A regional air quality forecasting system over Europe: the MACC-II daily ensemble production


1 Groupe d’étude de l’Atmosphère Météorologique/Centre National de Recherches Météorologiques, CNRS-Météo-France, UMR3589, Toulouse, France
2 European Centre for Medium-range Weather Forecasts, Reading, UK
3 Swedish Meteorological and Hydrological Institute, Norrkoping, Sweden
The MACC-II daily ensemble production

V. Marécal et al.

Received: 28 December 2014 – Accepted: 2 February 2015 – Published: 11 March 2015

Correspondence to: V. Marécal (virginie.marecal@meteo.fr)

Published by Copernicus Publications on behalf of the European Geosciences Union.
Abstract

This paper describes the pre-operational analysis and forecasting system developed during MACC (Monitoring Atmospheric Composition and Climate) and continued in MACC-II (Monitoring Atmospheric Composition and Climate: Interim Implementation) European projects to provide air quality services for the European continent. The paper gives an overall picture of its status at the end of MACC-II (summer 2014). This system is based on seven state-of-the-art models developed and run in Europe (CHIMERE, EMEP, EURAD-IM, LOTOS-EUROS, MATCH, MOCAGE and SILAM). These models are used to calculate multi-model ensemble products. The MACC-II system provides daily 96 h forecasts with hourly outputs of 10 chemical species/aerosols (O$_3$, NO$_2$, SO$_2$, CO, PM$_{10}$, PM$_{2.5}$, NO, NH$_3$, total NMVOCs and PAN + PAN precursors) over 8 vertical levels from the surface to 5 km height. The hourly analysis at the surface is done a posteriori for the past day using a selection of representative air quality data from European monitoring stations.

The performances of the system are assessed daily, weekly and 3 monthly (seasonally) through statistical indicators calculated using the available representative air quality data from European monitoring stations. Results for a case study show the ability of the median ensemble to forecast regional ozone pollution events. The time period of this case study is also used to illustrate that the median ensemble generally outperforms each of the individual models and that it is still robust even if two of the seven models are missing. The seasonal performances of the individual models and of the multi-model ensemble have been monitored since September 2009 for ozone, NO$_2$ and PM$_{10}$ and show an overall improvement over time. The change of the skills of the ensemble over the past two summers for ozone and the past two winters for PM$_{10}$ are discussed in the paper. While the evolution of the ozone scores is not significant, there are improvements of PM$_{10}$ over the past two winters that can be at least partly attributed to new developments on aerosols in the seven individual models. Nevertheless, the year to year changes in the models and ensemble skills are also linked to
the variability of the meteorological conditions and of the set of observations used to calculate the statistical indicators. In parallel, a scientific analysis of the results of the seven models and of the ensemble is also done over the Mediterranean area because of the specificity of its meteorology and emissions.

The system is robust in terms of the production availability. Major efforts have been done in MACC-II towards the operationalisation of all its components. Foreseen developments and research for improving its performances are discussed in the conclusion.

1 Introduction

The chemical composition of the air close to the Earth’s surface, generally referred as “air quality” (AQ), directly affects human and animal health and also the vegetation. For instance, ozone has a known impact on the respiratory system (e.g. WHO, 2004) and on the vegetation development (e.g. Fuhrer and Booker, 2003). Recently, the World Health Organisation reported that in 2012 around 3.7 million people deaths are attributable to ambient air pollution (http://www.who.int/phe/health_topics/outdoorair/databases/en/). This is why air quality has become a major concern, starting in the 1970’s, in particular in Europe (e.g. WHO, 2013). Since the Helsinki Protocol in 1985, many regions and countries, including the European Union countries, have progressively put in place tools to regulate and to control the emissions of the main air pollutants. This has led to an important effort to monitor the air composition near the surface but also to develop air quality forecasting systems in experimental or operational modes (see reviews by Ebel et al., 2005; Menut and Bessagnet, 2010). These tools can be used in cases of high pollution episodes for the information to people and to take emergency measures to prevent harming effects. They can also be used for policy makers for the regulations on air pollutant emissions and for monitoring the effect of these regulations on air quality (episodes and also background pollution).

The main pollutants under focus for air quality are ozone, nitrogen oxides ($\text{NO}_x = \text{NO}_2 + \text{NO}$), sulphur dioxide ($\text{SO}_2$), Volatil Organic Compounds (VOCs), ammonia
(NH\textsubscript{3}), particulate matter, heavy metals (Pb, Cd, Hg) and Persistent Organic Pollutants (POPs, e.g. pesticides and dioxine). Ozone is a secondary pollutant meaning that it is not emitted but produced from gaseous precursors (mainly VOCs and NO\textsubscript{x}) originating from both natural and anthropogenic sources. Particulate matters (PM) correspond to small size aerosols. PMs are categorised as PM\textsubscript{10} (size < 10 µm), PM\textsubscript{2.5} (size < 2.5 µm) and PM\textsubscript{1} (size < 1 µm). These categories were chosen because of their known effects on health. In PMs, there is no distinction between primary (dust, sea salts, black carbon and organic carbon) and secondary aerosols formed from gaseous precursors such as SO\textsubscript{2}, DMS, H\textsubscript{2}S, NH\textsubscript{3}, NO\textsubscript{x} and VOCs.

Besides the development of surface measurement networks for these main pollutants, there has been a sustained research effort on the atmospheric chemistry modelling for air quality forecasting purposes. Regional and local air quality forecasting systems (Kukkonen et al., 2012; Zhang et al., 2012) rely on limited area models that can be based either on an off-line or an on-line approach to take into account the effect of meteorological conditions on air composition. Off-line chemistry models, known as Chemistry Transport Models (CTMs), use the meteorological parameters from the analyses or the forecasts provided by a separate numerical weather prediction model. On-line models are meteorological models in which chemical variables and processes are included (Baklanov et al., 2014). On-line models have the capability to represent the feedback of the chemical composition on meteorological parameters but they are computationally demanding by design. This is why CTMs are generally preferred for operational air quality forecasting systems.

The chemical composition of air depends on many processes that need to be well represented in models in order to provide reliable air quality forecasts (e.g. Rao et al., 2011). The composition near the surface is very much driven by emissions but also by chemical processes (gaseous/heterogeneous reactions and photolysis) including the production of secondary pollutants, by the advection by winds, by the diffusion in the planetary boundary layer, by the scavenging by rain and by the dry deposition at the surface. Each of these processes has its own uncertainty. These uncertainties come,
on one hand, from the limit of our current knowledge and, on the other hand, from
the need to simplify the process representation in models because of computational
constraints. In meteorology and climate studies, and more recently in atmospheric
dispersion and chemistry modelling, the approach based on a multi-model ensemble of
forecasts has been developed to improve their quality through statistical approaches.
The methods vary from very simple such as the average or the median to more elabo-
rated such as weighted averages based on past scores or Bayesian models or spectral
methods (e.g. Delle Monache, 2006; Riccio et al., 2007; Potempski et al., 2010; Gal-
marini et al., 2013).

The European Union is very much involved in air quality issues not only through
a series of protocols on emissions and consecutive political actions but also by sup-
porting research activities aiming at developing tools for air quality monitoring for Eu-
rope. These activities were initiated in the GEMS (Global and regional Earth-system
(Air) Monitoring using Satellite and in-situ data, FP6, 2005–2009; Hollingworth et al.,
2008) and PROMOTE (ESA PROtocol MOniToring for the GMES Service
ued in the MACC (Monitoring Atmospheric Composition and Climate, FP7, 2009–
2011), MACC-II (Monitoring Atmospheric Composition and Climate: Interim Imple-
mentation, FP7, 2011–2014) and now MACC-III (Monitoring Atmospheric Composition
and Climate-III, H2020, 2014–2015) projects. One of the major achievements done in
GMES, MACC and MACC-II for European AQ objectives is the development and the
exploitation of a pre-operational analysis and forecasting system run on a daily basis.
This system is based on the combined use of an ensemble of air quality models. GEMS
involved 10 research and operational models. Evolving towards a pre-operational sys-
tem, the MACC/MACC-II/MACC-III ensemble is, since 2009, based on seven following
state-of-the-art regional CTMs that are all developed and run in Europe: CHIMERE,
EMEP (MSC-W version), EURAD-IM, LOTOS-EUROS, MATCH, MOCAGE and SILAM.
They are used to produce a multi-model ensemble for major monitored pollutants. Al-
though each of these models can perform very well on particular days in particular
areas, the ensemble approach provides on average forecasts and analyses of better quality than any of the individual models. It also gives an indication of the uncertainties through the spread between the models. Similarly to meteorological forecasts, the quality of the AQ forecasts needs to be routinely evaluated to provide information to users on its reliability. The performance of the individual and ensemble forecast products is evaluated on a daily basis from comparisons with available surface observations from the European AQ station network. Additionally the system has been providing birch pollen forecasts at surface during the pollen season since 2013.

The objective of the paper is to give a description of the pre-operational analysis and forecasting system in place within MACC and MACC-II to provide AQ services for the European continent and of its performances. Since the system continuously evolves with time, we present here the configuration at the end of the MACC-II project (summer 2014) with short information on recent upgrades done before the end of 2014. An overview of the analysis and forecasting system, including the seven models and the Ensemble, is provided in Sect. 2. Section 3 is devoted to the system performances on case studies and on a seasonal basis. Section 4 gives a summary and the perspective of short and mid-term developments of the MACC-II system and associated research.

