condensation of low-volatility organic compounds. Thus, the spatial variability of organic carbon is between that of purely primary and secondary pollutants. Major primary sources of organic carbon include diesel and gasoline-burning engines, biomass burning and some industrial processes, so organic carbon will be found in urban and rural background environments.

Due to the high disaggregation and the specifically detailed emissions implemented in the HERMES model, the impact of principal highways is noticeable, (e.g. Madrid-Sevilla, Barcelona-Bilbao, the A-7 Mediterranean highway) where annual mean values range from 0.5 to 2 μ g m⁻³.

In urban environment (eg. Zaragoza, Valencia, Málaga, Sevilla, Madrid and Barcelona) the seasonal trend of TC is characterized by a winter maximum (Rodríguez et al., 2002 at Barcelona site and Querol et al., 2008 at many sites), indicating a major local anthropogenic origin. Reduced dispersion conditions in the boundary layer together with enhanced condensation due to the low temperature explain the seasonal evolution of carbon in the urban environment. The urban plumes from the two largest Spanish cities (Madrid and Barcelona) reach the highest TC concentrations ($\sim 5 \ \mu g \ m^{-3}$). In both regions, on-road traffic constitutes the main source of primary pollutants in the region (Gonçalves et al., 2009) and TC follows the same pattern as NO₂, as shown in Baldasano et al. (2011) and in Fig. S3, Supplementary material.

In Madrid during winter the TC dispersion follows a southwestern direction conditioned by the barrier of Central System and the canalization of Tajo valley dominated by the predominant Atlantic advection episodes. The urban plume reaches the highest concentration at the urban nuclei ($\sim 7 \ \mu g \ m^{-3}$), moving toward Toledo (south) and reaching Guadalajara (east) to a lesser extent ($\sim 1.5 \ \mu g \ m^{-3}$). In contrast in summer, the urban plumes is less disperse, determined by the developed of the IP Thermal Low over the central plateau which couple with several deep convective cells that inject aged pollutant aloft for the Madrid area.

A different TC pattern is observed in Barcelona. The TC dispersion shows a perpendicular flow to the coast dominated by the northwestern winds. The dispersion pattern is intensified in summer due to the frequent regional recirculation events characterized by sea breeze over the Mediterranean (Baldasano et al., 1994; Millán et al., 1997). The very complex coastal terrain induces mesoscale phenomena which control the superficial wind flows. Sea-breezes and mountain valley winds contribute to the accumulation and recirculation of air masses. The littoral mountain chain (1000–1500 m height) acts as a barrier, recirculating TC flow toward the Mediterranean Sea channeled by the river valleys. Also the high isolation favors the formation of SOA.

The industrialized area of Tarragona and Castellón and the urban area of Valencia, all located along the Mediterranean coast, present significant levels of TC also affected by mesoscale phenomena dominated by sea-breezes which determine TC flow perpendicular to the coast. In northern Spain, the urban and industrial areas of Bilbao show significant TC levels dispersed along an estuary that runs almost 16 km from the center of the city to the sea and is aligned in an SE–NW direction.

In contrast, background levels of carbonaceous aerosols exhibited a light spatial variability, but a seasonal trend characterized by a summer maxima ~0.5 μ g m⁻³. An important fraction of SOA could arise from bVOC emissions given their increase in the warm season due to enhanced plant transpiration (Rodríguez et al., 2002; Keenan et al., 2009) and also from biomass burning from agricultural and domestic sources.

Similarly to nm-sulfate, the major shipping route originating from the northern Atlantic, toward northern Africa is shown to have a notable impact on TC concentrations, which is intensified in summer due to the increase of maritime traffic and the favored photochemistry. Concentrations of TC range from ~1.5 $\mu g~m^{-3}$ near the coastlines of Portugal and Gulf of Cadiz to ~2 $\mu g~m^{-3}$ over the Alboran Sea. Background concentration of TC over the sea ranges between ~0.7 $\mu g~m^{-3}$ and <0.2 $\mu g~m^{-3}$ over the Mediterranean and Atlantic Sea, respectively, owing to distinct Atlantic-Mediterranean regimes of Spain.

