



Supplement of

Can we use modelling methodologies to assess airborne benzo[*a*]pyrene from biomonitors? A comprehensive evaluation approach

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23 Pine needles analysis and quantification

24

25 Duplicate samples of 5 g of needles underwent ultrasonic extraction (USE) with a mixture of
26 hexane:dichloromethane (1:1) as solvent and were subsequently cleaned-up using 5g alumina
27 solid-phase extraction (SPE) cartridges from International Sorbent Technology (Mid
28 Glamorgan, UK), using the same solvent for elution. After blowing down to dryness and
29 solvent change to hexane, chromatographic analysis of BaP was done in a Varian CP-3800
30 gas chromatograph (Lake Forest, CA, USA) coupled to a Varian 4000 mass spectrometer in
31 Portugal and a Trace GC 2000 Series gas chromatograph from ThermoQuest (Waltham, MA,
32 USA) coupled to a Finnigan Trace MS 2000 Series mass spectrometer in Spain. However,
33 the operation was similar in both cases, namely using electron impact ionization (70 eV), a
34 J&W Scientific (Folsom, CA, USA) 30 m × 0.25 mm I.D. DB-5 column coated with 5%
35 diphenylpolydimethylsiloxane (film thickness 0.25 µm) and the same oven temperature
36 program. The injector, transfer line and ion source temperatures were also the same (280, 250
37 and 200 °C, respectively). Finally, the acquisition was made in single ion monitoring (SIM)
38 mode using deuterated PAHs as surrogate standards. BaP was identified and quantified using
39 retention time and up to three ions, with perylene-d₁₂ acting as surrogate standard and
40 anthracene-d₁₀ as internal standard to look for GC-MS errors.

41 Linear behaviour between 0.01 and 1 mg L⁻¹ and good chromatographic resolution was
42 obtained for BaP, with a limit of detection below 0.10 ng g⁻¹ (dry weight). The BaP
43 concentrations were calculated in dry weight, after determining the water content of the
44 needles for each species (Table S1). This information is needed for the estimates of air
45 concentrations from the levels found in pine needles, as detailed below.

46

47 Table S1. Characteristics of the four pine needle species employed in this study.

	<i>P. pinea</i>	<i>P. pinaster</i>	<i>P. halepensis</i>	<i>P. nigra</i>
Mean mass of one needle (g)^a	0.06	0.13	0.018	0.035
Mean surface area (m²×10⁻⁶)^a	545	815	254	366
Lipid content (mg g⁻¹, dw)	121.95	182.93	105.56	104.26
Water content (% mass)	59	59	46	53

48 ^a Data taken from Daligault (1991) and Moro (2006)

49

50 **Modelling experiment**

51

52 Table S2. Set of parameterisations used in the WRF+CHIMERE modelling system

WRF	CHIMERE
Microphysics → WSM3	Chemical Mechanisms → MELCHIOR2
PBL → Yonsei University	Aerosol chemistry → Inorganic (thermodynamic equilibrium
Radiation → CAM	with ISORROPIA) and organic (MEGAN SOA scheme)
Soil → Noah LSM	aerosol chemistry
Cumulus → Kain-Fritsch	Natural aerosols → dust, re-suspension and inert sea-salt
	BC → LMDz-INCA+GOCART

53

54 The Advanced Research Weather Research and Forecasting (WRF-ARW) Model v3.1.1
 55 (Klemp et al., 2007; Skamarock et al., 2008) is used to provide the meteorology to the
 56 chemistry transport models. WRF is a fully compressible, Eulerian non-hydrostatic model
 57 that solves the equations that govern the atmospheric motions. 33 vertical layers on sigma
 58 coordinates cover from the ground level up to 10 hPa. Microphysical processes are treated
 59 using the single-moment 3-class scheme described in Hong et al. (2004). The sub-grid-scale
 60 effects of convective and shallow clouds are resolved by a modified version of the Kain-
 61 Fritsch scheme based on Kain and Fritsch (1993). The Noah land surface model was used to
 62 solve the soil processes on 4 layers to a depth of 2m (Chen and Dudhia, 2001a; 2001b). The
 63 vertical sub-grid-scale fluxes caused by eddy transport in the atmospheric column are
 64 resolved by the Yonsei University non-local planetary boundary layer scheme (Noh et al.,
 65 2003). Finally, radiation was treated through the Community Atmospheric Model (CAM) 3.0
 66 radiation scheme (Collins et al., 2006).