2 Description of the analysis and forecasting systems

2.1 General description of the system

The MACC-II air quality system aims at providing analyses and forecasts of the main pollutants at the regional scale over the European continent: from 25° W to 45° E and from 30 to 70° N. Each of the 7 models is run with its own horizontal and vertical resolutions, with the horizontal resolutions varying between ∼20 and ∼10 km. This range of resolutions is not designed to reproduce local aspects of air pollution but to provide concentrations of pollutants at the regional scale that can then be used in particular as boundary conditions for AQ forecasts at finer resolution.
The range of the forecasts is 96 h from 00:00 UTC on Day0 with hourly outputs on 8 vertical levels (surface, 50, 250, 500, 1000, 2000, 3000 and 5000 m). Before mid-May 2014, only surface, 500, 1000 and 3000 m were produced. The species concentrations available from the forecasts are firstly \( \text{O}_3 \), \( \text{NO}_2 \), \( \text{SO}_2 \), \( \text{CO} \), \( \text{PM}_{10} \) and \( \text{PM}_{2.5} \), called core species hereafter. The core species are monitored in Near Real-Time (NRT) by European air quality stations and forecasts can therefore be evaluated routinely against these observations. Forecasts of birch pollen concentrations at surface are also produced during the pollen season (1 March to 30 June) since 2013. This product is not discussed in this paper since its description and validation is detailed in Sofiev et al. (2015). Additionally, since mid-May 2014, the production has been extended to other species or aggregation of species (\( \text{NO} \), \( \text{NH}_3 \), PAN + PAN precursors, total Non-Methane Volatile Organic Compounds). These new species are mainly designed for the use as initial and/or boundary conditions mainly for finer scale models designed for local AQ purposes. The analysis at the surface for Day0–1 is run daily a posteriori on Day0 using the assimilation of the hourly data from the AQ monitoring stations available in Europe between 00:00 UTC and 23:00 UTC on Day0–1. All 7 models do not produce yet the analysis for all the 6 core species.

Table 1 gives the portfolio of the regional data products. All models do not provide yet all additional species and vertical levels but this is planned to be completed in 2015. Table 2 gives the current times of delivery of the ensemble numerical data products. These production times have been shifted earlier since summer 2014 in order to fulfil users’ needs, in particular Day0 and Day1 forecasts, that are the mostly used products, are now available at 07:00 UTC. This has been made possible by an earlier delivery of the forecasts of each of the 7 models and by replacing the bulk 96 h processing of the ensemble by a processing 24 h per 24 h. The delivery time of the analysis has also been shifted earlier in June 2014.

The NRT hourly observations of \( \text{O}_3 \), \( \text{NO}_2 \), \( \text{SO}_2 \), \( \text{CO} \), \( \text{PM}_{10} \) and \( \text{PM}_{2.5} \) from the European AQ monitoring stations are used for model assimilation to produce the daily analyses and also for the forecast and analysis evaluation. From 2009 until recently, they
were gathered country by country through bilateral agreements. Since 2014, a new system has been set to retrieve these observations from the centralised European Environment Agency (EEA) database that collects the NRT data from the European countries. The delivery time of the observations to EEA is earlier and there is on average more data available than when gathering them country by country as before, although there is a large variability from one day to another on the number of data available. For the use in the production of the analyses, we chose after a dataflow monitoring of the EEA database a cut-off time at 07:00 UTC on Day0 for the dataset covering Day0–1. At this time of the day, more than 90% (on average) of all data are available. The 07:00 UTC cut-off time is therefore a compromise between having enough data available for the model assimilation and a reasonable production time for the ensemble analysis that is currently at 14:30 UTC. This production time is still too late for the forecasts to be initialised from the analysis, meaning that the forecast and the analysis products are currently run in two separate chains for each model. For the product evaluation, the observations covering Day0–1 available in the EEA database at 23:00 UTC on Day0 are used since there is less constraint on the time of delivery of evaluation products and, at 23:00 UTC, the dataset is almost complete. Because the NRT AQ observations used are not validated data, sorting procedures are applied to reject unrealistic observations by use of blacklists and/or thresholds. Moreover, only the data representative of the horizontal resolution of the regional models are selected. Such a procedure is necessary because currently there is no uniform and reliable metadata on site representativeness available for all regions and countries. The data selection follows the work that has been carried by Joly and Peuch (2012) to build an objective classification of sites, based on past validated measurements available in the Airbase database (EEA). Stations are classified between 1 and 10 depending on the characteristics of their series of measurements (diurnal cycle, “week-end effect” and high frequency variability with periods lower than 3 days) from most rural (class 1) to most polluted (class 10). The classification is used in order to restrict to the sites that have a sufficient spatial representativeness with respect to the model resolution (∼10–20 km), corresponding
to stations from classes 1 to 5. This leads to a typical number in summer 2014 of: \( \sim 600 \) sites for ozone, \( \sim 500 \) sites for NO\(_2\), \( \sim 150 \) sites for SO\(_2\), \( \sim 40 \) sites for CO, \( \sim 400 \) sites for PM\(_{10}\), \( \sim 150 \) sites for PM\(_{2.5}\). All these data are used for the verification of the forecast products. For the analyses, \( \sim 80 \% \) of these data is used in the assimilation. The remainder is kept aside for the verification of the analyses which is currently under development.

The sources of uncertainties in regional AQ forecast and analyses are the following: quality of the emissions used, meteorological forcings, representation of the atmospheric physical and chemical processes and boundary conditions for the chemical species. The approach chosen in MACC-II to minimize uncertainties is to use the same set of best available emissions over Europe, high quality meteorological forecasts and chemical boundary conditions in all seven chemistry-transport models. Therefore the variability between the forecasts of the seven models used in the ensemble comes mainly from differences in the models on the treatment of the chemical processes (homogeneous and heterogeneous, photolysis), the advection, the convective transport, the turbulent mixing and the wet and dry depositions. Other differences are on the use of different vertical and horizontal grids. For the production of the analysis, each model uses its own assimilation system.

The inventory used for anthropogenic emissions was built primarily for modelling purposes in the frame of the MACC-II project (Kuenen et al., 2014). This is an updated version of the MACC inventory (Kuenen et al., 2011). Its resolution is \( 1/8^\circ \) longitude \( \times \) \( 1/16^\circ \) latitude which is approximately \( 7 \text{km} \times 7 \text{km} \) and it covers the UNECE-Europe for the years 2003 to 2009. The 2009 inventory is currently used in the MACC-II daily production. An important upgrade of the MACC-II inventory compared to earlier MACC inventory is the provision of a particulate matter split between Elemental Carbon, Organic Carbon, SO\(_4\), Na and other aerosols. More details on this inventory can be found in Kuenen et al. (2014). For the biogenic sources, each model deals with its own emissions based on dynamical parameterizations and/or inventories that will be detailed in the following individual model description sub-sections. Additionally, emis-
sions from fires are taken into account using the GFASv1.1 product (Kaiser et al., 2012) available daily at 0.1° × 0.1° resolution. GFASv1.1 is based on fire radiative power retrievals from data of the Moderate Resolution Imaging Spectroradiometer (MODIS) instruments aboard the Terra and Aqua satellites. The GFAS product for Day0–1 is available around 06:00 UTC on Day0. This is soon enough to be used in the daily analysis individual production chains. At the time the individual forecasts begin for Day0, only the fire emissions from Day0–2 are available. To have less time gap between the fire emissions and the starting time of the regional forecast runs (usually around 20:00 UTC), an additional fire emission product available around 20:30 UTC on Day0–1 using satellite observations from 15:00 UTC on Day0–2 to 15:00 UTC on Day0–1 is currently under test. In the forecasts, a persistence of the fire emissions of 3 days is assumed. This is a rounded average of the fire duration obtained by Turquety et al. (2014) from Euro-Mediterranean region from MODIS MCD64 product (Giglio et al., 2010) in the period 2003–2012.

The meteorological fields used to force the 7 CTMs are from the operational IFS (Integrated Forecasting System) daily meteorological forecasts of the European Centre for Medium-Range Weather Forecasts (ECMWF). The IFS forecast starting at 12:00 UTC on Day0–1 is used for the MACC-II air quality 96 h forecast starting at 00:00 UTC on Day0. For the analysis on Day0–1, the IFS forecast starting at 00:00 UTC on Day0–1 is used.

The regional domain boundary conditions for the aerosols and gaseous species are provided by the MACC-II global assimilation and forecasting system. This forecasting system is an extension of the ECMWF meteorological IFS running at lower resolution, providing concentrations of dust, sea salt, organic matter, black carbon and sulphate aerosols (Morcrette et al., 2009; Benedetti et al., 2009) that are used to force the aerosols in the regional CTMs at the boundaries. At the end of MACC-II project (summer 2014), for the chemical species the IFS was two-way coupled to the off-line MOZART global chemical transport model (CTM). This allowed assimilation of satellite data for O₃, NO₂, and CO in the IFS itself, while the detailed chemical processes
were handled in the MOZART model (Flemming et al., 2009; Stein et al., 2012; Inness et al., 2013). Since 18 September 2014, the MACC-II global assimilation and forecasting system has been upgraded to a fully integrated system for aerosols and chemical species. Instead of the coupling with the MOZART model, the chemistry is now treated on-line in the IFS using chemistry modules based on the TM5 model (Huijnen et al., 2010). This new system is named Composition-IFS (C-IFS) and is further described in Flemming et al. (2014). The chemical mechanism in the TM5 operational version of C-IFS is based on a modified version of the Carbon Bond 5 (CB05) scheme (Williams et al., 2013; Yarwood et al., 2005).

Based on all the inputs described above, each of the centres in charge of the 7 models runs its production locally and transfers its forecast and analysis files to Météo-France (referred to central production centre hereafter). For each timestep (one hour) of the daily forecasts and analyses, the individual model fields are interpolated on a common regular 0.1° × 0.1° grid over the MACC-II European domain, if not yet on this common grid. It is from these re-gridded fields that the ensemble and verification products are calculated.

The general organisation of the MACC-II air quality forecasts and analysis system is summarized in Fig. 1. Tables 3 and 4 give the general features of the seven individual models and of their analysis system. A short description of the seven individual models and of the ensemble is given in the following sections. More details can be found in the MACC-II 6-monthly reports (http://www.gmes-atmosphere.eu/documents/maccii/deliverables/ens/).

2.2 CHIMERE forecast and analysis system

CHIMERE is an Eulerian chemistry-transport model able to simulate concentrations fields of gaseous and aerosols species at a regional scale (Menut et al., 2013a). The model is developed under the GPL licence (http://www.lmd.polytechnique.fr/chimere/). CHIMERE is used for analysis of pollution events, process studies (Bessagnet et al.,
2009; Beekmann and Vautard, 2010), experimental and operational forecast (Rouil et al., 2009), regional climate studies and trends (Colette et al., 2011), among others.

