

1 Can biomonitors effectively detect airborne 2 benzo[a]pyrene? An evaluation approach using modelling

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12 13 **Abstract**

14 Biomonitoring data available on levels of atmospheric polycyclic aromatic hydrocarbons
15 (PAHs) in pine needles from the Iberian Peninsula was used to estimate air concentrations of
16 benzo[a]pyrene (BaP) and, at the same time, fuelled the comparison with chemistry transport
17 model representations. Simulations with the modelling system WRF+EMEP+CHIMERE
18 were validated against data from the European Monitoring and Evaluation Programme
19 (EMEP) air sampling network and using modelled atmospheric concentrations as a consistent
20 reference in order to compare the performance of vegetation-to-air estimating methods. A
21 spatial and temporal resolution of 9 km and 1 hour was implemented. The field-based
22 database relied on a pine needles sampling scheme comprising 33 sites in Portugal and 37
23 sites in Spain complemented with the BaP measurements available from the EMEP sites. The
24 ability of pine needles to act as biomonitoring markers for the atmospheric concentrations of
25 BaP was estimated converting the levels obtained in pine needles into air concentrations by
26 six different approaches, one of them presenting realistic concentrations when compared to
27 the modelled atmospheric values. The justification for this study is the gaps still existing in
28 the knowledge of the life cycles of semi-volatile organic compounds (SVOCs), particularly
29 the partition processes between air and vegetation. The strategy followed in this work allows

- 1 the effective estimation by the model of concentrations in air and vegetation and of the best
- 2 approaches to estimate atmospheric levels from values found in vegetation.
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- 4

1 **1 Introduction**

2 Semi-volatile organic compounds (SVOCs) are widespread chemicals that even at low
3 concentrations possess carcinogenic capacity (Baussant et al., 2001) and ecotoxicity (Solé,
4 2000) due to their persistence in different environmental matrices (air, soil, water, living
5 organisms). In particular, polycyclic aromatic hydrocarbons (PAHs) are originated by natural
6 and anthropogenic combustion processes or released from fossil fuels (Mastral and Callén,
7 2000) and can be transported in the atmosphere over long distances in gaseous phase or as
8 particulate matter (Baek et al., 1991). The lighter PAHs (2 or 3 aromatic rings) exist mainly
9 in the former, whereas the heavier (5 to 6 rings) consist almost entirely of the latter
10 (Bidleman, 1988), and this is the case of 5-ringed benzo[*a*]pyrene (BaP), arguably the most
11 studied PAH. BaP is the reference for PAH air quality standards, as defined by the European
12 Commission (Directive 2008/50/EC), which sets a limit of 1 ng m⁻³ over a 1-year averaging
13 period (European Commission, 2008).

14 The establishment of strategies for sampling and modelling of SVOCs in the atmosphere
15 aiming at the definition and validation of their spatial, temporal and chemical transport
16 patterns can be achieved by an integrated system of third-generation models that represent
17 the current state of knowledge in air quality modelling and experimental data collected in
18 field campaigns (Jiménez-Guerrero et al., 2008; Morville et al., 2011). The modelling
19 methods currently applied for SVOCs use very simple mass balance techniques or have
20 deterministic approaches, reflecting the complexity to characterise adequately the chemical
21 transport processes. These limitations urge for more experimentally-based information, hence
22 the need to combine field-based campaigns and modelling to address the problem properly
23 (Jakeman et al., 2006), including multi-matrix approaches whenever possible.

24 Moreover, measurements of pollutants such as PAHs are labour-intensive compared to those
25 of criteria air contaminants such as ozone and particulate matter, and the processes governing
26 their atmospheric fate and representation within chemistry transport models (CTMs) are not
27 yet well understood (Galarneau et al., 2013), particularly in terms of uncertainties associated
28 with the emissions and re-emissions from sinks, partition patterns, volatility and fate of
29 SVOCs, among others. A number of atmospheric modelling studies have tried to characterise
30 the levels and spatial-temporal patterns of PAHs (most of them focusing on BaP) using
31 CTMs both on global (Sehili and Lammel, 2007; Lammel et al., 2009; Friedman and Selin,
32 2012) and regional scales (Matthias et al., 2009; Aulinger et al., 2011; Bieser et al., 2012;

1 San José et al., 2013). These authors identify a lack of measurement data in Europe to
2 evaluate the behaviour of the CTMs against observations. For example, Bieser et al. (2012)
3 use only six European Monitoring and Evaluation Programme (EMEP) stations (four in the
4 Scandinavian region) and six additional sites in Germany and the UK to evaluate their year
5 2000 simulations. Bernalte et al. (2012) also highlight the importance of studies on PAHs
6 over the Western Mediterranean (Iberian Peninsula) in order to increase the knowledge of the
7 ambient levels in this region. For that purpose, San José et al. (2013) conducted a 12-week
8 modelling study supported by a field campaign to describe the behaviour of their
9 WRF+CMAQ simulations, but using only a single location in Spain.

10 Hence, there is a strong need to have trustful information on the atmospheric levels of
11 compounds like BaP and other SVOCs, in particular in areas with limited information, like
12 over the Iberian Peninsula. In that sense, vegetation species can play a decisive role as
13 biomonitors of the incidence and chemical transport of atmospheric pollutants (Maddalena et
14 al., 2003). Coniferous trees are particularly important, given their worldwide distribution and
15 specific characteristics. However, even if some studies report geographical or temporal
16 patterns of PAHs in coniferous needles (Weiss et al., 2000; Hwang and Wade, 2008;
17 Lehndorff and Schwark, 2009; Augusto et al., 2010; Ratola et al., 2010; Amigo et al., 2011;
18 Ratola et al., 2012) only a few deal with their air-vegetation distribution (St-Amand et al.,
19 2009a; 2009b). In addition, to our knowledge there is no study regarding the simultaneous
20 use of field and modelling data to assess the distribution of PAHs between air and pine
21 needles. Consequently, if trustful estimates of the atmospheric incidence could be obtained
22 from vegetation, the abundance of biomonitors such as pine needles would provide essential
23 information about the regional and global atmospheric behaviour of persistent contaminants.

24 Under these premises, the WRF+CHIMERE modelling system, coupled to BaP emission data
25 from EMEP was run and evaluated for the Iberian Peninsula. The modelled depositions were
26 compared to data from biomonitoring campaigns carried out along 70 sites, to assess the
27 ability of the model to reproduce BaP canopy deposition. Monitoring data from EMEP
28 (Torseth et al., 2012) was used to validate the modelled atmospheric BaP climatologies
29 (2006-2010). A total of six approaches were tested to estimate the conversion of BaP levels
30 from vegetation into air. To achieve this, the atmospheric levels from these approaches were
31 evaluated against the modelled air concentrations.

