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## 10 **Responses to Editor's and Referee #2's Comments**

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12 We thank the Editor and Referee #2 for their time and consideration. The worthy and  
13 constructive comments helped us to improve the manuscript.

14

15 The Responses are structured as follows: we begin with the Editor's and Referee #2's  
16 comments (pages 2–4) and then follow with the Authors' answers (pages 5–7), dividing into  
17 four parts, (i) EF derivation; (ii) ORCHIDEE LAI simulation; (iii) MODIS reference; (iv)  
18 Sindelarova reference.

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1 **Editor 's Comments**

2  
3 Dear Palmira Messina et al.,

4  
5 While both referees found the manuscript interesting, they were also very critical on some key  
6 aspects of the paper. These included the derivation of emission factors for different plant  
7 functional types, and the validity of the leaf area index calculated by ORCHIDEE. The referee  
8 #2 was not satisfied with the revised version (see comments below). Thus I subject the paper  
9 to major revision, after which it will be reconsidered.

10 The derivation of EFs needs to be revised with more systematic approach. I agree with the  
11 referees that “qualitative and comparative method” sounds very subjective. As authors  
12 mention in their response to the referee #2 that the modification of EFs “represents an  
13 important part of the upgrade of the emission module”, being more systematic and transparent  
14 in the methodology is crucial. The comments by the referee #2 offer good guidelines in this:  
15 Take into consideration only primary publications (no reviews, no proceeding modeling  
16 studies), select the papers to be used (weed the papers for obvious outliers and otherwise  
17 suspicious data), calculate mean EF (or median, and possibly standard deviation). Finally  
18 describe the method and show which papers you have actually used.

19 According to the studies pointed out by the referee, the areas in which the ORCHIDEE and  
20 MODIS LAI deviate most (lower boreal, temperate, and tropics) are the ones where MODIS  
21 is thought to be most certain. Thus the impact of the deviation of LAIs should be addressed in  
22 a more rigorous way e.g. by running the model by forcing the LAI to match the MODIS  
23 values geographically and analyze the effect on the VOC fluxes.

24  
25 Sincerely,

26  
27 Janne Rinne

1 **Comments of Referee #2:**

2

3 I am still confused about the derivation of EFs. It is now said that the authors used a  
4 ‘qualitative and comparative method’, which generally means to me that there is no rule  
5 behind the selection. Later on it is stated that for isoprene they ‘principally use the most  
6 recent papers’, which would be okay if the given examples for larch (here 5 references  
7 between 2002 and 2012 are used, outliers disregarded) and also for BoBrDe (2003 – 2012)  
8 wouldn’t proof differently. The example of larch indicates another oddity here: the authors  
9 explain which papers were reviewed (10) and which were actually used to determine the  
10 arithmetic average (5). However, in the reference list of table 3 all the references which are  
11 discarded for good reasons are still listed. So are these (bad) references still used in the  
12 derivation of some of the other EFs?? Interestingly, for monoterpenes outliers are also  
13 ‘officially’ disregarded although ‘generally’ any paper is taken into account. Overall, I still  
14 cannot see which references are actually used for EF derivation and the explanations give the  
15 impression of justifying a selection that has done with no clear rules in mind. I can understand  
16 the difficulties in determining ‘objective’ values and I hope not to be overly picky here being  
17 educated by the authors that ‘there is no universal agreement on parameterization’ but why  
18 don’t disregard any clear outliers as well as any reference with unclear sources or which refers  
19 to measurements that are already considered? In Table 3, the references used for each PFT  
20 should be clearly differentiated, possibly with numbers given behind the respective EF value.

21

22 From my given options to proceed with the LAI problem in the simulations (1 Improve the  
23 LAI simulations, 2 Improve the argumentation to a degree that the reader can accept  
24 ORCHIDEE simulations as equally likely as MODIS data, or 3 Run all simulations with  
25 MODIS derived values only), the authors choose no. 2. The arguments are 1) that also  
26 MODIS data don’t provide the ‘real’ LAI, and 2) that it is the intention of the article to study  
27 the weaknesses in LAI modelling, so that they can be resolved in a future model version of  
28 ORCHIDEE. I am a bit surprised about the reasoning given that the Maignan et al. study  
29 which investigated LAI and suggested improvements already 5 years ago seemingly was not  
30 considered. Indirectly supporting Maignan et al., Fang et al. 2013 found that the largest  
31 uncertainties of MODIS data are in the Australian southwest, the Rocky Mountains and in

1 lower boreal regions while the tropics and temperate regions are relatively ‘certain’.  
2 Unfortunately these are the regions that deviate most with ORCHIDEE simulations and are  
3 most important for BVOC emissions. So I only can support the view of the authors that the  
4 weaknesses of LAI simulations have to be analysed and improved but I think that this should  
5 be done in a more differentiated way than with a mere global reduction/increase of simulated  
6 LAI – and much preferably in a separate study preceding the BVOC analysis.

7 By the way: The referenced MODIS data are indicated to be obtained from Yuan et al. 2011,  
8 which however, only supplies a tool for improving MODIS data and doesn’t provide a global  
9 dataset. So this doesn’t seem to be the correct reference.

10

11 Leaf age, drought stress and CO<sub>2</sub> activity modifications are explained but I couldn’t find the  
12 give reference (Sindelarova et al. 2014) in the reference list. I guess it is Sindelarova et al.  
13 2014. In this publication the CO<sub>2</sub> effect is determined to increase isoprene emission by 2.7  
14 percent based on the concentration of the year 2003. It is indeed probably less if the average  
15 during the simulation period is concerned.

16

17 Fang, H., et al. (2013), Characterization and intercomparison of global moderate resolution  
18 leaf area index (LAI) products: Analysis of climatologies and theoretical uncertainties, *J.*  
19 *Geophys. Res. Biogeosci.*, 118, 529–548, doi:10.1002/jgrg.20051.

20

21 Sindelarova K, Granier C, Bouarar I, Guenther A, Tilmes S, Stavrakou T, Müller JF, Kuhn U,  
22 Stefani P, Knorr W. 2014. Global dataset of biogenic VOC emissions calculated by the  
23 MEGAN model over the last 30 years. *Atmospheric Chemistry and Physics*, 14: 9317-9341.

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## 1 **Authors' Response to comments on EF derivation**

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3 The comments of Editor and Referee #2 have rightly highlighted the delicate points of the  
4 methodology used to derive the EFs and the weak points of its description. All global model  
5 developers in this field face the difficulty of choosing EFs. Indeed, fixing a single EF value  
6 per PFT at global scale is extremely difficult, considering the temporal, geographical, and  
7 plant-to-plant variability of EFs, as well as the potential error in EF measurements. Moreover,  
8 reviewing the literature to assign new EFs is a substantial and time-consuming part of the  
9 work, but often does not lead to a very satisfactory result, with the overall number of data that  
10 can eventually be used (actually providing emission factors, with proper units and in  
11 appropriate standard conditions, etc.) being reduced. Considering a simple average of EFs  
12 published would not make real sense without taking into account the surface of the plant  
13 species investigated, the representativeness of this species for the corresponding PFT, or the  
14 EF geographical variability, information that we often do not have access to. Consequently,  
15 no totally satisfactory and robust approach has emerged yet in the scientific community, and  
16 this parameter is necessarily affected by a significant uncertainty. We would like to clarify  
17 that one objective of this work is not the derivation of EF itself, trying to build a full and  
18 robust statistical method, but to update the ORCHIDEE emission scheme, where EF  
19 derivation is one of the various but more controversial developments, as we explain in section  
20 2.2.1. We agree that the EF update has to be presented more clearly.

21  
22 Following the remarks of the Editor and Referee #2 and in order to make our explanation  
23 clearer, we carefully and extensively checked the derivation of EFs and the presentation of  
24 our methodology, adding the papers used eventually and deleting the ones that we excluded.  
25 We then re-wrote section 2.2.1 and added a Supplementary Material listing all data used to  
26 derive EFs, specifying for which emitted compound and which PFT they are considered. We  
27 tried, in the new version, to be clearer and more rigorous, and we hope that these changes will  
28 meet the Editor's and Referee #2's expectations and suggestions.

## 1 **Author's response on ORCHIDEE LAI simulation**

2  
3 Using a BVOC emission scheme embedded in the ORCHIDEE global vegetation model, one  
4 aim of our work is to assess to which extent the uncertainty in the leaf area index (LAI) can  
5 impact BVOC emission estimates, and not, as stated by Referee #2, to study the weaknesses  
6 in LAI modelling. As the LAI absolute value is a strong driver of BVOC emissions, we first  
7 focused our analysis mostly on this variable, performing sensitivity tests, varying the LAI  
8 calculated by ORCHIDEE by a factor of 0.5 and 1.5. In the present study, we modified the  
9 BVOC emission module and not the other components of the model, which are taken from the  
10 latest and most updated ORCHIDEE version. The vegetation core of this model certainly has,  
11 as in any other model, some limitations and room for improvement. However, as far as the  
12 LAI is concerned, this would imply potentially modifying the phenology of PFTs considered,  
13 the description of the soil hydrology, and then the carbon uptake by plants, developments  
14 which are not trivial and are beyond the scope of this work. We mentioned the study of  
15 Maignan et al. (2011) to underline some weak points, already investigated, of ORCHIDEE in  
16 modelling the LAI. However, the work by Maignan et al. (2011) focused on the plant  
17 phenology, and especially on the annual and interannual variability of LAI, while our  
18 sensitivity analysis focuses on its absolute value. Since the work performed by Maignan et al.  
19 (2011), some improvements have been regularly carried out in the ORCHIDEE vegetation  
20 model, with benefits to the newer versions, including the one we used. Having vegetation and  
21 BVOC emissions consistently calculated online in ORCHIDEE based on the same  
22 environmental conditions (temperature, radiation, precipitations, CO<sub>2</sub>, etc.) makes it an  
23 appropriate and interesting tool, especially for long-term studies (future scenarios, land-use  
24 change) that could not be addressed by forcing vegetation characteristics, and for instance  
25 LAI. Contrary to what is stated by Referee #2 (“*Maignan et al., Fang et al. 2013 found that*  
26 *... tropics and temperate regions are relatively ‘certain’*”), Fang et al. (2013) themselves  
27 show in Fig. 4 of their paper (*LAI uncertainty maps for MODIS, GEOV1, and JRC-TIP from*  
28 *2003 to 2010*, page 537) that the highest uncertainties are in the tropical areas and in northern  
29 temperate and boreal regions during the summer season (“*and show that the tropical ... and*  
30 *boreal regions ... have higher uncertainties than the other areas*”, page 536).

