

1 Evaluation of European air quality modelled by CAMx 2 including the volatility basis set scheme

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30 **Abstract**

31 Four periods of EMEP (European Monitoring and Evaluation Programme) intensive
32 measurement campaigns (June 2006, January 2007, September-October 2008 and February-
33 March 2009) were modelled using the regional air quality model CAMx with VBS (Volatility
34 Basis Set) approach for the first time in Europe within the framework of the EURODELTA-
35 III model intercomparison exercise. More detailed analysis and sensitivity tests were
36 performed for the period of February-March 2009 and June 2006 to investigate the
37 uncertainties in emissions as well as to improve the modelling of organic aerosols (OA).
38 Model performance for selected gas phase species and PM_{2.5} was evaluated using the
39 European air quality database Airbase. Sulfur dioxide (SO₂) and ozone (O₃) were found to be
40 overestimated for all the four periods with O₃ having the largest mean bias during June 2006
41 and January-February 2007 periods (8.9 ppb and 12.3 ppb mean biases, respectively). In
42 contrast, nitrogen dioxide (NO₂) and carbon monoxide (CO) were found to be underestimated
43 for all the four periods. CAMx reproduced both total concentrations and monthly variations
44 of PM_{2.5} for all the four periods with average biases ranging from -2.1 µg m⁻³ to 1.0 µg m⁻³.
45 Comparisons with AMS (aerosol mass spectrometer) measurements at different sites in
46 Europe during February-March 2009 showed that in general the model over-predicts the
47 inorganic aerosol fraction and under-predicts the organic one, such that the good agreement
48 for PM_{2.5} is partly due to compensation of errors. The effect of the choice of volatility basis
49 set scheme (VBS) on OA was investigated as well. Two sensitivity tests with volatility
50 distributions based on previous chamber and ambient measurements data were performed. For
51 February-March 2009 the chamber-case reduced the total OA concentrations by about 42% on
52 average. On the other hand, a test based on ambient measurement data increased OA
53 concentrations by about 42% for the same period bringing model and observations into better
54 agreement. Comparison with the AMS data at the rural Swiss site Payerne in June 2006 shows
55 no significant improvement in modelled OA concentration. Further sensitivity tests with
56 increased biogenic and anthropogenic emissions suggest that OA in Payerne were affected by
57 changes in emissions from residential heating during the February-March 2009 whereas it was
58 more sensitive to biogenic precursors in June 2006.

59

60 **1 Introduction**

61 Air pollution is known to cause damage to human health, vegetation and ecosystems. It is one
62 of the main environmental causes of premature death. Only in Europe, more than 400,000
63 premature deaths were estimated in 2011 with PM_{2.5} (particles less than 2.5 µm in
64 aerodynamic diameter) having the highest relative risk for health damage (WHO, 2014a). Air
65 quality models help understanding the processes taking place between emission sources and
66 pollutant concentrations at receptor sites. They are very useful to define control strategies for
67 future legislation. In spite of large improvements in recent years, Chemical Transport Models
68 (CTMs) have still some uncertainties (Solazzo et al., 2012a). Various air quality model
69 intercomparison exercises were successfully carried out over the last decades to determine
70 uncertainties in chemical and physical processes governing particulate matter and its
71 precursors (Solazzo et al., 2012a; Bessagnet et al., 2014). However, a large variability in
72 particulate matter concentrations was found between different models indicating process
73 parameterization as one of the main reasons for such discrepancies. Moreover, recent studies
74 based on AMS (Aerosol Mass Spectrometer) measurements at different sites in Europe,
75 revealed that the organic fraction dominates the non-refractory PM₁ composition (Crippa et
76 al., 2014). Organic aerosol (OA) can be found in the atmosphere from direct emission by
77 various sources, such as fossil fuel combustion by road vehicle engines or residential wood
78 combustion. Direct emissions of OA are typically referred to as primary organic aerosol
79 (POA) whereas gas-to-particle conversion is referred to as secondary organic aerosol (SOA).
80 Formation mechanisms of SOAs are not very well known yet and their representation in
81 CTMs is still challenging (Hallquist et al. 2009; Fountoukis et al., 2011; Bergstrom et al.,
82 2012; Li et al., 2013; Langmann et al., 2014; Tsigaridis et al., 2014). In one of our recent
83 aerosol modelling studies we compared model PM_{2.5} prediction with PM₁ AMS
84 measurements for different sites (Payerne and Zürich) and periods (summer and winter) in
85 Switzerland. We found that particulate matter was generally well reproduced by the model
86 with the SOA fraction being under-predicted and POA over-predicted (Aksoyoglu et al.,
87 2011). Traditional CTMs treat POA as non-volatile. Some studies however have revealed the
88 semi-volatile nature of POA, through its dynamic equilibrium of organic aerosol with its gas
89 phase, and the importance of semi-volatile (SVOC) and intermediate volatility (IVOC)
90 organic compounds as SOA precursors (Donahue et al., 2006; Robinson et al., 2007; Cappa
91 and Jimenez, 2010). To describe the absorptive partitioning and ongoing oxidation of the
92 atmospheric material, a volatility basis set (VBS) where organic species are organized into

93 surrogates according to their volatility was developed (Donahue et al., 2011, 2012a,b). Air
94 quality models updated with VBS scheme started being used (Lane et al., 2008; Murphy and
95 Pandis, 2009; Hodzic et al., 2010; Fountoukis et al., 2011; Bergström et al., 2012; Murphy et
96 al., 2012; Jo et al., 2013; Zhang et al., 2013; Athanasopoulou et al., 2013; Fountoukis et al.,
97 2014). Bergström et al. (2012) reported an EMEP model study over Europe for the 2002-2007
98 period using different assumptions regarding partitioning and aging processes. They could not
99 reproduce the measured OA levels in winter suggesting that residential wood combustion
100 inventories might be underestimated in different parts of Europe. Fountoukis et al. (2014)
101 applied the PMCAMx model to simulate EUCAARI (Kulmala et al., 2009, 2011) and EMEP
102 (Tørseth et al., 2012) campaigns in Europe. They could reproduce most of PM₁ daily average
103 OA observations within a factor of two, with the February-March 2009 period having the
104 largest discrepancies. Zhang et al. (2013) deployed the CHIMERE model with the VBS
105 framework during the MEGAPOLI summer campaign in the Greater Paris region for July
106 2009. They found a considerable improvement in predicted SOA concentrations which might
107 be even overestimated depending on the emission inventory used. In our study, we applied
108 the regional air quality model CAMx with the VBS scheme for the first time in Europe within
109 the framework of EURODELTA-III model intercomparison exercise. In addition to the base
110 case configuration used in the exercise, more sensitivity tests with the VBS scheme for winter
111 and summer episodes were performed together with a general evaluation of the four EMEP
112 field measurement campaigns.

113 **2 Method**

114 **2.1 The EURODELTA-III exercise**

115 The EURODELTA-III (EDIII) framework is a European model intercomparison exercise
116 between several modelling teams sharing both efforts and technical knowledge in order to
117 reduce model uncertainties and to improve understanding of the performances. It contributes
118 to the scientific work of the United Nations Economic Commission for Europe (UNECE)
119 Task Force on Measurement and Modelling (TFMM) within the Convention on Long-range
120 Transboundary Air Pollution (CLRTAP). In the first phase of the EDIII exercise, 4 periods of
121 the EMEP field measurement campaigns were chosen in order to evaluate the model results:

- 122 • 1 June – 30 June 2006
- 123 • 8 January – 4 February 2007
- 124 • 17 September – 15 October 2008

125 • 25 February – 26 March 2009

126 Multiple models were applied on a common domain and driven with the same input data
127 provided by the National Institute for Industrial Environment and Risks (INERIS). However,
128 for some models, different meteorology, boundary conditions and emissions data such as
129 biogenic emissions were used (Bessagnet et al., 2014).

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131 **2.2 Modelling method**

132 **2.2.1 CAMx**

133 The Comprehensive Air quality Model with extensions, CAMx-VBS (CAMx5.41_VBS,
134 kindly provided by ENVIRON before its public release) was used in this study. The model
135 domain consisted of one grid with a horizontal resolution of $0.25^\circ \times 0.25^\circ$. The latitude and
136 longitude grid extended from 25.125°W to 45.125°E and 29.875°N to 70.125°N resulting in
137 281×161 grid cells covering the whole of Europe. Hourly four-dimensional meteorological
138 fields for wind speed and direction, pressure, temperature, specific humidity, cloud cover and
139 rain required by CAMx simulations were calculated from ECMWF IFS (Integrated Forecast
140 System) data at 0.2° resolution. Vertical diffusivity coefficients were estimated following the
141 Kz approach of O'Brien (1970) using PBL depth profiles as available in IFS data. CAMx
142 simulations used 33 terrain-following σ -levels up to about 8000 m above ground level, as in
143 the original IFS data. The lowest layer was about 20 m thick. MACC (Monitoring
144 Atmospheric Composition and Climate) reanalysis data were used to initialize initial and the
145 boundary condition fields (Benedetti et al., 2009; Inness et al., 2013). Elemental carbon,
146 organic aerosol, dust and sulfate were used to model aerosol species at the boundaries of the
147 domain. One half of the OA was assumed to be secondary organic aerosol (SOA) and the
148 other half primary organic aerosol (POA), as recommended in the EDIII exercise. Photolysis
149 rate inputs were calculated using the TUV radiative transfer and photolysis model
150 (Madronich, 2002). The required ozone column densities to determine the spatial and
151 temporal variation of the photolysis rates were extracted from TOMS data (NASA/GSFC,
152 2005). Removal processes as dry and wet deposition were simulated using the Zhang
153 resistance model (Zhang et al., 2003) and a scavenging model approach for both gases and
154 aerosols (ENVIRON, 2011), respectively. For the gas phase chemistry the Carbon Bond
155 (CB05) mechanism (Yarwood et al., 2005) with 156 reactions and up to 89 species was used.

156 Partitioning of inorganic aerosols (sulfate, nitrate, ammonium, sodium and chloride) was
157 performed using the ISORROPIA thermodynamic model (Nenes et al., 1998). Aqueous
158 sulfate and nitrate formation in cloud water was simulated as well using the RADM aqueous
159 chemistry algorithm (Chang et al., 1987).

160 **2.2.2 Emissions**

161 **Anthropogenic emissions**

162 Annual total gridded anthropogenic emissions were prepared and provided by INERIS for the
163 EDIII exercise, which is based on a merging process of data-bases from different sources, i.e.
164 TNO-MACC (Kuenen et al., 2011), EMEP (Vestreng et al., 2007), GAINS (The Greenhouse
165 Gas and Air Pollution Interactions and Synergies). For specific countries where TNO-MACC
166 emissions were missing (Iceland, Liechtenstein, Malta and Asian countries), the EMEP $0.5^\circ \times$
167 0.5° emissions were used and re-gridded using adequate proxies such as “artificial land-use”
168 and EPER (European Pollutant Emission Register) data (<http://www.eea.europa.eu/>) for
169 industries. Total primary particle emissions were made available by EMEP in two different
170 size ranges: below $2.5\mu\text{m}$ (fine) and between $2.5\mu\text{m}$ and $10\mu\text{m}$ (coarse). Total emissions were
171 later split to estimate the amount of elemental carbon, and organic matter for each of the 10
172 SNAP codes (Selected Nomenclature for Air Pollution) and country. The final emission
173 inventory thus compiled consisted of 6 gas species namely methane, carbon monoxide,
174 ammonia, sulfur oxides, non-methane volatile organic compounds and nitrogen oxides and 6
175 categories of particulate matter classes: fine elemental carbon (EC2.5), coarse elemental
176 carbon (EC10), fine primary organic material (fine POA), coarse primary organic material
177 (coarse POA), fine other primary particulate material (non-carbonaceous) and coarse other
178 primary particulate material (non-carbonaceous). $\text{PM}_{2.5}$ and PM_{10} emissions were provided by
179 EMEP and they were split to elemental carbon and organic matter using the fractions given by
180 IIASA (International Institute for Applied Systems Analysis) for each source and country.
181 Total non-methane volatile organic compounds were split for the CB05 mechanism using the
182 recommendations of Passant (2002). Hourly, weekly and monthly time profiles as in the
183 EURODELTAII exercise were applied to total annual anthropogenic emissions.