32

1 **2 Experimental section**

2 **2.1 Pine needles sampling**

3 The Iberian Peninsula, located in the SW of Europe, has an area close to 600,000 km² and a
4 population of almost 60 million, the majority of which distributed along the Atlantic and
5 Mediterranean coastlines, except for some important conurbations such as Madrid, Seville or
6 Zaragoza. Forests (with several pine species commonly present) are scattered through the
7 whole territory. Mountainous areas follow the same trend, with the most elevated chains
8 found in the Northern borders (Pyrenees and Cantabria) and in the south (Sierra Nevada).
9 Rural activities can be found almost everywhere, but are particularly important for the
10 economy in the central plateau, where population density is scarcer. A representation of the
11 different land uses in the target domain as represented by the WRF+CHIMERE modelling
12 system can be found in Ratola and Jiménez-Guerrero (2015). In this study, and according to
13 their availability, needles from *Pinus pinaster*, *Pinus pinea*, *Pinus halepensis* and *Pinus nigra*
14 with up to 1.5 years of exposure to contamination were collected from the bottom and outer
15 branches, placed in sealed plastic bags, kept from light and frozen until extraction. The
16 sampling campaigns were carried out in 33 sites in Portugal and 37 in Spain, in both cases
17 including urban, industrial and rural or remote areas. For further description of these
18 campaigns, the reader is referred to Ratola et al. (2009; 2012).

19 **2.2 Pine needles analysis and quantification**

20 The analytical procedure used to quantify the levels of PAHs (BaP included) in pine needles
21 was reported previously (Ratola et al., 2009; 2012). A brief description of the methodology
22 and of some characteristics of the pine needles from the different species can be found in
23 Supporting Information.

24 **2.3 Methods for the estimation of BaP air concentrations from vegetation.**

25 Given the lack of information on atmospheric concentrations of BaP in the sampling sites
26 chosen for this study, an estimation of those values from data provided by biomonitoring
27 studies with vegetation (coniferous needles in this case) was required. Resorting to literature,
28 six approaches (four of them using the same main calculation method, varying only one

1 parameter) were tried and the resulting estimated BaP concentrations compared with the
2 modelling experiments.

3 **Approach 1a.** This approach is based on the studies by St. Amand and co-workers (2007;
4 2009a; 2009b), who measured the levels of PBDEs and PAHs in vegetation (Norway spruce
5 needles in this case) and in the surrounding atmosphere (both gas-phase and particulate
6 material) and presented a strategy to estimate the air concentrations from those in vegetation
7 and vice-versa. In brief, the atmospheric concentration of SVOCs (C_a) estimated from the
8 levels in vegetation can be determined by the contribution of particle-bound (C_p) and
9 gaseous (C_g) phases. In the case of BaP, being a high molecular weight PAH, the gas-phase
10 contribution is negligible, which means ϕ (ratio between particle and particle+gas phases) ≈ 1
11 and C_a can be given by:

$$12 \quad C_a = C_p = (C_{vp} * m) / (A * v_p * t) \quad (1)$$

13 where C_{vp} - contribution of particle-bound deposition processes to the total concentration in
14 vegetation (ng g^{-1}); m - dry weight of pine needles (g); A - total surface area (m^2) of
15 vegetation (in our study, pine needles); v_p - particle-bound deposition velocity (m h^{-1}); t -
16 environmental exposure time of pine needles (h) with C_p expressed in ng m^{-3} . Since it was
17 impossible to calculate v_p for our samples, due to the lack of information on the atmospheric
18 concentrations, in this first approach the value calculated by St. Amand et al. (2009a) for
19 Norway spruce (*Picea abies*) needles was used: 10.8 m h^{-1} . Values of the mass and total
20 surface area for the pine needles studied are presented in Table S1. The exposure time was
21 estimated considering that the new needles sprung out on April 15 and counting the hours
22 from this day to the sampling date.

23 **Approaches 1b, 1c and 1d.** These approaches follow the same strategy, only with different
24 v_p values calculated from studies in literature reporting BaP concentrations in air and pine
25 needles (from *Pinus sylvestris* trees in cases 1b and 1c and a coniferous forest in 1d).
26 Approach 1b refers to the work by Klánová et al. (2009) and the estimated v_p (BaP) is 0.0039
27 m h^{-1} , while approach 1c comes from the work by Tremolada et al. (1996), with v_p (BaP) =
28 0.0263 m h^{-1} . For the 1d approach, it was considered the deposition velocity Horstmann and
29 McLachlan (1998) found for BaP over a coniferous forest canopy: 2.196 m h^{-1} . As can be
30 seen, the variability of v_p is evident, not only considering different species of vegetation, but
31 also using the same species in different locations. In the case of approaches 1b and 1c,
32 Klánová et al. (2009) sampled remote areas whereas Tremolada et al. (1996) considered

1 more urbanised locations, which may justify the higher deposition velocity in the latter case.
2 Differences in the uptake of PAH by different pine species in the same sampling sites are
3 also described in literature (Piccardo et al., 2005; Ratola et al., 2011).

4 **Approach 2.** This approach follows the work of Tomashuk (2010), which used
5 biomonitoring results in *Pinus nigra* needles and in turn profits from a study by Simonich
6 and Hites (1994). In the latter, an air-vegetation partition coefficient (K_v) is defined by:

$$7 \ln K_v = (1000/T) * \text{slope} - 35.95 \quad (2)$$

8 with T – air temperature (K); slope – calculated by Simonich and Hites (1994) for some
9 PAHs. And from K_v , the air concentration of PAHs (C_a) can be estimated by (in ng m^{-3}):

$$10 C_a = C_v / (K_v * \text{lipid}) \quad (3)$$

11 with C_v – concentration in the vegetation (ng g^{-1} , dw); lipid – lipid content per dry weight of
12 pine needles (mg g^{-1} , dw). Values of the lipid content for the pine needles studied are
13 presented in Table S1.

14 **Approach 3.** Chun (2011) measured PAH concentrations in *Pinus koraiensis* needles and the
15 surrounding air and came up with the following correlation between $\log K_{oa}$ and C_v/C_a :

16 From acenaphthylene to chrysene:

$$17 C_a = C_v / \exp [(\log K_{oa} - 7.9603) / 0.4557] \quad (4)$$

18 with C_a – concentration in air (ng m^{-3} , dw); C_v – concentration in the vegetation (ng/g , dw)

19 From chrysene to benzo(ghi)perylene (the equation used to calculate BaP concentrations):

$$20 C_a = C_v / \exp [(\log K_{oa} - 12.18) / (-0.2272)] \quad (5)$$

21 $\log K_{oa}$ is a temperature-dependent coefficient, and was calculated using the following
22 equation:

$$23 \log K_{oa} = A + (B/T) \quad (6)$$

24 where coefficients A and B are given by Odabasi et al. (2006) and the temperature (T) in
25 each site was the mean from the three months previous to sample collection, since it
26 corresponded to the intervals of exposure between campaigns (with a seasonal periodicity for
27 most sampling points). The equilibrium between air and pine needles is still not completely
28 understood and can be a slow process for compounds with high $\log K_{oa}$ such as BaP
29 (Mackay, 1991) and it may not be possible to acknowledge if “non-equilibrium” conditions
30 or alternative processes (Tremolada et al., 1996).