67 WRF was coupled off-line to CHIMERE. Atmospheric concentrations of BaP have been
 68 calculated using CHIMERE chemistry transport model (v2008b), coupled off-line to WRF
 69 outputs and EMEP emissions. This CHIMERE version includes gaseous and particulate BaP
 70 and its degradation by OH radicals, which represents over 99% of the degradation path for
 71 BaP (Bieser et al., 2012). For further details on the model options, the reader is referred to
 72 Menut et al. (2013). MELCHIOR2 gas-phase mechanism is implemented within CHIMERE.

73 The chemistry transport model includes aerosol and heterogeneous chemistry; distinguishes
74 among different chemical aerosol components, namely nitrate, sulphate, ammonium,
75 elemental and organic carbon with three subcomponents (primary, secondary anthropogenic
76 and secondary biogenic) and marine aerosols. Unspecified primary anthropogenic aerosols
77 and aerosol water are additionally kept as separate components. The model considers the
78 thermodynamic equilibrium using the ISORROPIA model (Nenes et al., 1998). Last, the
79 aerosol microphysical description for CHIMERE is based on a sectional aerosol module
80 including 6 bins from 10 nm to 40 μm using a geometrical progression.

81 In the present work, simulations covered the period 2006-2010. Initial and boundary
82 conditions for WRF were provided by ERA-Interim reanalysis (Dee et al., 2011), while for
83 CHIMERE, the global climate chemistry model LMDz-INCA2 was used (96 x 72 grid cells,
84 namely 3.75° x 2.5° in longitude and latitude, with 19 sigma-p hybrid vertical levels, Szopa et
85 al. (2009) developed by the Laboratoire des Sciences du Climat et l'Environnement (LSCE).
86 Climatic monthly mean data are interpolated in the horizontal and vertical dimensions to
87 force the major chemical concentrations at the boundaries of the domain. A detailed
88 description of the INteractive Chemistry and Aerosol (INCA) model is presented by
89 Hauglustaine et al. (2004) and Folberth et al. (2006). Because the contribution of long-range
90 transport on ground level concentrations (those considered in this work) can be considered as
91 negligible, the influence of using climatological boundary conditions is limited and
92 overwhelmed by local processes.

93 Anthropogenic emissions for the entire period of simulations are derived from the EMEP
94 database (Vestreng et al., 2009) and disaggregated to the working resolution following spatial
95 proxy data, according to the methodology stated in Pay et al. (2010). For BaP emissions, data
96 have been obtained from the EMEP-MSCEAST web site (<http://www.msceast.org>). The
97 accuracy of simulations depends strongly on emission data and unfortunately there are strong
98 uncertainties in BaP emissions, by a factor of 2 to 5 (San José et al., 2013). According to
99 these authors, the main source of BaP is incomplete combustion processes of organic
100 material, in particular wood and coal in private households. Industrial heating and cookeries
101 as well as road traffic are also large sources of BaP, which is emitted in particle phase.

102 Natural emissions (of sea salt and dust) depend on meteorological conditions, and
103 consequently they are coupled hourly to WRF meteorological outputs. Biogenic emissions
104 were generated dynamically using MEGAN (Model of Emissions of Gases and Aerosols
105 from Nature) (Guenther et al., 2006) with the parameterized form of the canopy environment

106 model. The model estimates hourly isoprene, monoterpene, and other BVOC emissions based
107 on plant functional type and as a function of hourly temperature and ground level shortwave
108 radiation from WRF.