CHIMERE calculates and provides the atmospheric concentrations of tens of gas-phase and aerosol species over local (e.g. urban) to continental domains (from 1 km to 1° resolution). Vertically, the model is able to simulate the whole troposphere. The gaseous species are calculated using the MELCHIOR2 scheme and the aerosols using the scheme developed by Bessagnet et al. (2004). This module takes into account species such as sulphate, nitrate, ammonium, primary organic matter (POM) and elemental carbon (EC), secondary organic aerosols (SOA), sea salt, dust and water. These aerosols are represented using 8 bins, from 40 nm to 40 µm, in diameter. The life cycle of the aerosols is completely represented with nucleation of sulphuric acid, coagulation, adsorption/desorption, wet and dry deposition and scavenging. This scavenging is both represented by coagulation with cloud droplets and precipitation. The formation of SOA is also taken into account (Bessagnet et al., 2009).

Biogenic emissions are calculated using the MEGAN emissions scheme (Guenther et al., 2006) which provides fluxes of isoprene and monoterpenes. The mineral dust emissions are calculated using the (Alfaro and Gomes, 2001) scheme, forced by satellite soil and surface data (Menut et al., 2013b).

The CHIMERE assimilation system for operational products is based upon hourly optimal interpolation processing of surface observations for O\textsubscript{3} and PM\textsubscript{10} (Honoré et al., 2008). During MACC-II, an Ensemble Kalman filter has been also developed for ozone analysis (Gaubert et al., 2014).

2.3 EMEP forecast and analysis system

The EMEP/MSC-W model (hereafter referred to as “EMEP model”) has been developed at the EMEP Meteorological Synthesizing Centre-West at the Norwegian Meteorological Institute. The model has been publicly available as open source code since 2008, and a detailed description is given in Simpson et al. (2012).
The numerical solution of advection is based on Bott (1989). The turbulent diffusion coefficients are calculated for the whole 3-D model domain on the basis of local Richardson number, and the planetary boundary layer (PBL) height is calculated using methods described in Simpson et al. (2003). Dry deposition uses a resistance analogy combined with stomatal and non-stomatal conductance algorithms (Simpson et al., 2003; Tuovinen et al., 2004), whereas wet deposition uses scavenging coefficients applied to the 3-D rainfall, including both in-cloud and sub-cloud scavenging of gases and particles. The chemical scheme couples the sulphur and nitrogen chemistry to the photochemistry using about 140 reactions between 70 species (Andersson-Sköld and Simpson, 1999; Simpson et al., 2012).

The methodology for biogenic emissions builds on maps of 115 forest species generated by Köble and Seufert (2001). Emission factors for each forest species and for other land classes are based on Simpson et al. (1999), updated with recent literature (see Simpson et al. (2012) and references therein), and driven by hourly temperature and light using algorithms from Guenther et al. (1995). Other natural emissions include marine emissions of dimethyl sulphide, and \( \text{SO}_2 \) from volcanoes.

The standard model version distinguishes two size fractions for aerosols, fine aerosol (PM\(_{2.5}\)) and coarse aerosol (PM\(_{10}\) excluding PM\(_{2.5}\)). The aerosol components presently accounted for are sulphate, nitrate, ammonium, anthropogenic primary particulate matter and sea salt. Also aerosol water is calculated. The parameterisation of dry deposition for aerosols follows standard resistance formulations, accounting for diffusion, impaction, interception, and sedimentation. Wet scavenging is treated with simple scavenging ratios, taking into account in-cloud and sub-cloud processes. For secondary organic aerosol (SOA) the so-called “EmChem09soa” scheme is used, which is a slightly simplified version of the mechanism described by Bergström et al. (2012).

The EMEP data assimilation system (EMEP-DAS) is based on the 3-D-Var implementation for the MATCH model (Kahnert, 2008, 2009). The background error covariance matrix is estimated following the NMC method (Parrish and Derber, 1992). Currently, the EMEP-DAS delivers analyses for NO\(_2\), using NO\(_2\) columns of OMI and in
situ measurements of NO$_2$ surface concentrations. The assimilation window is 6 h, 4 times per day.

2.4 EURAD-IM forecast and analysis system

EURAD-IM is an Eulerian meso-scale chemistry transport model involving advection, diffusion, chemical transformation, wet and dry deposition and sedimentation of tropospheric trace gases and aerosols (Hass et al., 1995; Memmesheimer et al., 2004). It includes 3d-var and 4d-var chemical data assimilation (Elbern et al., 2007) and is able to run in nesting mode. EURAD-IM has been applied on several recent air pollution studies (Monteiro et al., 2012, 2013; Zyryanov et al., 2012; Elbern et al., 2011; Kanakidou et al., 2011).

The positive definite advection scheme of Bott (1989) is used to solve the advective transport. An Eddy diffusion approach is used to parameterize the vertical sub-grid-scale turbulent transport. The calculation of vertical Eddy diffusion coefficients is based on the specific turbulent structure in the individual regimes of the planetary boundary layer (PBL) according to the PBL height and the Monin–Obukhov length (Holtslag and Nieuwstadt, 1986). A semi-implicit (Crank–Nicholson) scheme is used to solve the diffusion equation.

Gas phase chemistry is represented by the Regional Atmospheric Chemistry Mechanism (RACM) (Stockwell et al., 1997) and an extension based on the Mainz Isoprene Mechanism (MIM) (Geiger et al., 2003). A two-step Rosenbrock method is used to solve the set of stiff ordinary differentials equations (Sandu et al., 2003; Sandu and Sander, 2006). Photolysis frequencies are derived using the FTUV model according to Tie et al. (2003). The radiative transfer model therein is based on the Tropospheric Ultraviolet-Visible Model (TUV) developed by Madronich and Weller (1990). The modal aerosol dynamics model MADE (Ackermann et al., 1998) is used to provide information on the aerosol size distribution and chemical composition. To solve for the concentrations of the secondary inorganic aerosol components, a FEOM (fully equivalent operational model) version, using the HDMR (high dimensional model representation)
technique (Rabitz et al., 1999; Nieradzik, 2005), of an accurate mole fraction based thermodynamic model (Friese and Ebel, 2010) is used. The updated SORGAM module (Li et al., 2013) simulates secondary organic aerosol formation. Biogenic emissions are calculated in the EURAD-IM CTM with the Model of Emissions of Gases and Aerosols from Nature (MEGAN) (Guenther et al., 2012).

The gas phase dry deposition modelling follows the method proposed by Zhang et al. (2003). Dry deposition of aerosol species is treated size dependent using the resistance model of Petroff and Zhang (2010). Wet deposition of gases and aerosols is derived from the cloud model in the EPA Models-3 Community Multiscale Air Quality (CMAQ) modelling system (Roselle and Binkowski, 1999).

The EURAD-IM assimilation system includes (i) the EURAD-IM CTM and its adjoint, (ii) the formulation of both background error covariance matrices for the initial states and the emission factors, (iii) the observational basis and its related error covariance matrix, and, (iv) the minimisation including the transformation for preconditioning. The quasi-Newton limited memory L-BFGS algorithm described in Nocedal (1980) and Liu and Nocedal (1989) is applied for the minimization. Following Weaver and Courtier (2001) with the promise of a high flexibility in designing anisotropic and heterogeneous influence radii, a diffusion approach for providing the background error covariance matrices is implemented.

2.5 LOTOS-EUROS forecast and analysis system

The 3-D chemistry-transport model LOTOS-EUROS (Schaap et al., 2008) is developed by the Dutch institutes TNO (www.tno.nl), RIVM (www.rivm.nl) and, more recently, KNMI (www.knmi.nl). It is used for regional-scale air-quality forecasts in Europe and the Netherlands (de Ruyter de Wildt et al., 2011). The LOTOS-EUROS model has participated in several international model intercomparison studies addressing ozone (Van Loon et al., 2007; Solazzo et al., 2012a) and particulate matter (Cuvelier et al., 2007; Vautard et al., 2007; Stern et al., 2008; Solazzo et al., 2012b). These studies have shown that the model has a performance comparable to other European regional mod-
els. In the past three years, three major updates of the LOTOS-EUROS model have been implemented, moving from version 1.7 to version 1.10. Detailed update information can be found on the model web page, http://www.lotos-euros.nl. Since the end of MACC-II, the latest update to v1.10 implemented operationally consists of changes in the SO$_2$ to SO$_4$ conversion rate, use of AQMEII conventions for the fine/coarse dust assignment, update of resistances for e.g. ozone (leading to an overall ozone increase), and improvement of the treatment of fire emissions.

The model extends up to 3.5 km above sea level, with three dynamic layers and a fixed 25 m thick surface layer. The lowest dynamic layer is the mixing layer, followed by two reservoir layers. The height of the mixing layer is obtained from the ECMWF meteorological input data used to drive the model. Transport is based on the monotonic advection scheme developed by Walcek (2000). Gas phase chemistry is described using the TNO CBM-IV scheme (Schaap et al., 2008). Hydrolysis of N$_2$O$_5$ is described following Schaap et al. (2004). Aerosol chemistry is represented using ISORROPIA-2 (Fountoukis and Nenes, 2007). The aerodynamic resistance is calculated for all land use types separately. Below cloud scavenging is described using simple scavenging coefficients for gases (Schaap et al., 2004) and particles (Simpson et al., 2003). Dry deposition is based on the well-known resistance approach, with the DEPAC parameterization for gases (Wichink Kruit et al., 2012) and the Zhang et al. (2001) parameterization for particles.

Biogenic isoprene emissions are calculated following the mathematical description of the temperature and light dependence of the isoprene emissions, proposed by Guenther et al. (1993), using the actual meteorological data. For land use the CORINE/Smiatek database has been enhanced using the tree species map for Europe made by Koeble and Seufert (2001). Total PM$_{10}$ in the LOTOS-EUROS model is composed of chemically unspecified primarily PM in the fine and coarse mode, black carbon, dust, ammonium, sulphate, nitrate and sea salt (Na in the fine and coarse mode).
The LOTOS-EUROS model is equipped with a data assimilation package with the ensemble Kalman filter technique (Barbu et al., 2008; Timmermans et al., 2009; Curier et al., 2012). Data assimilation for the MACC-II daily analyses is performed with surface ozone observations (Curier et al., 2012). An extension to other surface and satellite data is foreseen in the near future.

2.6 MATCH forecast and analysis system

The MATCH model (Multi-scale Atmospheric Transport and Chemistry model) has been developed at SMHI over the past 20 years and is applied for emergency purposes as well as for regional scale chemistry modelling (Langner et al., 1998; Robertson et al., 1999).