1 **2.4 Modelling experiment and validation**

2 In this study, the Weather Research and Forecasting (WRF) (Skamarock et al., 2008) and the
3 CHIMERE modelling system (Menut et al., 2013), with a resolution of 9 km for the entire
4 Iberian Peninsula coupled to EMEP BaP emissions (Vestreng et al., 2009), was run and
5 evaluated for the Iberian Peninsula in a simulation covering the years 2006 to 2010 on an
6 hourly basis. This CHIMERE version has been modified to include gaseous and particulate
7 BaP. Gas-phase degradation by OH radicals, which represents over 99% of the degradation
8 path for gas-phase BaP, was accounted for, with a $k_{OH} = 5.68 \times 10^{-11}$ (Schwarzenback et al.,
9 2003). But more importantly, the oxidation of particulate BaP with ozone was also included,
10 since the respective reaction rate is one order of magnitude higher than other degradation
11 processes, and can be considered the only effective degradation path for particulate BaP in the
12 atmosphere (Bieser et al., 2012). In this case, the reaction constant follows the approach of
13 Pöschl et al. (2001):

14

$$15 \quad k = k_{\max} [O_3]/(1 + K_{O_3}[O_3]) \quad (7)$$

16

17 being $k_{\max} = 0.015 \text{ s}^{-1}$ and $K_{O_3} = 2.8 \times 10^{-13} \text{ cm}^3$. A bias adjustment technique was applied and
18 is referred in Supporting Information, together with a description of the modelling set-up and
19 validation procedures (Table S2). All modelled concentrations presented in this work are bias-
20 adjusted.

21 The BaP concentrations in pine needles used in this work are taken from biomonitoring
22 campaigns previously performed in the Iberian Peninsula (Ratola et al., 2009; 2010; 2012).
23 These data were compared to the deposition over vegetal canopies as estimated by the
24 CHIMERE transport model. The dry deposition flux in CHIMERE is directly proportional to
25 the local concentration C of the target compound (in this case, BaP):

$$26 \quad F = -v_d \times C \quad (8)$$

27 where F represents the vertical dry deposition flux, the amount of material depositing to a
28 unit surface area per unit time. The proportional constant between flux and concentration, v_d ,
29 is known as the deposition velocity. The main factors governing dry deposition are the grade
30 of the atmospheric turbulence, the chemical properties of the species, and the nature of the
31 soil and the vegetation.

1 The deposition over vegetal canopies in CHIMERE for particles employs a resistance scheme
2 (Wesely, 1989). The dry deposition velocity follows the formulation of Seinfeld and Pandis
3 (1997):

$$4 \quad v_d = (1/(r_a + r_b + r_a \times r_b \times v_s)) + v_s \quad (9)$$

5 where r_a is the aerodynamic resistance (or aerodynamic drag) and r_b the resistance at the
6 quasi-laminar sublayer. The aerodynamics resistance is calculated as the integral of the
7 inverse of the diffusivity coefficient K_z up to the middle of the model surface layer, which
8 can be estimated using the analytical formulae of the surface-layer similarity profiles for K
9 (Seinfeld and Pandis, 1997) and v_s stands for the sedimentation velocity. For vegetal
10 canopies, as in our case, corrections have been implemented. These corrections are not
11 detailed in the CHIMERE manual (<http://www.lmd.polytechnique.fr/chimere/>), but rather
12 supported on the literature presented (Giorgi, 1986; Peters and Eiden, 1992; Zhang et al.,
13 2001). For this reason, and for the sake of brevity, the same strategy is adopted here and
14 readers are referred to those works for further details.

15

16 **3 Results and discussion**

17 **3.1 Model evaluation for vegetation and air levels**

18 The model climatologies for BaP in canopy deposition and air concentration were done under
19 the premise of constituting a base for a broad spectrum of studies within the air-vegetation
20 interactions. In fact, a description of these simulations was mentioned previously by Ratola
21 and Jiménez-Guerrero (2015). However, given the importance for the current study, a
22 summary is presented here, also considering a different perspective.

23 **3.1.1 Vegetation**

24

25 The modelled deposition over vegetal canopies was evaluated against observations compiled
26 from pine needles. Thus, the adequacy of the model's deposition velocity for the Iberian
27 Peninsula is assessed by a direct evaluation of the deposition velocity against observations.
28 This information is summarised in Table 1 and a point-to-point comparison is shown in
29 Supporting Information (Table S3). The samples were explicitly compared with the model
30 period corresponding to their effective exposure interval. Given the assumption that there is a

1 full uptake by the pine needles of the deposited BaP, the modelled deposition flux is
2 converted to pine needles concentration multiplying it by the respective time of exposure
3 (equivalent for the model and the pine needles). The results indicate an overall good ability
4 of the model to reproduce the vegetation's uptake of BaP, when compared to the
5 biomonitors. Generally, the modelled concentrations tend to be overpredicted DJF, MAM
6 and SON, when the deposited BaP is overestimated by 0.08 to 0.17 ng g⁻¹ (MFB up to
7 +17%). On the other hand, in summer (JJA) the model is likely to underpredict the measured
8 levels in vegetation (-0.41 ng g⁻¹, -39% as MFB), seemingly due to its tendency volatilise
9 SVOCs as a result of the high temperatures simulated over the Iberian Peninsula. The RMSE
10 remains under 1.5 ng g⁻¹ in all seasons (Table 1), indicating a close approach of the model to
11 the levels obtained in pine needles. Particularly noticeable is the accurate reproduction of the
12 spatial patterns. In fact, the estimates from the spatial correlation coefficient (which is
13 highest for MAM and lowest for SON, ranging from 0.77 to 0.87 for all seasons) indicate
14 that regardless of the model bias, the spatial reproducibility of the deposition patterns over
15 the Iberian Peninsula is very well reproduced in all seasons, capturing also the seasonal
16 distribution.