3.2.5. PM10 sea-salt aerosol (SS)

For SS, the modeling system reproduces correctly 70% of the observed values with a correctly temporal variability (r = 0.65) (Fig. 2g). 40% of modeled concentrations are within a factor of 2 with observations (Table 2). The modeling system presents a good skill to reproduce the amplitude of the daily variation of SS with σ (Mod./Obs.) = 1.17. Highest correlations are obtained during spring and winter when high wind velocities take place and the flux of SS is high. However, under stagnant situations, like summer over Mediterranean, the modeled sea-salt is reproduced with lower correlation and higher daily variability, r = 0.52 and σ (Mod./Obs.) = 1.44, respectively. As shown in Fig. 2g, at coastal stations the CALIOPE system tends to slightly overestimates SS, meanwhile at continental areas salts concentrations tends to be underestimated. Over the southern Atlantic and central Mediterranean the CALIOPE system performance agrees better with the observations. Part of the underestimation and the excessive variability modeled is because AERO4 considers coarse-mode aerosols as dry and inert. This approach does not allow important aerosol processes, such as the replacement of chloride by NO_3^- in mixed marine/urban air masses (Kelly et al., 2010) especially intense in summer when HNO₃ is released by the thermal instability of NH₄NO₃ (Querol et al., 2004). Moreover, degassing of Cl⁻ is not implemented in the model, and heterogeneous reactions are not taken into account. Transfer from PM10 to PM2.5 is also not considered in AERO4. Furthermore, only sea-salt emissions from open-ocean are included so the detected underestimation over coastal areas is in part due to the fact that surf-zone mechanism is not implemented (Athanasopoulou et al., 2008). The results subscribe the modeled SS calculation performed over Mediterranean Basin in Jiménez-Guerrero et al. (2011).

Fig. 7a and b shows the non-uniform behavior of SS in the surrounding Spanish oceans, showing a strong seasonality and dependence on the wind speed variation. Over the ocean, the maximum concentrations are simulated during winter. A winter maxima of $\sim 12 \ \mu g \ m^{-3}$ is detected in the northern Atlantic Ocean and Gulf of Lion, the first related to the occurrence of intense frontal systems and strong winds events that peak during this season (Alpert and Ganor, 1993; Saaroni et al., 1998). Zakey et al. (2008) state that the Gulf of Lion is a region of strong Mediterranean cyclogenesis, with maximum activity taking place in the winter and fall months (Camuffo et al., 2000; Cavaleri, 2005). Under this condition maximum SS concentrations over the continents are found along the Atlantic coastal area of Galicia ($\sim 3 \ \mu g \ m^{-3}$) and over the whole Mallorca (\sim 3 µg m⁻³) and Menorca (\sim 9 µg m⁻³) islands affected by the dynamics in the Gulf of Lion. A different SS pattern is observed in summertime accentuated by the Atlantic-Mediterranean regimes. Maximum concentrations are simulated over the eastern Strait of Gibraltar ($\sim 10 \ \mu g \ m^{-3}$). The Azores high-pressure system is displaced westward and it develops high western winds which are



Fig. 7. Annual average concentration (in μ g m⁻³) for PM10 sea-salt aerosol (SS) (a, b), and PM10 desert dust (DD) (c, d), in winter (right) and in summer (left) at lower-most level simulated by CALIOPE over Spain at a 4 km \times 4 km spatial resolution. Winter corresponds to January, February and March. Summer corresponds to June, July and August.

intensified when they reach the strait due to its complex topography inducing SS dispersion aligned in a west to east direction (Millán et al., 2002). On the other hand, along the Mediterranean coast SS presents the maximum concentration in summer ($\sim 2 \ \mu g \ m^{-3}$) when the intensive thermal inversions, the lower convective dynamics and the low dispersive local circulations intensifies sea breeze circulation over the coast and transport SS inland.

3.2.6. PM10 desert dust (DD)

Fig. 2h presents the comparisons of the modeled DD surface concentration from BSC-DREAM8b with the DD estimated according the methodology reported by Escudero et al. (2007b). Despite the episodic character of Saharan dust outbreaks and their dependency on the synoptic scale, when BSC-DREAM8b is compared with measurements it presents a rather high correlation coefficient (r = 0.47) and captures the amplitude of the daily variations of desert dust episodes with high confidence, $\sigma(Mod./Obs.) = 0.68$.

