



## Modelling of atmospheric concentrations of fungal spores: a two-year simulation over France using CHIMERE.

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### Abstract

Fungal spore organic aerosol emissions have been recognised as a significant source of particulate matter as PM<sub>10</sub>; however, they are not widely considered in current air quality models. In this work, we have implemented the parameterisation of fungal spore organic aerosol (OA) emissions introduced by Heald and Spracklen (2009) (H&S) and further modified by Hoose et al. (2010) in the CHIMERE regional chemistry-transport model. This simple parameterisation is based on two variables, leaf area index (LAI) and specific humidity. We have validated the geographical and temporal representativeness of this parameterisation on a large scale by using yearly polyol observations and primary biogenic organic aerosol factors from PMF analysis at 11 French measurement sites. For a group of sites in northern and eastern France, the seasonal variation of fungal spore emissions, displaying large summer and small winter values, is correctly depicted. However, the H&S parameterisation fails to capture fungal spore concentrations for a smaller group of Mediterranean sites with less data availability both in terms of absolute values as well as seasonal variability, leading to strong negative biases especially during the autumn and winter seasons occur. Two years of CHIMERE simulations with the H&S parameterisation have shown a significant contribution of fungal spore OA to PM<sub>10</sub> mass, lower than 10 % during winter, and reaching up to 20 % during summer in high emission zones, especially over large forested areas. In terms of contribution to organic matter (OM) concentrations, the simulated fungal spore contribution in autumn is as high as 40 % and reaches at most 30 % of OM for other seasons. As a conclusion, the fungal spore OA contribution to total OM concentrations is shown to be substantial enough to be considered as a major PM<sub>10</sub> fraction and shall then be included in state-of-the-art chemistry transport models. The H&S parameterisation shows satisfactory results over northern and eastern France, but may underestimate concentrations for Mediterranean areas that may indicate missing factors influencing emissions or a missing source of spores.



## 48 1. Introduction

50 Modelling of the organic matter (OM) fraction of PM<sub>10</sub> chronically  
51 underestimates in situ observations (Ciarelli et al., 2016; Pai et al., 2020). This  
52 underestimation can be attributed to several causes such as the complexity of the  
53 organic matter composition, which is not yet fully known, incomplete emission  
54 inventories or their inherent uncertainties, and poorly parametrised atmospheric  
55 chemical transformations.

56 It is therefore important to assess whether the primary source of organic  
57 aerosol, currently not considered in many models, can help to improve atmospheric  
58 aerosol modelling. Primary biogenic organic aerosols (PBOA) are mainly composed of  
59 microorganisms such as bacteria, fungi, fungal or bacterial spores, pollens or viruses  
60 and biological fragments such as plant debris or microbes (Després et al., 2012;  
61 Fröhlich-Nowoisky et al., 2016; Jaenicke et al., 2007). Their size varies from less than  
62 0.3 µm for viruses to about 100 µm for pollens (Després et al., 2012; Jones and  
63 Harrison, 2004; Shaffer and Lighthart, 1997). When looking at atmospheric particles  
64 with an aerodynamic diameter of less than 2.5 or 10 µm (which are the fractions  
65 routinely measured and studied for health risk assessment), it is possible to find  
66 viruses, bacteria (agglomerated or not) and spores; however spores, when produced  
67 by fungi, represent the major fraction in terms of mass (Elbert et al., 2007).

68 More specifically, fungal spores are emitted directly into the atmosphere during  
69 the fungal reproduction process when temperature and humidity conditions are  
70 favourable, but their emission can also be triggered by wind and rain (Elbert et al.,  
71 2007; Huffman et al., 2013; Jones and Harrison, 2004). Previous studies estimated  
72 that fungal spores can contribute to around 5 % and 10 % of the mass of respectively  
73 PM<sub>10</sub> and organic carbon, in urban and suburban areas (Bauer et al., 2002, 2008b). In  
74 specific environments such as tropical forests, the contribution of fungal spores can  
75 represent 45 % of the PM<sub>10</sub> mass (Elbert et al., 2007).

76 Fungal spores are susceptible to cause major health problems such as asthma,  
77 pulmonary obstruction, tuberculosis, meningitis and legionellosis (Douwes et al., 2003;  
78 Eduard et al., 2012; Fröhlich-Nowoisky et al., 2016; Ghosh et al., 2015; Pearson et al.,  
79 2015; Samaké et al., 2017). Some studies on PBOA have shown that aerosols emitted  
80 directly by fungi in the form of spores contribute significantly to the oxidative potential  
81 of aerosols (Samaké et al., 2017). Moreover, based on a positive matrix factorisation  
82 (PMF) analysis, Weber et al. (2021) derived a primary biogenic factor based on a large  
83 data set of speciated PM<sub>10</sub> aerosol measurements over France, including polyol  
84 measurements as a tracer for fungal spores. They found a high intrinsic oxidative  
85 potential by dithiothreitol (DTT) for this factor, equal to that of biomass burning, but  
86 lower than that of primary traffic emissions.

87 Literature review shows several parameterisations suitable for use of modelling  
88 primary biogenic aerosols emissions from fungal spores in the PM<sub>10</sub> size range in  
89 chemistry transport models. Samaké et al. (2019a) identified the parameters  
90 responsible for the variability for up to 82 % of the annual variability of polyols as a tracer  
91 of fungal spores for a temperate latitude site in an alpine environment, using multi-  
92 linear approaches. These variations were mainly explained by the mean night-time  
93 temperature (54 %) and LAI (37 %), and to a lesser extent by the atmospheric humidity  
94 (3 %) and the wind speed (2 %). The combined factor of LAI and wind speed explains  
95 the remaining variability (4 %). A first parameterisation for the treatment of fungal



96 scores in atmospheric models was proposed by Heald and Spracklen (2009) (H&S)  
98 and modified by Hoose et al. (2010). It estimates fungal spore emissions as a linear  
100 function of leaf area index (LAI) and specific humidity. In this formulation, the LAI is a  
102 proxy for the vegetation density and the specific humidity is a proxy for the water  
104 availability, but is also related temperature. The parameterisation proposed by Sesartic  
106 and Dallafor (2011) (S&D) suggests a different approach by varying emissions as a  
108 function of soil types, not relying on LAI, and therefore removing the seasonality  
inherently present in the H&S parameterisation. Hummel et al. (2015) compared these  
parameterisations across Europe and developed a new statistical model, based on the  
H&S parameterisation using LAI and specific humidity, to also include a linear  
dependence with temperature, and a threshold below which emissions are assumed  
to be zero.

In Hummel et al. (2015), the concentrations simulated with three  
parameterisations of H&S, S&D and Hummel were compared to measurements of  
fluorescent biological aerosol particles (FBAP) at four sites in several parts of Europe  
(Germany, Finland, UK, Ireland) for almost weekly time periods in July, August and  
October of 2010. This comparison was carried out using 1,536 hourly data points,  
1,200 of which came from the German and Finnish stations, each with 600 data points.  
At these two sites, one week in July, one week in August and 10 days in September  
were measured, unlike the UK and Irish sites, where the data was taken only for August  
2010. FBAP measurements are taken as a proxy for fungal spore emissions. By  
construction, the S&D parameterisation does not reproduce the observed daily and  
seasonal variability, while it is known that fungal spore emissions display a general  
summer maximum across Europe (Samaké et al., 2019a, b). On the contrary, the H&S  
and Hummel parameterisations include these temporal variations and therefore show  
better correlations with measured concentrations ( $R = 0.43$ ) compared to the S&D  
approach ( $R = -0.05$ ). The parameterisation by Hummel et al. (2015) showed a lower  
normalised mean bias (NMB = -43 %) compared to the H&S one (NMB = -44 %).