17 In terms of the modelled levels in canopies, Figure 1 shows that the deposition of BaP are
18 clearly lowest for JJA (under 3 ng g⁻¹ over most of the Iberian Peninsula) and have the
19 highest values in DJF and MAM (10-20 ng g⁻¹ over north-western Iberian Peninsula and the
20 Cantabria coast). But apart from the geographic distribution being closely related to the
21 emitting areas, the differences in the entrapment of PAHs by the different land uses can play
22 an equally significant role, as observed in the spatial uptake patterns shown in Figure 1. Even
23 if a discussion on the role of the different pine species is beyond the scope of this work,
24 several points were brought to our attention. For instance, it was shown previously that *P.*
25 *pinaster* needles have a superior uptake capacity towards PAHs than *P. pinea* (Ratola et al.,
26 2011) or *P. nigra* ones (Piccardo et al., 2005). The first two species have a strong
27 implantation in the forests of the Iberian Peninsula, but while *P. pinea* is more equally
28 distributed (although mainly present in the south and Mediterranean coast), *P. pinaster*
29 prevails in the north-west and Atlantic coast. This may be the reason why the model tends to
30 present higher deviations over the northernmost biomonitoring points (*P. pinaster*, MFB =
31 21%) than over eastern-southern areas, with predominant *P. pinea* (MFB = -17%), as shown
32 in Table S3 of the Supporting Information). It was also suggested that leaf surface properties
33 are more a function of the environmental exposure than of the plant response (Cape et al.,

1 1989). Given all these facts, both chemistry transport models and other parameterisations
2 face a huge task to represent the levels of pollutants in vegetation. In this sense, enhancing
3 the field experimental work on the uptake of these chemicals would be strongly beneficial.

4 **3.1.2 BaP air climatology**

5 As mentioned previously, studies in literature regarding the field monitoring of PAHs levels
6 in the Iberian Peninsula's vegetation are limited and, therefore, modelling strategies can
7 represent a valuable tool to assess BaP levels over the target region. The few existing studies
8 (described in Introduction) reflect two main points: the influence of local sources and the
9 variability of the uptake abilities of the different vegetation species. Since the main focus of
10 this work is on the climatologies of the atmospheric BaP levels, in order to assess the correct
11 reproducibility of their spatial-temporal patterns the WRF+CHIMERE BaP modelled
12 concentrations were evaluated against EMEP air quality data after the bias adjustment
13 explained in the Supporting Information.

14 According to Ratola and Jiménez-Guerrero (2015), the modelled atmospheric concentrations
15 of BaP present normalised biases that are under 30% over all the EMEP stations in the
16 Iberian Peninsula. The fact that both positive and negative biases were found for annual
17 mean concentrations indicates that the model is not generally inclined towards overprediction
18 or underprediction for all the domain of study. As depicted in Figure 2, the deviations only
19 range between +1.63 pg m^{-3} over the northern Iberian Plateau (Peñausende station, close to
20 the Spanish-Portuguese border) and -4.59 pg m^{-3} (San Pablo de los Montes station, in the
21 southern-central Iberian Plateau). The low biases obtained indicate that the model is
22 reproducing accurately the atmospheric concentrations of BaP, and therefore can be used as a
23 reference for the comparison with the levels of this compound obtained from air-vegetation
24 partition, as will be explained in detail below.

25 Modelled BaP concentrations in the atmosphere (Figure 3) achieve a maximum during the
26 winter months (DJF), that can reach over 300 pg m^{-3} in most polluted areas (NW Spain and
27 western coast of Portugal), while background areas hardly exceed 5 pg m^{-3} (lowest
28 concentrations in the SE Levantine coast). The highest BaP concentrations measured using
29 pine needles as the biomonitoring matrix and atmospheric concentrations simulated by the
30 model were found in urban and industrial settings, mainly distributed along the north-western
31 coast of the Iberian Peninsula (as also reported by Amigo et al., 2011 and Ratola et al., 2012)
32 followed by rural and remote areas. This reflects the accumulation of anthropogenic sources

1 like traffic, building heating or industrial processes involving combustions in the most
2 populated areas of the Iberian Peninsula. Due to the characteristics of such sources, a
3 tendency to seasonality can be anticipated as well. In the colder months, traffic and building
4 heating are increased and this is not only reflected by the field measurements (Ratola et al.,
5 2010), but also by the models, as shown in Figure 3.

6 Given that the model represents accurately the air climatologies of BaP, can we use its results
7 to evaluate the ability of the air/vegetation methods available in scientific literature to
8 estimate the atmospheric levels of BaP from biomonitoring databases? Being the accuracy of
9 the model to capture the air concentrations evaluated against EMEP air measurements, the
10 argument this work adopts is: since the model correctly captures air concentrations and
11 deposition (which have been previously assessed in Section 3.1.1), we can use the modelled
12 air concentrations as a reference to evaluate the fitness of the different vegetation-air
13 conversion approaches. Therefore, in the following section, the model concentrations have
14 been considered as a consistent reference (due to the low biases obtained) to act as a
15 reference to validate the approaches for this vegetation-to-air conversion.

16 **3.2 Comparison of vegetation-to-air approaches**

17 Databases on the atmospheric levels of SVOCs are already available, but the existing ones
18 (like EMEP) do not cover, for instance, the entire Iberian Peninsula for a climatologically-
19 representative period of time (apart from some isolated measurements). In terms of
20 vegetation, the scenario is even worse, but since the presence of SVOCs in such
21 environmental matrices (and in particular in pine needles) reflects entirely an entrapment
22 from the atmosphere (Hwang and Wade, 2008), these measured data can be used not only to
23 validate the model results in vegetation but also to complement the information gathered by
24 the direct atmospheric sampling. For that purpose, six approaches to convert the
25 concentrations found in the 70 sites where pine needles were collected into atmospheric
26 levels were compared to the reference provided by the CTM simulations. This hypothesis is
27 based on the fact that models represent correctly the measured atmospheric concentrations of
28 BaP over the Iberian Peninsula, taking into account the evaluation against EMEP field
29 measurements available. This hypothesis was forced by the lack of simultaneous samplings
30 of vegetation and air concentrations over the target area. Therefore, we used the following
31 methodology: (a) validate simulations with WRF+CHIMERE data against EMEP network
32 measurements, in order to check the ability of the CTM to reproduce atmospheric

1 concentrations over the entire Iberian Peninsula; (b) once proven that errors are acceptable
2 and that the model shows no trend bias, use modelled atmospheric concentrations as a
3 consistent reference that allows us to compare various vegetation-to-air estimating methods
4 and check which is the most suitable approach for the particular conditions of the area.

5 It is clear that given the numerous variables and conditions involved, the uptake processes of
6 compounds like PAHs by matrices such as pine needles is not entirely understood (Barber et
7 al., 2004). But the information we have so far indicates that pine needles are valid
8 biomonitors of atmospheric loads, but also can be used to assess the performance of different
9 methods to convert vegetation uptake levels into atmospheric concentrations. Thus, the
10 objective is to test the response of the six vegetation-to-air approaches detailed in section 2.3
11 through a field/model check in the sampling points chosen.