Fig. 8 shows the time series at four stations considered to be representative of the different geographical areas where Saharan dust presents a significant contribution: central (a), eastern (b), southwestern (c) and southeastern (d) Spain. BSC-DREAM8b is able to reproduce most of the Saharan dust outbreaks along the year 2004, both in time and value. Highest correlations were obtained for southeastern stations, like Víznar (r = 0.57), which are directly affected due to the proximity to the African continent and at high altitude. However, BSC-DREAM8b underestimates some dust events associated with meteorological situations when emissions come directly from the Atlas region. The complex orography of this region produces a mesoscale phenomenon that hampers the simulation of meteorological parameters at surface level (like winds and precipitation) and consequently, the dust emissions. The observed episode in middle March is difficult to reproduce by BSC-DREAM8b over central and southern Spain. During the 18th and 19th March, Saharan mineral dust followed an Atlantic trajectory from North Africa achieving the western part of the IP. The low Saharan dust observed before the aforementioned episode was caused by the arrival of Atlantic air masses preceding the northward high particulate flow. Another important dust event is underestimated by the BSC-DREAM8b during late July.

As shown in Fig. 7 modeled DD concentration shows a high spatial variability marked by a south-to-north gradient with minimum levels in the north ($<2 \ \mu g \ m^{-3}$ as annual average) and maxima in the southeastern Spain (\sim 7–9 µg m⁻³ as annual average) linked to frequent African dust plumes affecting IP (Rodríguez et al., 2001; Escudero et al., 2007b; Basart et al., 2009). These results agree with findings presented in Ouerol et al. (2009a). In addition, the DD contribution presents significant different patterns between winter (Fig. 8c) and summer (Fig. 8d) governed by synoptic situations as described Escudero et al. (2005). In winter African dust outbreaks were less frequent in 2004 (Fig. 8) and mainly affecting the southwestern Spain ($\sim 6 \ \mu g \ m^{-3}$) since a typical winter African transport scenario consists on the location of an anticyclone over the IP at surface levels. This creates a curved plume affecting the IP firstly on its western flank. Thus, high DD contribution during winter is found there ($\sim 5 \,\mu g \, m^{-3}$). On the other hand, in summer large amounts of DD are transported along southern and eastern Spain linked to frequent African dust outbreaks (Rodríguez et al., 2001; Escudero et al., 2005; Sicard et al., 2011; among others). Saharan dust reaches southern Spain when the north Atlantic anticyclone (Azores high) is displaced westward and the North African high is centered over Algeria (Rodríguez et al., 2001). In these episodes, DD is transported at high altitudes, between 1500 and 4000 m a.s.l (Talbot et al., 1986; Escudero et al., 2005; Pérez et al., 2004, 2006b; Olmo et al., 2008; Córdoba-Jabonero et al., 2011). The largest summer mean contribution of DD coincides with the Sierra Nevada mountain range with values reaching up to 40 μ g m⁻³. Such concentrations are the consequences of (1) the mountain range location within the main zone of dust (Middleton and Goudie, 2001) and (2) the existence of several peaks within the mountain range 3000 m a.s.l. which corresponds to the altitude range for Saharan dust transport (Córdoba-Jabonero et al., 2011). In a lesser extent, summer concentration of PM10 DD load reaches concentration between 9 and 15 $\mu g\ m^{-3}$ over southern, central and eastern Spain, including the Tramontane Mountain Range (Balearic Islands).

3.3. Modeled aerosol performance goals and criteria

Fig. 9 shows the annual MFE and MFB at each of the studied stations for PM10, PM2.5, nm-SO $_{4}^{2-}$, NO $_{3}^{-}$, NH $_{4}^{+}$, TC, SS and DD.



Fig. 8. Time series of observed (red diamond) and simulated (blue continuous line) daily PM10 desert dust load for the year 2004: at (a) Risco Llano, (b) Zarra, (c) Barcarrota, and (d) Víznar (Fig. 1 and Table 1 for the location of the sites). The continuous red lines represent the current limit values for PM10: the annual limit value of 40 μ g m⁻³ and the daily mean value of 50 μ g m⁻³ not to be exceeded 35 times in a year (European Commission, 2008).