The transport is described by a Bott-like mass conservative scheme (Bott, 1989; Robertson et al., 1999). For the vertical diffusion an implicit mass conservative scheme is used where the turbulent exchange coefficients for neutral and stable conditions are parameterized following Holtslag and Moeng (1991). In the convective case the turbulent Courant number is directly determined from the turn-over time in the atmospheric boundary layer.

The dynamical core of the model contains initialization and adjustment of the horizontal wind components based on a procedure proposed by Heimann and Keeling (1989). This is important to ensure mass conservative transport for interpolated input weather data, specifically for the transport scheme used.

Boundary layer parameterization is determined from surface heat and water vapour fluxes as described by van Ulden and Holtslag (1985) for land surfaces, and Burridge and Gadd (1977) for sea surfaces. The boundary layer height is calculated from formulations proposed by Zilitinkevich and Moronov (1996) for the neutral and stable case and from Holtslag et al. (1995) for the convective case. These parameterizations drive the formulations for vertical diffusion and dry deposition where for the latter a resistance approach is used (Andersson et al., 2007). In-cloud and sub-cloud wet deposition is implemented following Andersson et al. (2007). The photochemistry scheme is to large
extent based on the EMEP chemistry scheme (Simpson et al., 1993), with some updates where a modified production scheme for isoprene is the most notable based on the so-called Carter-1 mechanism (Carter, 1996; Langner et al., 1998).

Aerosols are described for 4 bins and only for SIA, dust and primary organic compounds at the moment. Inclusion of SOA is under testing. Sea salt emissions are dynamically described following Foltescu et al. (2005). A module for wind driven dust emissions is under testing that follows Schaap et al. (2005).

A 3-D variational data assimilation scheme is used with spectral transformation (Kahnert, 2008). The limitation then is that background covariance structures are described as isotropic and homogeneous, however not necessarily the same for different wavenumbers, and derived from the so-called NMC-method (Parish and Derber, 1992). The advantage though is that the background error matrix becomes block diagonal and there are no scale separations as the covariance between spectral components are explicitly handled. The block diagonal elements are the covariance between wave components at different model layers and chemical compounds.

2.7 MOCAGE forecast and analysis system

The MOCAGE (Model Of atmospheric Chemistry At larGE scale) model (Josse et al., 2004; Dufour et al., 2004) has been developed at Météo-France since 2000. Its assimilation system has been developed jointly with CERFACS. This model and its assimilation system have been successfully used for tropospheric and stratospheric researches (e.g. Bousserez et al., 2007; Barré et al., 2013, 2014; Lacressonnière et al., 2014) and also for operational purposes (Rouil et al., 2009).

MOCAGE uses the semi-lagrangian advection scheme from Williamson and Rasch (1989) for the grid-scale transport, the parameterization of convective transport from Bechtold et al. (2001) and the turbulent diffusion parameterization from Louis (1979). Dry deposition is based on the approach proposed by Wesely (1989). The wet deposition by the convective and stratiform precipitations follows Mari et al. (2000) and Giorgi and Chameides (1986). MOCAGE includes the RACM scheme for tropospheric
chemistry (Stockwell et al., 1997) and the REPROBUS scheme for stratospheric chemistry (Lefèvre et al., 1994). Biogenic emissions in MOCAGE are fixed monthly biogenic emission from Guenther et al. (1995).

The aerosol module of MOCAGE follows a bin approach and includes so far the primary aerosols: dusts (Martet et al., 2009), sea salts, black carbon (Nho-Kim et al., 2005) and organic carbon. Recent updates of the primary aerosol module and corresponding evaluation can be found in Sič et al. (2014). Inorganic secondary aerosols have been developed recently but are not yet included in the MOCAGE production in MACC-II.

MACC-II operations use a variational assimilation system based upon MOCAGE and the PALM coupler, which has been developed during the ASSET European project (Geer et al., 2006; Lahoz et al., 2007). The system, recently renamed VALENTINA, has been used to compute global and regional re-analyses of atmospheric composition in multiple studies (El Amraoui et al., 2008; Massart et al., 2009; Barré et al., 2013, 2014; Emili et al., 2014). The assimilation algorithm employed for MACC-II analyses is a 3-D-Var with assimilation windows of one hour length (Jaumouillé et al., 2012), corresponding to the frequency of surface measurements. The assimilation has first been set for surface ozone analyses and in MACC-III it has been extended to surface NO$_2$. The specification of the background and observation errors is done based on the evaluation of historical time-series of observations and model values. The horizontal error correlation has a Gaussian shape and its typical length is set to 0.4° for ozone and 0.1° for NO$_2$, to account for the larger variability of NO$_2$ at fine spatial scales. The vertical error correlation length is set to 1 model grid point for all species (∼ 100 m). As a consequence, assimilation increments linked to surface observations are confined in the planetary boundary layer.

2.8 SILAM forecast and analysis system

SILAM is a meso-to-global scale dispersion model (Sofiev et al., 2008, see also the review Kukkonen et al., 2012, http://silam.fmi.fi) that is used for atmospheric composi-
The model has been applied with resolutions ranging from 1 km up to 3°, incorporates 8 chemical and physical transformation modules and covers the troposphere and the stratosphere. The model is publicly available since 2007 and is used as operational and research tool.

The model has two dynamic cores: Lagrangian (Sofiev et al., 2006), primarily used in emergency-type applications, and Eulerian (Galperin, 2000; Sofiev, 2002) used in atmospheric composition, climate, and air quality-related applications, including MACC-II. The MACC-II operational SILAM v.5.2 uses the simple dry deposition scheme of (Sofiev, 2000) for gases and new development for aerosols (Kouznetsov and Sofiev, 2012), which covers particle sizes from 1 nm up to ~50 µm of effective aerodynamic size. The wet deposition scheme used in MACC-II simulations calculates the 3-D removal coefficient and distinguishes between sub- and in-cloud scavenging, large-scale and convective precipitations, as well as between rain and snow (Sofiev et al., 2006). Boundary layer parameterization follows (Sofiev et al., 2010), whereas in the free troposphere and the stratosphere turbulence is computed following IFS approach and corresponding turbulent length scale.

Two chemical schemes are used: the CBM-4 gas-phase chemistry mechanism and own development for heterogeneous chemical transformations and inorganic aerosol formation after (Sofiev, 2000). Aerosols in SILAM are represented via sectional approach with species-specific size spectra. The aerosol species include primary anthropogenic aerosols, divided into PM$_{2.5}$ and PM$_{10}$, secondary inorganic aerosols (sulphates, nitrates and ammonia), and sea salt aerosols.

The forecasts utilise the BVOC emission term based on NatAir project results (Poupkou et al., 2010) and own development for the sea-salt emission (Sofiev et al., 2011).

The data assimilation system of SILAM consists of 3-D-VAR and 4-D-VAR modules (Vira and Sofiev, 2012). The MACC-II near-real time analysis suite uses the 3-D-VAR method and assimilates hourly surface observations of NO$_2$, O$_3$ and SO$_2$. PM observ-
Observations have been assimilated in reanalysis simulations (Vira and Sofiev, 2015). The 4-D-VAR methodology is utilised in re-analysis mode for pollen.

The model evolution from the MACC-II v.5.2 towards v.5.4 that will become operational in early-2015, include several important updates. The dry deposition scheme will follow the resistance analogy with extensions after (Simpson et al., 2003). Wind-blown dust will be included via lateral boundary conditions in the next release of operational SILAM v.5.4, together with a secondary organic aerosol module and fire emission.

### 2.9 ENSEMBLE forecast and analysis system

To process the ensemble, all seven individual models are firstly interpolated in a common $0.1^\circ \times 0.1^\circ$ horizontal grid. For each grid point, the ensemble model (referred as the ENSEMBLE hereafter) value is calculated as the median value of the individual model forecasts or analyses available. The median is defined as the value having 50% of individual models with higher values and 50% with lower values. This method is rather insensitive to outliers in the forecasts or analyses and is very efficient computationally. These properties are useful from an operational point of view. The method is also little sensitive if a particular model forecast or analyses is occasionally missing. The performances of the median ensemble are discussed in Sect. 3. For the forecasts, the ensemble is produced for all levels and all species (core and additional). For the analyses, the individual assimilation systems do not produce analyses for all species yet. At the end of MACC-II, ozone was the only species that is produced by 6 of the models. This is why the ensemble analysis in MACC-II was only calculated for ozone. It will be extended to NO$_2$ in 2015 since more models are now producing NO$_2$ analyses.
3 Evaluation of the performances of the system

3.1 General description

The evaluation of the performances of a forecast system is a necessary step rating its quality and thus proving its usefulness. The MACC-II air quality forecasts are evaluated against the NRT AQ surface monitoring data detailed in Sect. 2.1. Note that this set of data is fully independent of the forecast since the analyses assimilating the NRT AQ data are produced too late to be used to initialise the forecasts. The tools to assess the performances of the analyses are not yet in place but this is planned to be ready in 2015. Since the focus of the MACC-II regional system over Europe is on air quality, meaning air composition close to the surface, no column observations (ground based or from satellite) or upper air in situ measurements (i.e. on board aircraft) are used operationally to evaluate the system performances.

The forecast performances are measured using the five statistical indicators detailed in the Appendix: the mean bias (MB), the root mean square error (RMSE), the modified normalized mean bias (MNMB), the fractional gross error (FGE) and the correlation. These statistical measures, when taken together, provide a valuable indication of the model performances. Taylor diagrams are also used to combine root mean square errors and correlations.

The performances of the MACC-II regional AQ forecasts are assessed operationally by several means:

– on a daily basis with plots of statistical indicators and charts available on the MACC-II regional website (http://macc-raq.gmes-atmosphere.eu/),

– on a 6 monthly basis in reports including plots of statistical indicators over two periods of 3 months (winter + spring or summer + autumn) and analysis of these indicators (http://www.gmes-atmosphere.eu/documents/maccii/deliverables/ens/).

Additionally, on a 6 monthly basis, reports are especially dedicated to the scientific analysis of the forecasts of the 7 models and of the ensemble in the Mediterranean area.
The Mediterranean area is recognized as challenging for models, in particular under summer conditions with very active photochemistry and because of its large variety of emission sources.

The performances of the NRT analysis are not presented in this paper since there is only an ensemble production of one species (ozone) and the daily verification procedure against an independent dataset was not yet in place at the end of MACC-II project.