1 In order to strengthen and complete our analysis in a more rigorous way, we performed  
2 additional simulations, as suggested by Referee #2 and underlined by the Editor. For this  
3 work, we ran ORCHIDEE, forcing the LAI with the MODIS values, and therefore by-passing  
4 the LAI calculated online. We report and detail the results of this experiment in the new  
5 section 3.4.3 and in one additional figure (Fig. 13). Moreover, we emphasize and clarify the  
6 differences of the LAI approach between ORCHIDEE and MEGAN in section 2.5.

7

### 8 **Authors' response to comments on MODIS reference**

9 The Yuan et al. (2011) paper is the reference that the authors mention themselves on the  
10 website where the MODIS LAI (global) data can be downloaded. Please see here (bottom of  
11 the webpage):

12 <http://globalchange.bnu.edu.cn/research/lai/>

13

### 14 **Authors' response to comments on Sindelarova reference**

15 At the end of section 2.4 we changed the reference *Sinderalova et al., (2014)* with the correct  
16 one: *Sindelarova et al., (2014)*.

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# Global Biogenic Volatile Organic Compound emissions in the ORCHIDEE and MEGAN models and sensitivity to key parameters

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## 1 Abstract

2 A new version of the Biogenic Volatile Organic Compounds (BVOC) emission scheme has  
3 been developed in the global vegetation model ORCHIDEE (Organizing Carbon and  
4 Hydrology in Dynamic EcosystEm), that includes an extended list of biogenic emitted  
5 compounds, updated emission factors (EFs), a dependency on light for almost all compounds  
6 and a multi-layer radiation scheme. Over the 2000–2009 period, using this model, we estimate  
7 mean global emissions of 465 Tg C yr<sup>-1</sup> for isoprene, 107.5 Tg C yr<sup>-1</sup> for monoterpenes, 38  
8 Tg C yr<sup>-1</sup> for methanol, 25 Tg C yr<sup>-1</sup> for acetone and 24 Tg C yr<sup>-1</sup> for sesquiterpenes. The  
9 model results are compared to state-of-the-art emission budgets, showing that the  
10 ORCHIDEE emissions are within the range of published estimates. ORCHIDEE BVOC  
11 emissions are compared to the estimates of the Model of Emissions of Gases and Aerosols  
12 from Nature (MEGAN), which is largely used throughout the biogenic emissions and  
13 atmospheric chemistry community. Our results show that global emission budgets of the two  
14 models are, in general, in good agreement. ORCHIDEE emissions are 8% higher for isoprene,  
15 8% lower for methanol, 17 % higher for acetone, 18% higher for monoterpenes and 39%  
16 higher for sesquiterpenes, compared to the MEGAN estimates. At the regional scale, the  
17 largest differences between ORCHIDEE and MEGAN are highlighted for isoprene in  
18 northern temperate regions, where ORCHIDEE emissions are higher by 21 Tg C yr<sup>-1</sup>, and for  
19 monoterpenes, where they are higher by 4.410 and 10.218 Tg C yr<sup>-1</sup> in northern and southern  
20 tropical regions compared to MEGAN. The geographical differences between the two models  
21 are mainly associated with different EF and Plant Functional Type (PFT) distributions, while  
22 differences in the seasonal cycle are mostly driven by differences in the Leaf Area Index  
23 (LAI). Sensitivity tests are carried out for both models to explore the response to key  
24 variables or parameters such as LAI and Light Dependent Fraction (LDF). The ORCHIDEE  
25 and MEGAN emissions are differently affected by LAI changes, with a response highly  
26 depending on sensitive to the compound considered. Scaling the compound. LAI is scaled by a  
27 factor of 0.5 and 1.5 changes, changing the isoprene global emission by -21% and +8% for  
28 ORCHIDEE and -15% and +7% for MEGAN, and affects affecting the global emissions of  
29 monoterpenes by -43% and +40% for ORCHIDEE and -11% and +3% for MEGAN.  
30 Performing a further sensitivity test, forcing ORCHIDEE with the MODIS LAI, confirms the

1 | [high sensitivity of the ORCHIDEE emission module to LAI variation](#). We find that MEGAN  
2 | is more sensitive to variation in the LDF parameter than ORCHIDEE. Our results highlight  
3 | the importance and the need to further explore the BVOC emission estimate variability and  
4 | the potential for using models to investigate the ~~estimatedestimate~~ uncertainties.

## 6 | **1 Introduction**

7 | The terrestrial biosphere emits large amounts of Volatile Organic Compounds (VOCs) in  
8 | particular terpenoids, such as isoprene, monoterpenes, and sesquiterpenes, and oxygenated  
9 | hydrocarbons such as methanol, acetone, formaldehyde, acetaldehyde, acetic acid, or formic  
10 | acid (Laothawornkitkul et al., 2009; Guenther et al., 2012; Penñelas and Staudt 2014). On the  
11 | global scale, the ecosystem contribution to VOC emissions is significantly higher than the  
12 | anthropogenic one, and accounts for 75–90% of the total emission (Guenther et al. 1995;  
13 | Lamarque et al., 2010). BVOCs play a central role in atmospheric chemistry, influencing the  
14 | oxidative capacity of the atmosphere (Arneth et al., 2011; Taraborrelli et al., 2012), leading to  
15 | the production of tropospheric ozone in the presence of nitrogen oxides (Von Kuhlmann et al.,  
16 | 2003; Mao et al., 2013), and influencing the tropospheric carbon monoxide budget (Pfister et  
17 | al., 2008). Additionally, BVOCs and their oxidation products lead to the formation and  
18 | growth of more than 50% of the Secondary Organic Aerosols (SOA) (Kanakidou et al., 2005;  
19 | Goldstein and Galbally, 2007; Van Donkelaar et al., 2007; Engelhart et al., 2008; Hallquist et  
20 | al., 2009; Acosta Navarro, et al., 2014; Tsigaridis et al., 2014). Under appropriate atmospheric  
21 | conditions, BVOCs can contribute to a significant fraction of particles that evolve into cloud  
22 | condensation nuclei (CCN) (Riipinen et al., 2012), even enhancing the droplet number  
23 | concentration in clouds (Topping et al., 2013).

24 | Despite numerous measurements and the progressive understanding of the processes  
25 | underlying their production, BVOC emission estimates are still highly uncertain and vary  
26 | significantly (Steiner and Goldstein, 2007; Arneth et al., 2008; Simpson, et al., 2012;  
27 | Sindelarova et al., 2014).

28 | Over the last 20–25 years, two main methods have been developed to derive BVOC  
29 | inventories: a top-down approach based on the inversion of satellite measurements, which  
30 | allows BVOC emissions to be indirectly derived (Palmer et al., 2006; Barkley et al., 2013),

1 and a bottom-up approach. The latter approach is the most widely used method for local-,  
2 regional- or global-scale studies and can be divided into two main categories:

3 (i) an empirical method, based essentially on Guenther et al. (1995), where the response of leaf  
4 emissions to environmental changes is modelled using algorithms combined in a  
5 multiplicative way (Guenther et al., 2006, 2012; Lathière et al., 2006, 2010; Steinbrecher et  
6 al., 2009; Oderbolz et al., 2013). Hereafter we refer to it simply as the *Guenther formulation*);

7 (ii) a process-based approach, where emissions are linked to the photosynthetic electron  
8 transport rate in chloroplasts (Niinemets et al., 2003a, b; Sitch et al., 2003; Keenan et al.,  
9 2009; Schurgers et al., 2009; Pacifico et al. 2011; Unger et al. 2013).