Fig. 9. Mean fractional error (MFE, left) and bias (MFB, right) by station as annual average compared to proposed performance goals and criteria by Boylan and Russell (2006). Plots are carried out for different species: PM2.5 and PM10 (a and b), nm-SO²₄, NO³₃, NH⁴₄ and TC (c and d), SS (e and f) and DD (g and h). Solid symbols represent statistics for the raw modeled outputs. Empty symbols indicate corrected values calculated from the statistics evaluation. The corrected factors are 3 for nm-SO²₄, 2 for NO³₃, 3 for NH⁴₄ and 4 for TC. Corrected PM2.5 and PM10 are calculated from the corrected chemical species.



Fig. 10. Seasonal average concentrations at lower-most level (μ g m⁻³) of (left) raw modeled PM2.5 and (right) corrected PM2.5 when multiplicative correction factors are applied to nm-SO₄²⁻, NO₅, NH₄⁺, and TC in the PM2.5 fraction. Seasons correspond to winter (a and b), spring (c and d), summer (e and f) and fall (g and h). Points represent measured seasonal averaged concentrations at the IDAEA-CSIC stations. The number of data points used to compute measured seasonal averaged concentrations is indicated in Table 2 in the column named Data. The number at bottom-left in each figure indicates the spatial correlation between model and observations.



Fig. 11. Seasonal average concentrations at lower-most level (μ g m⁻³) of (left) raw modeled PM10 and (right) corrected-PM10 (right) when multiplicative correction factors are applied to nm-SO₄²⁻, NO₃⁻, NH₄⁺, and TC in the PM2.5 fraction. Seasons correspond to winter (a and b), spring (c and d), summer (e and f) and fall (g and h). Points represent measured seasonal averaged concentrations at the IDAEA-CSIC stations. The number of data points used to compute measured seasonal averaged concentrations is indicated in Table 2 in the column named Data. The number at bottom-left in each figure indicates the spatial correlation between model and observations.

The Boylan and Russell (2006) criteria for MFB and MFE are not met for most of the stations for PM10 and PM2.5. Fig. 9 also shows the MFE and MFB for the aforementioned species when the multiplicative correction factors from the statistical evaluation are applied over individual species. When applied, most of the stations largely meet the model performance goals for all the species. For background stations the correction factors are effective to meet the goals although they result less effective at traffic and industrial stations.

For nm-SO^{2–}, NO₃, NH[±] and TC (Fig. 9c and d) the model performance criteria for MFE is met at 21 stations out of 32. Nevertheless for MFE only 50% of the stations meet the model performance criteria. Better results are obtained for nm-SO^{2–}₄ (-100% < MFB < -90%, MFE < 100%) than for NO₃, NH[±] and TC. The stations of Barcelona meet the criteria for MFB and MFE for OC + EC. When bias-correction factors are applied all the stations meet the criteria for MFB and MFB, and most of them meet the goals.

For SS only 8 stations out of 9 accomplished the criteria proposed by Boylan and Russell (2006). Most of the largest errors (MFE > 130%) and biases (MFB < -120%) correspond to inland stations (Fig. 9e and f). For DD (Fig. 9g and h) the central eastern and southern stations meet or nearly meet the Boylan and Russell criteria, 9 stations out of 16 corresponding to desert dust concentrations of ~ 16 µg m⁻³. The highest MFE and MFB are found in northern and northeastern stations associated to low DD concentrations (~9 µg m⁻³).

3.4. Spatial and seasonal patterns of PM2.5 and PM10

The present section provides a general description of the seasonal mean spatial distribution of PM size fractions (PM2.5 and PM10) across Spain. Note that the discussion performed over PM2.5 and PM10 takes into account the model-observations discrepancies highlighted in Sect. 3.2 for each compound by means the adjusting ratios estimated.