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187 **Biogenic emissions**

188 Biogenic VOC emissions were calculated using the Model of Emissions of Gases and
189 Aerosols from Nature MEGANv2.1 (Guenther et al., 2012). This model is driven by
190 meteorological variables such as hourly temperature, solar radiation, humidity, wind speed,
191 soil moisture and land cover data including leaf area index (LAI) and plant function type
192 (PFT) as available in the Community Land Model 4.0. 8-Days average satellite data at $0.25^\circ \times$
193 0.25° resolution were pre-processed and made available from the TERRA/MODIS satellite
194 system. Sixteen plant function types including needle-leaved evergreen, needle-leaved
195 deciduous, broad-leaved evergreen, broad-leaved deciduous, grass and crop for different
196 climatic zones were prepared for this study at $0.25^\circ \times 0.25^\circ$ resolution together with the global
197 emission factors of α -pinene, β -pinene, 3-carene, isoprene, limonene, 232-methylbutenol,
198 myrcene, NO_x , *t*- β -ocimene and sabinene. Common BVOC species such as isoprene, terpene,
199 sesquiterpene, xylene and toluene were obtained for each hour and cell in the domain.

200 **2.2.3 VBS scheme**

201 A new volatility basis set (VBS) scheme is available in the CAMx model to describe changes
202 in oxidation state and volatility. A total of four basis set simulates the evolution of organic
203 aerosol in the atmosphere (Koo et al., 2014). POA emissions were split in HOA-like and
204 BBOA-like emissions and allocated in two different basis sets. HOA-like emissions include
205 emissions from all SNAP sectors except SNAP2 (non-industrial combustion plants) and
206 SNAP10 (agriculture) which were assigned to BBOA-like emissions. Two other sets were
207 used in the model to allocate secondary organic aerosol from anthropogenic (i.e. xylene and
208 toluene) (ASOA) and biogenic (i.e. isoprene, monoterpene and sesquiterpene) (BSOA)
209 gaseous precursors. These two sets also allocate oxidation products of POA vapours, from
210 each of the two primary sets (HOA-like and BBOA-like). The 2D volatility space retrieved by
211 Donahue et al. (2011; 2012a,b) was used to distribute the organic molecular structures for
212 each of the volatility bins and different sets (Table S1). Five volatility bins represent the range
213 of semi-volatile organic compounds (SVOCs) ranging from $10^{-1} \mu\text{g m}^{-3}$ to $10^3 \mu\text{g m}^{-3}$ in
214 saturation concentrations (C^*). Oxidation processes are modelled by shifting C^* by a factor of
215 10 in the next lower volatility bin, increasing the oxidation state and reducing the carbon
216 number to account for fragmentation. OH reaction rates are assumed to be $4 \times 10^{-11} \text{ cm}^3$
217 $\text{molecule}^{-1} \text{ s}^{-1}$ for the reaction of semi-volatile primary vapors with OH and 2×10^{-11} for
218 further aging of ASOA and POA vapours from HOA-like emissions. More details about the

219 VBS parameterization in CAMx can be found in Koo et al. (2014). Further aging of BSOA is
220 not considered in this study based on previous modelling results showing over-prediction of
221 OA when such process is taken into account (Lane et al., 2008; Murphy and Pandis, 2009).
222 This implies that also further aging of POA vapours from BBOA-like emissions was not
223 considered since it is performed in the same basis set. In this work we focus on the effects of a
224 VBS framework on the total OA fraction. Aging processes and alternative VBS
225 implementations will be discussed together with SOA and POA components in a following
226 paper (Ciarelli et al. in prep). Three sensitivity tests were performed with different
227 assumptions on the volatility distributions (Table 1):

228 • **NOVBS:** Primary organic aerosol was assumed to be non-volatile. Biogenic (isoprene,
229 monoterpenes and sesquiterpenes) and anthropogenic (xylene, toluene and other
230 aromatics) volatile organic compounds (VOCs) were used as precursors for secondary
231 organic aerosol. Partitioning of condensable gases to secondary organic aerosol was
232 calculated using a semi-volatile equilibrium approach (Strader, 1999).

233 • **VBS_ROB:** Primary organic aerosol was assumed to be volatile and undergo
234 chemical oxidation. The volatility distribution estimated by Robinson et al. (2007) was
235 applied to HOA-like and BBOA-like emissions. Emissions of intermediate volatility
236 organic compounds (IVOCs) were assumed to be 1.5 times those of primary organic
237 aerosol (POA) as suggested by Robinson et al. (2007).

238 • **VBS_BC:** Primary organic aerosol was assumed to be volatile and undergo chemical
239 oxidation using the approach of Shrivastava et al. (2011) and Tsimpidi et al. (2010).
240 The total primary emissions are roughly 3 times higher than in **VBS_ROB**. Different
241 volatility distributions were applied for HOA and BBOA-like emissions. IVOCs were
242 assumed to be 1.5 times the amount of POA. This implies that for this scenario the
243 SVOC + IVOC mass added is equal to 7.5 times the initial amount of POA. This
244 represents the base case scenario used to evaluate gas phase and PM_{2.5} model
245 performance.

246 Based on the **VBS_BC** base case scenario, two other sensitivity tests were performed with
247 respect to emissions:

248 • **VBS_BC_2xBVOC:** Increased BVOCs emissions by a factor of 2.

249 • **VBS_BC_2xBBOA:** Increased BBOA-like emissions by a factor of 2.

250 **2.3 Statistical methods**

251 Statistical procedures as available in the Atmospheric Model Evaluation Tool (AMET, Apple
252 et al., 2010) were used in this study to evaluate model performance. Daily ambient
253 measurements of main gas phase species i.e. O₃, NO₂, CO, SO₂ and fine particulate matter
254 (PM_{2.5}) were extracted from the Airbase database in Europe and statistics reported in terms of
255 mean bias (MB), mean error (ME), mean fractional bias (MFB) mean fractional error (MFE)
256 and correlation coefficient (*r*).

257 Due to the coarse grid resolution, only rural-background stations, defined as stations far from
258 city sources of air pollution with pollution levels determined by the integrated contribution
259 from all sources upwind of the station (ETC/ACC, 2004/7), with at least 80% daily average
260 observations available were considered for the statistical analysis. For PM_{2.5} this resulted in
261 48 stations available for June 2006, 56 for January-February 2007, 90 for September-October
262 2008 and 110 stations for February-March 2009. PM_{2.5} components were further evaluated for
263 the February-March 2009 period where comprehensive high resolution AMS measurements at
264 11 European sites were available, i.e., at Barcelona, Cabauw, Chilbolton, Helsinki, Hyytiälä,
265 Mace Head, Melpitz, Montseny, Payerne, Puy de Dôme and Vavihill (Crippa et al., 2014).

266

267 **3 Results and discussions**

268 **3.1 Model evaluation**

269 Model performance metrics for gas phase species CO, NO₂, O₃ and SO₂ as well as for PM_{2.5}
270 are reported in Table 2 and they refer to the base case VBS_BC.

271 **NO₂ and O₃**

272 NO₂ was found to be under-predicted for all the four periods with mean fractional bias
273 between -54% and -28% and NO₂ concentrations being particularly under-predicted during
274 June 2006. Evaluation of the EURODELTA III model inter-comparison exercise showed that
275 all models performed similarly for NO₂ in terms of correlation with *r* values in the range 0.6-
276 0.7 and the spatial correlation was much higher in the range 0.7-0.9 for all models (Bessagnet
277 et al., 2016) with a general underestimation in the afternoon. The NO₂ performance could be
278 influenced by several factors:

279 - Uncertainties in the emission inventories. Although NO_x emission estimates in Europe
280 are thought to have an uncertainty of about $\pm 20\%$, the complete data set used in the
281 inventories has much higher uncertainty (Kuenen et al., 2014). A recent study
282 identified a significant discrepancy between emission estimates and actual flux
283 measurements, with the highest underestimation being a factor of two in central
284 London mainly due to under-representation of real world road traffic emissions
285 (Vaughan et al., 2016)

286 - The relatively coarse resolution of the domain which may result in too low NO_x
287 emissions or isolated local events that the model cannot resolve. We report daily
288 average time series of NO_2 for the period of Feb-Mar 2009 for stations in Table 2 as
289 well as daily average time series of NO_2 for stations not exceeding 5 ppb (which
290 represents 92% of the stations in Table 2) (Figure S1). The model performance for
291 NO_2 significantly improved when the 5 ppb threshold was applied to the dataset. An
292 emission map of NO for 1 March 2009 at 6 AM is reported in Figure S2. High
293 emissions of NO are predicted in the Benelux area, Po Valley, Germany and in some
294 of the eastern European countries. High NO emissions due to ship traffic are also
295 visible especially in the Mediterranean Sea

296 - Possible positive artefacts in the chemiluminescence methods for measuring NO_2 may
297 also occur when NO_2 is catalytically converted to NO on the molybdenum surface
298 leading to an over-prediction of measured NO_2 concentrations (Steinbacher et al.,
299 2007; Villena et al., 2012)

300 - Moreover, an evaluation of planetary boundary layer height (PBLH) within the EDIII
301 shows that although the PBLH was quite well represented in general in the ECMWF
302 IFS meteorological fields, CAMx tends to under-estimate the night-time minima and
303 to over-estimate some daytime peaks, over-predicting the dilution of day time NO_2
304 concentrations, whereas the wind speed was relatively well reproduced (Bessagnet et
305 al., 2016).

306 O_3 concentrations were found to be over-predicted for all the four periods with a mean
307 fractional bias ranging from 2% to 48%. Especially in June 2006, when the photochemical
308 activity is higher, the general under-prediction of NO_x in the whole domain reduces the O_3
309 titration potential during night time.

310 Model performance for O₃ is also strongly influenced by long-range transport especially
311 during the winter periods when the local chemical production of O₃ is limited. Figure S3
312 shows the model performance at the Mace Head station located on the west coast of Ireland
313 for all the four periods. Especially in January-February 2007 O₃ concentrations were found to
314 be over-predicted by about 10 to 20 ppb indicating that boundary conditions for O₃ were
315 probably not well represented. In June 2006 and September-October 2008 O₃ was relatively
316 well captured at Mace Head suggesting that the observed positive bias in O₃ concentrations
317 might arise from insufficient NO_x emissions to undergo titration during night time as well as
318 not correctly represented planetary boundary layer dynamics. In February-March 2009 the
319 model tends to under-predict the O₃ concentration at Mace Head and overall the O₃ model
320 performance shows the lowest bias (2%). Eventually, the under-prediction of O₃ in the
321 boundary condition may counteract the already mentioned deficiencies related to insufficient
322 NO_x emissions.