10 The models discussed in this study belong to the first category of bottom-up models.

11 BVOC emission modelling at the global scale is a complex issue, especially because of the  
12 number of variables and processes influencing the emission of these compounds, generally  
13 characterized by strong temporal and geographical variations. A critical point is the lack of  
14 information available at the global scale related to the various biomes, ~~making difficult an~~  
15 ~~would otherwise allow a more~~ accurate representation of the geographical distribution and of  
16 the seasonal variation of BVOC emissions (Peñuelas and Staudt, 2010). The basal EF for  
17 instance, defined as the emission at the leaf level under standardized environmental  
18 conditions of temperature and solar radiation (Guenther et al., 1995; Steinbrecher et al.,  
19 2009), shows large variability from one plant species to another. ~~Nowadays, Nowadays, a~~  
20 ~~large number of measurements is available for different plants and at various sites and~~ there is  
21 an increasing number of field campaigns that investigate, in addition to isoprene and bulk  
22 monoterpenes, many other important compounds for atmospheric chemistry, especially  
23 regarding the SOA formation, such as speciated monoterpenes and sesquiterpenes. More data  
24 and information are therefore available, allowing EF estimates for a wider range of BVOCs,  
25 despite the limitations ~~for modelling purposes~~ which ~~we will be discussed~~ ~~diseuss~~ in Sect.  
26 2.2.1. To calculate BVOC emissions, a single EF is usually assigned for each PFT, where one  
27 PFT represents a group of plants having the same phylogenetic, phenological and physical  
28 characteristics (Prentice et al., 1992). The choice of one single value for each PFT is  
29 especially difficult, as each PFT actually corresponds to several plant species, and EFs show,  
30 in general, a wide range of values among different plants (Kesselmeier and Staudt, 1999;

1 | Niinemets et al., 2011). Moreover, several measurements ~~even~~ show that the emission factors  
2 | are significantly sensitive to many processes, and parameters that are difficult to isolate and  
3 | linked to plant stress, such as drought periods, ozone exposure, insects, herbivores and  
4 | pathogen ~~attacks, attack~~ (for a review see Laothawornkitkul, et al., 2009 and Niinemets et al.,  
5 | 2010), making it ~~more~~ delicate to set EFs even for a single plant. In addition, the link between  
6 | EF variation and plant phenology is in general not taken into account, or is roughly described,  
7 | especially in models that adopt the empirical approach.

8 | In the early works focusing on BVOCs, isoprene was the only compound considered to be  
9 | both light and temperature dependent, while the other compounds were considered to be only  
10 | temperature dependent. More recent papers show a growing evidence of the dependency of  
11 | monoterpenes (Dindorf et al., 2006; Holzke et al., 2006; Šimpraga et al., 2013),  
12 | sesquiterpenes (Hansen and Seufert., 2003) and oxygenated BVOCs (Jacob et al. 2002, 2005;  
13 | Harley et al., 2007; Millet et al., 2008, 2010; Hu, et al., 2011; Wells et al., 2014) on radiation.  
14 | As proposed in Guenther et al. (2012), a general approach is now to consider, for each emitted  
15 | compound, an emission fraction which depends on both temperature and solar light, as done  
16 | for isoprene, with the remaining fraction dependent only on temperature. The Guenther et al.  
17 | (2012) approach considers only one value per emitted compound, whilst it has been shown  
18 | the LDF also depends on the plant species. For example, measurements of the diurnal cycle  
19 | for monoterpenes above Amazonian rainforest (Rinne et al. 2002; Kuhn et al., 2002) suggest  
20 | that emissions are dependent on both light and temperature, whilst the role of light in  
21 | influencing monoterpene emissions from boreal Scot pine forest is less clear (Taipale et al.,  
22 | 2011). Moreover, Staudt and Seufert (1995) and Loreto et al. (1996) show that monoterpene  
23 | emissions from coniferous trees are principally influenced by the temperature, while those  
24 | from Holm oak are predominantly controlled by a light-dependent mechanism. Owen et al.  
25 | (2002) find that, in the Mediterranean region, emissions of all compounds from *Quercus* sp.  
26 | are light dependent, the ocimene emitted by *Pinus pinea* is strongly correlated to light and an  
27 | apparent weak light dependency is exhibited by monoterpene emissions from *Cistus incanus*.  
28 | Ghirardo et al. (2010) provide the fraction of light-dependent monoterpene emission, being  
29 | 58% for Scots pine, 33.5% for Norway spruce, 9.8% for European larch, and 98–100% for  
30 | both Silver birch and Holm oak. Shao et al. (2001) and Steinbrecher et al. (1999) attribute for  
31 | Scots pine a value of 20–30% and 25–37%, respectively. Nevertheless, there is no general

1 agreement on the exact value of the temperature- and light-dependent fraction to assign for  
2 individual compound and PFT, as it appears also from the works mentioned right above.

3 Another crucial component in the estimation of BVOC emissions is the LAI, which can be  
4 either simulated using a vegetation model, or prescribed using values retrieved from satellite  
5 data or field measurements. Significant differences in terms of temporal and spatial  
6 distribution are found between the LAI estimated by measurements and the LAI calculated by  
7 models, with discrepancies of up to 100% at the global scale and more than 150% for specific  
8 biomes types (Garrigues et al., 2008; Pinty et al., 2011; Fang et al., 2012a, b). Consequently,  
9 the high uncertainty related to LAI affects the predicted regional and seasonal distribution of  
10 BVOC emissions.

11 According to our knowledge, most papers investigating BVOC emission sensitivity focus on  
12 the response of emissions to different experimental set-ups, changing, for instance, climate  
13 forcing and land use. For example, Oderbolz et al. (2013) pointed out the importance of the  
14 differences between the land-cover inventories, and of the uncertainties in the classification of  
15 land cover. Arneth et al. (2011) compared three vegetation models, changing the experimental  
16 set-up, such as the vegetation distribution and the climate forcings. Depending on the  
17 experiment considered, the total annual isoprene emissions were found to increase or decrease  
18 by more than 30%. Ashworth et al. (2010) investigated the impact of varying the climate  
19 forcing temporal resolution of isoprene emission in the MEGAN model, finding a variation of  
20 isoprene emissions of up to 7% at the global scale and up to 55% in some locations. Keenan  
21 et al. (2011) investigate the effect of canopy structure using different canopy models and they  
22 conclude that larger differences in the final emissions can be attributed to the use of different  
23 canopy models, rather than different emission model approaches. Nevertheless, very few  
24 studies have investigated the impact ~~on emissions~~ of the uncertainty of key  
25 parameters/variables, such as LAI, on emissions. One example is the work by Sindelarova et  
26 al. (2014) in which several simulations were performed with the MEGAN model to assess the  
27 sensitivity of isoprene emissions to many parameters and processes such as LAI, emission  
28 factors (EFs), CO<sub>2</sub> concentration, soil moisture, and the radiation scheme. The sensitivity  
29 simulations performed showed a variation in isoprene emissions of up to 50% at the global  
30 scale.

1 In the present work, our objectives are to (i) present the updated version of the emission  
2 module embedded in the dynamic global vegetation model ORCHIDEE, (ii) provide present-  
3 day estimates of global BVOC emissions for several relevant compounds (isoprene,  
4 monoterpenes, sesquiterpenes, methanol, acetone, formaldehyde, acetaldehyde, acetic acid,  
5 formic acid and the main speciated monoterpenes) using the new emission scheme, (iii)  
6 compare the ORCHIDEE results to the widely used emission model MEGAN, putting the two  
7 models under the same forcing conditions, but retaining their particular characteristics (see  
8 Sect. 2.5), in particular the emission scheme, classes and distribution of PFTs and LAI  
9 processing, and (iv) explore, at the global and regional scales, the BVOC emission sensitivity  
10 to EFs, LAI and LDF in ORCHIDEE and MEGAN, and to understand the reasons behind  
11 these discrepancies. ORCHIDEE is designed to provide past, present and future scenarios of  
12 emissions from vegetation, studying the links between climate, the plant phenology and  
13 emissions. It is therefore essential that the internal variability, weaknesses and inaccuracies of  
14 the emission module are extensively investigated. -The proper way to assess the correctness of  
15 a model ~~would be~~ to evaluate it against observations, as ~~it~~ is done, for example, for organic  
16 ~~aerosolsaerosol~~ by Mann et al. (2014) and Tsigaridis et al. (2014) and for tropical mountain  
17 forest carbon store by Spracklen and Righelato (2014). The evaluation of BVOC emission  
18 models against observations has already been carried out at ~~a~~-local and regional ~~scalesseale~~  
19 (i.e. Karl et al., 2007; Kuhl et al., 2007; Lathière et al., 2009; Smolander et al., 2014),  
20 demonstrating a good performance of the Guenther formulation. Nevertheless, given the  
21 ecosystem biodiversity, the huge variability of the parameters involved and the poor spatial  
22 and temporal coverage of BVOC emission observations, it is extremely difficult to infer ~~a~~  
23 ~~robustany~~ evaluation at global scale ~~from these tests~~. In such a context we can rely on model  
24 inter-comparison and sensitivity tests in order to assess the limitations and uncertainties of  
25 BVOC emission estimates, to relate them to particular key parameters/variables and to  
26 investigate their origin. In Sect. 2, the ORCHIDEE model and the updates from the previous  
27 version (Lathière et al, 2006), the MEGAN model and the technical details of the simulations  
28 are described. The comparison with other published estimates, the inter-comparison between  
29 the two models and the sensitivity tests carried out are extensively described in Sect. 3. The  
30 conclusions and future ~~directions are~~~~directionsare~~ provided in Sect. 4 and 5.

31

## 1 **2 Model developments and set-up**

### 2 **2.1 ORCHIDEE model: general description**

3 ORCHIDEE (Organizing Carbon and Hydrology in Dynamic EcosystEm) is a dynamic global  
4 vegetation model (Krinner et al., 2005; Magnan et al., 2011) that consists of two main parts:  
5 the carbon module STOMATE (Saclay-Toulouse-Orsay Model for the Analysis of Terrestrial  
6 Ecosystems) and the surface vegetation atmosphere transfer scheme SECHIBA  
7 (Schématisation des échanges hydriques à l'interface biosphere-atmosphère, in English:  
8 mapping of hydrological exchange at the biosphere/atmosphere interface).