3.2 Example of the forecast of two ozone episodes between 10 and 13 June 2014

The MACC-II AQ forecasting system and its performances are illustrated here for a case study of ozone pollution events that took place between 10 and 13 June 2014. There were two regional areas with high ozone concentrations (> 120 mg m\(^{-3}\)) occurring at the same time, one over Austria and surrounding regions (South of Germany and Hungary), and one over the South East of France and the North of Italy. This is illustrated by the surface station measurements shown in Fig. 2. The “Austrian” episode is highest on 11 June with values reaching 200 µg m\(^{-3}\) at Sopron (Hungary). The “French” episode peaks at 250 µg m\(^{-3}\) with daily maxima over 150 µg m\(^{-3}\) from the 10 to 13 June.

From the 7 model forecasts EPSgrams are built daily for 40 major cities in Europe. EPSgrams give a graphical representation of the spread of the 7 model forecasts and therefore an estimate of the uncertainty over the 4 days of the forecast. In Fig. 3, the EPSgrams are calculated and plotted for the same locations as the measurements shown in Fig. 2 from the forecast started on 10 June at 00:00 UTC. For the ozone peak event around Austria, there is a very good consistency in the daily variations provided by the seven models and the AQ station observations (see comparison between Figs. 2a and 3a, 2b and 3b). For the Hallein (Austria) station the threshold of 120 mg m\(^{-3}\) is reached by at least part of the 7 models everyday over the 4 days forecasts. For the Sopron (Hungary) station there is a main peak in the model on 11 June as
in the observations. For this “Austrian” ozone episode the spread between the 7 models is very reasonable (generally less than 30 µg m\(^{-3}\) for the 25–75 % range), showing the good consistency between the models with slightly more spread between the forecasts for the highest peak times. The median ozone value is generally lower during daytime peaks than the observations by 30 to 50 µg m\(^{-3}\) but the maxima of the 7 models are close to those observed. During nighttime, the median ozone is close to the measured values. For the ozone event in the south of France, the comparison shows also a good consistency between the models and the observations. Most of the 7 models are over the threshold of 120 µg m\(^{-3}\) for each of the 4 days forecasted, except at Plan d’Aups (Figs. 2d and 3d) where the very high ozone peak measured on 10 June (over 240 µg m\(^{-3}\)) is underestimated with a large spread between the 7 models. For the Sausset location (Figs. 2c and 3c), there is a good consistency between the models with generally small spread. Similarly to the Austrian area, the median ozone concentrations are lower than observed but the model maxima are generally close to those measured at the two French stations. Note that for both the “Austrian” and “French” ozone episodes, there is no significant degradation of the forecast skills at Day3 and Day4 indicating that uncertainty in ozone forecasts is more driven by inherent uncertainty in chemistry transport models and part of its input than by uncertainty of meteorological forecast.

For further evaluation of the forecast performances, Fig. 4 displays the maps of the 15h forecasts (i.e. valid for Day0 at 15:00 UTC) of the ozone at the surface from the ENSEMBLE with the available observations over-plotted with the same colour scale. Note that the set of observations available in NRT varied from day to day. Even if the comparison is limited by the missing observations, Fig. 4 shows that the median ensemble captures well the two ozone episodes.

### 3.3 Availability statistics

The MACC-II regional air quality forecasting and analysis system is currently under a pre-operational status that can be seen as the demonstrator of a future operational
system. Correct working of 7 model chains and of the Ensemble chain is monitored on working hours only since, at this stage, there is no funding yet for a 7 day/7 day 24 h/24 h control. Nevertheless, in its pre-operational configuration the production chains are reliable with availability in time (see Table 2) of the 7 individual forecasts and analysis generally above 85% during MACC-II. During the past year, the production suffered from failures because of the many changes that were applied to the individual and central systems to fit with fully operational standards (data format, file transfer, databases, processing softwares, …). The operationalisation being nearly fully settled, the reliability has been improved since the end of MACC-II (generally above 90%). The Ensemble forecast and analysis productions have been available 100% of the time since September 2012. This high performance was achieved because the ENSEMBLE can be produced even if all the 7 models are not available.

To illustrate the robustness of the MACC-II ensemble system in the case when one or more of the 7 models are missing for the production of the ENSEMBLE, we selected as an example the period of 9 to 15 June 2014 corresponding to the ozone episodes discussed in Sect. 3.2. Figure 5 displays the MB, MNMB, RMSE, FGE and correlation of ozone of the seven individual models and the ENSEMBLE calculated using the representative observations available over the whole European domain. These statistics are based on seven consecutive 96 h forecasts run every day from 9 June to 15 June. This figure shows that there is a spread of the 7 models and that the ENSEMBLE generally gives the best scores with MNMB between 0.2 and −0.1, FGE between 0.15 and 0.4 and correlations up to 0.75 during daytime. In order to assess the sensitivity of the ensemble performance to the number of models, the following ensembles are compared:

- “MEDIAN 7”, the operational ensemble method which is the median of the 7 models (= ENSEMBLE) as presented in Fig. 5,

- “MEDIAN 5”, built on 5 individual models, after filtering out the best and the worst models, according to the criterion described below,
– “MEDIAN 3”, built on 3 individual models, after filtering out the two best and the two worst models, according to the criterion described below,

– “1BEST”, the best model.

Since the relative performances of individual models vary in time and space, the criterion to order the 7 individual models from worst to best is measured by their RMSE over the 7 days of the verification between 12:00 UTC and 18:00 UTC (ozone peak time). This criterion is chosen on the basis that we look for the model best reproducing the high daytime ozone levels. RMSE is seen as the most objective criterion since MB and NMNB can include compensating effects and since there is a low spread between the models in the FGE. From this, the best model is displayed in purple colour in Fig. 5c and the worst in brown colour.

Results of the sensitivity experiments are shown in Fig. 6. This figure confirms that the median ensemble (MEDIAN 7) using all 7 models performs generally better than the best model on all statistical indicators. When only five models (excluding the best and the worst) are available to calculate the ensemble, all scores show only very slight differences with the ENSEMBLE (MEDIAN 7) based on 7 models. Going to only three models to calculate the ensemble (MEDIAN 3), leads to statistical indicators degraded compared to the ensemble from 7 (MEDIAN 7) or 5 (MEDIAN 5) models but performs generally better than the best model (1BEST). This indicates that using an ensemble of models, even if reduced, is more useful than using a single model event of very good quality. This also shows that with 5 models available (that may happen in case of problems of production of 2 of the 7 models), the ensemble is still robust compared to observations.

Note that we only illustrate here the behaviour of the ENSEMBLE over a short period of time, but these results are still true over longer time periods. In our tests we disregarded the worst (or 2 worst) and best (or two best) models on a RMSE criterion but Kioutsioukis and Galmarini (2014) showed that there is an impact of the quality of the models chosen on AQ ensemble performances. To assess this aspect, a more compre-
A comprehensive study could be done from the MACC-II ensemble of models that would cover ozone but also other species. The limitations of such a study would be that the current set of surface AQ observations used to assess the performances is not yet stable with time, that the seven models constituting the MACC-II ensemble are continuously evolving leading to changes of their performances and that the results may be different depending on the species considered.

3.4 Statistical performances of the Ensemble forecasts

Additionally to the production of daily skill scores, statistical indicators are calculated for ozone, NO$_2$ and PM$_{10}$ at surface on a seasonal basis since September 2009 for each of the seven models and for the ENSEMBLE. These skill scores and the analysis of their seasonal and year-to-year evolutions are presented in 6 monthly reports, each including 2 seasons. This now represents a long series of statistics that cannot be all presented in the paper. Here we only focus on the performances of the system for the past two years and on the two main pollutants for the season during which regulatory levels are more likely to be encountered: ozone in summer (1 June to 1 September) and PM$_{10}$ in winter (1 December to 1 March). The model statistical indicators are calculated against representative measurements from the European AQ surface station network as detailed in Sect. 2.1. So far, the data provision in NRT is not fully operational. Therefore, there is variability with time of the number of data available and of their location that should be taken into account when analysing the statistics. Also, the spatial coverage of the surface AQ network in Europe is very inhomogeneous with a high density of stations in France, Germany, UK, Belgium and the Netherlands. Thus the statistical indicators are more representative of the system skills for these countries.

The MB, MNMB, RMSE, FGE and correlation (defined in the Appendix) are shown in Fig. 7 for ozone in summer 2013 and 2014. For both years, a common feature is that there is no degradation of MB, MNMB, RMSE and FGE indicators from Day0 to Day3, meaning that the surface ozone concentration values are not significantly affected by the degradation of the meteorology with forecast time. In summer 2013, MB and MNMB
are consistent and show a general overestimation of ozone. In summer 2014, this positive bias is generally lower with a slight underestimation around 19:00 UTC but still an overestimation otherwise. For both years, there is a similar diurnal cycle of MB and MNMB showing minima around 19:00 UTC and maxima around 09:00 UTC. RMSE and FGE are also consistent with each other with best performances around 12:00 UTC and worse around 06:00 UTC. The shift of the time of the minima of the RMSE and FGE (12:00 UTC) compared to the minima of the MB and MNMB (19:00 UTC) shows that there are compensating errors in the ozone biases for both summer 2013 and summer 2014. The normalized indicators MNMB (FGE) are on average 0.24 (0.35) in 2013 and 0.14 (0.30) in 2014. As for MB and MNMB, RMSE and FGE in 2014 are lower compared to 2013 indicating a slight improvement of the ozone forecast. Their low values indicate that the ENSEMBLE forecasts perform well. The improvement of MB, MNMB, RMSE and FGE is not seen on the correlations that are slightly lower in summer 2014 than in summer 2013. Even if the ENSEMBLE provides ozone concentration values closer to the observations, their diurnal cycle has not been improved compared to observations. The evolution of the correlation with forecast time is similar for both years with highest values around 15:00 UTC (up to 0.70 in 2013 and up to 0.64 in 2014) and lowest at night (down to 0.46 in 2013 and down to 0.375 in 2014). This is consistent with the other statistical indicators showing lower performances during night. Correlations tend to decrease from Day0 to Day3. Apparently, correlations are more sensitive to the meteorological forecast skills than MB, MNMB, RMSE and FGE. In summary, the model ensemble exhibits good ozone skills for both summers, particularly during daytime, as illustrated by the low MNMB and FGE and high correlations. The changes in scores between 2013 and 2014 are not large enough to be regarded as significant because the observation datasets used for the verification are not exactly the same in 2013 and 2014 and since the meteorological conditions also varied.