323 **SO₂ and CO**

324 SO₂ concentrations were found to be slightly over-predicted for all the four periods with a
325 mean fractional bias ranging from 14% to 36% for SO₂. The daily variations of modelled and
326 measured SO₂ concentrations for February-March 2009 are reported as well in Figure S1
327 (lower-panel) for the stations in Table 2. In general, the daily variations of modelled and
328 measured SO₂ concentrations agree relatively well with each other throughout the period.

329 Most of the SO₂ emissions arise from high stack point sources which have injection heights of
330 a few hundred meters. It might be that the vertical distribution of SO₂ might affect the model
331 performance in particular near the harbors and coastal areas where ship emissions were
332 allocated in the second layer of the model domain (extending from ~20 to 50 m above ground
333 level) whereas they can reach up to 58 meters in deep draft vessels (SCG, 2004) and also
334 undergo plume rise. Insufficient conversion to sulfate or too low deposition processes might
335 also positively bias the model performance for SO₂.

336 CO was slightly under-predicted for all periods (mean fractional bias between -11% and -
337 31%), with highest values during the September-October 2008 period (-31%). The late
338 summer-fall period is known to be influenced by agricultural open field burning activities
339 which might be missing from standard emission inventories.

340 In general, for both SO₂ and CO, the model showed lower correlation coefficients with
341 respect to other gas-phase species (*r* values from 0.20 and 0.37 for CO and from 0.37 to 0.52
342 for SO₂).

343 **PM_{2.5}**

344 Of all investigated variables, CAMx shows the best statistical performance for PM_{2.5}. For all
345 four periods the acceptable model performance criteria recommended by Boylan and Russell
346 (2006) for aerosols were met (MFE ≤ +75 % and -60 % < MFB < +60 %). The fractional bias
347 ranges from less than 1% in September-October 2008 up to -13% in February-March 2009.
348 Also the recommended model performance goals (MFE ≤ +50% and -30% < MFB < +30%)
349 were met for all periods except for January 2007. Modelled average PM_{2.5} concentrations are
350 shown in Fig. 1. A different spatial distribution is seen for summer and winter. In June 2006
351 the model predicts higher concentrations in the southern part of the domain especially over
352 the Mediterranean Sea and North Africa (up to 35 µg m⁻³). On the other hand, the highest
353 concentrations were predicted in the Po valley area (above 40 µg m⁻³) and in the southern part
354 of Poland during January-February 2007. During the two colder periods (2007 and 2009)
355 elevated concentrations of around 15 µg m⁻³ are also visible close to urban areas such as Paris
356 and Moscow. Figure 2 shows PM_{2.5} variations at Airbase rural-background sites in terms of
357 medians, 25th and 75th percentiles. In all the four periods CAMx is able to reproduce the
358 observed monthly variation very well with some over-prediction occurring mainly from the
359 14th to the 17th of January 2007 and towards the end of 2008 period.

360 **3.2 Detailed evaluation of PM_{2.5} components in February-March 2009**

361 The modelled concentrations of non-refractory PM_{2.5} components were compared against
362 aerosol mass spectrometer measurements at eleven European sites for the February-March
363 2009 period (Crippa et al., 2014). Even though the AMS measures particles with a diameter *D*
364 < 1 µm, the difference between the non-refractory PM₁ and total PM_{2.5} mass is in general
365 rather small as shown in Aksoyoglu et al. (2011), at least for situations without exceedingly
366 high air pollution and situations when sea salt makes large relative contribution to PM_{2.5}. The
367 modelled average total non-refractory PM_{2.5} (sum of nitrate, sulfate, ammonium and OA)
368 concentrations match the measurements quite well with a few exceptions (Fig. 3 and Table 3).
369 The model is able to reproduce both high concentrations observed at the urban site Barcelona
370 and low ones at remote sites like Hyytiälä, Finland. Concentrations of inorganic aerosols are

371 over-predicted and OA are under-predicted at most of the stations (with similar behavior
372 during the other investigated periods, Figure S4 and Figure S5). Very similar results were also
373 presented by other recent studies (Knote et al., 2011). The effect of different schemes to treat
374 OA is discussed in Sect. 3.3. At the Cabauw site nitrate was the most dominant species
375 (Mensah et al., 2012). Especially at this site the model strongly over-predicts in particular the
376 nitrate (NO_3^-) fraction (by a factor of 3). A sensitivity test with 50% reduction in ammonia
377 emissions significantly improved the modelled NO_3^- concentrations at almost all sites (Table
378 S2) suggesting potential uncertainties in NH_3 emissions and their seasonal variability. Other
379 potential reasons for the over-prediction of NO_3^- could be related to uncertainties in removal
380 process of HNO_3 as well as dry deposition velocity of NH_3 . Substantial over-predictions were
381 found at the higher altitude site of Montseny and Puy de Dôme when compared with first
382 model layer concentrations (ca. 200 and 800 meters a.s.l. respectively at these sites). These
383 sites located at about 720 and 1465 meters a.s.l., are sometimes not within the PBLH during
384 winter periods. At the Montseny site, the relatively coarse resolution of the model could also
385 influence model performance since the site is located in a complex area about 50 km north-
386 east of Barcelona (Pandolfi et al., 2014). Sulfate concentrations (SO_4^{2-}) were over-predicted at
387 almost all sites and especially at Mace Head suggesting that long-range transport of SO_4^{2-}
388 might be positively biased.

389 Modelled and observed hourly concentrations of NO_3^- , SO_4^{2-} , ammonium (NH_4^+) and OA at
390 Payerne are reported in Fig. 4 for March 2009 together with meteorological parameters in Fig.
391 S6. The model was able to reproduce the meteorological parameters very well for most of the
392 time. The temperature was slightly under-predicted at both night and day-times (with a
393 maximum of -2°C) whereas both the monthly variation and the absolute values of wind speed
394 and specific humidity were reproduced well with a few under-predictions of high wind-speed
395 (6th and 11th of March and towards the end of the simulation). The model was able to capture
396 the three NO_3^- and NH_4^+ peaks observed around the 7th, 18th and 23rd of March with a general
397 slight over-prediction throughout the whole period. Indeed, the under-prediction in
398 temperature during day and night time could partially explain the over-prediction of the NO_3^-
399 fraction with more NO_3^- partitioning to the aerosol phase which also apply to the other
400 stations used in this study. An evaluation of modelled temperature at the European scale for
401 the February-March 2009 period confirmed that the model systematically under-predicted the
402 2 meter surface temperature (Bessagnet et al., 2014). All the inorganic components were over-
403 predicted during the first four days of March 2009 with a peak around the 3rd of March,

404 indicating that the PBLH was probably not correctly reproduced by the model during this
405 period. Although the temporal variation was captured, concentrations of OA were under-
406 predicted throughout all the simulation ($4.1 \mu\text{g m}^{-3}$ and $1.8 \mu\text{g m}^{-3}$ observed and modelled
407 average concentrations). Analysis of the OA fraction is discussed in the next section.

408 **3.3 Organic aerosols**

409 **3.3.1 Sensitivity of OA to the VBS scheme**

410 In this section, effects of different parameterizations of the organic aerosol module on the
411 modelled OA concentrations are discussed. The scatter plots in Fig. 5 show a comparison of
412 daily average OA concentrations against the same AMS measurements as in Table 3 during
413 February-March 2009. Statistics for each scenario are reported in Table 4. When the semi-
414 volatile dynamics of primary organic aerosol is not taken into account (scenario NOVBS), the
415 model under-predicts OA concentrations (MFB: -66%) with an observed and modelled
416 average concentrations of $3.0 \mu\text{g m}^{-3}$ and $1.2 \mu\text{g m}^{-3}$ respectively. In the VBS_ROB scenario
417 POA emissions are allowed to evaporate following the volatility distribution proposed by
418 Robinson et al. (2007) and to undergo chemical oxidation. In this case modelled OA
419 concentrations decrease by about 42% with respect to NOVBS, predicting an average OA
420 concentration of $0.7 \mu\text{g m}^{-3}$. On the other hand, the VBS_BC scenario improves the OA
421 model performance increasing the OA concentrations by about 42% with respect to NOVBS.
422 Predicted OA concentrations are found to be $1.7 \mu\text{g m}^{-3}$ on average (MFB: -47%). Similar
423 behavior during winter periods was also shown in recent studies where the same VBS scheme
424 was applied in the U.S. domain (Koo et al., 2014). Figure 6 shows the modelled total OA
425 concentration over Europe using NOVBS, VBS_ROB and VBS_BC scenarios. The model
426 predicts high OA values in the Eastern part of the domain as well as over Portugal, France and
427 the Po Valley (VBS_BC). Some hot-spots around large urban areas are also visible, i.e., Paris
428 and Moscow. Higher OA concentrations in the southern part of the domain are observed in
429 the VBS_BC case, likely because of higher temperature and more OH radicals available in
430 that part of the domain leading to an increase in the total organic mass upon reaction with
431 organic vapours. This is in line with the results of Fountoukis et al. (2014) for the February-
432 March 2009 period even though their study predicts lower concentration over the Po valley.
433 Even though model input data and parameterizations are not the same, the VBS_BC case in
434 particularly, uses a very similar volatility distribution as in Fountoukis et al. (2014). Our study

435 predicts relatively lower OA concentrations (MFB: -0.47, MFE: 0.79) compared to those
436 reported by Fountoukis et al. (2014) (MFB: 0.02, MFE: 0.68) for February-March 2009.
437 Unlike Fountoukis et al. (2014) our study does not include fire emissions and marine organic
438 aerosol which may partially explain the differences. Figure 7 shows hourly modelled and
439 observed OA concentration at Payerne for March 2009 and June 2006. In March 2009
440 VBS_ROB results are lower than those in NOVBS whereas OA concentrations in VBS_BC
441 case are higher (see Supplementary Fig. S8 and Table S3 for average concentrations and
442 statistics). In June 2006, the OA mass in VBS_ROB is lower than those in NOVBS while
443 VBS_BC predicts similar concentrations as the NOVBS scenario ($2.4 \mu\text{g m}^{-3}$ and $2.6 \mu\text{g m}^{-3}$,
444 respectively, Fig. S9 and Table S4). It has to be noted that the NOVBS scenario predicts
445 slightly lower OA concentration for June 2006 in Payerne with respect to our previous
446 application (Aksoyoglu et al., 2011), mainly because of a different biogenic model being used
447 which yields lower monoterpene and sesquiterpene emissions. Since both BVOCs and
448 BBOA-like emissions are highly uncertain, sensitivity tests with increased biogenic and
449 anthropogenic emissions were performed and results discussed in the next section (3.3.2).