109

110 **Model validation**

111

112 EMEP stations are located at a minimum distance of approximately 10 km from large
113 emission sources and thus assumed to fit the resolution of the model used for regional
114 background concentrations (Torseth et al., 2012). Thus, as reported by Ratola and Jiménez-
115 Guerrero (2015), results from the EMEP monitoring data were used to characterize the ability
116 of the model to reproduce present air BaP levels and variability. The available stations
117 running in the Iberian Peninsula in the 2006-2010 time frame were: Niembro (2006-2010),
118 Campisabalos (2007-2008), O Saviñao (2007), Víznar (2008-2010), Peñausende (2008-
119 2009), Barcarrota (2008), Zarra (2008), San Pablo de los Montes (2009-2010), Mahón (2010)
120 and Els Torms (2010). In all of them, BaP measurements are available as weekly or monthly
121 averages. The results have been compared to the available periods for observations. Although
122 it was possible to find some data from air monitoring stations from the Generalitat de
123 Catalunya and the Comunitat Valenciana, not all of them presented climatologically
124 representative series. Thus, also to maintain a wider geographical coverage with under the
125 same sampling and analytical framework to ensure the homogeneity of the data.

126 For the evaluation of canopy deposition and atmospheric concentrations, a number of
127 statistical parameters have been selected (Figure S1). Spatial correlation coefficient (r), root
128 mean square error (RMSE) and mean bias (MB) values are commonly used by the modelling
129 community and have therefore been selected according to the criteria of Pay et al. (2010),
130 who use them to evaluate a modelling system for Europe. Moreover, Boylan and Russell
131 (2006) suggest that the mean normalised bias error (MNBE) for each model-observed pair by
132 the observation is a useful parameter, but may not be appropriate for evaluating particulate
133 matter and their components. These authors suggested the mean fractional bias (MFB) and
134 the mean fractional error (MFE) instead, indicating that model performance goal would be
135 met when both the MFE and MFB are less than or equal to 50% and $\pm 30\%$, respectively, and
136 the model performance criterion when $MFE \leq 75\%$ and $MFB \leq \pm 60\%$. These criteria and
137 goals have been selected to provide the metrics for the WRF+EMEP+CHIMERE evaluation

138 of BaP. Annual and seasonal mean statistics are computed, with seasons corresponding to
 139 December, January and February (DJF, winter), March, April and May (MAM, spring), June,
 140 July and August (JJA, summer) and September, October and November (SON, autumn).

141

$$\begin{array}{ll}
 \text{MOD MEAN} & \frac{1}{N} \sum C_{mod} \\
 \text{(modelled concentrations)} & \\
 \text{OBS MEAN} & \frac{1}{N} \sum C_{obs} \\
 \text{(pine needle concentrations)} & \\
 \text{BIAS} & \frac{1}{N} \sum (C_{mod} - C_{obs}) \\
 \\
 \text{RMSE} & \sqrt{\sum \frac{(C_{mod} - C_{obs})^2}{N}} \\
 \text{(root mean square error)} & \\
 \text{MFB} & \frac{1}{N} \sum \left(\frac{(C_{mod} - C_{obs})}{\left(\frac{C_{mod} + C_{obs}}{2}\right)} \right) \\
 \text{(mean fractional bias)} &
 \end{array}$$

142

143 Figure S1. Main statistical parameters used in model validation

144 As our aim is to have the best approximation of atmospheric BaP levels through modelling
 145 procedures, to serve as a reference pseudo-reality to estimate the most accurate vegetation-to-
 146 air conversion method, the multiplicative ratio bias-correction technique has been applied
 147 following the methodology of Borrego et al. (2011). The correction factor is calculated as the
 148 quotient between the additions of observed and modelled concentrations at a particular hour
 149 of the n previous days. Borrego et al. (2011) and Monteiro et al. (2013) recommend a four-
 150 day training period ($n=4$). However, given the limited availability of EMEP data (only on a
 151 weekly basis), a four-week training period has been chosen here instead as a compromise
 152 between having a sufficiently long timeframe to gather adequate statistics but not as much as
 153 to mask seasonal variations.

154

155 **Results**

156 Table S3. Parameters of the modelled deposition over vegetal canopies evaluated against
 157 observations compiled from pine needles, for all the sampling points (mean concentrations in
 158 ng g^{-1}).