Figure 8 shows the ensemble scores for the two past winters (2012–2013 and 2013–2014) for PM$_{10}$. Since there were much fewer observations available at 00:00 UTC compared to other times of the day, the values given at the forecasts times of 0, 24,
48, 72 and 96 h show a specific behaviour that is not analysed since not significant. The MB and MNMB both indicate a low bias in the ENSEMBLE. This can be linked to the fact that not all individual models include secondary inorganic aerosols and/or secondary organic aerosols. These missing aerosol components lead to a general underestimation of PM$_{10}$. Nevertheless, there has been a significant reduction of this low bias between winter 2012–2013 and winter 2013–2014 that can be attributed, at least partly, to the improvements of the aerosol representation in several of the seven models contributing to the ENSEMBLE. This improvement can also be seen to a lesser extent on the RMSE and FGE. For both winters, MB, MNMB, RMSE and FGE are best during daytime (particularly at 09:00 UTC and 15:00 UTC) with consistent diurnal variations. The normalized indicators MNMB (FGE) are on average $-0.28$ ($-0.11$) in winter 2012–2013 and $0.58$ ($0.53$) in winter 2013–2014. The decrease of skills with forecast time is again significant for the correlations. Values of the correlation for both winters are fairly low (0.43 at maximum) compared to ozone correlations in summer. This means that the seven forecasts on which the ENSEMBLE is calculated are less skillful in modeling the aerosols than ozone. This is a common feature of most chemistry models since there are still large uncertainties on primary aerosol emissions and processes of production and evolution of secondary aerosols, particularly of secondary organic aerosols. Moreover, because of the operational context of MACC-II production, the seven forecasts models are optimized to run in short times. This prevents them to include the very detailed processes, thus computationally costly, that can be afforded in research mode.

### 3.5 Example of the specific evaluation for the Mediterranean area

Within the European continent, the Mediterranean area is characterized by special features – high emission densities due to concentration of human activities in surrounding coastal areas, intense photochemistry, high background pollution, small scale meteorology, . . . – that make air quality forecasting specially challenging. This is why work has been specifically carried out to evaluate the seven models and the ENSEMBLE
in this region. This is complementary to the systematic daily and seasonal evaluation performed over the whole European continent. Its aim is not about scoring the system but on a better scientific understanding of the behaviour of the seven models and the ENSEMBLE in the Mediterranean region. This work is based, firstly, on two high resolution models run daily over eastern (Greece) Mediterranean and western (Spain) areas and surface station measurements that are not used in the operational MACC evaluation and, secondly, on scientific analyses of case studies.

For the Eastern Mediterranean area, the LAP-AUTH forecasting system is run daily. It consists of the Weather Research and Forecasting mesoscale meteorological model (WRF version 3.2) (Skamarock et al., 2008) and the chemistry transport model Comprehensive Air quality Model with extensions (CAMx version 5.30) (ENVIRON, 2010). The anthropogenic emission data, used as CAMx input data, are from Kuenen et al. (2014) for the reference year 2009. Anthropogenic emissions data are temporally processed using the Model for the Spatial and tEmporal diStribution of emissionS (MOSESS) (Markakis et al., 2013). The emissions originating from natural sources are calculated with the use of the emission model namely NEMO (Natural Emission MOdel) (Markakis et al., 2009). Wind erosion dust, sea salt and biogenic NMVOCs emissions are calculated using the WRF model meteorology. The air quality forecasting system derives meteorological initial and boundary conditions from the operational 12:00 UTC forecast of ECMWF while chemical boundary conditions derived from the IFS–MOZART global model forecast and replaced by C-IFS from September 2014. The domain of the WRF-CAMx implementation is the South-East Europe/Eastern Mediterranean region from 18–30° N and 34.9–44.5° E. The grid resolution is 10 km × 10 km. The air quality modelling system runs on a daily basis in order to produce 72 h air quality forecasts. For the verification, the WRF-CAMx, the ENSEMBLE and the seven models are compared with available air quality data from the GMEECC (Greek Ministry of Environment Energy and Climatic Change) air pollution monitoring network as well as from the background station of Finokalia, operated by the University of Crete (Greece).
AEMET runs a version of the MOCAGE (Josse et al., 2004) model at 0.05° horizontal resolution in the Western Mediterranean coast, daily up to 48 h using the ENSEMBLE forecasts as chemical lateral boundary conditions. Meteorological forcings for the high resolution domain come from operational HIRLAM run every 6 h at AEMET (Navascues et al., 2013). Emissions over land in this domain come from the GEMS-TNO inventory (Visschedijk et al., 2007). The domain is 44–36° N–5° W–5° E. The ENSEMBLE has been compared to the AEMET forecasts and to observations from EMEP/GAW Spanish stations and from different local and regional Air Quality Monitoring networks. From these high resolution daily forecasts, a collection of case studies in which high resolution could have been an advantage, has been selected and analysed. These comparisons show the high variability of results between model forecasts depending on the location, time and day, whereas, sometimes, model forecast agreement is quite noticeable.

We are presenting here a brief summary of the analysis of the case study that occurred between the 15 and the 18 July 2013, when high values of ozone were measured in many Spanish Air Quality Monitoring Stations due to very strong solar radiation and high temperatures together with persistent anticyclonic conditions and very weak pressure gradients. Ozone concentrations at surface above 140 µg m⁻³ were not rare at the stations used in this period and values above 120 µg m⁻³ were common.

Figure 9 shows two maps with the 18 July 2013 ENSEMBLE and AEMET model at H+18 forecasts and the observations over-plotted using the same colour intervals. The ENSEMBLE forecasts generally fit well to the measurements. The main characteristic of the ENSEMBLE forecasts (left) is that it is too smooth to capture all the small scale features occurring in reality because of its horizontal resolution (~15 km). As an example, we can look at Fig. 9 in which the Madrid area has been magnified to observe how ozone values between 100 and 160 µg m⁻³ were measured by different Air Quality Networks (belonging to Madrid Regional Authorities and Madrid City Council) whereas in the ENSEMBLE forecasts all the concentrations are lying in the 100–120 µg m⁻³ interval. In the same period, the AEMET forecasts provides values in this area with
a higher spread, between 100 and 160 µg m\(^{-3}\) which fits better to observations. Something similar can be observed in the Eastern Spain area, also magnified in the same figure. Nevertheless, as mentioned, the quality of the ENSEMBLE forecasts is generally very good and the verification scores of the forecasts calculated for the whole period of the project show, most of the time, better results for the ENSEMBLE than for the AEMET forecasts. The limitations of the verification carried out (only the 7 EMEP background air quality stations within the domain have been considered) and the different high resolution emission inventories used in AEMET and ENSEMBLE can be part of the reason for these different results.

Another product we have started to generate at the end of the project is the behaviour of forecasts of the seven models together with the ENSEMBLE and the AEMET forecasts against observations from the EMEP Air Quality Network. An example is presented in Fig. 10. In this figure, we can see the ozone forecasts at ES10 station which is located at Cabo de Creus in the Northeastern corner of Spain (42.32° N, 3.32° E). We observe that the spread between the seven model forecasts in the H + 24 to H + 48 forecast period from the 9 April 2014 is fairly low with most of the members producing similar forecasts. It changes quickly on the next day at the same place with the seven models providing very different concentrations leading to a high spread. We have also observed differences in the spread of the members at other locations on the same day and forecast time. More generally this pattern with very different spreads (ranging from low to high) depends on the case studies: day, time period and location. The analysis of the spread between different model forecasts in the same period can help modellers to understand how their models behave in the Mediterranean area.

4 Conclusion and future developments

In this paper, we give an overview of the current state and performances of the forecasting system for European air quality that was put in place in the framework of MACC project and continued during MACC-II project and now in the MACC-III project.
Its strength comes from the fact that it is based on an ensemble of seven state-of-the-art chemistry-transport models (CHIMERE, EMEP, EURAD-IM, LOTOS-EUROS, MOCAGE, MATCH, SILAM) that are developed and run by recognized institutes in Europe. It also relies on good quality inputs for meteorological forcings, emissions and chemical boundary conditions. It provides daily 4 days forecasts for 6 major pollutants ($O_3$, $NO_2$, $SO_2$, $CO$, $PM_{10}$ and $PM_{2.5}$) and birch pollen during pollen season, and also additional species for downscaling air quality modeling purposes. The production also includes hourly analysis for the previous day. Daily statistical performances of the forecasts against available European air quality monitoring station are processed on daily, weekly and 3 monthly bases, giving an objective assessment of the products to users. They are also used to monitor the seasonal and yearly evolutions of the forecast scores.

Because of the resolution of the seven models (10 to 20 km), this system is not designed and do not attempt to forecast very local concentrations. Nevertheless, the ENSEMBLE has the capability to forecast pollution episodes at the regional scale as illustrated over 10 to 13 June 2014. The ENSEMBLE also shows good statistical performances for the main regulatory pollutants on a seasonal basis, as illustrated on ozone in summer and $PM_{10}$ in winter. These performances are directly linked to the skills of the seven individual models and to the use of a multi-model ensemble approach. The main improvements of the individual models between 2013 and 2014 were mainly on the aerosols representation. Scientific evaluation of the seven models and of the ENSEMBLE is also done for the Mediterranean region because of its specificities (emissions, population, meteorology, photochemistry, ...). Another important point to note is that major efforts have been put during MACC-II towards the full operationalisation of the system in order to improve its robustness.

The regional air quality production was extended during MACC-II and further developments are underway to improve the quality, the variety and the timeliness of its products based on users’ feedbacks. In the very short term, the ENSEMBLE analysis that is currently only provided for ozone will be extended to $NO_2$ from January 2015 and verification statistics with independent data will be produced. It is also planned to
shift the ENSEMBLE analysis production time earlier, at 11:00 UTC from early 2015 following the users’ recommendation. One planned change in the mid-term will be to have all individual models run at a \(\sim 10\) km horizontal resolution. This should improve the performances of the system compared to observations. Also the regional production benefits and will continue to benefit from the evolutions and improvements of the global production such as the use of the newly operated C-IFS (fully coupled chemistry to the IFS meteorological model) since September 2014 for regional boundary conditions for chemical species and aerosols. In parallel, a dedicated fire emission product for regional forecast purposes, available earlier than the current operational product, is progressively implemented in the seven models and its usefulness will be assessed.