450

451 **3.3.2 Sensitivity of OA to BBOA-like and BVOC emissions**

452 Emissions of BVOCs compounds (i.e. monoterpenes, isoprene and sesquiterpenes) were
453 doubled in scenario VBS_BC_2xBVOC, whilst primary organic aerosol emissions from
454 SNAP2 and SNAP10 (BBOA-like) were doubled in scenarios VBS_BC_2xBBOA, with other
455 emissions and processes represented as in VBS_BC. Figure 8 shows modelled and observed
456 OA daily average concentrations for the VBS_BC, VBS_BC_2xBVOC and
457 VBS_BC_2xBBOA scenarios across the sites. Statistics for each scenario are reported in
458 Table 5. Increasing biogenic emissions by a factor of two during February-March 2009
459 resulted in almost no change in the predicted total OA ($1.7 \mu\text{g m}^{-3}$ and $1.8 \mu\text{g m}^{-3}$ for the
460 VBS_BC and VBS_BC_2xBVOC scenarios, respectively). On the other hand, doubling the
461 BBOA-like emissions (VBS_BC_2xBBOA) during the same period strongly increased the
462 predicted OA mass (up to $2.8 \mu\text{g m}^{-3}$ on average). As a result the mean fractional bias
463 decreased further, from -47% to -12% averaged across the sites. This could eventually
464 confirm other studies where substantial under-predictions in residential wood burning
465 emissions were underlined (e.g., Bergström et al., 2012). A few points above the 2:1 lines in
466 VBS_BC_2xBBOA mainly belong to the sites of Montseny, Puy de Dôme and Helsinki.

467 During winter periods, it is likely that elevated stations such Montseny and Puy de Dôme are
468 most of the time above the PBLH, as suggested by previous studies for Puy de Dôme (Freney
469 et al., 2011), whereas model concentrations are extracted from the first layer of the model. In
470 Helsinki, BBOA emissions seem to be overestimated or the dispersion underestimated in the
471 model.

472 Comparison with a warmer period in June 2006 is reported as well for Payerne where AMS
473 measurements were also available (Fig.9). In February-March 2009 increasing BBOA-like
474 emissions (VBS_BC_2xBBOA) reduced the fractional bias from -85% in VBS_BC to -37%
475 (Table S3) with an over-prediction occurring during 1-5 of March (Fig. 9, upper panel). As
476 already discussed in Section 3.2, it is likely that the vertical mixing processes were not
477 correctly represented by the model since also the inorganic components were over-predicted
478 for the same period. Almost no change in the predicted OA mass was found when biogenic
479 emissions were doubled (scenario VBS_BC_2xBVOC) (Fig. 9, upper panel) due to lower
480 BVOCs emission during winter periods. Increasing BVOCs emissions in June 2006 increased
481 the predicted OA mass at Payerne site especially during the 12-16 June and towards the end
482 of the simulation period, where higher concentrations and temperature (Fig. S7) were also
483 observed (Fig. 9, lower panel). In contrast, similar OA concentrations were predicted in
484 Payerne for VBS_BC and VBS_BC_2xBBOA during June 2006 (with averages of $2.4 \mu\text{g m}^{-3}$
485 and $2.8 \mu\text{g m}^{-3}$ respectively). This is in line with a very recent source apportionment study
486 based on ACSM (aerosol chemical speciation monitor) measurements performed in Zürich for
487 13 months (February 2011 - February 2012) which revealed substantial differences between
488 the winter (February-March) and summer (June-August) f_{44} / f_{43} space (organic mass fraction
489 measured at mass to charge ratio 44 and 43) indicating that summer OOA (oxygenated
490 organic aerosol) is strongly influenced by biogenic emission and winter OOA by biomass
491 burning emission (Canonaco et al., 2015). Increased OA concentrations at Payerne in June
492 2006 with increased biogenic emissions were also found in other modelling studies.
493 Bergström et al. (2012) used the VBS framework with different assumptions regarding aging
494 processes and compared the model results for June 2006 with the AMS results at Payerne. In
495 their study the total OA was found to be under-predicted with lower bias observed when
496 aging processes were taken into account and biogenic emissions were increased by a factor of
497 3. Even though their model differs from ours in various aspects (number of volatility bins,
498 aging processes parameterization and input data) in two of their scenario without aging of
499 biogenic SOA Bergström et al. (2012) predicted an average OA concentration ranging from

500 2.6 $\mu\text{g m}^{-3}$ to 3.4 $\mu\text{g m}^{-3}$ which is similar to our base case VBS_BC and VBS_BC_2xBVOC
501 scenario (2.4 $\mu\text{g m}^{-3}$ and 3.4 $\mu\text{g m}^{-3}$, respectively, Table S4).

502 **3.3.3 OA components in summer and winter**

503 Comparisons of the primary and secondary organic fraction at the rural site of Payerne during
504 summer (June 2006) and winter (February-March 2009) periods are reported in Figure 10.
505 During the winter period the VBS scheme better reproduced the primary and secondary
506 organic aerosol components compared to the NOVBS case. In particular, For the VBS_ROB
507 base case, total OA concentrations were lower compared to the NOVBS case, consistent with
508 the study of Woody et al. (2016) where the same VBS scheme was applied to the US domain.
509 The total OA concentrations in the base case (VBS_BC) and in the scenario with increased
510 biomass burning emissions (VBS_BC_2xBBOA) were higher compared to NOVBS case,
511 even though SOA and POA fractions were not correctly reproduced. Higher contribution from
512 the primary fraction during winter periods was also predicted by the study of Koo et al., 2014
513 which deployed the same VBS scheme. Eventually, this might indicate that biomass burning
514 precursors might be missing in this study, or that the oxidation pathways of primary organic
515 material need to be improved in the model (up to 86% of the reacted primary organic material
516 is still allocated in the primary set as oxidation proceeds, directly increasing the POA
517 fraction).

518 Different behavior was observed for the summer period where the larger contribution of SOA
519 to the total OA retrieved from measurements is also reproduced by the model, even though
520 the total OA concentration was still underestimated. These results for summer are also in line
521 with the study of Koo et al. (2014) for summer periods in the US domain using the same VBS
522 scheme.

523

524 **4 Conclusions**

525 A modelling study using the regional air quality model CAMx with VBS (Volatility Basis
526 Set) scheme was performed for the first time in Europe within the EURODELTA-III model
527 intercomparison exercise. An evaluation for the main gas phase species and $\text{PM}_{2.5}$ for four
528 different periods was performed using the European air quality database Airbase as well as
529 AMS (Aerosol Mass Spectrometer) measurements. The period in February-March 2009 was
530 further analyzed in more detail using different assumptions regarding the volatility of emitted

531 organic aerosol and emissions of precursor. The main findings of this study are summarized
532 below:

- 533 • Although total PM_{2.5} mass concentrations and its variations were well reproduced by
534 the model in all four periods, comparisons with AMS measurements for the February–
535 March 2009 period revealed that the good agreement between model and
536 measurements was most of the time due to overestimation of the inorganic fraction,
537 especially NO₃⁻, and underestimation of OA. Sensitivity tests with reduced NH₃
538 emissions generally reduced the positive bias in NO₃⁻ suggesting potential
539 uncertainties in NH₃ emissions and their seasonal variability.
- 540 • In general, for all the four periods, the model under-predicted NO₂ and CO
541 concentrations. On the other hand, O₃ was found to be over-predicted likely because
542 of insufficient NO_x to undergo titration during night-time chemistry or not well
543 captured vertical mixing processes and concentrations at the boundaries. SO₂ was
544 over-predicted presumably mainly because of uncertainties in high stack point sources
545 representation in the model or too low deposition processes.
- 546 • Including evaporation and oxidation processes of primary organic particles with the
547 volatility distribution proposed by Robinson et al. (2007) lowered the modelled OA
548 mass both in winter and summer periods. On the other hand, the adjustment of the
549 scheme by Robinson et al. (2007) suggested by Shrivastava et al. (2011) and Tsimpidi
550 et al. (2010) brings model and observations into better agreement by reducing the
551 negative bias for OA by about 29% (MFB) in winter.
- 552 • Sensitivity tests with increased BVOCs and BBOA-like emissions suggested that
553 emissions from residential heating represent an important contributor to total OA
554 during winter periods (February-March 2009). The model under-predicted the winter
555 OA concentrations (MFB -47% for base case VBS_BC) more than gas phase
556 pollutants e.g. NO₂ (Table 2). Eventually, increasing BBOA-like emissions by a factor
557 of 2 brought model and observation to a reasonably good agreement even though the
558 model still under-predicts the OA fraction (-12% MFB). This underlines the necessity
559 to better constrain emission inventories with a focus on residential heating. Also the
560 implementation of the VBS scheme for domestic wood burning, which substantially
561 influences both the primary and secondary organic aerosol, should be evaluated.

562 • A summer period was simulated as well and results were compared at Payerne. In June
563 2006, the current VBS implementation could not explain the discrepancy between
564 modelled and observed OA. During this period the difference between the model and
565 measurements is likely to be related to BVOCs emissions which are uncertain and
566 difficult to constrain with measurements. In this case the model was sensitive to an
567 increase in biogenic emissions especially during periods with higher temperature and
568 OA concentrations. The latter could confirm the importance of BVOC precursors in
569 summer in Payerne and the way to correctly represent their evolution in the
570 atmosphere.

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574

575 **Acknowledgements**

576 We thank the EURODELTA-III modelling community, especially INERIS for providing
577 various model input data. We are grateful to ENVIRON for providing the CAMx-VBS code
578 before its public release. Calculations of land use data were performed with the Swiss
579 National Supercomputing Centre (CSCS). We thank D. Oderbolz for developing the
580 CAMxRunner framework to ensure reproducibility and data quality among the simulations
581 and sensitivity tests. We thank M. Tinguely for the visualization software. We also thank G.
582 Pirovano for helping with the pre-processing of Airbase data. This study was financially
583 supported by the Swiss Federal Office of Environment (FOEN). JLJ was supported by NSF
584 AGS-1360834 and EPA STAR 83587701-0. We thank D.A. Day for analysis on the DAURE
585 dataset. Erik Swietlicki for the Vavihill dataset, A. Kiendler-Scharr for Cabauw AMS data,
586 Evelyn Freney for the Puy de Dôme dataset.

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895

896 **5 Figures and Tables**

897 Table 1. Volatility distributions used for different scenarios.

898

Scenarios	POA emission sources	Emission fraction for volatility bin with C* of				
		0	1	10	100	1000
NOVBS (non-volatile CAMxv5.40)	HOA-like	1.00	-	-	-	-
	BBOA-like	1.00	-	-	-	-
VBS_ROB (Robinson et al., 2007)	HOA-like	0.09	0.09	0.14	0.18	0.5
	BBOA-like	0.09	0.09	0.14	0.18	0.5
VBS_BC (Tsimpidi et al., 2010 and Shrivastava et al., 2011)	HOA-like	0.40	0.26	0.40	0.51	1.43
	BBOA-like	0.27	0.27	0.42	0.54	1.50

899

900 Table 2. Model gas phase and PM_{2.5} performance for the EDIII field campaigns (base case
 901 VBS_BC).

Species	Number of sites	Observed mean (ppb) ($\mu\text{g m}^{-3}$ for PM _{2.5})	Modelled mean (ppb) ($\mu\text{g m}^{-3}$ for PM _{2.5})	MB (ppb) ($\mu\text{g m}^{-3}$ for PM _{2.5})	ME (ppb) ($\mu\text{g m}^{-3}$ for PM _{2.5})	MFB [-]	MFE [-]	r
June 2006								
CO	36	192.0	158.0	-34.2	80.7	-0.12	0.36	0.20
NO ₂	320	4.1	2.3	-1.9	2.2	-0.54	0.68	0.55
O ₃	460	42.3	51.2	8.9	10.8	0.21	0.24	0.57
PM _{2.5}	48	12.0	11.7	-0.3	4.5	-0.07	0.39	0.55
SO ₂	263	1.0	1.2	0.2	0.7	0.14	0.67	0.52
Jan-Feb 2007								
CO	45	248.0	191.0	-57.8	107.0	-0.11	0.37	0.21
NO ₂	337	6.5	4.4	-2.2	3.2	-0.28	0.57	0.68
O ₃	455	23.5	35.8	12.3	12.6	0.48	0.49	0.61
PM _{2.5}	56	11.7	12.8	1.0	6.1	-0.04	0.56	0.69
SO ₂	271	1.3	1.7	0.4	1.1	0.36	0.75	0.46
Sep-Oct 2008								
CO	53	208.0	136.0	-72.0	91.4	-0.31	0.48	0.27
NO ₂	370	5.3	3.7	-1.7	2.5	-0.28	0.56	0.62
O ₃	465	24.3	32.5	8.2	9.6	0.32	0.37	0.50
PM _{2.5}	90	13.0	14.1	1.0	5.7	<0.01	0.46	0.76
SO ₂	256	0.9	1.1	0.2	0.8	0.25	0.74	0.37
Feb-Mar 2009								
CO	57	262.0	170.0	-91.6	119.0	-0.26	0.48	0.37
NO ₂	380	6.0	3.9	-2.0	2.8	-0.33	0.56	0.61
O ₃	488	32.7	33.0	0.2	7.1	0.02	0.23	0.55
PM _{2.5}	110	15.1	13.0	-2.1	6.4	-0.13	0.50	0.71
SO ₂	257	1.0	1.3	0.3	0.9	0.23	0.76	0.45

902

903

904

905 Table 3. Statistical analysis of nitrate, ammonium, sulfate and organic aerosol in base case
 906 (VBS_BC) for February-March 2009 at different AMS sites.