As fungal spores make a significant contribution to  $PM_{10}$  and are rarely included  
in chemistry transport models (CTM), the aim of our study is to integrate them into the  
state-of-the-art Chemistry Transport Model CHIMERE (Menut et al., 2021), to evaluate  
the model performance with field measurements, and to infer the spatio-temporal  
variability of their occurrence. This could lead to improved modelling of  $PM_{10}$   
concentrations, of organic matter, and of other pollutants such as secondary biogenic  
compounds or even oxidative potential. This study will focus on France, displaying one  
of the largest database of chemically speciated PM measurements in Europe (Favez  
et al., 2021). Interestingly, France has a wide range of climatic variability (oceanic,  
semi-oceanic, continental, mountainous, Mediterranean), making it possible to  
compare fungal spore modelling results under various climatic conditions. To assess  
the modelling of fungal spores, measurements of polyols were used, specifically  
mannitol and arabitol, since many studies indicate that they are specific tracers of this  
PBOA fraction (Bauer et al., 2008a; Gosselin et al., 2016; Samaké et al., 2019a).  
Furthermore, we compared our CTM results to the concentrations of organic matter  
ascribed to this primary biogenic source using the receptor model Positive Matrix  
Factorisation (PMF) in previous work.

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## 2. Material & methods

### 146 2.1. Observations

#### 148 2.1.1. PM<sub>10</sub> and Organic matter measurements

150 The PM<sub>10</sub> mass concentration data have been obtained from continuous  
152 measurements performed by French regional air quality monitoring networks (AQMN).  
154 A total of 699 air quality stations performed measurements in metropolitan France  
156 during the period of the study, restricted to 2013 and 2014, including fixed and mobile  
158 stations. These observations have been achieved by AQMN using two types of  
160 automated analysers during this period: tapered element oscillating microbalances  
162 equipped with filter dynamic measurement systems (TEOM-FDMS, Thermo Scientific),  
164 and beta radiation absorption analysers (Met One BAM 1020 and ENVEA MP101M).  
166 These measurements have been conducted in accordance with standard procedures  
168 described in CSN EN 16450. As described by Favez et al. (2021), the aim of the  
170 aerosol characterisation program (CARA) is to develop knowledge of the chemical  
composition and contribution of atmospheric particle sources. This work is enriched by  
research programmes, with data from some of them being used in this study. In CARA  
and other programs, the chemical analysis of (PM<sub>10</sub>) filter samples has been performed  
following relevant European standard methods. Briefly, for datasets used herein,  
organic carbon was measured by thermo-optical analysis using the EUSAAR2 protocol  
(Cavalli et al., 2010). Sugars were measured by liquid chromatography using pulsed  
amperometric detection (Verlhac et al., 2013; Yttri et al., 2015). The measurement  
protocols have been detailed in previous studies (Samaké et al., 2017, 2019a, b;  
Weber et al., 2021). The analysed species include mannitol and arabitol, which  
currently make up for a large fraction of organic sugars (Elbert et al., 2007) and are  
used as a tracer for fungal spore emissions.

172 In summary, for the datasets used in the present study, PM<sub>10</sub> organic matter  
174 observations were performed at 13 different stations for a total of 2,227 daily filter  
176 samples, including 1,497 data on polyols on 11 sites. The locations of these sites are  
illustrated in Figure 1, while Table 1 provides details on the number of data points  
available per station and their temporality.

#### 178 2.1.2. OC apportionment based on filter samples

180 Positive Matrix Factorisation (*PMF*) is one of the most widely used techniques  
182 for identifying factors contributing to aerosol concentrations using field measurements  
(Belis et al., 2020; Hopke et al., 2020; Karagulian et al., 2015; Paatero and Tapper,  
1994). This receptor model commonly uses off-line chemical speciation measured on  
filters and factor-specific tracers as input data. The correlation matrices allow the  
184 identification of the species co-emitted with the tracers and thus determine the  
contribution of the factors to the PM<sub>10</sub> concentrations. For this study, *PMF* analysis  
186 were previously performed with a harmonised methodology (Weber et al., 2021),  
providing source apportionment results for a total number of 842 daily samples  
188 collected at 7 sites from early 2013 to the end of 2014. *PMF* results at all sites include  
a factor which can be attributed to PBOA because of the large concentrations of the  
190 two polyols in this factor, representing more than 90 % of them. However, this PBOA  
factor may also contain biogenic secondary organic aerosols (BSOA) since it is  
192 sometimes associated with BSOA tracers, such as 3-MBTCA (resulting from  $\alpha$ -pinene



194 oxidation) or 2-MTs (resulting from oxidation of isoprene) (Borlaza et al., 2021).  
 195 Therefore, we propose here to use the PBOA factor as an upper boundary for fungal  
 196 spore concentrations (see section 3.2).

198 **Table 1 : Summary of organic matter and polyols filter-based observations as well as primary biogenic**  
 199 **factor derived from PMF analysis available for this study over the years 2013 and 2014 at different French**  
 200 **sites (within the PM<sub>10</sub> fraction). The measurement period and geographical coordinates are also indicated.**

Stations	Coordinates (latitude ; longitude)	Measurement period	PMF	Polyols
Aix-en-Provence	43.53 ; 5.44	18.07.2013 – 13.07.2014	56	117
Andra-OPE	48.55 ; 5.46	01.01.2013 – 29.12.2014	/	98
Grenoble	45.16 ; 5.74	02.01.2013 – 29.12.2014	237	238
Lens	50.44 ; 2.83	05.04.2013 – 26.09.2014	167	138
Marseille	43.30 ; 5.39	01.06.2014 – 31.12.2014	/	95
Nice	43.70 ; 7.29	04.06.2014 – 31.12.2014	77	89
Nogent-sur-Oise	49.28 ; 2.48	02.01.2013 – 31.12.2014	155	220
Port-de-Bouc	43.40 ; 4.98	01.06.2014 – 31.12.2014	79	80
Revin	49.91 ; 4.63	02.01.2013 – 26.09.2014	/	168
Roubaix	50.71 ; 3.18	20.01.2013 – 08.09.2014	/	159
Strasbourg	48.59 ; 7.74	02.04.2013 – 31.12.2014	71	95