Continuous research is pursued to improve the seven individual models and their assimilation systems. In particular, there is an important effort on the use of new satellite data or combinations of satellite data with surface measurements in the assimilation systems. Also, there is on-going work on ensemble methods in order to extract as much value as possible from the seven model forecasts. Alternative methods to the median are currently tested: application of weights on the individual models at each grid point related to the performances from the day before or spectral decomposition (Galmarini et al., 2013). The results of these alternative methods applied to the MACC-II multimodel ensemble will be the subject of a forthcoming paper. Another goal in MACC-II was the start of research and developments for the modelling of CO\(_2\) in the regional models in view of potential future high-resolution surface CO\(_2\) flux inversion products over Europe. This work will be pursued.

In the next few years, the availability of more daily European surface observations, in a wider European area (i.e. from more countries) and at earlier time is foreseen. More data on a wider area would comfort the strength of the statistical product evaluation. The continuous improvement of the quality of the surface monitoring data is also important for performance evaluation. Earlier availability of the surface station data would give the opportunity of an earlier production of the analyses with the goal of using the analyses as the initial state for the forecasts.
Other studies will be conducted on the possibility to provide complementary indicators such as the exceedences in ozone or PM$_{10}$. In the future, the production could be extended to other types of pollens than birch. There are currently some developments to test olive, grass and ambrosia pollens based on work done at the Finnish Meteorological Institute. Also, the possibility to produce additional species will be considered for users running forecast systems at finer scales than the MACC-II system, such as the concentrations of different types of aerosols.

Appendix: Statistical indicators

The forecast performances are measured using five statistical indicators: the mean bias, the root mean square error, the modified normalized mean bias, the fractional gross error and the correlation.

The mean bias captures the average deviations between two datasets and is defined as:

$$MB = \frac{1}{N} \sum_{i} (f_i - o_i)$$

where $f_i$ and $o_i$ are the forecast value at the observation location and the observation value, respectively.

The root mean square error combines the spread of individual error and is defined as:

$$RMSE = \sqrt{\frac{1}{N} \sum_{i} (f_i - o_i)^2}$$

It should be noted that the RMSE is strongly dominated by the largest values, due to the squaring operation. Especially in cases where prominent outliers occur, the usefulness of the RMSE is questionable and the interpretation becomes more difficult.
and RMSE are not dimensionless variables, but have the same dimension as the modelled/observed quantity and requires knowledge of typical mean values. By scaling the MB and RMSE to the observations these metrics can be made relative, dimensionless, and hence more appropriate for use as a score. This is relevant when comparing bias and RMSE of atmospheric species whose concentrations can vary by orders of magnitude. This is why the modified normalized mean bias (MNMB) and the fractional gross error (FGE) are also used. MNMB is defined as:

\[
MNMB = \frac{2}{N} \sum_{i} \left( \frac{f_{i} - o_{i}}{f_{i} + o_{i}} \right)
\]

This gives a measure of forecast bias bounded by the values –2 to +2. It performs symmetrically with respect to under and over-prediction of the observations, which is a desirable feature.

FGE is defined as:

\[
FGE = \frac{2}{N} \sum_{i} \left| \frac{f_{i} - o_{i}}{f_{i} + o_{i}} \right|
\]

FGE gives a measure of the overall forecast error. This is proposed in addition to the more traditional RMSE, because due to the squaring procedure the RMSE gives the largest weight to the (possibly spurious) largest observations. FGE is bounded between 0 and 2.

In addition, the correlation coefficient is needed to indicate the extent to which patterns in the forecast match those in the observations. The correlation coefficient \( R \) between the forecast and observed values is defined as:

\[
R = \frac{\frac{1}{N} \sum_{i}(f_{i} - \bar{f})(o_{i} - \bar{o})}{\sigma_{f}\sigma_{o}}
\]
where \( \bar{f} \) and \( \bar{d} \) are the mean values of the forecast and observed values and \( \sigma_f \) and \( \sigma_o \) are the corresponding SDs. The correlation coefficient has a maximum value of unity when, for each observation site, \((f_i - \bar{f}) = c(o_i - \bar{d})\), where \( c \) is a positive constant. In this case the two datasets have the same pattern of variation but are not identical unless \( c = 1 \) for all sites.

**Acknowledgements.** This study was funded by the European Commission under the EU Seventh Research Framework Programme (grant agreement no. 283576, MACC II). In-situ air quality data were provided by the European Environment Agency. Additional financial support at national level was given by the French Ministère de l’écologie, du développement durable et de l’énergie through the ADONISS project. This work was granted access to the HPC resources of CCRT under the allocation 2013–6695 made by GENCI (Grand Equipement National de Calcul Intensif). IASI has been developed and built under the responsibility of the Centre National d’Etudes Spatiales (CNES, France). Developments of SILAM system were supported by projects ASTREX and IS4FIRES of Academy of Finland. NO\(_2\) column retrievals from AURA/OMI and METOP/GOME-2, and CO profiles from TERRA/MOPITT have been used for assimilation in some of the individual models.

**References**


Friese, E. and Ebel, A.: Temperature dependent thermodynamic model of the system $\text{H}^+\text{NH}_4^+\text{Na}^+\text{SO}_4^{2-}\text{NO}_3^-\text{Cl}^-\text{H}_2\text{O}$, J. Phys. Chem. A, 114, 11595–11631, 2010.


Fountoukis, C. and Nenes, A.: ISORROPIA II: a computationally efficient thermodynamic equilibrium model for $K^+\text{Ca}^{2+}\text{Mg}^{2+}\text{NH}_4^+\text{Na}^+\text{SO}_4^{2-}\text{NO}_3^-\text{Cl}^-\text{H}_2\text{O}$ aerosols, Atmos. Chem. Phys., 7, 4639–4659, doi:10.5194/acp-7-4639-2007, 2007.


2783


2784


2788


Table 1. Portfolio of the MACC-II regional data products. Each product is provided once daily.
Core species correspond to \( \text{O}_3 \), \( \text{NO}_2 \), \( \text{CO} \), \( \text{SO}_2 \), \( \text{PM}_{10} \), \( \text{PM}_{2.5} \). Additional species correspond to \( \text{NO} \), \( \text{NH}_3 \), PAN+PAN precursors, total Non-Methane Volatile Organic Compounds. Birch pollen concentrations are only available during season from 1 March to 30 June each year. Old levels refer to surface, 500, 1000, 3000 and 5000 m, corresponding to the production before mid-May 2014. All levels refers to surface, 50, 250 m, 500, 1000 m, 2000, 3000 and 5000 m, produced from mid-May 2014. The analysis is run a posteriori on Day0 for Day-1 (00:00 to 24:00 UTC).

<table>
<thead>
<tr>
<th>Model name</th>
<th>Forecast or Analysis</th>
<th>Species</th>
<th>Time span</th>
<th>Vertical levels</th>
<th>Format</th>
</tr>
</thead>
<tbody>
<tr>
<td>CHIMERE</td>
<td>Forecast</td>
<td>core + additional</td>
<td>0 to 96 h, hourly</td>
<td>All levels</td>
<td>Netcdf</td>
</tr>
<tr>
<td>CHIMERE</td>
<td>Analysis</td>
<td>( \text{O}<em>3 ), ( \text{PM}</em>{10} )</td>
<td>0 to 96 h, hourly</td>
<td>Surface</td>
<td>Netcdf</td>
</tr>
<tr>
<td>EMEREP</td>
<td>Forecast</td>
<td>core + additional</td>
<td>0 to 96 h, hourly</td>
<td>All levels</td>
<td>Netcdf</td>
</tr>
<tr>
<td>EMEREP</td>
<td>Analysis</td>
<td>( \text{NO}_2 )</td>
<td>–24 to –1 h, hourly</td>
<td>Surface</td>
<td>Netcdf</td>
</tr>
<tr>
<td>EURAD-IM</td>
<td>Forecast</td>
<td>Birch pollen</td>
<td>0 to 96 h, hourly</td>
<td>All levels</td>
<td>Netcdf</td>
</tr>
<tr>
<td>EURAD-IM</td>
<td>Analysis</td>
<td>Birch pollen</td>
<td>0 to 96 h, hourly</td>
<td>Surface</td>
<td>Netcdf</td>
</tr>
<tr>
<td>EURAD-IM</td>
<td>Analysis</td>
<td>( \text{O}_3 ), ( \text{NO}_2 ), ( \text{CO} ), ( \text{SO}<em>2 ), ( \text{PM}</em>{10} )</td>
<td>–24 to –1 h, hourly</td>
<td>Surface</td>
<td>Netcdf</td>
</tr>
<tr>
<td>LOTSOS-EUROS</td>
<td>Forecast</td>
<td>Birch pollen</td>
<td>0 to 96 h, hourly</td>
<td>Old levels</td>
<td>Netcdf</td>
</tr>
<tr>
<td>LOTSOS-EUROS</td>
<td>Analysis</td>
<td>Birch pollen</td>
<td>0 to 96 h, hourly</td>
<td>Surface</td>
<td>Netcdf</td>
</tr>
<tr>
<td>MATCH</td>
<td>Forecast</td>
<td>Birch pollen</td>
<td>0 to 96 h, hourly</td>
<td>All levels</td>
<td>Netcdf</td>
</tr>
<tr>
<td>MATCH</td>
<td>Analysis</td>
<td>Birch pollen</td>
<td>0 to 96 h, hourly</td>
<td>Surface</td>
<td>Netcdf</td>
</tr>
<tr>
<td>MOCAGE</td>
<td>Forecast</td>
<td>core + additional (except ( \text{NH}_3 ))</td>
<td>0 to 96 h, hourly</td>
<td>All levels</td>
<td>Netcdf</td>
</tr>
<tr>
<td>MOCAGE</td>
<td>Analysis</td>
<td>Birch pollen</td>
<td>0 to 96 h, hourly</td>
<td>Surface</td>
<td>Netcdf</td>
</tr>
<tr>
<td>SILAM</td>
<td>Forecast</td>
<td>core</td>
<td>0 to 96 h, hourly</td>
<td>All levels</td>
<td>Netcdf</td>
</tr>
<tr>
<td>SILAM</td>
<td>Analysis</td>
<td>Birch pollen</td>
<td>0 to 96 h, hourly</td>
<td>Surface</td>
<td>Netcdf</td>
</tr>
<tr>
<td>SILAM</td>
<td>Analysis</td>
<td>( \text{O}_3 ), ( \text{NO}_2 ), ( \text{SO}_2 )</td>
<td>–24 to –1 h, hourly</td>
<td>Surface</td>
<td>Netcdf</td>
</tr>
<tr>
<td>ENSEMBLE</td>
<td>Forecast</td>
<td>Birch pollen</td>
<td>0 to 96 h, hourly</td>
<td>All levels</td>
<td>Netcdf + Grib2</td>
</tr>
<tr>
<td>ENSEMBLE</td>
<td>Analysis</td>
<td>( \text{O}_3 )</td>
<td>–24 to –1 h, hourly</td>
<td>Surface</td>
<td>Netcdf + Grib2</td>
</tr>
</tbody>
</table>
### Table 2. Time of delivery of the ENSEMBLE numerical products. Core species for the analysis is restricted to ozone only.