SITE	LATITUDE	LONGITUDE	PINE SPECIES	BIAS	MFB	OBS. MEAN	MOD. MEAN
Alcolea de Cinca	42.03	-1.56	<i>Pinus pinea</i>	-0.63	-95.41%	0.98	0.35
Alcoutim	37.47	-7.47	<i>Pinus pinea</i>	0.11	26.63%	0.81	0.92
Antuã 1	40.69	-8.52	<i>Pinus pinea</i>	-0.17	24.18%	2.71	2.53
Barcelona	41.39	2.11	<i>Pinus pinea</i>	-2.53	-105.46%	3.66	1.13
Beja	38.01	-7.87	<i>Pinus pinea</i>	-0.29	20.86%	1.02	0.73
Braga	41.56	-8.40	<i>Pinus pinea</i>	0.71	31.72%	0.96	1.67
Castelo Branco	39.83	-7.50	<i>Pinus pinea</i>	0.60	31.72%	0.81	1.41
Coimbra	40.21	-8.42	<i>Pinus pinea</i>	0.54	32.59%	0.62	1.16
El Bocal	41.57	-0.69	<i>Pinus pinea</i>	-0.49	-33.85%	1.71	1.21
El Prat	41.30	2.10	<i>Pinus pinea</i>	-0.38	-16.77%	2.44	2.06
Évora	38.58	-7.91	<i>Pinus pinea</i>	-1.13	6.74%	1.33	0.21
Faro	37.02	-7.94	<i>Pinus pinea</i>	-1.53	7.34%	1.85	0.32
Leiria	39.75	-8.80	<i>Pinus pinea</i>	0.34	29.56%	0.76	1.10
Lisboa	38.72	-9.14	<i>Pinus pinea</i>	-4.73	5.32%	5.37	0.64
Loulé	37.13	-8.10	<i>Pinus pinea</i>	-1.90	10.17%	2.56	0.65
Maleján	41.82	-1.55	<i>Pinus pinea</i>	-0.77	-91.95%	1.22	0.45
Miranda de Ebro 1	42.68	-2.95	<i>Pinus pinea</i>	-0.25	-70.21%	0.49	0.23
Monteagudo	41.96	-1.69	<i>Pinus pinea</i>	-0.34	-26.47%	1.46	1.12
Movera	41.64	-0.80	<i>Pinus pinea</i>	-0.01	-0.61%	1.22	1.21
Outão	38.49	-8.98	<i>Pinus pinea</i>	2.11	35.21%	1.53	3.64
Portalegre	39.30	-7.43	<i>Pinus pinea</i>	-0.01	24.89%	1.24	1.23
Porto 1	41.18	-8.60	<i>Pinus pinea</i>	1.08	31.13%	1.66	2.74
Praia Verde	37.18	-7.48	<i>Pinus pinea</i>	-0.22	17.50%	0.47	0.25
Quintãs 1	40.58	-8.63	<i>Pinus pinea</i>	0.80	33.80%	0.74	1.53
Santarém	39.24	-8.69	<i>Pinus pinea</i>	-0.73	16.55%	1.44	0.71
Sines	37.96	-8.81	<i>Pinus pinea</i>	0.03	25.51%	0.75	0.78
Souselas	40.29	-8.41	<i>Pinus pinea</i>	1.58	29.94%	3.20	4.78
Torres de Segre	41.54	0.51	<i>Pinus pinea</i>	-0.11	-7.74%	1.46	1.35
Vic	41.94	2.25	<i>Pinus pinea</i>	-0.71	-21.37%	3.66	2.95
Villodas	42.83	-2.78	<i>Pinus pinea</i>	1.91	98.82%	0.98	2.88
Antuã 2	40.69	-8.52	<i>Pinus pinaster</i>	-0.67	22.50%	3.71	3.03
Bragança	41.81	-6.76	<i>Pinus pinaster</i>	0.23	26.96%	1.37	1.60
Caminha	41.87	-8.86	<i>Pinus pinaster</i>	0.54	29.23%	1.33	1.87
Estarreja	40.77	-8.57	<i>Pinus pinaster</i>	1.34	31.68%	1.83	3.17
Fóia	37.31	-8.61	<i>Pinus pinaster</i>	0.84	35.29%	0.60	1.44
Guarda	40.54	-7.27	<i>Pinus pinaster</i>	0.66	29.41%	1.55	2.21
Leça	41.22	-8.71	<i>Pinus pinaster</i>	-0.63	23.80%	6.85	6.22
Mirandela	41.37	-7.14	<i>Pinus pinaster</i>	-1.14	18.88%	2.89	1.76
Porto 2	41.18	-8.60	<i>Pinus pinaster</i>	1.20	28.27%	3.66	4.86
Quintãs 2	40.58	-8.63	<i>Pinus pinaster</i>	-0.14	24.13%	2.07	1.93
Rio de Onor	41.94	-6.61	<i>Pinus pinaster</i>	0.73	31.06%	1.14	1.87
Torre	40.31	-7.58	<i>Pinus pinaster</i>	0.32	29.64%	0.71	1.03
Vide	40.29	-7.78	<i>Pinus pinaster</i>	1.19	65.60%	1.22	2.41
Vila Real	41.30	-7.74	<i>Pinus pinaster</i>	2.17	32.42%	2.57	4.74
Arazuri	42.81	-1.72	<i>Pinus nigra</i>	0.14	20.40%	0.64	0.78
Briñas	42.59	-2.84	<i>Pinus nigra</i>	1.30	75.67%	1.06	2.36
La Bordeta	41.60	0.62	<i>Pinus nigra</i>	-0.32	-117.75%	0.43	0.11
Miranda de Ebro 2	42.67	-2.09	<i>Pinus nigra</i>	-0.10	-27.59%	0.43	0.32
Nestares	43.00	-4.15	<i>Pinus nigra</i>	0.00	-0.10%	0.43	0.43
Urdiáin	42.90	-2.14	<i>Pinus nigra</i>	0.61	83.80%	0.43	1.04
Amposta	40.72	0.58	<i>Pinus halepensis</i>	-0.60	-39.43%	1.83	1.23
Andosilla	42.37	-1.94	<i>Pinus halepensis</i>	0.38	29.17%	1.10	1.48
Caldearenas	42.40	-0.50	<i>Pinus halepensis</i>	0.01	3.14%	0.37	0.38
Cascante	41.98	-1.68	<i>Pinus halepensis</i>	-0.44	-63.19%	0.92	0.48
Cuarte de Huerva	41.61	-0.92	<i>Pinus halepensis</i>	-0.33	-21.98%	1.65	1.32