9 STOMATE describes processes such as photosynthesis, carbon allocation, litter  
10 decomposition, soil carbon dynamics, maintenance and growth respiration. A completely  
11 prognostic plant phenology including leaf critical age, maximum LAI (leaf area index),  
12 senescence, plant tissue allocation, and leaf photosynthetic efficiency that varies depending  
13 on the leaf age, is also taken into account. The soil water budget and the exchanges of energy  
14 and water between the atmosphere and the biosphere are calculated in SECHIBA (Krinner et  
15 al., 2005). The Choisnel hydrological scheme is used with a two-meter soil column  
16 represented by two moisture layers: a superficial layer and a deep layer (Ducoudré et al.  
17 1993). The biogenic emission scheme, of which we present a new version, is embedded in this  
18 module (Lathière et al., 2006).

19 In ORCHIDEE, ecosystems are represented by 13 Plant Functional Types (PFTs, listed in  
20 Table 1). Each PFT is representative of a specific set of plant species that are grouped  
21 according to plant physiognomy (tree or grass), leaf shape (needleleaf or broadleaf),  
22 phenology (evergreen, summergreen or raingreen) and photosynthesis type for crops and  
23 grasses (C3 or C4). The main biophysical and biogeochemical processes for each PFT are  
24 described in Krinner et al. (2005) and in Maignan et al. (2011). For our study, the global  
25 vegetation distribution is prescribed for all runs using appropriate forcings, as described in  
26 paragraph 2.4.

## 1 2.2 BVOCs in ORCHIDEE: module improvements

2 The BVOC module is extensively updated, considering recent findings regarding emission  
3 schemes and field measurements. The new BVOC emission scheme is a development of the  
4 module implemented in ORCHIDEE by Lathière et al. (2006) and based on the model  
5 presented by Guenther et al. (2012). It now provides a multi-layer canopy model, where  
6 radiation is calculated following the scheme proposed by Spitter et al. (1986a, b) and the one  
7 already used in ORCHIDEE for the calculation of photosynthesis. The canopy is considered  
8 split *vertically* into several LAI layers, the number of which (up to 17) depends on the LAI  
9 value. Emissions are calculated for each layer through consideration of the sunlit and shaded  
10 leaf fractions and the light extinction and light diffusion through canopy. In a second step they  
11 are vertically summed, providing a single value for each PFT and grid point.

12 The emission flux  $F$  of a specific biogenic compound  $c$ , for a given PFT  $i$  at a LAI layer  $l$  is  
13 calculated following the Eq. (1):

$$14 \quad F_{c,i}(l) = LAI_i(l) \cdot SLW_i \cdot EF_{c,i} \cdot CTL_c(l) \cdot L_c \quad (1)$$

15 where  $LAI_i(l)$  is the leaf area index expressed in  $m^2 m^{-2}$  at a particular LAI layer and PFT,  
16  $SLW_i$  is the specific PFT leaf weight in  $g m^{-2}$ ,  $EF_{c,i}$  is the basal emissions at the leaf level for  
17 an individual compound and PFT at standard conditions of temperature ( $T = 303.15$  K) and  
18 photosynthetically active radiation ( $PAR = 1000 \mu mol m^{-2} s^{-1}$ ), expressed in  $\mu gC g^{-1} h^{-1}$ .  
19  $CTL_c$  is the emission activity factor, depending on the emitted compounds, that takes into  
20 account the deviation from the standard conditions related to temperature and PAR, and it is  
21 extensively described in the second part of the present paragraph.  $L_c$  is the activity factor  
22 simulating the impact of leaf age on emissions and is considered for isoprene and methanol.  
23 The total emission per grid cell is obtained by summing  $F_{c,i}(l)$  over the layer  $l$  and averaging  
24 the emission contribution of each individual PFT, weighted by PFT fractional land coverage.  
25 Further details on the original version of the emission module are given in Lathière et al.  
26 (2006).

27 Table 2 summarises the principal modifications compared to the previous module version. In  
28 particular, we (i) added new emitted compounds, (ii) estimated the emissions using a multi-



1 layer radiation scheme that calculates diffuse and direct components of light at different LAI  
 2 levels, (iii) inserted a dependence on light for almost all compounds, and (iv) updated the EFs.  
 3 Eight speciated monoterpenes ( $\alpha$ -pinene,  $\beta$ -pinene, limonene, myrcene, sabinene, camphene  
 4 3-carene, t- $\beta$ -ocimene) and bulk sesquiterpenes are now included in the updated ORCHIDEE  
 5 emission module. We chose these compounds because measurements have shown that they  
 6 are emitted from vegetation in the greatest abundance and because of their importance in  
 7 atmospheric chemistry, in particular regarding secondary organic aerosol formation.

8 We mentioned that, the emission module has also been modified to include a light  
 9 dependency for almost all compounds emitted. In the previous module version, indeed,  
 10 isoprene was the only compound dependent on both light and temperature, while the others  
 11 were only dependent on temperature. As detailed in Sect. 1, most recent field campaigns  
 12 highlight, for a large number of plants, the dependency of monoterpenes, sesquiterpenes and  
 13 oxygenated BVOC emissions on radiation as well. To adopt a detailed parameterisation is not  
 14 yet possible, cause to data lacking at global scale. Therefore, in the new emission module we  
 15 consider the approach described in Guenther et al. (2012), even if it is rather oversimplified.  
 16 BVOCs are now modelled to consider both light-dependent and light-independent emission  
 17 processes, and the response to temperature and light (CTL) is calculated for individual  
 18 compounds at each LAI layer (l):

$$19 \quad CTL_c(l) = (1 - LDF_c) \cdot CTLI_c + LDF_c \cdot CTLD \cdot CL(l) \quad (2)$$

20 LDF<sub>c</sub> is the light-dependent fraction of the emission, specified for each compound emitted  
 21 (Table 2). To chose the LDF value for monoterpenes, we rely on Dindorf et al. (2006), Holzke  
 22 et al. (2006), Guenther et al. (2012) and Šimpraga et al. (2013). Other LDF values were based  
 23 on Guenther et al. (2012). CTLI<sub>c</sub> is the temperature-dependent emission  
 24 response that is not light dependent and depends on individual compounds.  
 25 CTLD and CL are the temperature and light responses for the light-dependent fraction,  
 26 respectively, and are the same functions as in the previous version of the emissions module.  
 27 For all details we refer to Guenther et al. (1995).and Lathièrè et al. (2006). CTLI is equal to:

$$28 \quad CTLI = \exp(\beta(T - T_0)) \quad (3)$$

1 Where  $\beta$  is the empirical coefficient of the exponential temperature response and it is now  
2 defined as in Guenther et. al (2012) (Table 2).

### 3 **2.2.1 Emission Factor update**

4 EF determination represents one of the greatest sources of uncertainty in the quantification of  
5 BVOC emissions (Niinemets et al., 2011). Several measurement campaigns were carried out  
6 over the last decade, giving important new insights and information for re-examining the  
7 emission factors used in the emission module and correcting them accordingly. Nevertheless,  
8 the methodology to assess EFs is still under debate within the scientific community.  
9 Assigning EFs, especially on the global scale, is very tricky. In the ideal case, for each  
10 compound emitted, we should consider the EFs of all plants belonging to one particular PFT  
11 and the land cover of each plant. We could then, for each PFT and compound, make averages  
12 weighted by plant land cover, thus obtaining an average EF for each PFT and emitted  
13 compound. Unfortunately, there are not yet enough observations available to use such a  
14 methodology.

15 Several aspects make it difficult to find a good strategy to assign EFs. First, sources of  
16 information regarding EFs are very heterogeneous such as bibliographical reviews, article  
17 presenting punctual or fairly widespread measurement campaigns, and modelling  
18 experiments, making the selection of papers to use especially tricky. When a large range of  
19 EF values is documented for one particular plant species, it is not obvious whether this range  
20 is actually representative of a natural (geographical or species-to-species) variability, and can  
21 therefore be considered as valid, or originates from technical difficulties or improvements  
22 (and, in this case, if preference should be given to more recently published papers). A further  
23 difficulty is linked to the high number of plant species that can be combined together into one  
24 PFT, in comparison to the relatively small proportion of plant species and/or measurement  
25 sites worldwide that could be investigated, despite numerous and crucial field studies.  
26 Moreover, our EF review shows that EFs are highly variable from one plant to another, even  
27 if the plants belong to the same PFT. In this context, it is difficult to assign a single EF per  
28 each PFT which integrates this variability adequately. Lastly, the procedure itself used to  
29 determine EFs from field measurements adds another source of uncertainty. Indeed, EFs are  
30 derived by adjusting the measured flux at leaf level in standard conditions of

1 photosynthetically active radiation (PAR) and temperature, using algorithms such as Guenther  
2 et al. (1995). However, there is no universal agreement on the parameterization of these  
3 algorithms (Tarvainen et al., 2005; Duhl et al., 2008; Kim et al., 2010; Bracho-Nunex et al.,  
4 2011; Fares et al., 2011).