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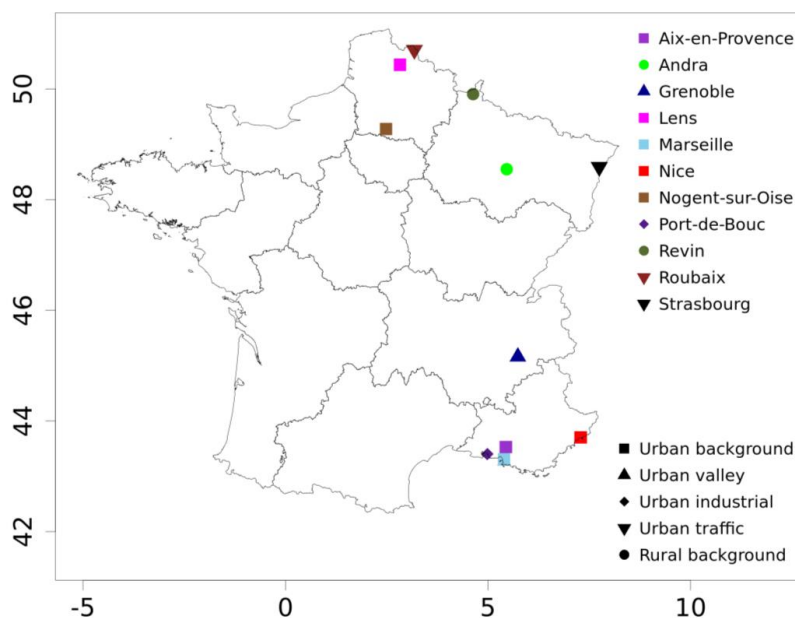


Figure 1 : Location and type of sites for PM<sub>10</sub> organic matter and polyol measurements from filters as well as primary biogenic from PMF over the years 2013 and 2014.

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206 Sites are distributed over different geographical areas (Figure 1) in in the  
northeast and southeast of France, including cities from the Channel region (Lens,  
208 Roubaix, Nogent sur Oise) to the German border (Strasbourg), remote rural sites  
located in between (Revin and Andra-OPE) as well as sites an Alpine urban station  
210 (Grenoble) and sites near the Mediterranean Sea (Aix-en-Provence, Marseille, Nice,  
Port-de-Bouc). These sites are classified as rural background (Andra-OPE, Revin),  
212 urban background (Aix-en-Provence, Grenoble, Lens, Marseille, Nogent-sur-Oise,  
Nice, Petit Quevilly, Talence), traffic sites (Roubaix, Strasbourg), urban industrial (Port-  
214 de-Bouc). It is thought that the varied characteristics of the observational sites can give  
us an unprecedented possibility of evaluation of the simulated spore emissions and  
216 concentrations.

## 218 2.2. Regional Modelling

### 220 2.2.1. The chemistry transport model CHIMERE

CHIMERE is an eulerian state-of-the-art regional chemistry transport model  
222 (Menut et al., 2021). It is used operationally by the French platform PREV'AIR (Rouil  
et al., 2009) and the Copernicus Atmospheric Modelling System (CAM5) (Marécal et  
224 al., 2015) to forecast and monitor air quality. The version v2020r3 of CHIMERE has  
been used in this work (Menut et al., 2021).

226 The EMEP anthropogenic emissions inventory with a resolution of 10 km<sup>2</sup>  
provides input data for anthropogenic emissions based on the methodology described  
228 in Vestreng (2003). Biogenic VOC emissions are computed by CHIMERE based on  
the Model of Emissions and Gases and Aerosols from Nature MEGAN 2.1 algorithm  
230 (Guenther et al., 2012). The gas phase chemistry is provided by the Melchior2  
mechanism (Derognat et al., 2003). The ISORROPIA thermodynamic model is used to  
232 compute the formation of inorganic aerosols based on the approach described in  
Fountoukis and Nenes (2007). For organic aerosol formation and volatilisation of  
234 primary organic aerosol, the volatility basis set (VBS) for the organic species as  
described in Cholakian et al. (2018) was activated.

236 Chemical boundary conditions with a 3-hour temporal resolution are from the  
CAM5 project (Marécal et al., 2015), together with the chemical fields for the model  
238 upper boundary at the 500 hPa level. The WRF model is used for meteorological  
forcing (Skamarock et al., 2008). For the emissions of biogenic volatile organic  
240 compounds (VOC) as well as for the parameterisation of the emissions of primary  
organic aerosols, we use the LAI (Leaf Area Index) obtained from the observations of  
242 the MODIS instrument with a frequency of 8 hours and a native resolution of 30  
seconds for each year (Sindelarova et al., 2014). The simulation has been carried out  
244 during years 2013 and 2014 on a western European domain, with a 9 x 9 km<sup>2</sup> horizontal  
resolution. It is run on 9 vertical hybrid levels from ground to an upper height of 500  
246 hPa, the height of the first layer being around 20 meters.

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**Table 2 : Parameterisations initially considered for the present work.**

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Name in this work	Variables	Reference
H&S	Leaf area index (LAI), specific humidity ( $q_v$ )	(Heald and Spracklen, 2009; Hoose et al., 2010)
S&D	Land use classes	(Sesartic and Dallafor, 2011)
Hummel	Leaf area index (LAI), specific humidity ( $q_v$ ), surface temperature (T)	(Hummel et al., 2015)
Janssen statistical	Leaf area index (LAI), specific humidity ( $q_v$ ), and wind friction speed ( $u^*$ )	(Janssen et al., 2021)

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### 2.2.2. Parameterisation of fungal spore OA emissions

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262 In the Introduction section, we have presented several parameterisations of  
 264 fungal spores, and which are listed in Table 2. Among the three of them (H&S, S&D  
 266 and Hummel) compared by Hummel et al. (2015) to observations, the S&D  
 268 parameterisation showed the worst statistical agreement, and also is based on  
 270 seasonally fixed land-use parameters. It was therefore discarded. Among the two  
 272 better performing parameterisations, we preferred the H&S parameterisation. This is  
 274 because in Hummel's approach, the inclusion of a temperature-dependent and  
 276 vegetation-independent term leads to significant fungal spore emissions under high  
 278 temperature conditions even at places where LAI is small and therefore no large  
 280 emissions are expected. This yields to large emissions especially over Southern  
 Europe which are not confirmed by measurements. Finally, two recent  
 parameterisations by Janssen et al. (2021) have been developed over the eastern  
 United States using measurements of spore concentrations consider LAI, specific  
 humidity and wind friction velocity in the first case, and a spore population model in the  
 second. Comparisons with annual measurements of fluorescent primary organic  
 aerosols at German, Finnish and Colorado sites show similar correlations between  
 these two parameterisations and that of H&S (Janssen et al., 2021). With respect to  
 these simulations, we preferred the simpler H&S parameterisation. This  
 parameterisation was integrated in our simulations for its robustness at different sites  
 and it has been set-up specifically for temperate latitude European conditions.

282 Equation 1 shows the fungal spore emission flux  $F_{H\&S}$  (unit: number of spores  
 284  $m^{-2} s^{-1}$ ) varying as a function of leaf area index **LAI** and specific humidity  $q_v$ . The  
 constant  $c$ , equal to 2315, introduced by Hoose et al. (2010) accounts for fungal spore  
 emission fluxes with an aerodynamic diameter of 3  $\mu m$  instead of 5  $\mu m$  (which was  
 initially estimated).