907

Site	Mean observed ($\mu\text{g}/\text{m}^3$)	Mean modelled ($\mu\text{g}/\text{m}^3$)	MB $\mu\text{g m}^{-3}$	ME $\mu\text{g m}^{-3}$	MFB [-]	MFE [-]
NO_3^-						
Barcelona	3.6	5.8	2.19	3.98	0.35	0.98
Cabauw	2.2	6.7	4.49	4.58	0.85	1.01
Chilbolton	2.7	4.0	1.33	2.21	0.02	0.76
Helsinki	1.0	1.9	0.93	1.30	0.29	0.92
Hyytiälä	0.2	1.0	0.75	0.83	0.21	1.09
Mace Head	0.6	1.7	1.11	1.12	0.14	0.70
Melpitz	3.1	4.3	1.25	2.41	0.35	0.71
Montseny	3.1	5.9	2.83	4.31	0.38	1.00
Payerne	3.9	5.7	1.81	2.83	0.34	0.61
Puy de Dôme	0.9	2.7	1.81	2.17	1.13	1.30
Vavihill	2.8	3.7	0.89	2.17	0.14	0.78
NH_4^+						
Barcelona	1.6	2.5	0.92	1.41	0.42	0.71
Cabauw	1.0	2.7	1.73	1.75	0.95	0.97
Chilbolton	1.3	2.0	0.68	1.02	0.39	0.61
Helsinki	0.8	1.3	0.52	0.59	0.51	0.60
Hyytiälä	0.4	0.8	0.43	0.48	0.55	0.70
Melpitz	1.4	2.1	0.72	1.11	0.45	0.69
Montseny	1.7	2.6	0.92	1.58	0.39	0.74
Payerne	1.7	2.5	0.80	1.15	0.36	0.56
Puy de Dôme	0.7	1.2	0.51	0.87	0.83	1.07
Vavihill	1.6	1.9	0.38	0.90	0.17	0.56
SO_4^{2-}						
Barcelona	2.7	2.3	-0.44	1.25	-0.19	0.48
Cabauw	1.0	2.1	1.13	1.34	0.73	0.85
Chilbolton	1.3	2.2	0.91	1.33	0.45	0.70
Helsinki	2.4	2.2	-0.24	0.92	-0.04	0.43
Hyytiälä	1.4	1.7	0.26	0.73	0.09	0.58
Mace Head	0.4	1.2	0.83	0.89	1.04	1.12
Melpitz	1.1	2.2	1.15	1.40	0.54	0.76
Montseny	1.4	2.3	0.97	1.19	0.55	0.64

Payerne	1.1	2.1	1.06	1.16	0.62	0.70
Puy de Dôme	0.4	1.1	0.77	0.82	1.14	1.19
Vavihill	1.6	2.3	0.73	1.05	0.18	0.54

OA

Barcelona	8.2	3.1	-5.11	5.15	-0.80	0.82
Cabauw	1.2	1.1	-0.14	0.53	-0.13	0.50
Chilbolton	2.4	0.7	-1.70	1.70	-1.09	1.10
Helsinki	2.7	2.9	0.26	1.64	0.08	0.62
Hyytiälä	1.3	1.0	-0.28	0.52	-0.48	0.60
Mace Head	0.8	0.4	-0.38	0.43	-0.29	0.70
Melpitz	1.5	0.5	-0.95	0.98	-0.94	0.97
Montseny	3.1	3.9	0.88	1.88	0.31	0.57
Payerne	4.1	1.8	-2.33	2.43	-0.85	0.90
Puy de Dôme	0.6	1.4	0.78	0.96	0.68	0.91
Vavihill	3.9	1.4	-2.53	2.53	-1.04	1.04

908

909 Table 4. Statistical analysis of OA for NOVBS, VBS_ROB and VBS_BC scenarios for the 11
910 AMS sites for February-March 2009.

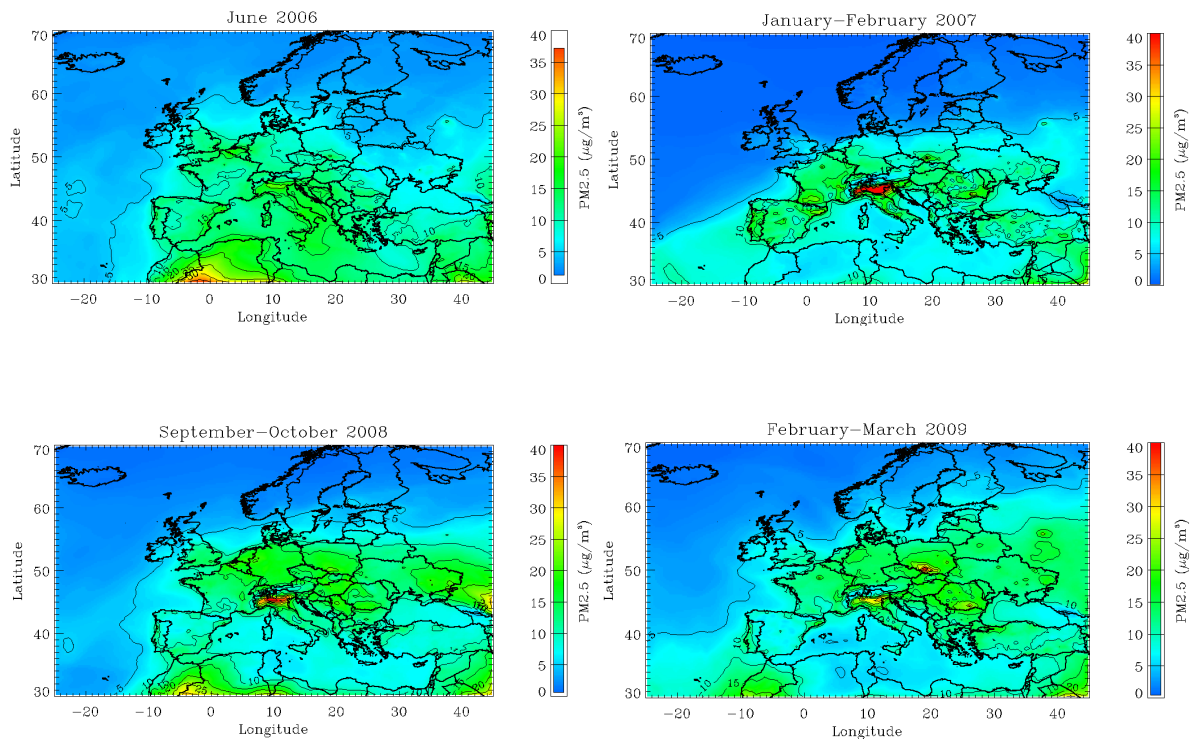
Scenario	Mean observed OA ($\mu\text{g m}^{-3}$)	Mean modelled OA ($\mu\text{g m}^{-3}$)	MB ($\mu\text{g m}^{-3}$)	ME ($\mu\text{g m}^{-3}$)	MFB [-]	MFE [-]
NOVBS	3.0	1.2	-1.8	2.0	-0.66	0.88
VBS_ROB	3.0	0.7	-2.3	2.4	-1.08	1.19
VBS_BC (base case)	3.0	1.7	-1.2	1.8	-0.47	0.79

911

912 Table 5. Statistical analysis of OA for VBS_BC, VBS_BC_2xBVOC and VBS_BC_2xBBOA
913 scenarios for the 11 AMS sites for February-March 2009.

Scenario	Mean observed OA ($\mu\text{g m}^{-3}$)	Mean modelled OA ($\mu\text{g m}^{-3}$)	MB ($\mu\text{g m}^{-3}$)	ME ($\mu\text{g m}^{-3}$)	MFB [-]	MFE [-]
VBS_BC (base case)	3.0	1.7	-1.2	1.8	-0.47	0.79
VBS_BC_2xBVOC	3.0	1.8	-1.2	1.8	-0.46	0.78
VBS_BC_2xBBOA	3.0	2.8	-0.1	1.9	-0.12	0.69

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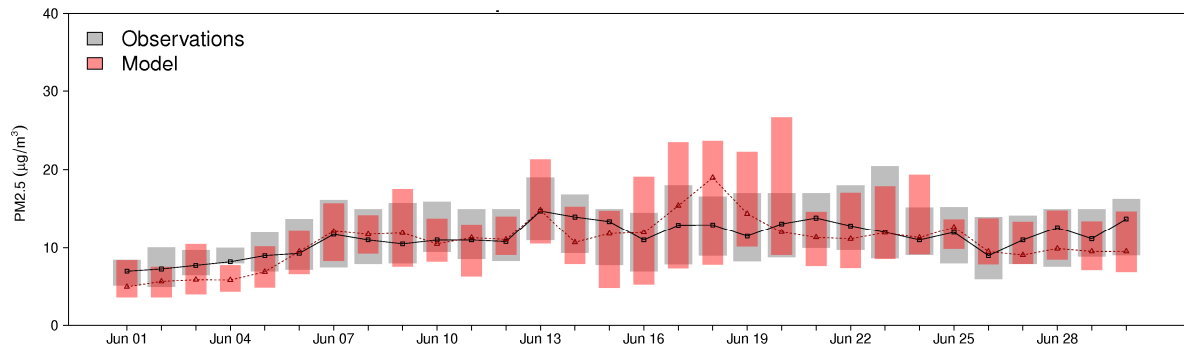
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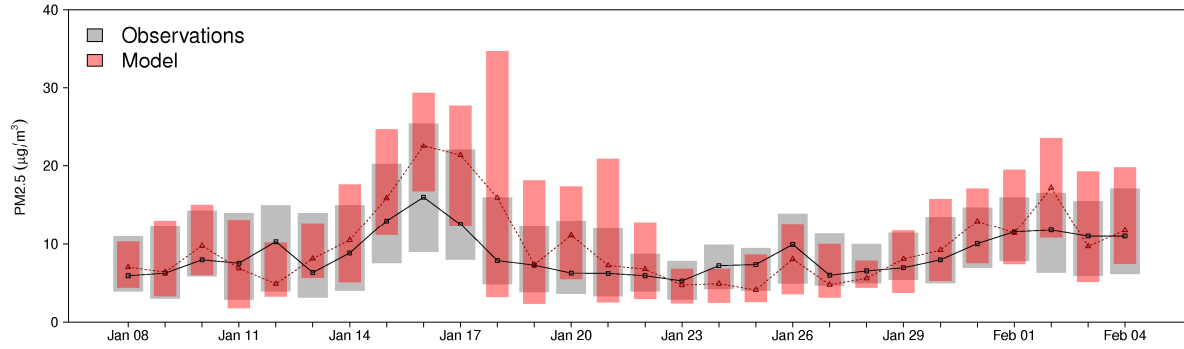
918

919 Figure 1. Modelled average PM_{2.5} concentrations for June 2006, January-February 2007,
 920 September-October 2008 and February-March 2009 (top to bottom) based on the base case
 921 (VBS_BC). Note that the color scale was limited to maximum of 40 $\mu\text{g}/\text{m}^3$ to facilitate
 922 comparison of the panels.