5 All these aspects underline the challenge and uncertainty of assigning one fixed EF value for  
6 each PFT in global models (Kesselmeier and Staudt, 1999, Niinemets et al., 2010 and Arneth  
7 et al. 2011), also considering that the emission estimates are very sensitive to changes in EF.

8 In this particular context, we try to establish a sufficiently consistent methodology and we  
9 follow the guidelines below to update the EFs in the ORCHIDEE emission scheme. All the  
10 values and related references used to define the new EFs are provided in Tables S1-S10 (one  
11 table for each compound) of the Supplementary Material.

12 a) First, we select only papers that provide EFs in the proper units ( $\mu\text{gC g}^{-1} \text{h}^{-1}$ ) and for  
13 standard conditions such as defined in ORCHIDEE (PAR =  $1000 \mu\text{mol m}^{-2} \text{s}^{-1}$ , temperature =  
14 30 °C), unless the information needed to convert the EFs accordingly was given.

15 b) When the most recent papers agree on a specific EF range, we discard the old references if  
16 the EF value is significantly different. In other cases all the works collected are taken into  
17 account.

18 c) First for each paper we gather all the values available per ORCHIDEE PFT and per emitted  
19 compound. In there are more values per paper, we calculate the average in order to have one  
20 EF per compound, PFT and paper.

21 d) Then, for each compound and each PFT, we choose an EF that is in the range of the  
22 collected values and is the closest to the average and median calculated. When one EF value  
23 cannot be clearly assigned, we take a value between the average or the median and the  
24 previous ORCHIDEE EF values (Lathière et al., 2006). Considering the high sensitivity of the  
25 emission module to EF variation, in order to avoid unreliable estimate, in the case of  
26 ambiguity, for the highly emitted compounds, in particular for isoprene, a more conservative  
27 approach is adopted and the EF values of the previous version are kept.

1 e) In choosing the new EFs, in the case of very few or inconclusive information, EF  
2 variability between the different PFTs of the old version of ORCHIDEE (Lathière et al.,  
3 2006) and/or MEGAN (Guenther et al., 2012) is taking into account.

4 f) For each compound we check *a posteriori* that the new set of EFs provides a regional  
5 distribution which is consistent with the orders of magnitude expected and given in the  
6 literature. Only for monoterpenes for tropical PFTs we replace the first value selected (2.5  
7  $\mu\text{gC g}^{-1} \text{h}^{-1}$ ) with the current value (2.0  $\mu\text{gC g}^{-1} \text{h}^{-1}$ ).

8 Table 3 shows the new and old EFs used in the emission module, and Table 4 shows the EF  
9 values for each speciated monoterpene as a percentage of the bulk monoterpene EF value. As  
10 shown in Table 3, the revision leads to the modification of almost all EFs. In some cases, the  
11 EF differences in comparison with the previous version are very significant. Regarding  
12 isoprene, boreal needleleaf deciduous PFT is now recognized as a less important emitter (EF  
13 = 8  $\mu\text{gC g}^{-1} \text{h}^{-1}$  in the old version and EF = 0.5  $\mu\text{gC g}^{-1} \text{h}^{-1}$  in the new one). The new EF is  
14 decided considering the EF proposed by Guenther et al. (2006) (0.003  $\mu\text{gC g}^{-1} \text{h}^{-1}$ ), Guenther  
15 et al. (2012) (0.002  $\mu\text{gC g}^{-1} \text{h}^{-1}$ ), Steinbrecher et al. (2009) and Karl et al., 2009 (0.44  $\mu\text{gC g}^{-1}$   
16  $\text{h}^{-1}$ ), Smiatek and Steinbrecher (2006) (0.10  $\mu\text{gC g}^{-1} \text{h}^{-1}$ ) and Klinger et al. (2002) (2.23  $\mu\text{gC}$   
17  $\text{g}^{-1} \text{h}^{-1}$ ) (more details in the Supplement). Our choice is confirmed by Ruuskanen et al.  
18 (2007), who assign a contribution of less than 3% of the VOC emission to isoprene, 2-methyl-  
19 3-buten-2-ol (hereafter referred to it simply as MBO) and 1,8-cineole, for larch, which is the  
20 major component of boreal needleleaf deciduous PFT.

21 Furthermore, we consider boreal broadleaved deciduous trees to be a higher emitter of  
22 isoprene than in the previous model version (now EF = 18  $\mu\text{gC g}^{-1} \text{h}^{-1}$ , while before EF = 8  
23  $\mu\text{gC g}^{-1} \text{h}^{-1}$ ), since the papers collected propose particularly high values, such as Guenther et  
24 al. (2012) (22.7  $\mu\text{gC g}^{-1} \text{h}^{-1}$ ), Guenther et al. (2006) (30.8  $\mu\text{gC g}^{-1} \text{h}^{-1}$ ), Stewart et al. (2003)  
25 (33.9  $\mu\text{gC g}^{-1} \text{h}^{-1}$ ) and Smiatek and Steinbrecher (2006) (18.8  $\mu\text{gC g}^{-1} \text{h}^{-1}$ ). For  
26 monoterpenes, we assign a significantly higher EF (from 0.8  $\mu\text{gC g}^{-1} \text{h}^{-1}$  to 2.0  $\mu\text{gC g}^{-1} \text{h}^{-1}$ )  
27 to tropical broadleaf evergreen and deciduous PFTs. For MBO, we reduce the EF for the  
28 temperate needleleaf evergreen PFT from 20  $\mu\text{gC g}^{-1} \text{h}^{-1}$  to 1.4  $\mu\text{gC g}^{-1} \text{h}^{-1}$  (Tarvainen et al.,  
29 2005; Hakola et al., 2006; Chang et al., 2009; Kim et al., 2010).

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~~EFs represent one of the greatest sources of uncertainty in the quantification of BVOC emissions (Niinemets et al., 2011). Several measurement campaigns were carried out over the last decade, giving important insights and information for re-examining thoroughly the emission factors used in the emission module and correcting them accordingly. Nevertheless the methodology to assess EFs is still under debate within the scientific community.~~

~~Assigning EFs, especially on the global scale, is very tricky. In the ideal case, for each compound emitted, we should consider the EFs of all plants belonging to one particular PFT and the land cover of each plant. We could then, for each PFT and compound, make averages weighted on plant land cover, thus obtaining an average EF for each PFT and emitted compound. Unfortunately, there are not yet enough observations available to use such a methodology.~~

~~There are several factors that make it difficult to find a good strategy to assign EFs valid for all compounds:~~

- ~~1. — depending on the compound and the PFT, the number of measurements available differs considerably, and the statistical accuracy of the EFs may therefore be very variable;~~
- ~~2. — in some cases, the most recent measurements contradict the older ones, therefore it is reasonable to consider only the most recent data. However, in other cases the difference between recent and older measurements is not so clear, therefore it is not easy to understand if it is better to consider less recent measurements in the evaluation of EFs;~~
- ~~3. — considering the values of EFs that we collected from the literature, we note that they are actually often related to a small number of plant species from mostly the same measurement sites. The values found could not be considered as a significant representative set for the PTFs at the global scale;~~
- ~~4. — in many papers focussing on modelling, the EFs presented are either taken directly from previous models, or are based on a review or on measurements available. In this context, it is very difficult to make consistent averages and understand which values found should be taken into account.~~

1 Taking all this into account we decided to proceed as follows.

2 ~~As a general rule, and based on an extensive review of publications, we select papers, in which~~  
3 ~~it is possible to convert the EFs into the units and at the standard conditions that are~~  
4 ~~considered in ORCHIDEE (PAR = 1000  $\mu\text{mol m}^{-2} \text{s}^{-1}$ , temperature = 30 °C). We do not~~  
5 ~~always perform an average over all values collected, but we use a qualitative and comparative~~  
6 ~~method to justify the EFs.~~

7 ~~In the case of isoprene, we principally consider the most recent papers, the ones that present~~  
8 ~~new measurements or original review. The review carried out for EFs confirms that the values~~  
9 ~~used in the previous version (Lathière et al., 2006) are consistent with the latest~~  
10 ~~measurements. Only for certain PFTs it is necessary to change the value of EF. Indeed,~~  
11 ~~isoprene has already been widely measured for several years, while other BVOCs have been~~  
12 ~~documented only more recently.~~

13 ~~In the case of the other compounds, since there are fewer papers and the information is not so~~  
14 ~~well consolidated, we adopt a similar strategy but we are less restrictive in paper choice. In~~  
15 ~~general, we perform averages considering the different values from all papers collected, and~~  
16 ~~we compare these averages to the older values in ORCHIDEE. Whenever big differences~~  
17 ~~between the new value and the old one were found, we look in detail at the various papers to~~  
18 ~~see if there are some outliers, and if so, we do not consider them in the EF evaluation.~~