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$$F_{H\&S} = c \frac{LAI}{5} \frac{q_v}{1.5 \times 10^{-2}} \quad (1)$$



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290 Fungal spore number concentrations are transformed into mass using an  
291 aerosol density of 1 which is used as reference density for the definition of aerodynamic  
292 diameter. All mass is attributed to organic matter. Within CHIMERE, fungal spores OA  
293 are prescribed as a new species considered as chemically inert in our simulation, but  
294 they can influence the condensation of semi-volatile secondary organic compounds  
295 (as part of the organic aerosol phase) and act as cloud condensation nuclei (Patade  
296 et al., 2021). However, no conclusive laboratory data are available to include such  
297 processes in a model. Other processes considered in the model apart from emissions  
298 are transport, and size-resolved dry and wet deposition with characteristics like that of  
299 primary anthropogenic aerosols.

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### 301 3. Results

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303 We will present here the results of the two-year long simulations containing  
304 fungal spores' organic aerosol. Our initial analysis delves into the variability of  
305 simulated emissions and concentration patterns, along with their impact on the  
306 simulated PM<sub>10</sub> levels. We will then present an assessment of the simulated  
307 concentration fields with respect to polyol observations as well as primary biogenic  
308 organic burden as determined by the source apportionment studies.

#### 309 3.1. Simulated two years of fungal spore primary organic aerosol

310 Figure 2 presents the seasonal variation in emissions and concentrations of  
311 fungal spore primary organic aerosols for the years 2013 and 2014 averaged, as well  
312 as that of LAI and specific humidity, obtained from our simulations. As parametrised,  
313 emissions are largely driven by vegetation density (represented here by the LAI) with  
314 emission structures that follow the distribution of the main French forest areas. Major  
315 forested areas and emission hotspots are seen over the Massif Central (centred at 2  
316 °E, 45.5 °N), the Jura (6 °E, 47 °N), the lower parts of the Alps (7 °E, 46 °N) and  
317 Pyrenees (0 °E, 43 °N), and the Landes Forest (-1 °W, 45 °N). Specific humidity, which  
318 is the other parameter used explicitly in the flux calculation (equation 1), is more  
319 homogeneous and its signature on the fluxes of spore emissions is less easily  
320 identifiable. LAI and specific humidity show the same seasonal cycles with higher  
321 values in summer and lower values in winter when the vegetation density and water  
322 content of the colder atmosphere are lowest. We can therefore hypothesise that LAI  
323 and specific humidity are responsible for much greater fungal spore emissions in  
324 summer than in winter.

325 Concentrations of atmospheric spores are found to be highly correlated with  
326 emissions, both spatially and on a seasonal scale. Small differences can be explained  
327 by transport and deposition processes. For instance, due to advection, contrasts in  
328 concentrations are less pronounced than those in emissions. Hummel et al. (2015)  
329 assumes that the lifetime of fungal spores is of about 5 hours in the atmospheric  
330 boundary layer. This short lifetime means that there is a small chance of long-distance  
331 transport, which explains the closeness of local concentrations to emission sources.  
332 In our simulations, the total deposition flux of fungal spores can reach a maximum of  
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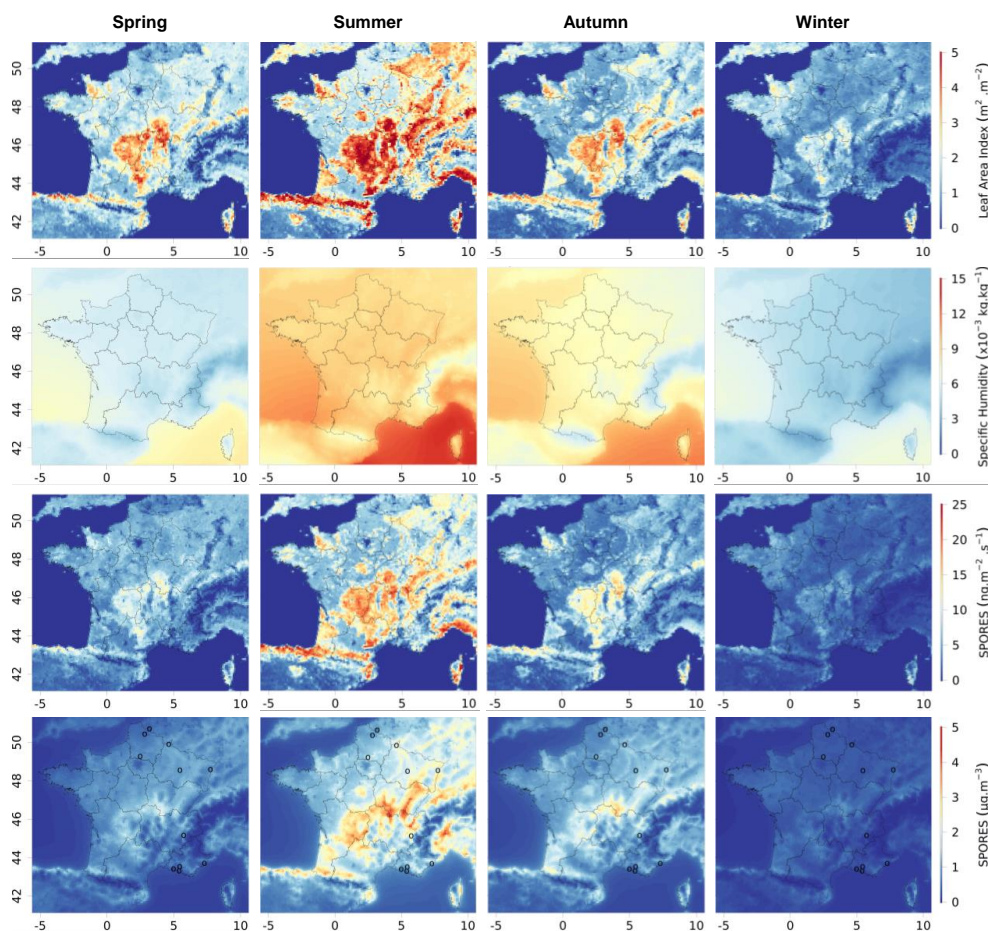
10  $\text{ng m}^{-2} \text{s}^{-1}$  on average over the two years, with 8  $\text{ng m}^{-2} \text{s}^{-1}$  for dry deposition and 5  
336  $\text{ng m}^{-2} \text{s}^{-1}$  for wet deposition. In summer, this total spore deposition reaches a maximum  
of 20  $\text{ng m}^{-2} \text{s}^{-1}$  in France, around 12  $\text{ng m}^{-2} \text{s}^{-1}$  in the Massif Central, while spore  
338 emissions peak in this area at 25  $\text{ng m}^{-2} \text{s}^{-1}$  on average over the summer period. The  
difference in emissions and deposition is therefore significant, confirming also partial  
340 transport out of source regions.