Fig. 10, left column, presents the seasonal mean concentration simulated with CALIOPE for PM2.5. From the statistical evaluation of the chemical compounds, multiplicative bias-correction factors where estimated for nm-SO $_4^2$ -, NO $_3^2$, NH $_4^+$, and TC with the objective to reduce the annual mean bias. Fig. 10, right column, shows the modeled PM2.5 when the aforementioned factors are applied for each species over the whole domain. Furthermore, seasonal mean observations are also represented in both cases. In order to quantify the CALIOPE performance to reproduce the observed spatial and seasonal variability of PM2.5, the spatial correlation (*r*) is shown bottom-left. PM10 seasonal and spatial variability is discussed analogously in Fig. 11.

When the factors are applied individually to PM2.5 compounds, spatial correlations improve during all the seasons. The largest improvements are found in summer (from 0.48 to 0.70) associated with the increased formation of secondary aerosols, mainly nm- SO_4^{2-} and TC as shown in Sect. 3.2.4. According to PM10, modeled corrected concentrations only improve correlation coefficients during winter (from 0.60 to 0.66) and spring (from 0.59 to 0.68). There is an increase in the observed PM2.5 and PM10 levels during summertime, which is more pronounced in the coarse fraction than in the finer fraction due to the contribution of coarser size of the resuspended dust and summer-nitrate particles (Querol et al., 2009b). The applied factors do not correct these compounds, given that coarse NO₃ is not implemented in the current version of CMAQv4.5 and only re-suspended material from road traffic is included. Overall, corrected-PM10 and PM2.5 from CALIOPE increase correlation spatially and seasonally when multiplicative correction factors for SIA and TC are applied. However, further analyses should be carried out on the coarse fraction.

Note that when the model is corrected, systematic high PM2.5 values (~10 μ g PM2.5 m⁻³) are found along the Mediterranean coast, Guadalquivir and Tajo valleys, where the impact of principal highways (Barcelona-Bilbao, the A-7 Mediterranean highway) and industrial sources (Huelva, Valencia, Tarragona) are noticeable. The highest PM2.5 concentrations are located around the biggest Spanish cities of Madrid and Barcelona (>15 μ g PM2.5 m⁻³). During summer PM2.5 background level increase, and the concentration pattern presents a low spatial variability (~10 μ g PM2.5 m⁻³). Similar patterns are found for corrected modeled PM10, although concentrations are dominated by the DD contribution in the southern and eastern IP.

4. Conclusions

The CALIOPE system has been evaluated in terms of PM levels (PM2.5 and PM10) and chemical compounds (nm-SO₄²⁻, NO₃⁻, NH₄⁺, TC, SS and DD) at ground level. Even more, CALIOPE annual results over the full year 2004 have been used to describe the seasonal pattern of PM and their chemical composition in Spain, increasing the knowledge on aerosol distribution.

The model performance presents wide variations among the different geographical areas in Spain. These variations also depend on the type of environment. The evaluation depicts that CALIOPE can reproduce the variability of PM2.5 and PM10 and their main aerosols compounds over Spain (0.45 < r < 0.64). However, concentrations are underestimated. At urban and suburban stations PM2.5 and PM10 levels present bias of $-15 \ \mu g \ PM2.5 \ m^{-3}$ and $-25 \ \mu g \ PM10 \ m^{-3}$, where 40% and 30% of these biases are due to the underestimation in the regional background concentrations.

For the first time, the BSC-DREAM8b model coupled within CALIOPE system has been evaluated in terms of DD surface concentration in the PM10 fraction. The BSC-DREAM8b reproduces most of the Saharan dust outbreaks in 2004, in time and value, in areas of central, southwestern and southeastern Spain. However the BSC-DREAM8b presents difficulties to reproduce episodes where Saharan mineral dust followed an Atlantic trajectory from North Africa achieving the western IP. These later episodes are less intense than the summer ones. Concerning SS in PM10, CALIOPE reproduces correctly 70% of the daily observed concentration with a correctly temporal variability. At coastal stations the modeling system tends to simulate slightly higher SS than that measured meanwhile at continental areas SS tends to be underestimated. For both natural aerosol, DD and SS, CALIOPE accomplishes the model performance criteria proposed by Boylan and Russell (2006).