12 Results (Table 2) reveal that approach 1d is the best fit to convert the levels measured in
13 vegetation into air concentrations, when compared to the outcome provided by the model.
14 This approach was used by Ratola and Jiménez-Guerrero (2015) to assess differences
15 between pine species in modelling simulations as the deposition velocity is in this case
16 defined for an entire forest canopy and not for a given species. This general characteristic is
17 seemingly giving this approach an advantage in terms of the vegetation-to-air calculations.
18 The MFB ranges from -19% for spring (MAM) to a slight overestimation during winter
19 (DJF, +9%), being the biases under 3 pg m^{-3} for all seasons. These errors are relatively low
20 bearing in mind the diversity of the sampling sites considered in this work. Previous works
21 have demonstrated the seasonal variability of PAHs uptake by pine needles (Hwang and
22 Wade, 2008; Ratola et al., 2010), with the highest levels occurring in winter and the lowest in
23 summer. However, these differences are much more visible in the lighter PAHs (the ones in
24 the gas-phase), given the stronger affinity of the pine needles waxy layer towards their
25 entrapment, when compared to the particulate PAHs.

26 Being one of the latter, BaP in pine needles may not experience the same level of seasonal
27 variation as in the atmosphere, even if it presents a similar trend. These seasonal differences
28 can be much stronger in the atmosphere, due to the fluctuation of the emission rates from
29 winter to summer. It is then not surprising that the model underestimates the atmospheric
30 concentrations of BaP measured in the colder months and overestimates them in the warmer
31 ones, since in this case the field values are obtained from the levels found in the pine needles.
32 Approach 1d is also the best representation for this seasonal variability (estimated as the
33 standard deviation between approaches and the CTM). Additionally, this approach shows the

1 best air/vegetation relationship simulated by the model, with the rest of the methods
2 providing unrealistic concentrations when compared to the measurements in EMEP stations
3 and modelling results. In fact, approaches 1a and 2 tend to underestimate the modelled
4 concentrations by a factor up to 10, yielding negative biases for all seasons. The rest of the
5 approaches greatly overestimate the levels of BaP (by a factor of 100 in the case of 1c and 3
6 and of 1000 in approach 1b). These large variations are mainly caused by the difference in
7 the deposition velocities used in each approaches 1a to 1d (from 10.8 m h^{-1} in 1a to 0.0039 m
8 h^{-1} in 1b) and in completely different vegetation-to-air estimation strategies in approaches 2
9 and 3. The deposition velocity has an important role in one of the three methodologies for
10 estimating air concentrations from vegetation (methodology which derives into approaches
11 1a to 1d), but it allows precisely to understand the differences that may occur when
12 conditions are changed (different species, different locations, different times of the year in
13 the same locations, different affecting sources, etc).

14 With respect to the temporal correlation coefficients, since approaches 1a to 1d present the
15 same value (0.51), as they rely on the same calculations (only changing the deposition
16 velocity). This is an acceptable description of the temporal variability observed in all sites.
17 Approach 2 is not able to reproduce these time series (correlation coefficient of -0.55), but,
18 interestingly, it is approach 3 that presents the best correlation (0.80). In this latter case,
19 although the bias for the BaP concentrations is quite high, the r value can be related with the
20 different uptake efficiencies pine needles show for gas-phase or particulate PAHs. The two
21 equations suggested by Chun (2011) to relate concentrations of PAHs in needles and air
22 separate the lighter from the heavier ones. So even if the actual concentrations are not very
23 well described, the temporal air-needles synergies may better projected by this approach in
24 this particular case.

25 Finally, spatial correlation coefficients (which provide a simulation for the adequate
26 representation of the BaP spatial patterns over the Iberian Peninsula) are correctly reproduced
27 by all approaches (Table 2). The highest value is seen for winter in approach 2 ($r=0.68$) and
28 for the rest of the seasons, approaches 1a-1d present the higher correlation coefficients (from
29 0.67 in JJA to 0.85 in MAM). Approach 3 generally offers the lowest spatial correlation
30 coefficients for all seasons, except in summer. The fact that the lowest r values are generally
31 found for winter and summer (also the extremes of BaP concentrations in the environment),
32 highlights the limitations of the model to represent these extremes.

1 Ideally, the air levels SVOCs are measured in the field using expensive active air sampling
2 equipment which also require permanent power supply while operating. Thus, these devices
3 only exist in certain parts of the world, which does not allow a proper coverage of the global
4 presence of such contaminants, which naturally hinders the efforts of modelling estimation as
5 well. As mentioned above, as living structures vegetation matrices have morphological,
6 physical and chemical behaviour that depends on many parameters, even within the same
7 species. Thus, the equations describing the air-vegetation partition suffer from these effects
8 when a broad solution is searched for. Again in ideal terms, only a direct comparison of field
9 campaigns and active air sampling performed in the same spots is bound to achieve some
10 accuracy, if it includes a seasonal framework as well. In fact, the main approaches presented
11 in this work derive from these type of combined studies. But when it is impossible to have
12 simultaneous active air and biomonitoring sampling models can help us to assess if the
13 assumptions we are working with are sound, if a previous validation with the field-based air
14 concentrations is successful (as is the case in our study). Naturally, there is a concern that the
15 uncertainty associated to all the steps involved may affect the conclusions of a study like this.
16 Even if a detailed analysis would be extremely complex and out of the scope of this work, the
17 main source of uncertainty of our global process can be identified: the emission inventories
18 for PAHs, as stated by San José et al. (2013). In general, this uncertainty was estimated to be
19 within a factor of 2 to 5 (Berdowski et al., 1997), much larger than any other uncertainty
20 associated to the validation process and rest of steps. For instance, EMEP individual
21 measurements should have a precision within $\pm 10\%$ and the data quality objectives for the
22 sampling and chemical analysis set a combined uncertainty between 15 and 25% (EMEP,
23 2001). Also, the analytical methodology to quantify BaP in pine needles have similar
24 precision values (Ratola et al., 2009). The contribution of these processes to the global
25 uncertainties would be reduced in comparison to the BaP emissions.

26 **4 Conclusions**

27 This work proved the good performance of pine needles as biomonitors of the BaP
28 atmospheric concentrations. Results show that the WRF+CHIMERE modelling system
29 reproduces accurately not only the atmospheric presence of BaP, with deviations below 0.4
30 ng g^{-1} , but also the spatial and temporal patterns of its concentrations over the vegetation in
31 the Iberian Peninsula (biases lower than 30% for all stations and seasons). From the six
32 methods tested to convert vegetation levels (in pine needles) into atmospheric concentrations,
33 approach 1d showed the most accurate results, followed by approach 1a, when compared to

1 modelling results and observations from EMEP. However, these results should not be
2 interpreted as a ranking of the general performance of the approaches. For instance, given
3 that approaches 1a, 1b, 1c and 1d only differ on the deposition velocity considered for BaP,
4 we can conclude that approach 1d is the one representing more closely the particular
5 conditions of the target area. Nevertheless, for other locations and frameworks, further
6 research should be conducted to verify these conclusions. Another very important aspect to
7 take into account is that none of the studies where the available approaches were reported
8 used needles from the same pine species of the current study nor was located in areas of
9 similar climatic or geographical conditions. These facts can considerably alter the uptake
10 conditions of the pollutants, hence the different deposition rates reported.