19 ~~Table 3 show the new and old EFs used in the emission module and Table 4 presents EF~~  
20 ~~values for each speciated monoterpene as a percentage of the bulk monoterpene EF value. As~~  
21 ~~shown in Table 3, the revision leads to the modification of almost all EFs. In some cases, the~~  
22 ~~EF differences in comparison with the previous version are very significant. Regarding~~  
23 ~~isoprene, boreal needleleaf deciduous PFT is now recognized as a less important emitter (EF~~  
24 ~~= 8  $\mu\text{gC g}^{-1} \text{h}^{-1}$  in the old version and EF = 0.5  $\mu\text{gC g}^{-1} \text{h}^{-1}$  in the new one). We based the~~  
25 ~~choice on papers focussing on reviewed or measured EFs, such as Guenther et al. (2006) (EF~~  
26 ~~= 1.44  $\mu\text{gC g}^{-1} \text{h}^{-1}$ ), Guenther et al. (2012) (0.002  $\mu\text{gC g}^{-1} \text{h}^{-1}$ ), Steinbrecher et al. (2009) (EF~~  
27 ~~= 0.44  $\mu\text{gC g}^{-1} \text{h}^{-1}$ ), and Smiatek and Steinbrecher (2006) (EF = 0.09  $\mu\text{gC g}^{-1} \text{h}^{-1}$ ) and~~  
28 ~~Klinger et al. (2002) (EF = 0.52  $\mu\text{gC g}^{-1} \text{h}^{-1}$ ). All these values are much lower than those~~  
29 ~~assigned by Lathière et al. (2006), and their average is 0.5  $\mu\text{gC g}^{-1} \text{h}^{-1}$ , which we set as the~~  
30 ~~new value. In this case, we do not consider the other papers where EFs are directly taken from~~

1 ~~previous models or for which the source of information was not clear. Our choice is~~  
2 ~~confirmed by Ruuskanen et al. (2007), who assign a contribution of less than 3% of the VOC~~  
3 ~~emission to isoprene, 2-methyl-3-buten-2-ol (hereafter we refer to it simply as MBO) and 1,8-~~  
4 ~~cinole, for larch, which is the major component of boreal needleleaf deciduous PFT.~~

5 ~~Furthermore, we now consider boreal broadleaved deciduous trees to be a higher emitter of~~  
6 ~~isoprene than in the previous model version (now  $EF = 18 \mu\text{gC g}^{-1} \text{h}^{-1}$ , while before  $EF = 8$~~   
7  ~~$\mu\text{gC g}^{-1} \text{h}^{-1}$ ), since most of the papers collected propose particularly high values, such as~~  
8 ~~Levis et al. (2003) ( $24 \mu\text{gC g}^{-1} \text{h}^{-1}$ ), Arneeth et al. (2011) ( $45 \mu\text{gC g}^{-1} \text{h}^{-1}$ ), Guenther et al.~~  
9 ~~(2006) ( $42.3 \mu\text{gC g}^{-1} \text{h}^{-1}$ ) and Guenther et al. (2012) ( $22.7 \mu\text{gC g}^{-1} \text{h}^{-1}$ ). For monoterpenes, a~~  
10 ~~significantly higher EF (from  $0.8 \mu\text{gC g}^{-1} \text{h}^{-1}$  to  $2.2 \mu\text{gC g}^{-1} \text{h}^{-1}$ ) is now assigned to tropical~~  
11 ~~broadleaf evergreen and deciduous PFTs. For MBO the EF for the temperate needleleaf~~  
12 ~~evergreen PFT is reduced from  $20 \mu\text{gC g}^{-1} \text{h}^{-1}$  to  $1.4 \mu\text{gC g}^{-1} \text{h}^{-1}$  (Tarvainen et al., 2005;~~  
13 ~~Hakola et al., 2006; Chang et al., 2009; Kim et al., 2010).~~

14 ~~Our review analysis confirms a large variability in EFs, even among plants that are usually~~  
15 ~~represented by one single PFT in global vegetation models (characterized by the same~~  
16 ~~physiognomy, leaf shapes and photosynthesis type). It is therefore a source of high~~  
17 ~~uncertainty to assign one fixed EF value for each PFT in global models, as also pointed out by~~  
18 ~~Kesselmeier and Staudt (1999) and Arneeth et al. (2011). Moreover, the procedure used to~~  
19 ~~determine emission factors from field measurements adds an additional source of uncertainty.~~  
20 ~~Indeed EFs are derived by adjusting the measured flux at leaf level at a standard conditions of~~  
21 ~~light photosynthetically active radiation (PAR) and temperature, using algorithms such as~~  
22 ~~Guenther et al. (1995). However, there is no universal agreement on the parameterization of~~  
23 ~~these algorithms (Tarvainen et al., 2005; Duhl et al., 2008; Kim et al., 2010; Bracho Nunex et~~  
24 ~~al., 2011; Fares et al., 2011).~~

### 26 **2.3 MEGAN description**

27 The Model of Emissions of Gases and Aerosols from Nature (MEGAN) is a modelling system  
28 for the estimation of emission fluxes of biogenic organic compounds from terrestrial  
29 vegetation. The basis of the model is a simple mechanistic approach established by Guenther

1 et al. (1991, 1993, 1995), which links emissions with the main environmental driving factors  
2 such as solar radiation and leaf temperature. Further development of the algorithm led to the  
3 inclusion of leaf ageing, soil moisture impact on the emissions, and effects of the loss and  
4 production of compounds within a forest canopy (Guenther et al., 2006). The current version  
5 of the model, MEGANv2.1, also includes a full canopy module. The model calculates light  
6 and temperature conditions inside a canopy by evaluating the energy balance on five canopy  
7 levels. Additionally, emissions of each compound are considered to have light-dependent and  
8 light-independent components defined by the light dependent fraction (LDF). For a detailed  
9 description of emission equations and parameterization we refer to Sect. 2 in Sindelarova et al  
10 (2014) and Guenther et al. (2012).

11 MEGANv2.1 is available either as a stand-alone version or embedded in the Community  
12 Land Model version 4 (CLM4) (Lawrence et al., 2011) of the Community Earth System  
13 Model (CESM) (Gent et al., 2011). When operating in the stand-alone version, the driving  
14 variables, such as meteorological input data, vegetation description and leaf area index, need  
15 to be provided by the user. When running MEGAN inside CLM4, the input data can be  
16 provided by the CESM atmospheric and land surface models on-line at each time-step. In this  
17 work, we use the stand-alone model version of MEGANv2.1, hereafter simply referred to as  
18 MEGAN.

19 MEGAN estimates emissions of 19 chemical compound classes, which are then redistributed  
20 into 147 final output model species, such as isoprene, monoterpene and sesquiterpene species,  
21 methanol, carbon monoxide, alkanes, alkenes, aldehydes, ketones, acids and other oxygenated  
22 VOCs. Although the input parameters, such as vegetation description and emission potentials,  
23 can be defined by the user, MEGAN comes with a default definition of PFTs and the emission  
24 factors assigned to them. The vegetation distribution is described with fractional coverage of  
25 16 PFT classes consistent with those of the CLM4 model (Lawrence and Chase, 2007). The  
26 emission potential of each modelled species is calculated based on the PFT coverage and  
27 emission factor of each PFT category. For several VOC compounds, emission potentials can  
28 be defined in the form of input maps. Emission potential maps with global coverage and high  
29 spatial resolution for isoprene, main monoterpene species and MBO are provided together  
30 with the MEGAN code.



1 MEGAN is widely applied for the estimation of biogenic VOC emissions at both regional and  
2 global scales (e.g., Guenther et al., 2006, 2012; Müller et al., 2008; Millet et al., 2010;  
3 Sindelarova et al., 2014; Situ et al., 2014; Stavrakou et al., 2014) and serves for the evaluation  
4 of the impact of BVOCs on atmospheric chemistry by coupling the model with chemistry  
5 transport models (e.g. Heald et al., 2008; Pfister et al., 2008; Emmons et al., 2010; Fu and  
6 Liao, 2012; Tilmes et al., 2015).

## 7 **2.4 Model set-up and sensitivity tests**

8 The objectives of the group of simulations are: (i) to provide global estimates of BVOC  
9 emissions for a large variety of compounds over the 2000–2009 period, (ii) to investigate the  
10 differences and similarities between the ORCHIDEE and MEGAN results regarding the  
11 spatial, inter-annual and inter-seasonal variability of emissions, (iii) to analyze the response of  
12 BVOC emissions to the variation of some key variables and parameters such as the LAI and  
13 LDF. Table 5 summarizes the simulations performed in this study and their principal  
14 characteristics.

15 We carried out a total of 54 sets of runs:

- 16 1. ~~Two~~two simulations for the 2000–2009 period performed by both models using each  
17 model's standard configuration, but with the same climatology (ORC\_CRU and MEG\_CRU).
- 18 2. ~~One~~one simulation for the 2000–2009 period with MEGAN using the LAI estimated  
19 by ORCHIDEE (MEG\_CRULAI).~~.)~~
- 20 3. ~~Four~~four simulations for the year 2006 by both models, using the ORCHIDEE LAI  
21 scaled by a factor 0.5 and 1.5, respectively (ORC\_LAI05, ORC\_LAI15, MEG\_LAI05 and  
22 MEG\_LAI15).
- 23 4. One simulation for the year 2006 forcing ORCHIDEE with the MODIS LAI used in  
24 MEGAN standard configuration.
- 25 5. ~~Two~~two simulations for the year 2006 performed by both models, where we output  
26 two test species, the first one totally dependent on light (LDF=1) and the second one totally  
27 independent on light (LDF=0) (ORC\_LDF and MEG\_LDF). The output time frequency is one  
28 hour for this run.

~~The run sets 3 and 4 are carried out for the year 2006, which is estimated as an averaged year regarding the BVOC emissions calculated by MEGAN and ORCHIDEE in the 10 years of simulation.~~

All simulations are performed at the global scale with a spatial resolution of  $0.5^\circ \times 0.5^\circ$ . We use the CRU-NCEP v5.2 meteorological forcing database (<http://dods.extra.cea.fr/data/p529viov/cruncep>) providing temperature, pressure, humidity, wind speed and shortwave solar radiation. This forcing is based on the 6-hourly  $2.5^\circ$  NCEP/NCAR re-analysis (Kalnay et al., 1996) combined with the CRU TS 2.1 monthly anomalies (Mitchell and Jones, 2005).