Despite little transport, spore concentrations at locations up to a few hundred  
342 kilometers away can be similar in mass and temporal variation, explained by similar  
meteorological conditions and leaf area index, leading to simultaneous emissions  
344 (Samaké et al., 2019b). Seasonal averages of fungal spore concentrations can reach  
values of several  $\mu\text{g m}^{-3}$  over large geographical areas, especially over the forested  
346 areas in the southern part of France (Massif Central). This is significant in view of the  
PM<sub>10</sub> concentration there and consistent with previous studies (Heald and Spracklen,  
348 2009). For instance, fungal spore OA contributes to about 20 % of PM<sub>10</sub> mass on  
summer averaged over the Massif Central (Figure 3). On the contrary, during winter,  
350 fungal spore concentrations remain always below 0.5  $\mu\text{g m}^{-3}$ , and do not contribute  
much to PM<sub>10</sub>, with a value always below a few percent. Spring and autumn are  
352 intermediate, both in terms of fungal spore OA concentrations and contributions to  
PM<sub>10</sub>. With the lower formation of BSOA compared to summer, the contribution of  
354 fungal spores to OM is largest in autumn, when it can reach around 40 %. It can reach  
about 30 % in other seasons with some geographical disparities. Despite low  
356 emissions and concentrations, the contributions of spores to concentrations of biogenic  
organic aerosols (BOA) is greatest in winter, reaching up to 70 %, due to the very low  
358 contribution of secondary biogenic organic aerosols during this period, in contrast to  
the summer period.

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Figure 2 : Leaf Area Index (LAI), specific humidity, mean seasonal emissions and seasonal concentrations of fungal spores modelled with CHIMERE for 2013 and 2014 in France respectively from top to bottom, by season for spring (March to May), summer (June to August), autumn (September to November), and winter (December to February), respectively from left to right. The circles represent the location of the measurement sites

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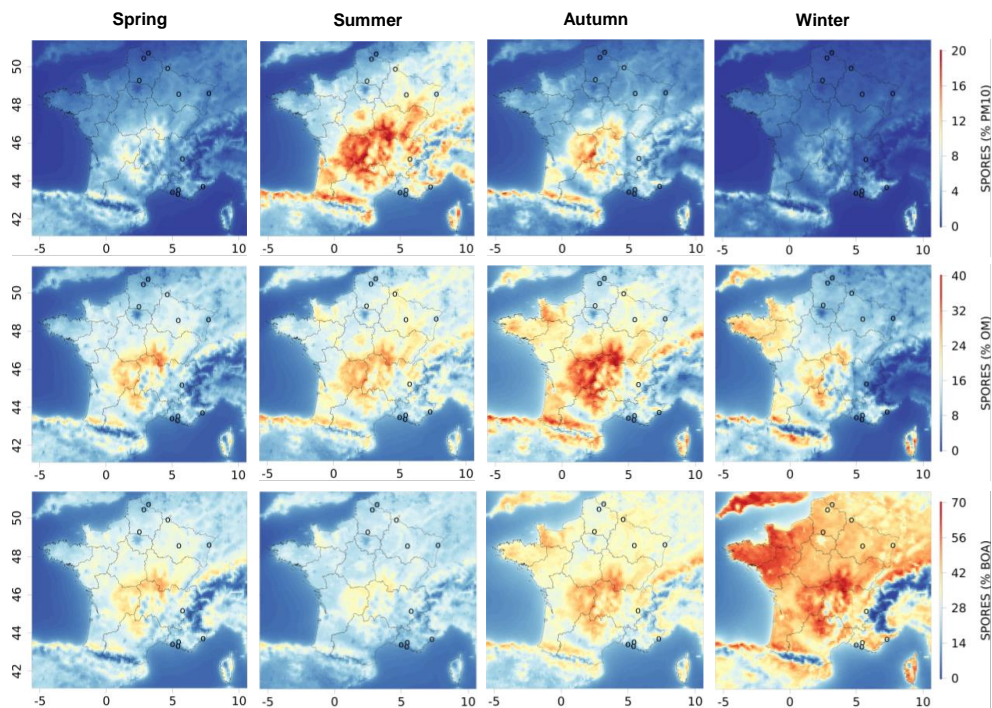
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Figure 3 : Contribution of fungal spores organic matter to PM<sub>10</sub>, OM and biogenic organic aerosols (BOA) modelled with CHIMERE for 2013 and 2014 in France respectively from top to bottom, by season for spring (March to May), summer (June to August), autumn (September to November), and winter (December to February), respectively from left to right. The circles represent the location of the measurement sites.

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406 **3.2. Comparison of fungal spore simulations to observations**

408 **3.2.1. General comparison for the entire data set**

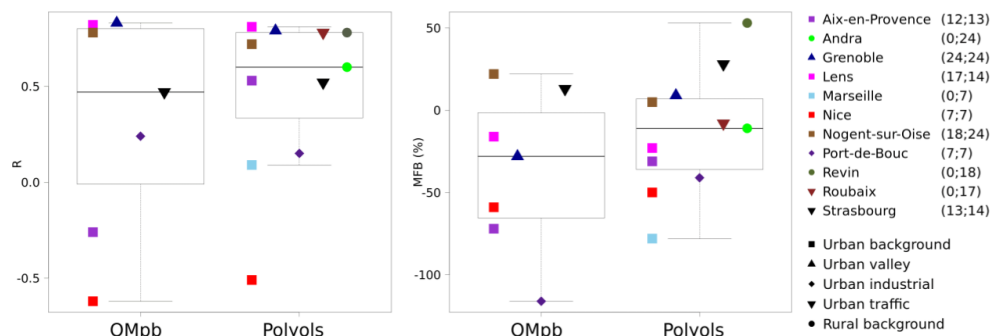
408 In order to compare simulations with observations, we can rely on two types of  
410 datasets available for several sites (as described respectively in sections 2.1.1 and  
412 2.1.2): first the polyol concentrations, and second the total OM concentration within the  
414 primary biogenic factor (OMpb) derived from PMF analysis of PM<sub>10</sub> filter samples. For  
416 these comparisons to be meaningful, we need to convert the simulated fungal spore  
418 organic aerosol concentrations into polyol ones. Bauer et al. (2008b) derived a  
420 conversion factor for this purpose, for the temperate latitude continental urban site of  
422 Vienna. For the sum of arabitol and mannitol, which are the two sugar alcohol species  
424 measured in our data base, the latter authors found an average mass of 2.9 (2 – 4.2)  
426 pg per fungal spore. Elbert et al. (2007) assumed an average mass of a fungal spore  
428 of 65 pg. Combining these values yields to a polyol / fungal spore mass ratio of 4.5  
430 (3.1 – 6.5), which will be used for the comparisons that follow. This is coherent with the  
432 work of Heald and Spracklen (2009), who used this same combination of values in  
434 order to derive their initial estimation of the mass of fungal spore emissions from multi-  
436 site polyol measurements.

436 We can first obtain a general picture of the performances of the model by  
438 studying the correlations and biases for all of sites with polyol measurements. For the  
440 169 polyol monthly averages from 11 sites, the median mean fractional bias (MFB) is  
442 slightly negative (-11 %), but with a large range of values for individual sites ranging  
444 from -78 % to +53 % (Figure 4, Table S1)<sup>1</sup>. Using the lower and upper boundaries for  
446 the conversion factor between mass of spore and mass of polyols (3.1 and 6.5,  
448 respectively), the corresponding median MFB values would be -47 % and +26 %. As  
450 a conclusion, within the range of quantified uncertainties, the median MFB for monthly  
452 polyol means of -11 % is statistically close to zero. A bias calculation performed directly  
454 with the 1497 daily means shows very similar results, with a median MFB of -11 %  
456 (range for the -81 % to +49 %). This is not surprising, since the comparison of monthly  
458 means has been based only on days for which observations were available.