11 Arguably, it could be said that when the model is taken as the reference, the deposition
12 velocity in the best approach is not the most adequate for the Iberian Peninsula, but rather the
13 one closer to the approximation of the deposition over vegetal canopies included in the CTM.
14 This suggestion can be rebutted given that the model results were validated against the field
15 data available from the EMEP air sampling stations, proving that the approximation of the
16 model is indeed the most satisfactory for the conditions of this area (and, therefore, so are
17 those of approach 1d). Another unprecedented perspective introduced by this work is that,
18 contrary to the few similar studies found in literature, instead of studying isolated episodes of
19 contamination, the simulations cover a large period (2006-2010). This highlights a climatic
20 viewpoint to the problem of BaP on a regional scale, and was not done previously (at least
21 over the Iberian Peninsula).

22 Considering that the theoretical principles of the three methodologies chosen in this work
23 that led to the air-vegetation partition calculations are valid worldwide and having some of
24 the parameters missing for our sampling domain, we had to resort to the ones existing in
25 literature. With more similar studies in the future we can head towards a much better
26 reproducibility and robustness of the modelling strategies. Our aim was to open a possible
27 path for it and the results are encouraging. But if fieldwork continues to be as scarce as it is
28 nowadays, the journey will be necessarily slower than we hope for.

29 The relevance of these findings open the possibility that pine needles can be used to assess
30 the temporal and spatial behaviour of BaP or other priority pollutants under completely
31 innovating perspectives. Namely allowing a reliable understanding of the air quality in areas
32 where common air sampling devices are unavailable. The comparison of levels within a
33 regional scale will enable the strong enhancement of the knowledge available so far in the

1 scientific literature for studies on atmospheric chemistry and transport of trans-boundary
2 SVOCs, which is scarce (even more if we consider model validation against experimental
3 data). Despite these promising results, further research is still needed and should be devoted
4 to: (a) study the applicability of the methods tested to different areas (both geographically
5 and in terms of land use) and (b) assess the performances of different vegetation species and
6 their ability to act as biomonitors of the atmospheric presence of several classes of hazardous
7 compounds.

8

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17

18 **Supporting information available**

19 Information on pine needles characteristics, sampling, analytical methodology, as well as on
20 the modelling and vegetation-to-air estimation strategies. This material is available free of
21 charge via the Internet.

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- 8

1 **TABLE AND FIGURE CAPTIONS**

2

3 **Table 1.** Seasonal evaluation of WRF+CHIMERE modelled BaP depositions results (over
4 vegetal canopies) against concentrations found in pine needles.

5

6 **Table 2.** Results from the comparison of BaP concentrations in air obtained by the chemistry
7 transport models (CTM) simulations and those estimated from pine needle levels by several
8 approaches

9

10 **Figure 1.** Seasonal distribution of modelled deposition of BaP on vegetation (ng g^{-1}) over the
11 domain covering the Iberian Peninsula: (from top-down and left-right): winter (DJF), spring
12 (MAM), summer (JJA) and autumn (SON) climatologies for the period 2006-2010.

13

14 **Figure 2.** BaP annual mean concentrations (pg m^{-3} , shaded) and biases for EMEP stations
15 (pg m^{-3} , circles) using the available information for the period 2006-2010.

16

17 **Figure 3.** BaP climatologies (pg m^{-3}) over the Iberian Peninsula (from top-down and left-
18 right): winter (DJF), spring (MAM), summer (JJA) and autumn (SON) for the period 2006-
19 2010.

20

21

22

1 Table 1. Seasonal evaluation of WRF+CHIMERE modelled BaP depositions results (over
 2 vegetal canopies) against measured concentrations found in pine needles.

	DJF	MAM	JJA	SON
MFB (%)	-2.17	16.77	-39.23	5.28
RMSE (ng g⁻¹)	1.26	1.45	0.84	1.97
BIAS (ng g⁻¹)	0.10	0.08	-0.41	0.17
OBS MEAN±STD DEV (ng g⁻¹)	1.67±1.66	2.39±2.17	1.25±0.90	1.85±1.64
MOD MEAN (ng g⁻¹)	1.76±1.70	2.48±2.37	0.84±0.64	2.02±1.42
SPATIAL CORR COEF (<i>r</i>)	0.86	0.87	0.85	0.77

3 DJF – December, January and February; MAM – March, April and May; JJA – June, July and August;
 4 SON – September, October and November; MFB - mean fractional bias; RMSE - root mean square
 5 error; OBS - pine needle concentrations; STD DEV – standard deviation; MOD - modelled
 6 concentrations; CORR COEF – correlation coefficient