~~The run sets 3 and 4 are carried out for the year 2006, which is estimated as an averaged year regarding the BVOC emissions calculated by MEGAN and ORCHIDEE in the 10 years of simulation.~~

For the ORCHIDEE model a spin-up of 20 years is first performed to balance the leaf stock. The spin-up is based on a 10-year loop using meteorological forcing for the year 1989, followed by a 10-year simulation from 1990 to 1999. In ORCHIDEE, the global vegetation distribution for the 13 PFTs is prescribed using the Land-Use History (LUHa.rc2) related to the year 2000 (Hurtt et al. 2006). ~~The database can be found in [http://dods.extra.cea.fr/work/p86ips1/IGCM/BC/SFR/OL2/PFTmap\\_1850to2005\\_AR5\\_LUHa.rc2](http://dods.extra.cea.fr/work/p86ips1/IGCM/BC/SFR/OL2/PFTmap_1850to2005_AR5_LUHa.rc2).~~ ~~The database can be found in [http://dods.extra.cea.fr/work/p86ips1/IGCM/BC/SFR/OL2/PFTmap\\_1850to2005\\_AR5\\_LUHa.rc2](http://dods.extra.cea.fr/work/p86ips1/IGCM/BC/SFR/OL2/PFTmap_1850to2005_AR5_LUHa.rc2).~~ In MEGAN the distribution for the 16 PFTs is consistent with the Community Land Model v4 (Lawrence and Chase, 2007) and related to the year 2000. Table 1 gives the global surfaces covered by the different PFTs in ORCHIDEE and MEGAN. For the present work, MEGAN in the standard configuration is forced by the LAI retrieved by MODIS (Yuan et al., 2011; <http://globalchange.bnu.edu.cn/research/lai/>).

~~We present here the results of MEGAN forced either by the LAI retrieved by MODIS (Yuan et al., 2011) or by the LAI provided by the ORCHIDEE simulation (see Table 5 for all simulation details).~~

In ORCHIDEE, the activity factor ( $L_c$ ) is kept as in Lathière et al. (2006), considering four leaf age classes (new, young, mature and old leaves). For methanol,  $L_c$  is equal to 1 for new and young leaves and equal to 0.5 for mature and old leaves, while for isoprene,  $L_c$  is equal to

1 | 0.5 for new and old leaves and equal ~~to 1 to~~ 1.5 for young and mature leaves. In MEGAN, the  
2 |  $L_c$  values are taken from Table 4 in Guenther et al. (2012); in particular, for isoprene,  $L_c$  is  
3 | equal to 0.05, 0.6, 1 and 0.9, and for methanol it is equal to 3.5, 3.0, 1.0, and 1.2 for the four  
4 | leaf age classes. For both models, no soil moisture activity factor is taken into account. The  
5 | annual  $CO_2$  concentration varies along the simulation from a value of 368 ppm in 2000 to 385  
6 | ppm in 2009. In ORCHIDEE, the variation of  $CO_2$  concentration can indirectly impact on the  
7 | BVOC emission as it affects leaf growth, while in MEGAN, a  $CO_2$  inhibition factor on  
8 | isoprene emission based on Heald et al. (2009) is activated. As the  $CO_2$  variation in this 10-  
9 | year simulation is low, the inhibition effect is considered insignificant  
10 | (~~Sindelarova~~~~Sinderalova~~ et al. 2014) in this context. For ORCHIDEE, LDF and the  $\beta$   
11 | coefficient values are given in Table 2. For MEGAN, the values of LDF and  $\beta$  are those  
12 | presented in Table 4 in Guenther et al. (2012).

## 13 | 2.5 Differences between ORCHIDEE and MEGAN emission algorithms

14 | While starting from a similar approach the ORCHIDEE and MEGAN emission modules  
15 | differ significantly in their parameterization and variable description and calculation. We list  
16 | below the main differences.:

17 | 1) One of the principal differences in the two emission schemes is the approach on LAI.  
18 | ORCHIDEE calculates the LAI at each model time step for each PFT and grid cell, taking  
19 | into account a full plant phenology scheme and the environmental condition (temperature,  
20 | radiation, precipitations,  $CO_2$ , etc.), while MEGAN stand-alone version used in this study,  
21 | does not compute the LAI, rather, it has to be provided as an external forcing averaged over  
22 | the vegetated part of the grid cell.

23 | 2) In~~1)~~ in ORCHIDEE, the formulation of CTLD and CL is the same as in Guenther et al.  
24 | (1995) (see Eq. 9 and 10), while in MEGAN it is defined by Eq. (8), (9), and (10) in Guenther  
25 | et al. (2012). In particular in Guenther et al. (2012) the parameters of the CTLD formulation  
26 | vary according to the average solar radiation over the past 24h and 240h, and this dependence  
27 | is different for diffuse and direct radiation. We calculate the CTLD obtained with this  
28 | formulation considering different incoming solar radiations and we observe that the CTLD for

1 direct light is around twice that for diffuse light. In ORCHIDEE the CTLD parameters are  
2 fixed and are the same for diffuse and direct radiation.;

3 ~~3) The~~ radiation scheme in ORCHIDEE and MEGAN is based on the same approach  
4 (Spitter et al., 1986 a,b-), but the parameterization and formulation used are different. For  
5 example, the number of vertical layers and their distribution over the LAI significantly differ  
6 between the two models: up to 17 in ORCHIDEE and up to 5 in MEGAN. MEGAN also  
7 takes into account the infrared radiation in emission calculation.;

8 ~~4) The PFT~~ classes and their distribution are not the same in the two models  
9 (Table 1) and they are not interchangeable without significantly modifying the models.;

10 ~~4) LAI is considered in a different way in the two models. ORCHIDEE calculates the LAI at~~  
11 ~~each model time step for each PFT and grid cell, taking into account a full plant phenology~~  
12 ~~scheme. MEGAN, on the other hand, does not compute the LAI, rather, it has to be provided~~  
13 ~~as an input averaged over the vegetated part of the grid cell;~~

14 ~~5) In~~ ORCHIDEE, emissions are calculated for each PFT using the associated EF and LAI.  
15 Next, they are averaged over the grid cell, considering the PFT land cover surface, as  
16 described in Sect. 2.2. In MEGAN, vegetated emission potential (EP) is calculated over the  
17 grid cell and multiplied by the average LAI over the vegetated part of the grid cell. In  
18 MEGAN, vegetated potential emission maps are provided for isoprene,  $\alpha$ -pinene,  $\beta$ -pinene, 3-  
19 Carene, limonene, myrcene,  $t$ - $\beta$ -ocimene and sabinene, while for the other compounds EPs  
20 are calculated starting from the EFs per PFT and the PFT land cover distribution. This is a  
21 significantly different approach. However, for ORCHIDEE, we find that global emissions  
22 calculated using the EP and LAI per grid cell (the MEGAN approach) are only 5-12% lower  
23 in comparison with the emissions calculated in the standard way. Isoprene presents the lowest  
24 differences and monoterpenes the highest.;

25 ~~In~~ the ORCHIDEE model, the dependence ~~on LAI~~ of the light independent emission on  
26 LAI is linear, as shown in the Eq. (1) and (2) of the present work. Whereas in MEGAN, the  
27 dependence on LAI is given by the  $\gamma_{LAI}$  factor that is equal to  $(0.49 \cdot LAI)/(1+0.2 \cdot LAI^2)^{0.5}$   
28 (Guenther et al., 2006). The implications of this difference are detailed in Sect. 3.4.2.;

1 7) ~~In~~ MEGAN, leaf age classes are derived from consideration of the variation between the  
2 LAI value of the current and preceding month, following a highly parameterised scheme. In  
3 ORCHIDEE, leaf age classes are calculated on-line considering the plant leaf growth and leaf  
4 turnover at each model time step (30 minutes).

5 8) ~~In~~ ORCHIDEE, hydrological processes are explicitly calculated, as briefly described in  
6 Sect. 2.1.

7 9) In ORCHIDEE, the air temperature is used to compute emission, while in MEGAN the leaf  
8 temperature is considered.

9

## 10 **3 Results**

### 11 **3.1 Global budgets**

12 As already discussed at the end of the introduction, the validation of BVOC emissions at the  
13 global scale is a complex issue because of the poor data coverage in many regions and the  
14 general lack of year-round measurements. Satellite observations provide very useful  
15 information, especially regarding the order of magnitude and the seasonal and regional  
16 variability of emissions, but the most abundant VOC species are not directly measured (such  
17 as isoprene and monoterpenes). Satellite measurements are also subject to large uncertainties  
18 arising from difficulties in the retrieval of the atmospheric concentration of short-lived  
19 compounds from space or in separation of the different sources (for instance terrestrial  
20 biogenic, anthropogenic, oceanic etc.) and the various compounds themselves. Global  
21 emission estimates are generally performed using models, or from the application of inverse  
22 modelling techniques that combine the measurements (from satellite, ground or aircraft  
23 measurements) and models, providing emissions for compounds such as methanol (Jacob et  
24 al. 2005; Millet et al., 2008; Stavrou et al., 2009; Hu, et al., 2011; Wells et al., 2012, 2014)  
25 and acetaldehyde (Jacob et al. 2002; Millet et al, 2010). Isoprene emissions have also been  
26 inferred from satellite formaldehyde concentration (Shim et al., 2005; Palmer et al., 2006;  
27 Stavrou et al., 2011; Barkley et al., 2013; Bauwens et al., 2013; Stavrou et al., 2014).