458 Next, simulated fungal spore OA is compared to OM in the primary biogenic  
460 factor (OMpb) (see section 2.1.2). Our simulations show a median MFB bias of -28 %  
462 and a range from -116 % to +22 % for different sites (Figure 4, data from 98 monthly  
464 means for 7 sites). A negative bias is expected for this comparison, since the PMF  
466 factor is likely to include OM contributions from BSOA in addition to that from fungal  
468 spores (see section 2.1.2). As a result of this bias analysis with two different types of  
470 observations (polyols, OMpb), we do not observe the presence of a systematic bias for  
472 our fungal spore OA simulations for the ensemble of French sites. This agrees with  
474 (Hummel et al., 2015), who also could not conclude on a significant bias of the H&S  
476 parameterisation.

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<sup>1</sup> Note that following the definition of MFB (see SI), a relative difference between simulations and observations of a factor 2 (1/2) corresponds to a MFB of 67 % (-67 %). Thus in very crude manner, the simulations with sites with the largest and lowest MFB show a factor 2 difference with observations.



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**Figure 4 :** Comparisons of simulated monthly mean concentrations of fungal spore OA to OM from PMF primary biogenic factor (OMpb) and polyols measurements (sum of mannitol and arabitol). Biannual mean correlation and mean fractional bias (MFB) are respectively illustrated at left and right side. Ranges between minimal and maximal values, and medians for respectively 7 and 11 sites. The number of monthly data for OMpb and polyols are noted next to the station list out of a total of respectively 98 and 169 monthly data.

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The data sets were also used in order to asses if the H&S parameterisation is able to reproduce the daily and the monthly average time variation (Figure 4 and Figure S1, Table S1). For daily polyol averages, the median correlation between simulations and measurements is 0.43 with a range from -0.19 to 0.57 for the 11 sites. The median correlation is increased to 0.60 when looking at monthly averages with a large range from -0.51 to 0.83. Expectedly, for many, but not all sites, the parameterisation better depicts the seasonal variation (with larger summer and lower winter values) compared to the daily variations (Figure S1). We will discuss this result further on a site-by-site basis in section 3.2.2. Finally, for comparison with the same polyol data set, daily and monthly mean fractional error (MFE) are respectively 0.79 and 0.56 at all sites (Table S1, Figure S2). The root mean square error (RMSE) was also calculated for estimating the error (Table S1) the median RMSE calculated with daily polyol data is  $0.04 \mu\text{g m}^{-3}$  and  $0.03 \mu\text{g m}^{-3}$  with monthly data.

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478 **3.2.2. Comparison of time series at selected sites**

480 In this section, we evaluate the robustness of our simulations as a function of  
482 the period of the year. To do so, comparisons are conducted between model outputs  
484 and polyol observations, which are available for more measurement sites than sites  
486 with PMF results. These comparisons especially aim at understanding the large ranges  
488 of biases and correlations encountered in the previous section. Figure 5 shows  
490 observed and estimated monthly mean polyol (sum of arabitol and mannitol  
492 concentrations) at the sites with the most data (> 130 daily data) during years 2013  
and 2014, namely Grenoble, Lens, Nogent-sur-Oise, Revin and Roubaix. These sites  
also have the advantage of being of different types, respectively urban background in  
an Alpine valley, urban background, urban background, rural background, and road  
traffic. The time series for the other sites are shown in Figure S3, Figure S4, Figure  
S5. We indicate both the simulated monthly means using data from all days, and only  
for days for which filter samples are available.

Differences between simulations and measurements are small (<10 %) for most  
of the values, which underlines the robustness of the model for monthly averages.  
Figure 5 shows simulated seasonal cycles coherent with that in Figure 2 which reflects  
the dependence of the simulated emissions on the LAI. We observe the maximum  
monthly values for the summer months with a difference in structure between 2013  
and 2014: while in 2013 the simulated maximum occurs in July for all of the sites, in  
2014, it occurs in September at least for the sites in Northern France (Roubaix, Lens,  
Nogent-sur-Oise, Andra-OPE, Strasbourg). The highest summer measurement values  
of polyols (0.1 – 0.15  $\mu\text{g m}^{-3}$  corresponding to 2 - 3  $\mu\text{g m}^{-3}$  of OMpb for monthly  
averages) are of course simulated on the sites where the regional LAI are the strongest  
(e.g. Grenoble, Andra-OPE, Revin, Strasbourg, Nogent-sur-Oise) as opposed to Lens,  
Roubaix, Marseille, Aix, Port-de-Bouc for which the LAI of the adjacent regions are  
lower. However, none of the measurement sites are located within the area of large  
simulated fungal spore OA concentrations over the Massif Central. Comparisons  
between simulations and observations show a remarkable agreement especially in the  
seasonal variation for the stations in the northern part of France (Lens, Roubaix, Revin,  
Nogent-sur-Oise), resulting in monthly correlation coefficients (R) of respectively 0.78,  
0.83, 0.78 and 0.72. Specifically, the gradual increase in polyols (and related fungal  
spores OM) from March to July is very well simulated, except for Revin for which  
summer concentrations are overestimated. MFB values vary between -23 % for Lens  
and +53 % for Revin.

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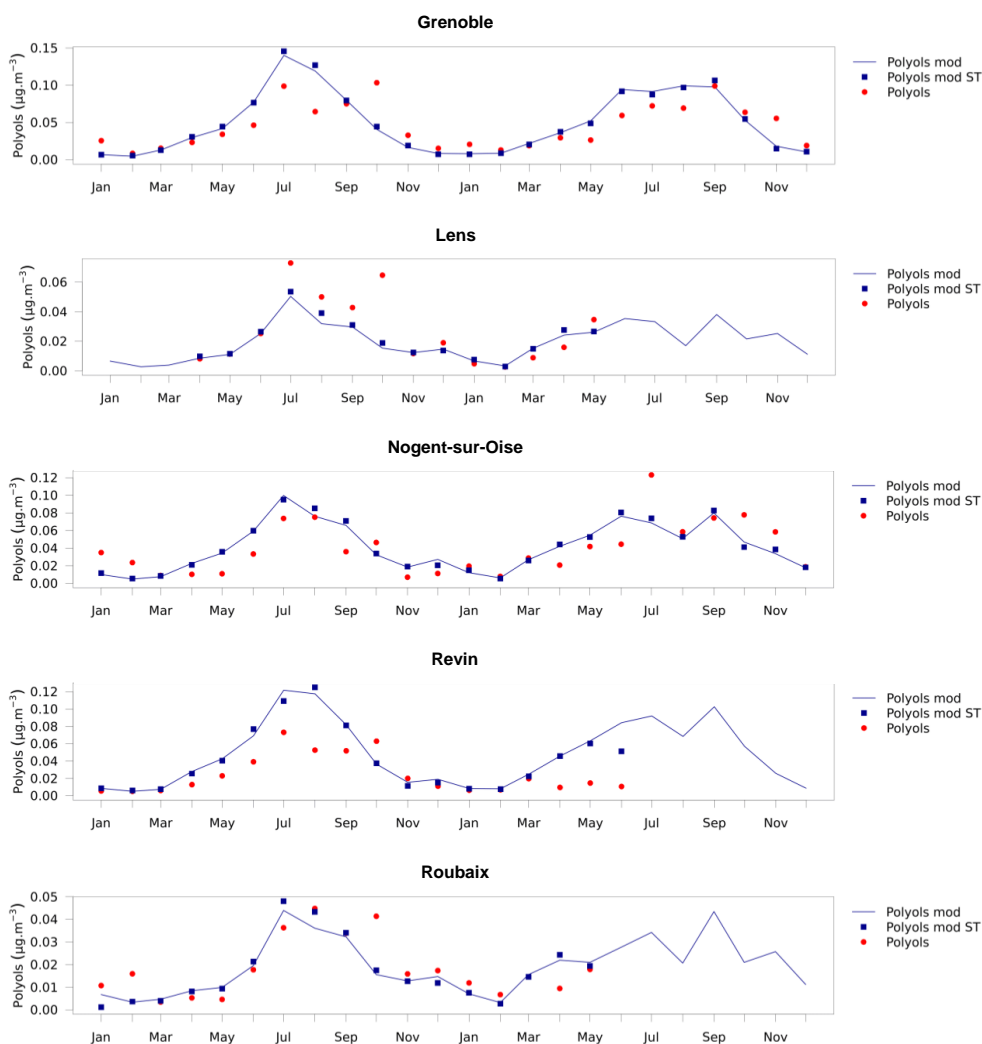
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**Figure 5 :** Timeseries of monthly-mean polyol concentrations over 2013 and 2014 modelled by CHIMERE (blue line), measured at the sites (red dots) and modelled by CHIMERE using the same timebase as the measurements (blue squares). The simulated values by CHIMERE have been derived by using a 4.5 % conversion factor between fungal spore OA and sum of polyols. Only the sites with the highest numbers of daily data used to calculate the monthly averages are shown (Grenoble, Lens, Nogent-sur-Oise, Revin, Roubaix).