<table>
<thead>
<tr>
<th></th>
<th>Forecast Day0 (0–24 h)</th>
<th>Forecast Day1 (25–48 h)</th>
<th>Forecast Day2 (49–72 h)</th>
<th>Forecast Day3 (73–96 h)</th>
<th>Analysis (−24–0 h)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Core species</td>
<td>07:00 UTC</td>
<td>07:00 UTC</td>
<td>08:00 UTC</td>
<td>09:00 UTC</td>
<td>14:30 UTC</td>
</tr>
<tr>
<td>Additional species</td>
<td>07:00 UTC</td>
<td>07:00 UTC</td>
<td>08:00 UTC</td>
<td>09:00 UTC</td>
<td>N/A</td>
</tr>
</tbody>
</table>
Table 3. General characteristics of the regional models at the end of MACC-II project.

<table>
<thead>
<tr>
<th>Model</th>
<th>Operated by</th>
<th>Horizontal resolution</th>
<th>Vertical levels</th>
<th>Top height</th>
</tr>
</thead>
<tbody>
<tr>
<td>CHIMERE</td>
<td>INERIS (Institut National de l’Environnement Industriel et des Risques) France</td>
<td>0.1° × 0.1°</td>
<td>8 levels</td>
<td>Top at 500 hPa</td>
</tr>
<tr>
<td>EMEP</td>
<td>MET Norway (Meteorologisk institutt) Norway</td>
<td>0.25° × 0.125°</td>
<td>20 levels</td>
<td>Top at 100 hPa</td>
</tr>
<tr>
<td>EURAD-IM</td>
<td>RIUUK (Rheinisches Institut Fuer Umwelt-Forschung an der Universitaet zu Koeln E. V.) Germany</td>
<td>15 km on a Lambert conformal projection</td>
<td>23 levels</td>
<td>Top at 100 hPa</td>
</tr>
<tr>
<td>LOTS-EUROS</td>
<td>KNMI (Koninklijk Nederlands Meteorologisch Instituut) Netherlands</td>
<td>0.25° × 0.125°</td>
<td>4 levels</td>
<td>Top at 3.5 km</td>
</tr>
<tr>
<td>MATCH</td>
<td>SMHI (Sveriges Meteorologiska och Hydrologiska Institut) Sweden</td>
<td>0.2° × 0.2°</td>
<td>52 levels</td>
<td>Top at 300 hPa</td>
</tr>
<tr>
<td>MOCAGE</td>
<td>Météo-France France</td>
<td>0.2° × 0.2°</td>
<td>47 levels</td>
<td>Top at 5 hPa</td>
</tr>
<tr>
<td>SILAM</td>
<td>FMI (Ilmatieteen Laitos) Finland</td>
<td>0.15° × 0.15°</td>
<td>8 levels</td>
<td>Top at 6.7 km</td>
</tr>
</tbody>
</table>
**Table 4.** Characteristics of the daily assimilation chains of the regional models at the end of MACC-II project.

<table>
<thead>
<tr>
<th>Model</th>
<th>Assimilation method</th>
<th>Observation assimilated</th>
<th>Species analysed</th>
</tr>
</thead>
<tbody>
<tr>
<td>CHIMERE</td>
<td>Optimal interpolation</td>
<td>O$<em>3$ and PM$</em>{10}$ from surface stations, NO$_2$ columns from OMI and NO$_2$ from surface stations</td>
<td>O$<em>3$, PM$</em>{10}$</td>
</tr>
<tr>
<td>EMEP</td>
<td>3-D- Variational</td>
<td>NO$_2$ columns from OMI and NO$_2$ from surface stations</td>
<td>NO$_2$</td>
</tr>
<tr>
<td>EURAD-IM</td>
<td>3-D- Variational</td>
<td>O$_3$, NO, NO$<em>2$, SO$<em>2$, CO, PM$</em>{10}$, PM$</em>{2.5}$ from surface stations, OMI and GOME-2 NO$_2$ column retrievals, MOPITT CO profiles</td>
<td>O$_3$, NO$_2$, SO$<em>2$, CO, PM$</em>{10}$</td>
</tr>
<tr>
<td>LOTOS-EUROS MATCH</td>
<td>Ensemble Kalman filter 3-D- Variational</td>
<td>O$_3$ from surface stations, O$<em>3$, NO$<em>2$, CO, PM$</em>{10}$, PM$</em>{2.5}$ from surface stations</td>
<td>O$_3$, O$<em>3$, NO$<em>2$, CO, PM$</em>{10}$, PM$</em>{2.5}$</td>
</tr>
<tr>
<td>MOCAGE</td>
<td>3-D- Variational</td>
<td>O$_3$ from surface stations, O$_3$, NO$_2$ and SO$_2$ from surface stations</td>
<td>O$_3$, O$_3$, NO$_2$, SO$_2$</td>
</tr>
<tr>
<td>SILAM</td>
<td>4-D-Variational</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
Figure 1. Schematic of the general organisation of the MACC-II air quality forecast and analysis system.
Figure 2. Ozone measurements from surface stations in µgm$^{-3}$ from 10 June 2014 at 00:00 UTC to 14 June 2014 at 00:00 UTC located (a) at 47.67°N/13.11°E (Hallein, Austria), (b) at 47.69°N/16.58°E (Sopron, Hungaria) (c) at 43.33°N/5.12°E (Sausset, France) and (d) at 43.34°N/5.73°E (Plan d’Aups, France).
Figure 3. EPSgrams giving median, 90% percentile, 75% percentile, 25% percentile, 10% percentile, minimum and maximum from 3 hourly outputs of the 96 h forecasts of the 7 models from 10 June 2014 at 00:00 UTC to 14 June 2014 at 00:00 UTC. Model outputs are interpolated at the location of the stations shown in Fig. 2. (a) at 47.67° N/13.11° E (Hallein, Austria), (b) at 47.69° N/16.57° E (Sopron, Hungaria) (c) at 43.33° N/5.12° E (Sausset, France) and (d) at 43.34° N/5.73° E (Plan d’Aups, France). The red dashed line corresponds to the 120 µg.m⁻³ threshold.
Figure 4. Maps of ozone concentrations at the surface in µg m⁻³ of the 15 h forecast for Day0 at 15:00 UTC of the median ensemble constructed with the 7 model forecasts. NRT AQ observations available (circles) for the same date/time are overlotted on the maps using the same colour scale. Top left: for 10 June 2014 at 15:00 UTC. Top right: for 11 June 2014 at 15:00 UTC. Bottom left: for 12 June 2014 at 15:00 UTC. Bottom right: for 13 June 2014 at 15:00 UTC.
Figure 5. Statistical indicators (see Appendix) for ozone as a function of the forecast time in hour for the median ensemble (in turquoise) and the seven models (other colours) compared to the hourly surface station measurements available for the period from the 9 to 15 June 2014 over the MACC-II European domain. (a) MB in \( \mu g \cdot m^{-3} \), (b) MNMB, (c) RMSE in \( \mu g \cdot m^{-3} \), (d) FGE and (e) correlation.
Figure 6. Statistical indicators (see Appendix) for ozone as a function of the forecast time in hour MEDIAN 7, MEDIAN 5, MEDIAN 3 and 1BEST (see text for their definition) compared to the hourly surface station measurements available for the period from the 9 to 15 June 2014 over the MACC-II European domain. (a) MB in $\mu$g m$^{-3}$, (b) MNMB, (c) RMSE in $\mu$g m$^{-3}$, (d) FGE and (e) correlation.
Figure 7. Statistical indicators (see Appendix) for ozone as a function of the forecast time in hour for the median ensemble compared to the hourly surface station measurements available for the period from 1 June at 00:00 UTC to 1 September at 00:00 UTC over the MACC-II European domain for 2013 (left column) and for 2014 (right column): MB (1st row in $\mu g m^{-3}$), MNMB (2nd row), RMSE (3rd row in $\mu g m^{-3}$), FGE (4th row) and correlation (5th row).
Figure 8. Statistical indicators (see Appendix) for PM$_{10}$ as a function of the forecast time in hour for the median ensemble compared to the hourly surface station measurements available for the period from the 1 December at 00:00 UTC to the 1 February at 00:00 UTC over the MACC-II European domain for winter 2012–2013 (left column) and for winter 2013–2014 (right column): MB (1st row in µg m$^{-3}$), MNMB (2nd row), RMSE (3rd row in µg m$^{-3}$), FGE (4th row) and correlation (5th row).
Figure 9. ENSEMBLE (left) and AEMET (right) surface ozone concentrations in $\mu$g m$^{-3}$ from a forecast (H + 18) started on 18 July 2013 at 00:00 UTC for the Western Mediterranean area. Observations from different air quality networks have been plotted on the map. The Madrid and the Eastern Spain areas appear magnified.
Figure 10. Ozone concentrations (in μg m⁻³) from a 48 h forecast of ENSEMBLE, AEMET and the 7 individual models at ES10 EMEP air quality station which is located at Cabo de Creus in the Northeastern corner of Spain (42.32° N, 3.32° E). The forecast is started on 9 April 2014. CHI, EMP, KNM, FMI, MFM, RIU, SMH, ENS, MACC3 and OBS correspond to CHIMERE, EMEP, LOTOS-EUROS, SILAM, MOCAGE, EURAD-IM, MATCH, ENSEMBLE, AEMET models and observations, respectively.