The most important underestimations are linked to nm-SO₄²⁻, fine NO $_3^-$, NH $_4^+$ and TC. nm-SO $_4^{2-}$ and NH $_4^+$ are underestimated in a factor of 3, which is resembled in the model performance for $NO_{\overline{3}}$ which is underestimated in a factor of 2. Most uncertainties in the equilibrium nm-SO₄²⁻/NO₃/NH₄⁺ are related to the treatment of NH₃ emission as it exhibits large spatial and temporal gradients. Annual mean concentration of TC at individual stations correlates by a factor of 4. These results agree with a large number of studies showing that air quality models generally underestimate organic aerosol concentrations, especially in urban air masses associate with uncertainties in secondary organic aerosol formation and bVOC emissions. Both temporal and spatial evaluations indicate that modeled skills for the aforementioned species improve when bias multiplicative bias-correction factors are applied over each compound. Temporally, corrected model results accomplish model performance criteria for MFB and MFE established by Boylan and Russell (2006) for each compound and total levels (PM2.5 and PM10) at each station. Spatially, maximum improvements are found in summer for PM2.5 where global correlation increases from 0.48 to 0.70.

Concerning the CALIOPE results across the whole Spanish domain, PM chemical compounds exhibit geographical variations as well as marked seasonal patterns which are in agreement with the literature findings nm-SO $_4^{2-}$ is found to be dependent mostly on the degree of industrialization of the regions. Fine NO_3^- concentration exhibits a winter maxima with no significant relationship with the distribution of emission sources, but highly dependent to the temperature and humidity. NH⁺ spatial variability is relatively small in winter, usually being dominated by its association with nm-SO $_4^{2-}$. TC presents a clear winter maximum over urban environments that show its high correlation with anthropogenic emissions. SS shows a strong seasonality and dependence on the wind speed variation. Saharan air masses reach Spain when the synoptic situation is governed by depressions located to the west or southwest of the IP (winter) or when the North African anticyclone shifts to the East or Southeast of the IP (summer). DD contribution decreases from southeastern to northeastern Spain, with maximum contribution in summer $(7-30 \ \mu g \ m^{-3})$ affecting more than 50% of the Spanish country.

PM2.5 and PM10 seasonal patterns registered maximum concentration in summer, associated with the higher frequency of African dust outbreaks, lower precipitation, higher resuspension due to soil dryness, increased formation of secondary aerosols and recirculation of air masses that prevent air renovation. Our conclusion agrees with Flaounas et al. (2009) who showed that in Mediterranean areas the use of large-scale dust emissions and transport models, such as the BSC-DREAM8b, provide more reliable PM10 simulations. Furthermore, modeled annual DD contribution in Spain subscribes the measurements which estimated that the mean annual net dust contribution to the annual PM10 mean recorded at rural background sites reached $< 2 \,\mu g \, m^{-3}$ in the northern Spain and $6 \ \mu g \ m^{-3}$ in the southeastern Spain (Querol et al., 2009a). On the other hand, PM10 and PM2.5 patterns present the highest variability during winter when intense pollution episodes of either anthropogenic (winter anticyclonic scenarios) or natural (African dust) origins take place. Local pollution episodes occur mainly in these periods: the anthropogenic emission increase and the atmospheric condition favors the accumulation of pollutants around the emission sources (development of intensive thermal inversion, lower convective dynamics and low dispersive local circulations).

The results of this study suggest an increase of the spatial coverage and temporal resolution of data sets on organic and inorganic aerosols. Furthermore, measurements of gas precursor like HNO₃, NH₃ and VOC are currently needed to test model results and identify sources of uncertainty in aeosol modeling.

The present analysis demonstrates that CALIOPE high-resolution modeling system (4 km \times 4 km, 1 h) is a useful tool which complements measurements for increasing our knowledge about PM in terms of levels and composition in (1) urban/industrial areas with a pervasive influence of anthropogenic emissions on a local scale and (2) areas with very complex terrains and meteorology like Spain. However, there are a number of possible improvements to be done in chemical transport models. Within CALIOPE, and because of the results presented here, future work should be devoted to the improvements of secondary organic aerosol formation and dynamic interactions of fine and coarse fractions by the implementation of the updated version of CMAQ (v4.7.1, aerosol module AERO5) and revision of ammonia and bVOC emission implemented in HERMES model.

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Appendix. Supplementary material

Supplementary data related to this article can be found online at doi:10.1016/j.atmosenv.2011.09.049.

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