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534 Correlations for eastern French sites are a bit lower, with 0.60 for the Andra-  
OPE site and 0.52 for Strasbourg with MFB respectively of -11 % and +28 %. For  
536 Grenoble, a city in SE France within the Alps, correlation is good ( $R = 0.79$ ) and bias  
is small (MFB = +9 %). For a group of sites in the south of France (Port-de-Bouc,  
538 Marseille, Nice), located less than 10 km from the Mediterranean Sea, the situation is  
singularly different, with strong underestimations in the simulation. It should also be  
540 noted that we have fewer observations at these sites (only seven monthly mean  
observations from June to December 2014), meaning that a full seasonal cycle was  
542 not obtained. Still, the simulated decline in autumn/winter (October to December)  
compared with summer (June to August) is not observed at these sites, resulting in low  
544 or even negative correlations for monthly means between -0.51 and 0.15 and negative  
biases (MFB values between -41 % and -78 %). Similarly, for Aix-en-Provence, some  
546 30 km inland, winter polyol levels are strongly underestimated, resulting in a MFB of -  
31 % and a correlation of 0.53.

### 548 3.3. Discussions

550 Overall, results obtained in this study demonstrate that the H&S  
parameterisation implemented into the CHIMERE model works remarkably well to  
552 reproduce the concentrations of fungal spore OA (or at least a proxy of these  
concentrations, with the polyols measurements) observed at sites located in the  
554 northern (Lens, Roubaix, Revin, Nogent-sur-Oise) and eastern (Andra-OPE,  
Strasbourg, Grenoble) parts of France. Indeed, the seasonal cycles observed at these  
556 sites and the intensity of the concentrations are remarkably well simulated by the model  
for the monthly averages. This gives great confidence in the ability of the H&S  
558 parameterisation to reproduce the fungal spore OA source over large parts of France.  
This extends the results from the earlier work of Hummel et al. (2015) based on an  
560 evaluation of 4 sites located in more northerly parts of Europe (Finland, Ireland, UK,  
Germany) limited to a week in the end of August, to more southerly regions, but still  
562 with temperate vegetation, and full seasonal cycles. For Europe, this extends also the  
results from Janssen et al. (2021) who implemented the H&S parameterisation into  
564 the global GEOS-Chem model. They compared the model output to yearly FBAP  
observations at the same sites in Finland and Germany and found rather similar  
566 seasonal variations with summer maxima and winter minima, although the simulated  
maximum occurred in June (2010), while it was observed in August. Note that Janssen  
568 et al. (2021) shows that the H&S parameterisation shows a strong overestimation of  
fungal spore numbers with respect to observations in the US.

570 Another remarkable fact is that positive results in our study have been obtained  
from sites with very different land-use typologies, ranging from traffic (Roubaix and  
572 Strasbourg) and urban background (Lens, Nogent-sur-Oise) to rural (Revin, Andra-  
OPE), or an urban background site within an Alpine valley (Grenoble). This can be  
574 explained by the fact that, due to low levels of long-distance transport, fungal spore  
OA seems to be controlled by the vegetation at local scale, as also pointed out already  
576 for Grenoble by Samaké et al. (2019a).

578 Despite these overall encouraging results, several limitations appear for our  
study. One is probably related to the simplification of using a unique LAI parameter



580 which cannot consider differences in vegetation typology. This may explain strong  
581 differences in MFB values between sites in NE France: Revin, located in a forest rich  
582 area in the Ardennes, shows a strong positive MFB of +53 % (the largest one  
583 encountered in our study), while the Andra-OPE site surrounded by extensive field  
584 crops shows an MFB of -11 %. For this latter site, we also can note that several  
585 observed daily peaks (in August 2013 and July 2014), as large as  $5 \mu\text{g m}^{-3}$  are not  
586 simulated. Such peaks may be related to agricultural activities such as harvesting as  
587 demonstrated by Samaké et al. (2019b) from the record of field work. In addition,  
588 atmospheric concentrations of fungal spores mainly come from plant host species  
589 (Samaké et al., 2020), so mechanised crop pruning and harvesting can have an impact  
590 on spore concentrations in rural areas. The processes which are known to trigger  
591 fungal spore emissions are not included specifically in the H&S parameterisation. In  
592 the context of this work, we did not seek to better characterise this potential missing  
593 source, but it is an interesting perspective for future work.

594 Our study also clearly shows the inability of the H&S parameterisation to  
595 correctly reproduce Ompb and polyol measurements for Mediterranean areas in  
596 Southern France, even though as noted before, our observational data base is weaker  
597 for this region. However, at these sites, analysis of the chemical composition of  
598 aerosols in the PM<sub>10</sub> fraction also showed poor simulation of the chemical species,  
599 suggesting a more global problem in the Mediterranean area. This could be explained  
600 by the specific dynamics in this sector (sea breezes, strong mistral-type winds) coupled  
601 with significant orography and heavy urbanisation. As a result, failure to take account  
602 of wind speed in the parameterisation of H&S may be a major cause of a lack of  
603 emission and concentration in the Mediterranean area. Again, this failure may also be  
604 related to the fact the LAI does not capture specific characteristics of Mediterranean  
605 type vegetation, and which are not included in the H&S parameterisation, mainly tested  
606 for sites mostly in northern Europe. In addition, it is striking that our simulations on  
607 Mediterranean sites, as expected still simulate weak autumn/winter emissions due to  
608 low LAI and specific humidity, but which are in contradiction to the still large observed  
609 concentrations. This could be due to a relatively stronger importance of soil related  
610 fungal spore emissions, which would be independent of LAI. Further, the drier and  
611 hotter Mediterranean climate could lead to relatively smaller emissions during dry  
612 summers and relatively larger emissions during winter still warm enough to allow for  
613 fungal spore emissions. It was observed by Samaké et al. (2019b) that the sudden  
614 and large decrease of the fall concentrations to the winter levels observed  
615 simultaneously in Grenoble and Chamonix (160 km apart) coincides with a first night  
616 temperature below +5 °C, which may be a threshold for the fungi population in this  
617 area. Such complex relationships would not be captured by the single specific humidity  
618 parameter which agglomerates information from relative humidity and temperature.