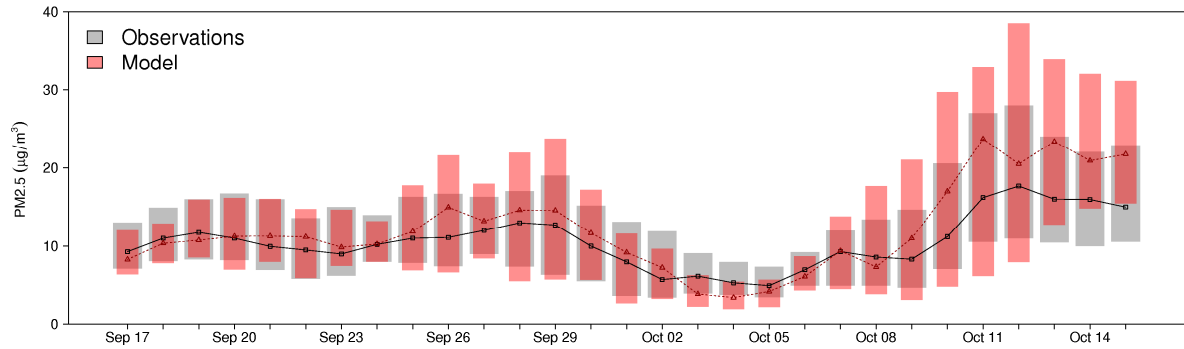
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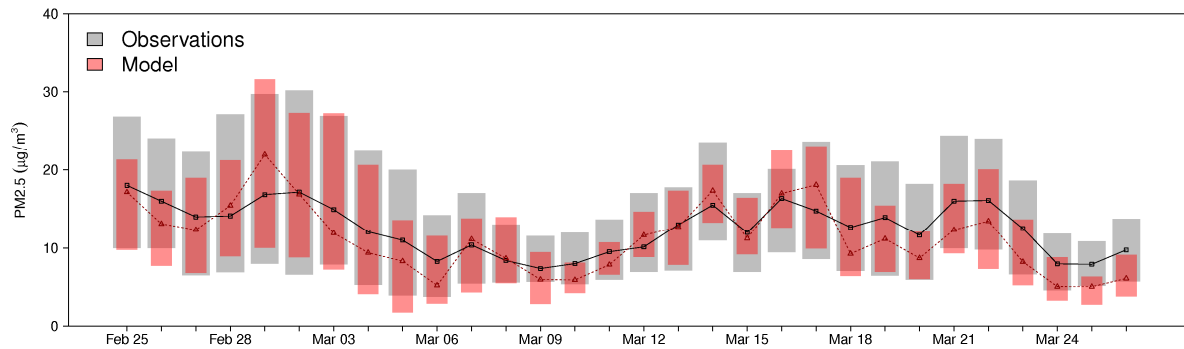
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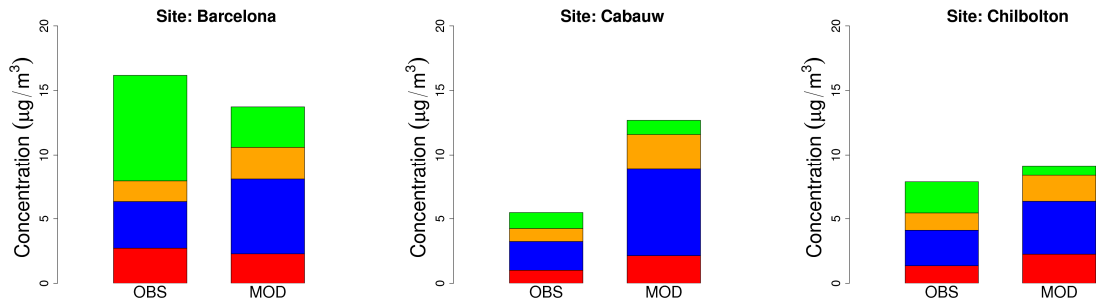


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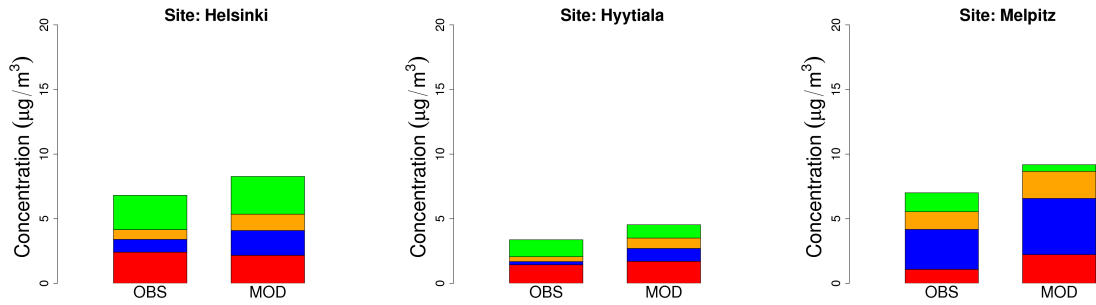
928 Figure 2. Comparison of modelled (red) and measured (grey) PM_{2.5} concentrations at AirBase
 929 rural background sites. The extent of the bars indicates the 25th and 75th percentile. The black

930 and red lines are observed and modelled median, respectively. The numbers of sites are 48,
931 56, 90, and 110 from top to down. Based on base case (VBS_BC).
932

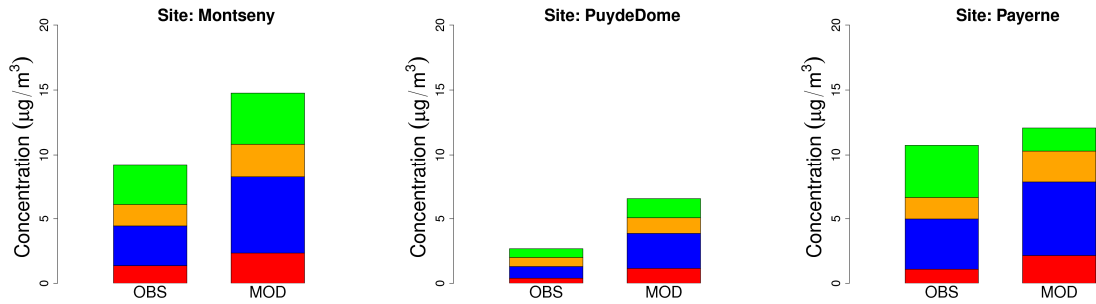
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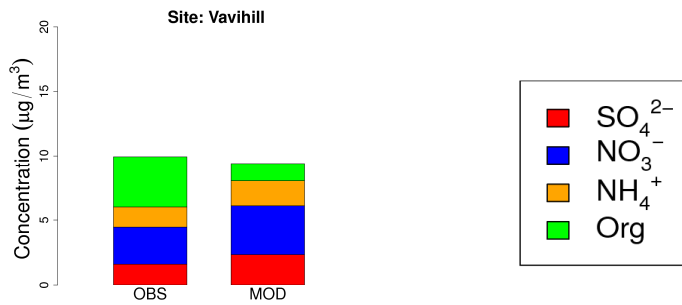
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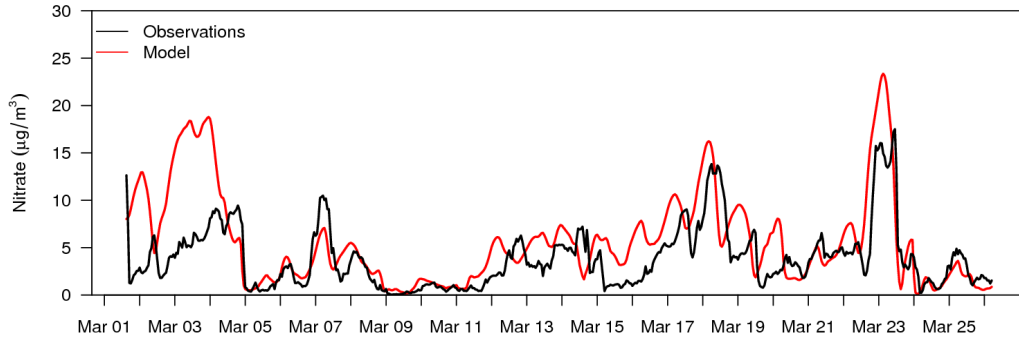


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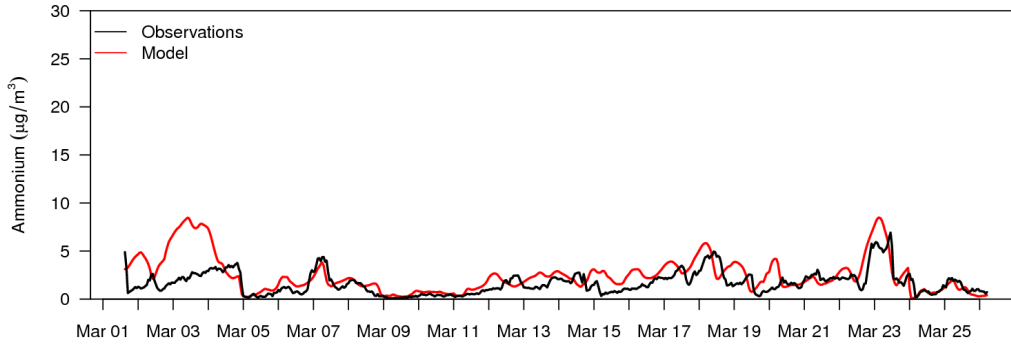


937 Figure 3. Comparison of observed (OBS) non-refractory PM₁ and modelled (MOD) non-
938 refractory PM_{2.5} at 10 AMS sites in Europe during February-March 2009. Mace head is
939 reported only in Table 3 since the ammonium component is not available.