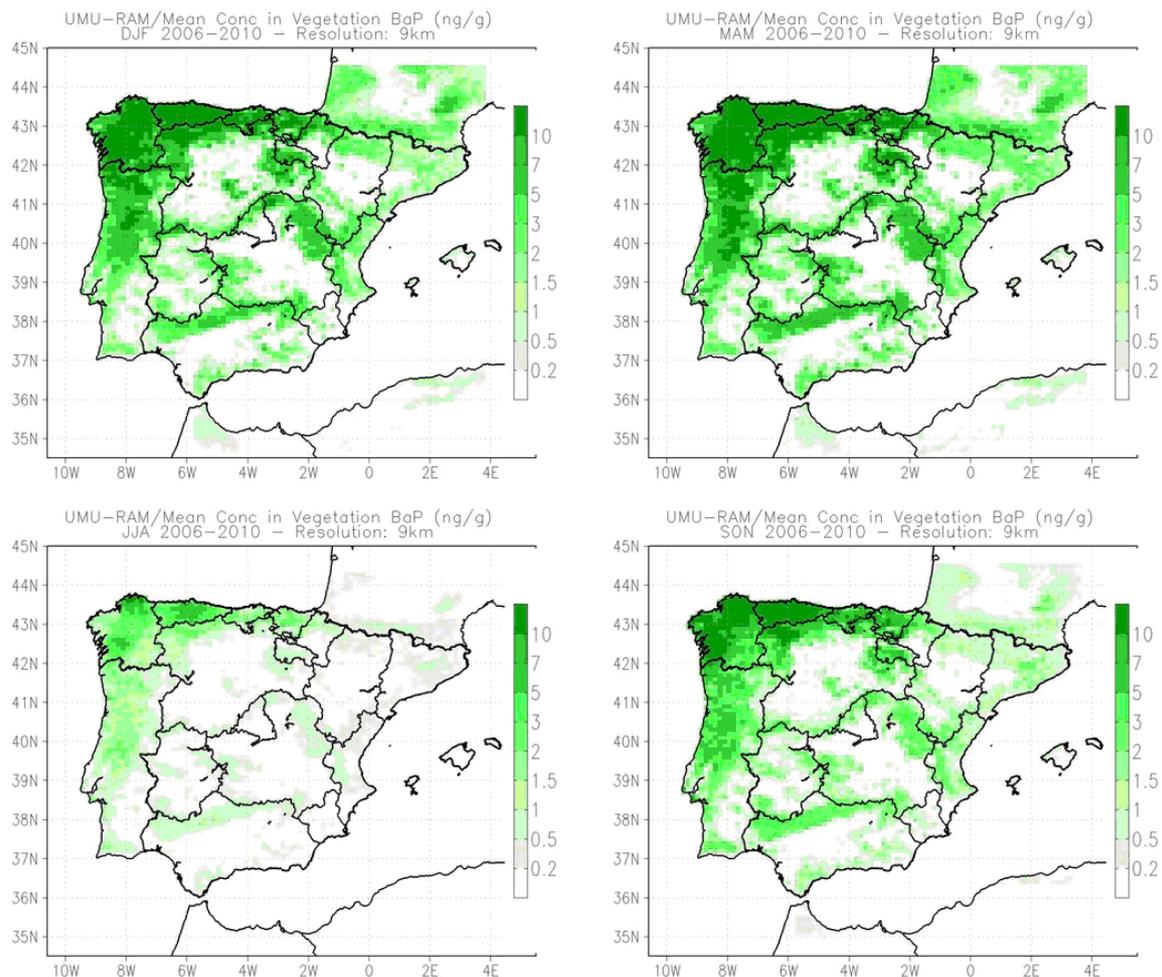
7

1 Table 2. Results from the comparison of BaP concentrations in air obtained by the chemistry
 2 transport models (CTM) simulations and those estimated from levels measured in pine
 3 needles by several approaches

	DJF	MAM	JJA	SON
CTM MEAN*±STD DEV (pg m ⁻³)	15.63±15.55	16.08±15.48	7.32±6.84	11.19±10.35
APPROACH 1a (TEMPORAL CORR. COEF.: 0.51)				
	DJF	MAM	JJA	SON
SPATIAL CORR. COEF.	0.57	0.85	0.67	0.80
MFB (%)	-125.46	-129.35	-125.75	-136.06
RMSE (pg m ⁻³)	19.09	16.14	8.11	14.57
BIAS (pg m ⁻³)	-12.70	-12.58	-6.01	-9.64
METHOD MEAN±STD DEV (pg m ⁻³)	3.31±3.24	3.51±3.21	1.31±1.01	1.55±1.21
APPROACH 1b (TEMPORAL CORR COEF: 0.51)				
	DJF	MAM	JJA	SON
SPATIAL CORR. COEF. (r)	0.57	0.85	0.67	0.80
MFB (%)	198.97	198.81	198.83	198.95
RMSE (pg m ⁻³)	12526.82	16294.77	4413.82	5197.87
BIAS (pg m ⁻³)	9203.00	9945.01	3815.12	4481.39
METHOD MEAN±STD DEV (pg m ⁻³)	9219±8358	9961±9722	3822±2890	4492±3424
APPROACH 1c (TEMPORAL CORR COEF: 0.51)				
	DJF	MAM	JJA	SON
SPATIAL CORR. COEF. (r)	0.57	0.85	0.67	0.80
MFB (%)	193.27	192.28	193.06	193.15
RMSE (pg m ⁻³)	1860.48	2420.65	653.60	765.74
BIAS (pg m ⁻³)	1361.62	1474.44	563.88	660.15
METHOD MEAN±STD DEV (pg m ⁻³)	1377.63±1347.92	1488.53±1400.05	571.20±431.94	671.34±511.74
APPROACH 1d (TEMPORAL CORR COEF: 0.51)				
	DJF	MAM	JJA	SON
SPATIAL CORR. COEF. (r)	0.57	0.85	0.67	0.80
MFB (%)	9.21	-18.99	-6.30	-15.58
RMSE (pg m ⁻³)	18.34	12.42	5.91	9.45
BIAS (pg m ⁻³)	0.08	-0.81	-0.84	-2.88
METHOD MEAN±STD DEV (pg m ⁻³)	15.94±15.60	15.27±14.86	6.48±4.96	8.31±8.19
APPROACH 2 (TEMPORAL CORR COEF: -0.55)				
	DJF	MAM	JJA	SON
SPATIAL CORR. COEF. (r)	0.68	0.89	0.35	0.76
MFB (%)	-179.73	-171.63	-115.84	-121.53
RMSE (pg m ⁻³)	21.01	19.09	8.22	13.70
BIAS (pg m ⁻³)	-15.33	-14.96	-5.81	-8.89
METHOD MEAN±STD DEV (pg m ⁻³)	0.68±0.60	1.13±1.06	1.51±1.15	2.30±2.24
APPROACH 3 (TEMPORAL CORR COEF: 0.80)				
	DJF	MAM	JJA	SON
SPATIAL CORR. COEF. (r)	0.26	0.48	0.65	0.41
MFB (%)	194.93	194.88	197.07	195.66
RMSE (pg m ⁻³)	1212.05	1166.83	897.97	916.64
BIAS (pg m ⁻³)	1283.79	1214.75	967.09	986.96
METHOD MEAN±STD DEV (pg m ⁻³)	1299.80±342.94	1230.83±333.38	974.41±36.72	998.15±41.59