619 Finally, it may be noted that marine sources could also contribute to enhanced  
620 polyol levels and organic aerosol at near coastal sites, although such sources are not  
621 considered in our simulation. For instance, Fu et al. (2013) reports that large mannitol  
622 concentrations, up to more than  $50 \text{ ng m}^{-3}$  over the Arctic Ocean, are comparable to  
623 the maximum concentrations observed at our Mediterranean coastal sites. They  
624 attribute this source to long range transport of fungal spores, despite the small  
625 transport distance at least in the boundary layer due to efficient dry deposition. Direct  
626 marine sources for polyols are an alternative explanation (algae, marine fungi).  
Particularly, mannitol can account up to 20-30 % of the dry weight of some algae



628 species and is likely to be an important source of carbon for marine heterotrophic  
630 bacteria (Groissillier et al., 2015). As a conclusion, the H&S parameterisation should not  
632 be applied for PBOA emissions in Mediterranean or marine areas, and further work is  
needed to better document PBOA concentrations and emission processes in such  
areas.

#### 634 4. Conclusions

636 In this work, we introduced the parameterisation proposed by Heald and  
Spracklen (2009) for fungal spore OA emissions and updated by Hoose et al. (2010)  
638 into the CHIMERE regional chemistry-transport model (Hereafter called H&S). The  
rationale for this work is to recognise the potentially important contribution of fungal  
640 spore to summertime PM<sub>10</sub> (Samaké et al., 2019b, a) that can fill in the missing part of  
the OM in chemistry transport models. The simplicity of the H&S parameterisation  
642 gives us specific advantages: a unique LAI parameter gives a slow varying emission  
potential, which is modulated with respect to meteorological conditions by specific  
644 humidity.

Here, we largely extend the geographical and temporal validity of this  
646 parameterisation, which has only been tested before for a limited dataset of  
observations at northern European locations during the end of summer, to a two-year  
648 dataset of seven sites over north-eastern France. Both polyols (more precisely sum of  
arabitol and mannitol observations), and a primary biogenic organic aerosol factor from  
650 PMF analysis show only limited biases for these sites, respectively +5 % and -2 %, in  
terms of MFB (from 4 sites only for the comparison with PMF analysis). These small  
652 biases, largely within the incertitude of the polyol/OM conversion factor and of the PMF  
factor, are a positive outcome of our study. In addition, for this group of sites, the  
654 seasonal variation of fungal spore emissions, displaying large summer and small  
winter values, is correctly depicted, as manifested in large monthly mean correlations  
656 (median 0.78, range from 0.52 to 0.83, from polyol measurements).

Still, and obviously, limitations can be noted, such as a wide range of biases for  
658 individual sites, with MFB values between -23 % and +53 % for polyol observations.  
This might be related to biome specific differences in the emissions only described by  
660 a single LAI parameter. The emission variability on a day-to-day basis is only partly  
expressed by the single specific humidity parameter (range of correlation coefficients  
662 between 0.31 and 0.57 for the polyol measurements at the 7 sites in North-eastern  
France). Here, using a more sophisticated combination of meteorological parameters  
664 would be desirable to improve the modelling, as for example in Janssen et al. (2021)  
including also maximum and minimum daily temperatures and friction velocity (even if  
666 these authors did not evaluate the capacity of such a combination to simulate the daily  
PBOA variation). One possible reason for the lack of correlation in daily time series is  
668 the impact of land-use dependent activities, such as annual harvest or tilling in  
agricultural areas.

670 For a smaller group of Mediterranean sites, with less observational data  
coverage, the H&S parameterisation failed to capture fungal spore emissions both in  
672 terms of absolute values and in seasonal variations, leading to strong negative biases  
especially during the autumn/winter seasons. As a conclusion, for this region the use  
674 of the H&S parameterisation in regional PM modelling may not consider certain factors



676 necessary for these specific sites. Additional efforts are required to enhance the model  
677 dynamics specifically over Mediterranean coastal environment. This includes  
678 extending the simulation of fungal spores over more extended periods in these  
679 locations. Furthermore, there is a need to better characterise a source of  
680 Mediterranean marine organic aerosol (AO) that is distinct from fungal spores but  
681 shares the emission of polyols. It is also necessary to have more measurement points  
682 in this specific area to be able to achieve a more concrete conclusion.

683 These two year-round CHIMERE simulations incorporating the H&S  
684 parameterisation revealed a significant contribution of fungal spore OA to PM<sub>10</sub> mass,  
685 which is of the order of one percent or less during winter, and up to 20 % during  
686 summer in high emission zones over forested areas such as the Massif Central. In  
687 terms of contribution to OM, the simulated autumn fungal spore contribution is even as  
688 high as 40 %. This large predicted fungal spore OA contribution over the Massif Central  
689 however still warrants confirmation by observations.

690 Finally, the projected impact of fungal spore organic aerosol (OA) suggests  
691 significant and seasonally variable contributions to both PM<sub>10</sub> and OA mass.  
692 Consequently, the simulation of spores should be included in state-of-the-art chemistry  
693 transport models. While the validity of the H&S parameterisation has been  
694 demonstrated with a good agreement with measurements across northern and eastern  
695 France, its application is cautioned against in Mediterranean regions.

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### Code and data availability

724 All measurement and PMF data for this paper are archived at the IGE, and are  
725 available on request from the corresponding authors (JLJ and GU). The codes and  
726 modelling data are available from the LISA authors (MV, GF, MB, GS).  
The model is available here: <https://www.lmd.polytechnique.fr/chimere/>  
728 The MODIS observations are available here :  
<https://modis.gsfc.nasa.gov/data/dataproduct/mod15.php>

730

### 732 Author contributions

JLJ and GU provided the PM<sub>10</sub>, polyol and PMF speciation data developed at the IGE  
734 for the PhD work of Abdoulaye Samaké and Samuel Weber. OF completed the data  
set with those obtained at the LCSQA during the CARA programme. FC developed the  
736 H&S parameterisation code at INERIS, GS adapted the code for a more recent version  
of the CHIMERE model at LISA. AC contributed to the LAI mapping. MV, GF, MB,  
738 designed the numerical experiments. MV performed the simulations, produced figures  
and tables, and wrote the paper. All co-authors contributed to the discussion of the  
740 results. MV prepared the paper with contributions from all co-authors. MV, MB, GF,  
GS, JLJ and GU designed the study. MV, MB, GF, GS, JLJ, GU, OF, FC and AC  
742 contributed to the writing of the article.

744

### Competing interests

746 The authors declare that they have no conflict of interest.

748

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756

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