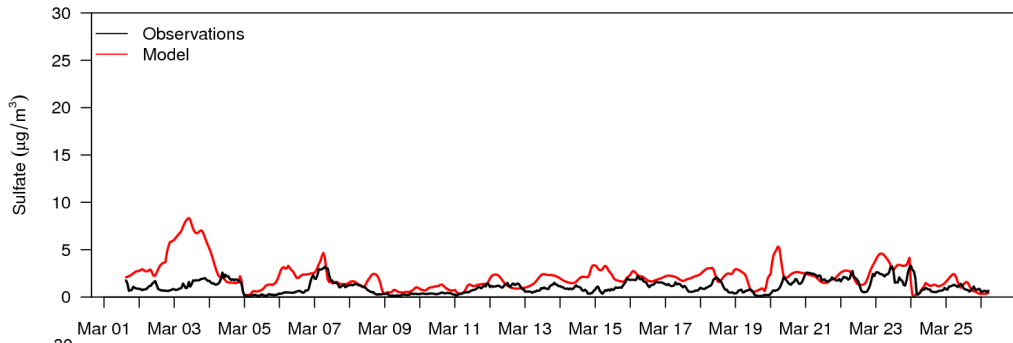
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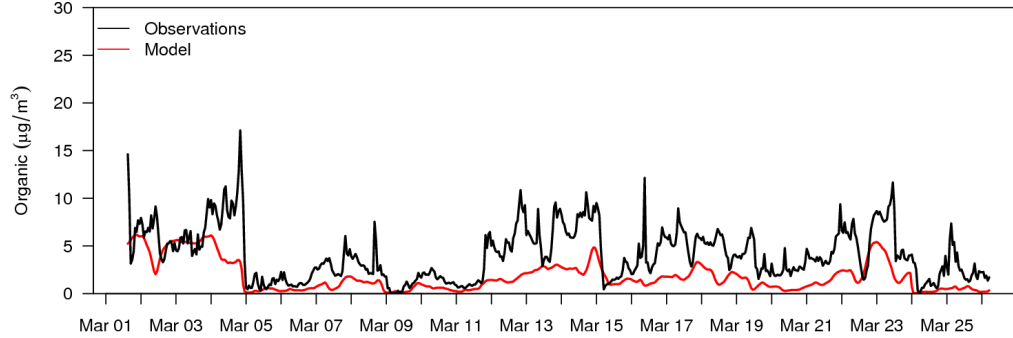
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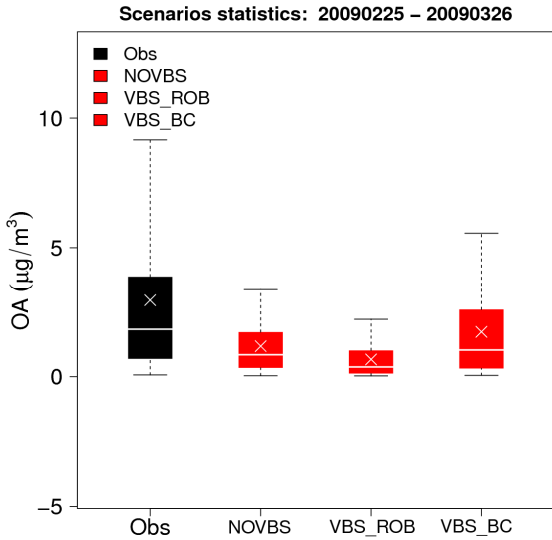
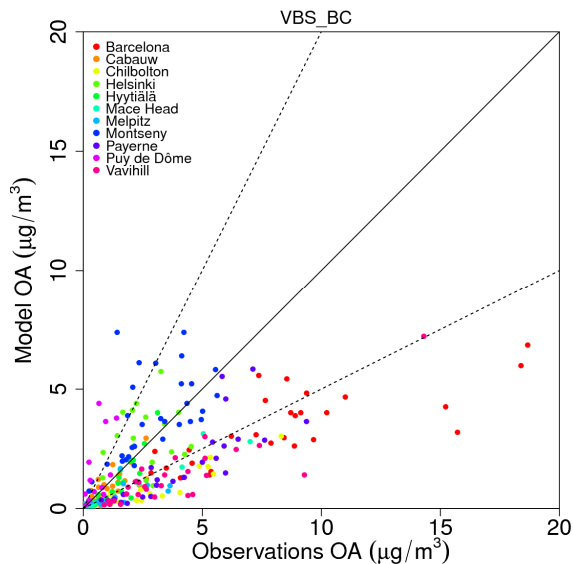
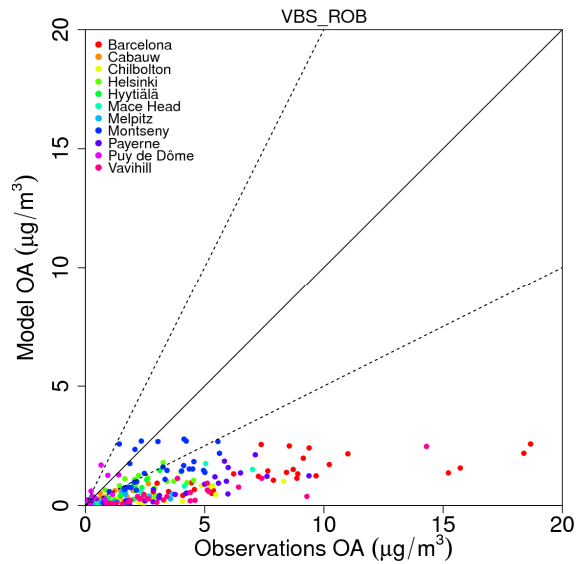
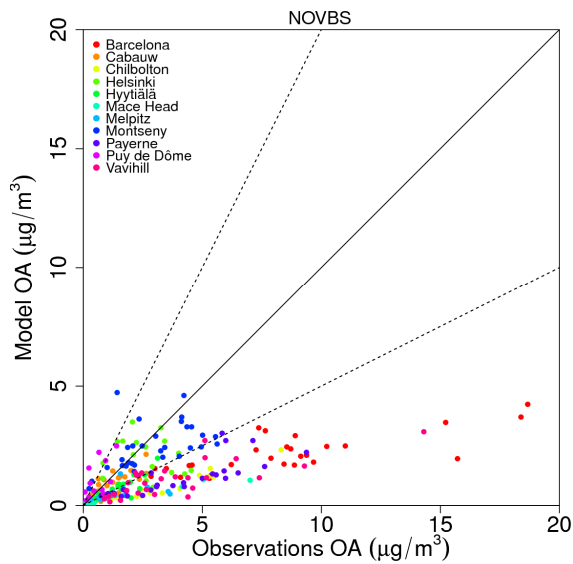
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945 Figure 4. Comparison of observed and modelled nitrate, ammonium, sulfate and organic
 946 aerosol at Payerne for March 2009.

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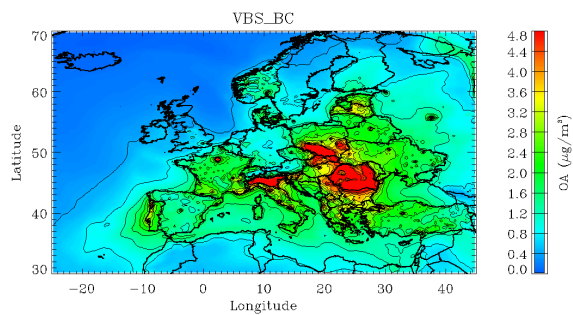
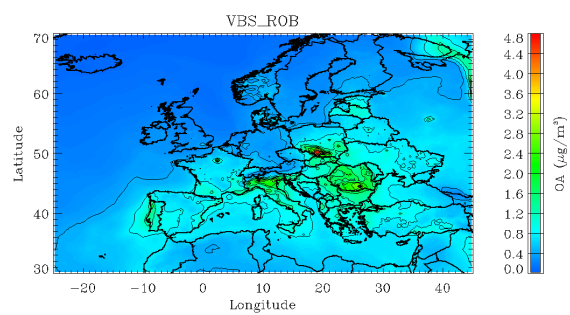
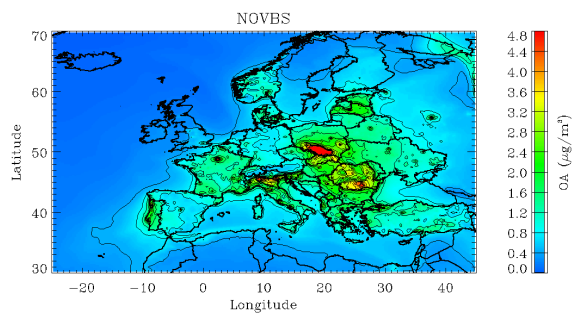


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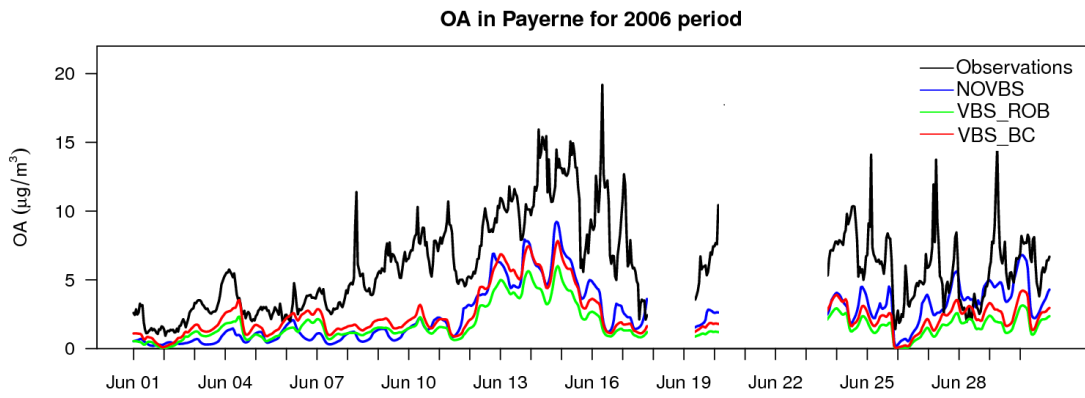
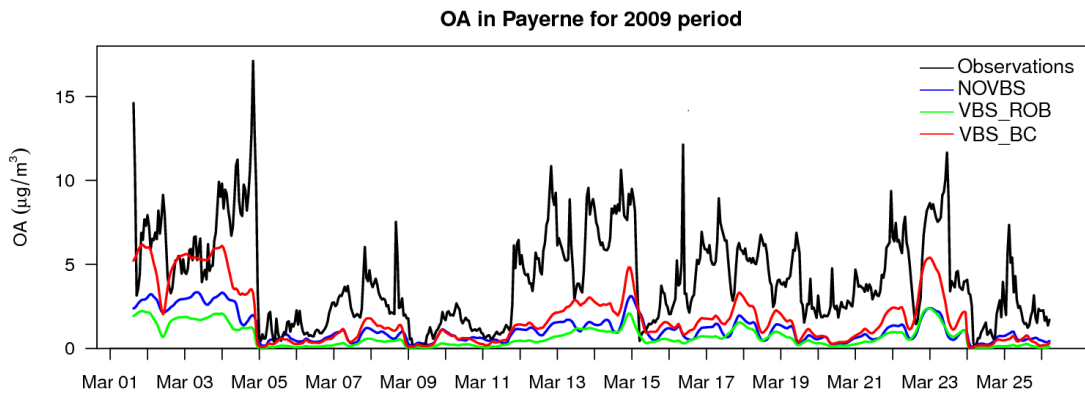
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950 Figure 5. OA daily average scatter plots for NOVBS, VBS_ROB and VBS_BC scenarios for
 951 February-March 2009 for stations in Table 3. Solid lines indicate the 1:1 line. Dotted lines are
 952 the 1:2 and 2:1 lines. Boxplots indicate medians, 5th, 25th, 75th and 95th quantiles for
 953 observations (black) and sensitivity tests (red). The crosses represent the arithmetic means. R^2
 954 is 0.55 for NOVBS, 0.64 for VBS_ROB and 0.59 for VBS_BC (excluding the elevated sited
 955 of Puy de Dôme and Montseny).