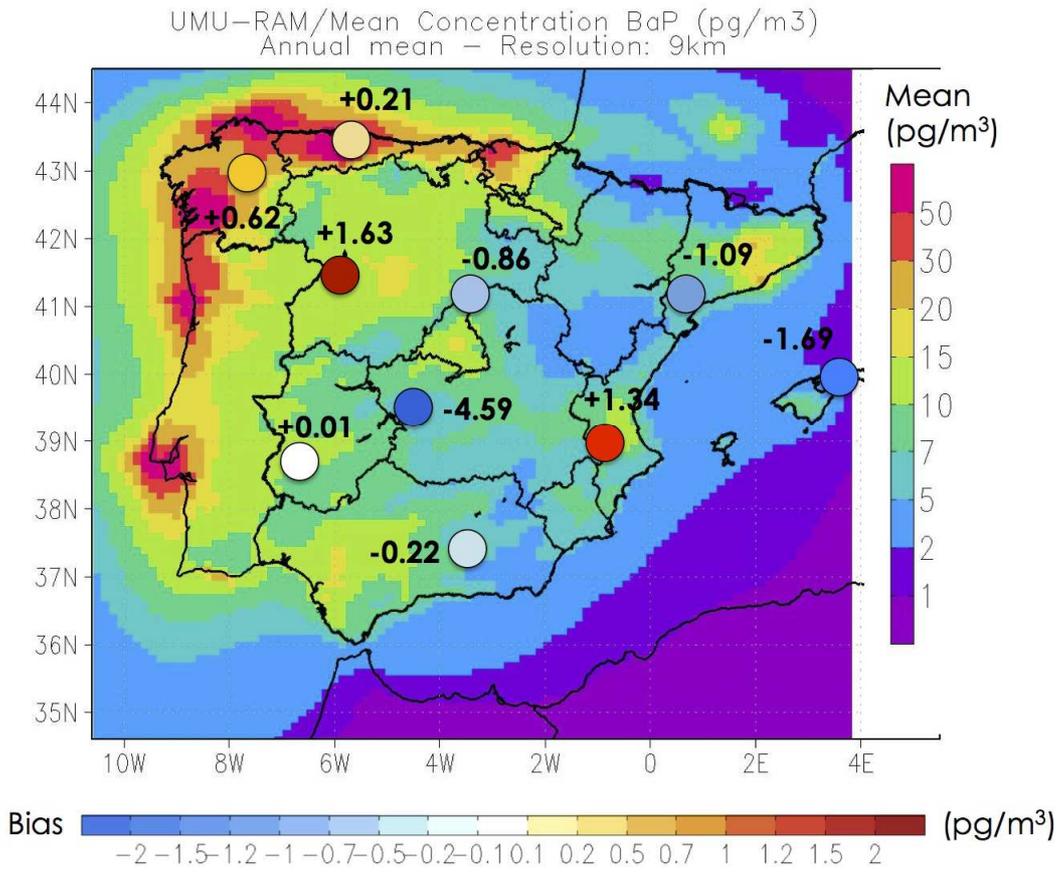
4 *Modelling results are considered as a consistent reference to compare the estimations from the different
 5 approaches. DJF – December, January and February; MAM – March, April and May; JJA – June, July and
 6 August; SON – September, October and November; CTM – chemistry transport model concentrations; STD
 7 DEV – standard deviation; CORR COEF – correlation coefficient; MFB - mean fractional bias; RMSE - root
 8 mean square error

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2 Figure 1. Seasonal distribution of modelled deposition of BaP on vegetation (ng g^{-1}) over the
3 domain covering the Iberian Peninsula: (from top-down and left-right): winter (DJF), spring
4 (MAM), summer (JJA) and autumn (SON) climatologies for the period 2006-2010.

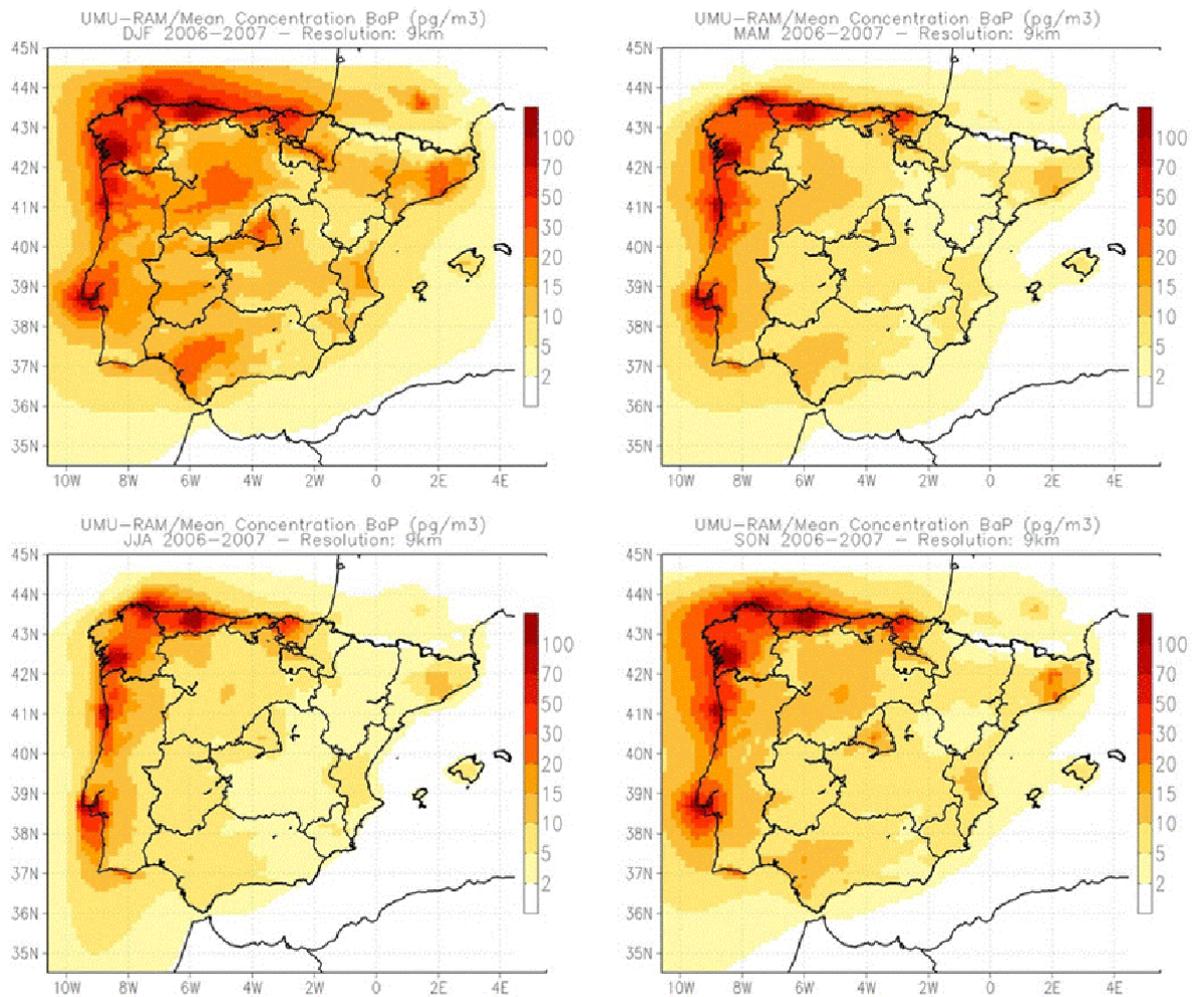
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2 Figure 2. BaP annual mean concentrations (pg m⁻³, shaded) and biases for EMEP stations (pg
3 m⁻³, circles) using the available information for the period 2006-2010.

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2 Figure 3. BaP climatologies (pg m⁻³) over the Iberian Peninsula (from top-down and left-
 3 right): winter (DJF), spring (MAM), summer (JJA) and autumn (SON) for the period 2006-
 4 2010.

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