159 Table S3. (cont.) Parameters of the modelled deposition over vegetal canopies evaluated
 160 against observations compiled from pine needles, for all the sampling points.
 161

SITE	LATITUDE	LONGITUDE	PINE SPECIES	BIAS	MFB	OBS. MEAN	MOD. MEAN
Deltebre	40.71	0.71	<i>Pinus halepensis</i>	-0.58	-37.70%	1.83	1.25
Estella/Lizarra	42.67	-2.03	<i>Pinus halepensis</i>	1.40	97.49%	0.73	2.13
Flix	41.23	0.55	<i>Pinus halepensis</i>	0.07	11.87%	0.55	0.62
Grisén	41.73	-1.18	<i>Pinus halepensis</i>	-1.22	-39.69%	3.67	2.45
Logroño 1	42.47	-2.44	<i>Pinus halepensis</i>	-0.44	-35.19%	1.47	1.03
Logroño 2	42.67	-2.42	<i>Pinus halepensis</i>	1.60	34.34%	3.85	5.45
Mollerussa	41.62	0.91	<i>Pinus halepensis</i>	-0.72	-77.74%	1.28	0.57
Puente La Reina	42.67	-1.82	<i>Pinus halepensis</i>	0.79	60.19%	0.92	1.71
San Adrián	42.33	-1.93	<i>Pinus halepensis</i>	-0.11	-8.91%	1.28	1.17
Sástago	41.32	-0.34	<i>Pinus halepensis</i>	-0.39	-72.55%	0.73	0.34
Tornabous	41.69	1.05	<i>Pinus halepensis</i>	-0.41	-77.45%	0.73	0.32
Tortosa	40.80	0.51	<i>Pinus halepensis</i>	-0.30	-16.01%	2.02	1.72
Tudela 1	42.07	-1.60	<i>Pinus halepensis</i>	-0.89	-35.90%	2.94	2.04
Tudela 2	42.08	-1.62	<i>Pinus halepensis</i>	-0.41	-25.29%	1.83	1.42
Villanueva de Gállego	41.77	-0.82	<i>Pinus halepensis</i>	-0.74	-31.16%	2.75	2.01

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