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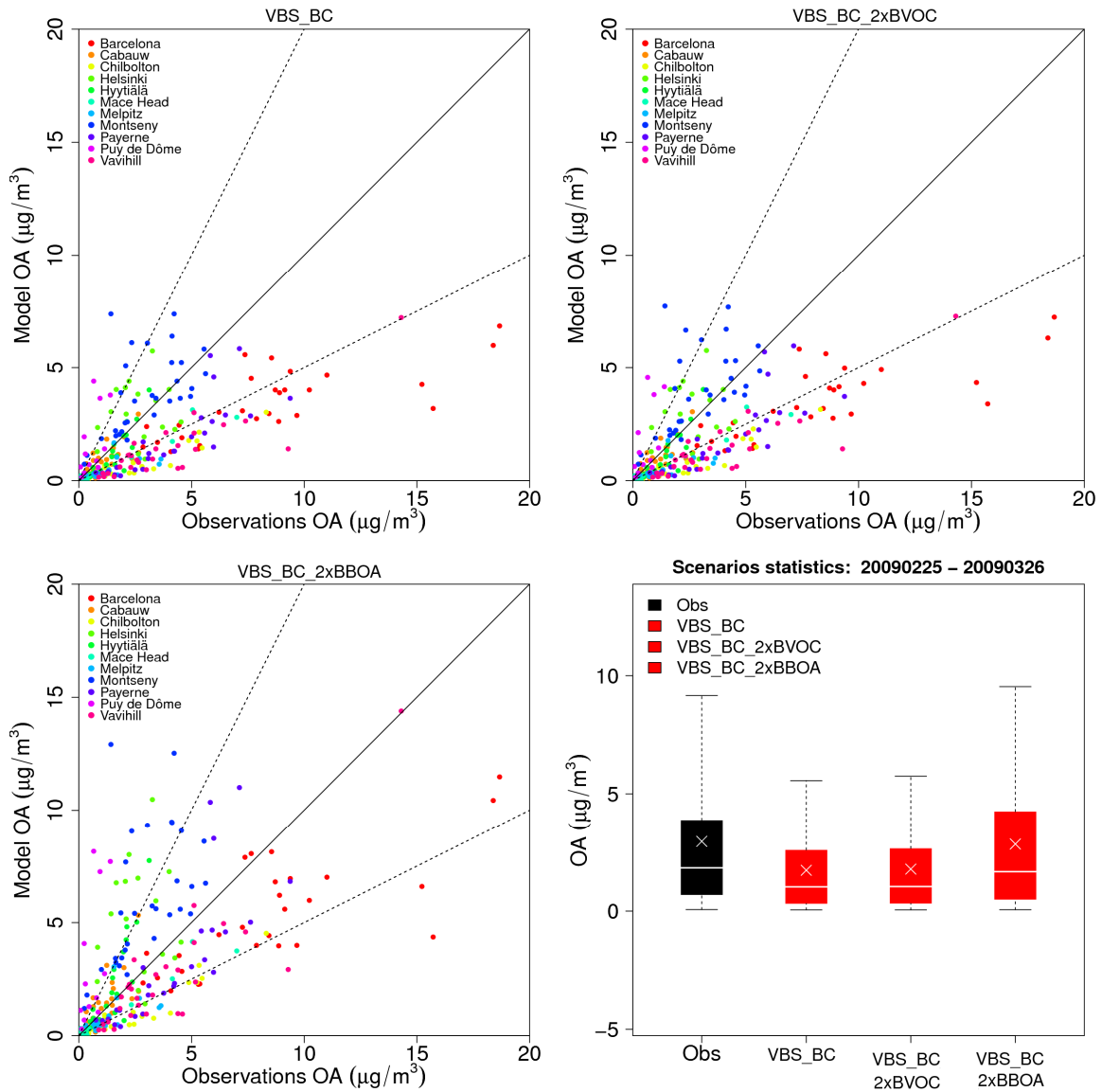


959 Figure 6. Predicted OA concentrations over Europe for the NOVBS, VBS_ROB and
 960 VBS_BC scenario in February-March 2009. Note that the color scale was limited to
 961 maximum of $4.8 \mu\text{g}/\text{m}^3$ to facilitate comparison of the panels.
 962



965 Figure 7. Predicted and observed total OA for scenarios NOVBS, VBS_ROB and VBS_BC in
 966 March 2009 (upper panel) and June 2006 (lower panel) at Payerne.

967

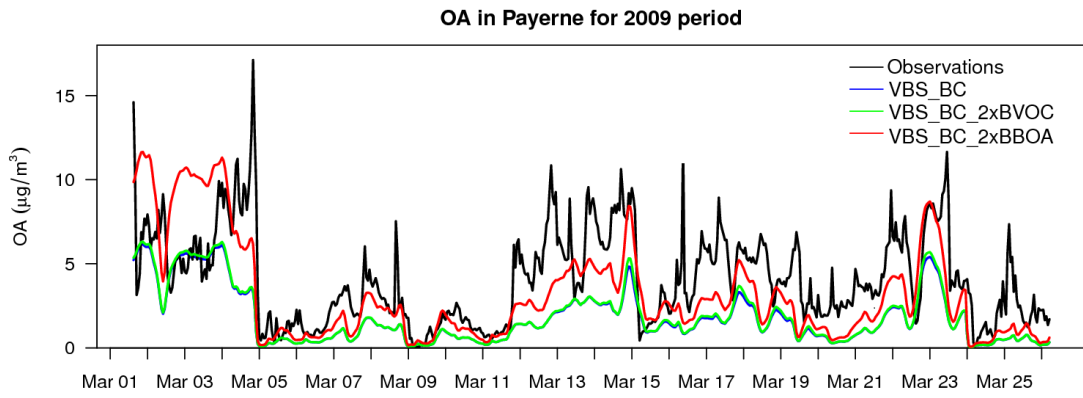


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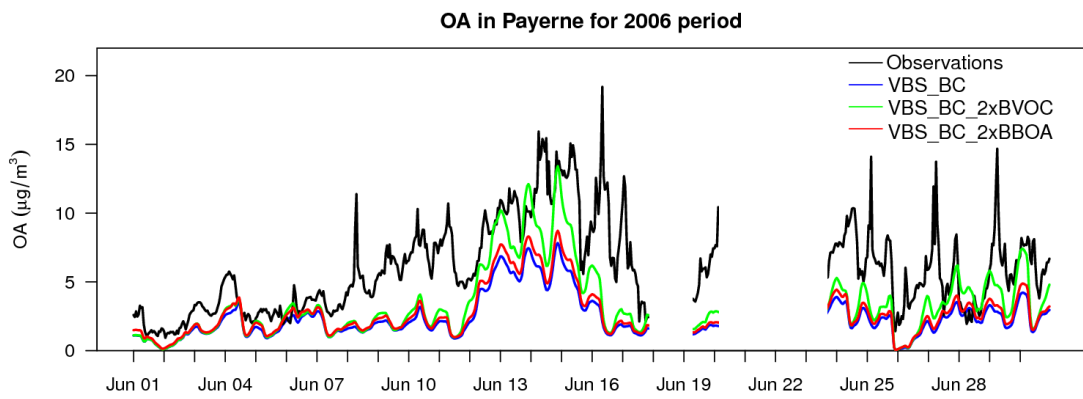
969

970 Figure 8. OA daily average scatter plots for VBS_BC, VBS_BC_2xBVOC and
 971 VBS_BC_2xBBOA scenarios for February-March 2009 for stations in Table 3. Solid lines
 972 indicate the 1:1 line. Dotted lines are the 1:2 and 2:1 lines. Boxplots indicate medians, 5th,
 973 25th, 75th and 95th quantiles for observations (black) and sensitivity tests (red). The crosses
 974 represent the arithmetic means.

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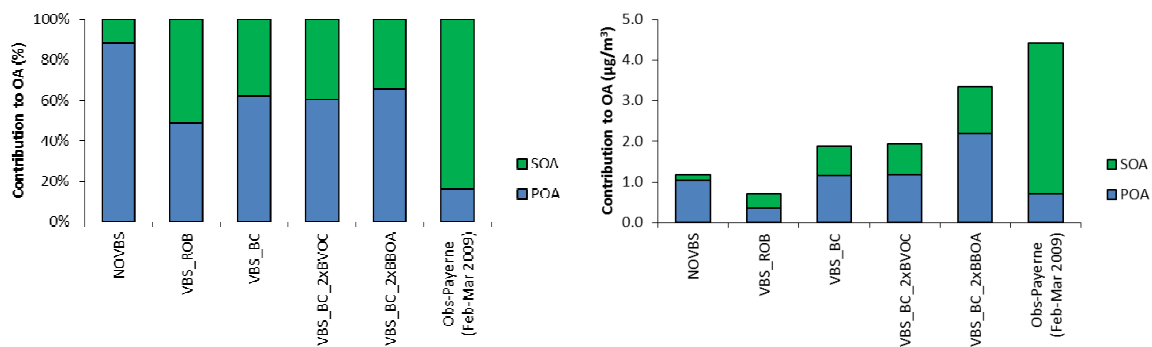


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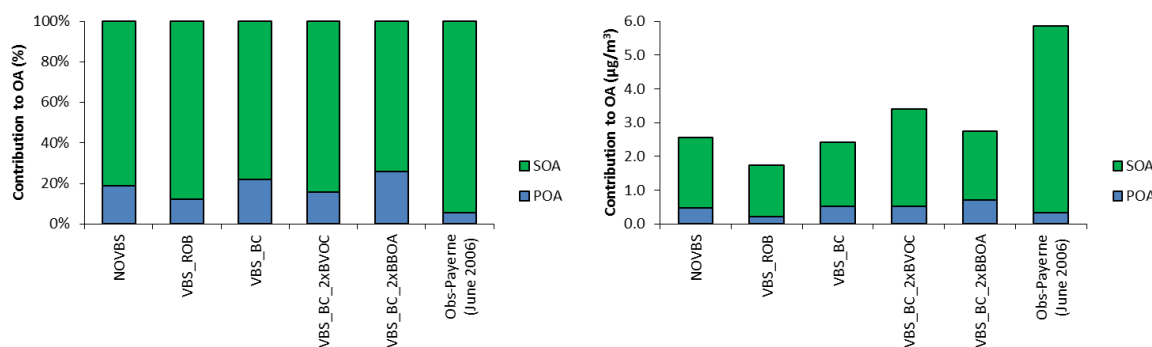
978 Figure 9. Predicted and observed total OA for scenarios VBS_BC, VBS_BC_2xBVOC and
 979 VBS_BC_2xBBOA in March 2009 (upper panel) and June 2006 (lower panel) at Payerne.

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983

984 Figure 10. Relative (left) and absolute (right) contribution of predicted and measured POA
 985 and SOA fractions to the total OA mass at Payerne for February-March 2009 winter period
 986 (upper-panel) and June 2006 (lower-panel) and different model scenarios. NOVBS:
 987 (traditional non-volatile POA), VBS_ROB (Robinson et al., 2007), VBS_BC (Tsimpidi et al.,
 988 2010, Shrivastava et al., 2011), VBS_BC_2xBVOC (increased biogenic emissions relative to
 989 VBS_BC), VBS_BC_2xBBOA (increased biomass burning emissions relative to VBS_BC),
 990 Obs-Payerne: AMS-PMF.

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