28 At the global scale, the main way to evaluate the results obtained in the present study is to  
29 compare them with the most recent emission budgets derived either from other model runs or  
30

1 from the inversion of satellite data. We have compared emissions from a large number of  
2 estimates published so far, over the 1980–2010 period, with the global emission budgets  
3 obtained from ORC\_CRU and MEG\_CRU simulations, the results of which are summarized  
4 in Fig. 1. The emissions, calculated by the earlier version of the emission module (black  
5 squares, Fig. 1) (Lathière et al., 2006), are particularly high, as already pointed out by  
6 Sindelarova et al. (2014). Methanol ( $106.1 \text{ Tg C yr}^{-1}$ ) and acetaldehyde ( $42.2 \text{ Tg C yr}^{-1}$ )  
7 emissions are twice as large, and formaldehyde emissions ( $10.0 \text{ Tg C yr}^{-1}$ ) are up to 5 times  
8 greater than the other estimates. The results of the new module version (ORC\_CRU, green  
9 stars) are more in the range of other published estimates. Although the MEG\_CRU simulation  
10 was carried out using the same MEGAN version as in Guenther et al. (2012) (blue hexagons,  
11 Fig. 1), there is a noticeable difference between the two emission budgets (especially for  
12 isoprene, monoterpenes and acetaldehyde), even when considering results for the same year  
13 (e.g. 2000). Using reanalysis provided by Qian et al. (2006) as climate forcings for the year  
14 2000, Guenther et al. (2012) report BVOC emissions of  $472 \text{ Tg C yr}^{-1}$  for isoprene,  $124 \text{ Tg C}$   
15  $\text{yr}^{-1}$  for monoterpenes (considering the speciated monoterpenes accounted in this work) and  
16  $11.5 \text{ Tg C yr}^{-1}$  for acetaldehyde. Our MEG\_CRU simulation estimates for 2000 are  $410 \text{ Tg C}$   
17  $\text{yr}^{-1}$ ,  $72 \text{ Tg C yr}^{-1}$ , and  $8.3 \text{ Tg C yr}^{-1}$  for isoprene, monoterpenes and acetaldehyde,  
18 respectively. As was already pointed out by Arneth et al. (2011), our results confirm that the  
19 differences between existing meteorological forcings can lead to substantial differences in the  
20 emission estimates (green triangles, first plot of Fig. 1).

21 Table 6 shows the annual emissions calculated by ORCHIDEE and MEGAN (ORC\_CRU and  
22 MEG\_CRU simulations) at the global scale and for the northern (lat: 0–30N) and southern  
23 (lat: 30S–0) tropics, the northern (lat: 30N–60N) and southern (lat: 30S–60S) temperate  
24 latitudes, and the northern boreal (lat: 60N–90N) regions, averaged over the 2000–2009  
25 period. At the global scale, the two models are in a good agreement. Isoprene is the main  
26 compound emitted with a global amount of  $465 \text{ Tg C yr}^{-1}$  for ORCHIDEE, accounting for  
27 61% of total BVOC emissions (estimated to  $757 \text{ Tg C yr}^{-1}$ ), and  $428 \text{ Tg C yr}^{-1}$  for MEGAN,  
28 accounting for 64% of total BVOCs (estimated at  $666 \text{ Tg C yr}^{-1}$ ). The following most  
29 abundant compounds are monoterpenes, accounting for 12% of the total for ORCHIDEE and  
30 11% for MEGAN, and methanol, accounting for 5% of the total BVOC emissions for

1 ORCHIDEE and 6% for MEGAN. Acetone, sesquiterpenes and acetaldehyde each represent  
2 1% to 4% of the total BVOCs for both models, while other compounds contribute to less than  
3 0.5%.

4 Compared to ORCHIDEE, MEGAN global emission are 8% lower for isoprene, 8% higher  
5 for methanol, 17% lower for acetone, 18% lower for monoterpenes, 39% lower for  
6 sesquiterpenes and 25% for MBO. Regarding speciated monoterpenes, major differences arise  
7 from  $\alpha$ -pinene (around 40%) while the relative difference between ORCHIDEE and MEGAN  
8 is between -8% and +16% for other compounds. The highest contribution to total emission is  
9 attributed to the tropical regions ranging between 34% and 50% for the southern tropics and  
10 between 31.5% and 39.5% for the northern tropics, depending on the compound (except  
11 MBO). Both models calculate the contribution of northern temperate regions to the total  
12 emission ranging from 6% to 24% and a contribution of less than 5% for southern temperate  
13 regions and northern boreal regions. For MBO, field campaigns measured significant  
14 emissions only for a few plant types such as Ponderosa and Scots pine (Kim et al., 2010;  
15 Tarvainen et al. 2005; Harley et al., 1998). The EF values in the ORCHIDEE and MEGAN  
16 models are consequently significant only for the PFTs representing these plants (TeNeEv and  
17 BoNeEv), leading to notable emissions in the temperate North latitudes and contributing 88%  
18 for ORCHIDEE and 63% for MEGAN of the global MBO emission.

19 At the regional scale, the largest differences between ORCHIDEE and MEGAN in terms of  
20 absolute values appear in the northern temperate regions for isoprene, where emissions are 21  
21  $\text{Tg C yr}^{-1}$  higher in ORCHIDEE. Indeed, the marked seasonal cycle of emissions for northern  
22 temperate latitudes implies that the largest differences between ORCHIDEE and MEGAN  
23 occur in summer. The differences between the two models are, in this case, directly linked to  
24 discrepancies in the EFs and in the occupying surface of the PFTs at these latitudes (see Fig.  
25 3, plots in the last row). In particular, in northern temperate region the highest discrepancies  
26 are mainly due to the different PFT surface coverage for grass and crop and the higher EFs  
27 values in ORCHIDEE in comparison to MEGAN. Actually, in ORCHIDEE C3Gr covers the  
28 42% of vegetated surface with an  $\text{EF} = 12 \mu\text{gC g}^{-1} \text{h}^{-1}$ , C3Ag covers the 18% with an  $\text{EF} = 5$   
29  $\mu\text{gC g}^{-1} \text{h}^{-1}$ , while in MEGAN the C3GrCool occupies the 20% with an  $\text{EF} = 2 \mu\text{gC g}^{-1} \text{h}^{-1}$ ,  
30 C3GrCold the 6% with an  $\text{EF} = 4 \mu\text{gC g}^{-1} \text{h}^{-1}$ , C3GrCool the 20% with an  $\text{EF} = 2 \mu\text{gC g}^{-1} \text{h}^{-1}$

1 and Crop the 23.2% with an EF = 0.12  $\mu\text{gC g}^{-1} \text{h}^{-1}$ . This example raises an important issue.  
2 Considering the EF assigned to C3Gr, we lowered its value with respect to the previous  
3 version, from 16 to 12  $\mu\text{gC g}^{-1} \text{h}^{-1}$ . These is a compromise value, chosen so that we do not  
4 excessively bias the emissions in other areas. C3Gr is, indeed, strongly present in other  
5 regions: 13% of northern tropical areas, 22% of southern tropical areas and 32% of the total  
6 vegetation surface. A more detailed description of the different crop and grass (in other words  
7 with a larger number of PFTs) could lead to more accurate results. The same consideration  
8 could be done for almost all the other PFTs.

9 This illustrates the strong impact of different choices in EF allocation, not only regarding  
10 global estimates, but also for geographical variation in emissions. For the other species the  
11 largest differences occur in tropical regions. For example, the emission differences between  
12 ORCHIDEE and MEGAN in the northern and southern tropics are  $-2.2 \text{ Tg C yr}^{-1}$  and  $-2.1$   
13  $\text{Tg C yr}^{-1}$  for methanol,  $4.3 \text{ Tg C yr}^{-1}$  and  $10.2 \text{ Tg C yr}^{-1}$  for monoterpenes and  $3.9 \text{ Tg C yr}^{-1}$   
14 and  $4.9 \text{ Tg C yr}^{-1}$  for sesquiterpenes.

### 15 **3.2 Inter-annual and inter-seasonal emission variations**

16 Fig. 2 shows the annual and monthly global emission budgets of ORC\_CRU and MEG\_CRU.  
17 The models have very similar annual trends and monthly variations for almost all compounds,  
18 illustrating that climate variables, in particular temperature and solar radiation, are the major  
19 driving factors at the global scale for inter-annual and inter-monthly variability.~~variabilities~~.

20 Nevertheless large differences appear for isoprene. The emissions in ORC\_CRU present a  
21 clear seasonal cycle with an emission maximum in July and August that is not simulated in  
22 MEG\_CRU results. Indeed, the major differences can be identified in July and August, when  
23 global emissions in MEG\_CRU are, on average, lower by  $11.5 \text{ Tg C month}^{-1}$  and  $9.0 \text{ Tg C}$   
24  $\text{month}^{-1}$  compared with ORC\_CRU. -The monthly zonal average for tropical, temperate and  
25 northern boreal latitudes regions are shown in Fig. 3. We observe, as mentioned in Sect. 3.1,  
26 that the ORCHIDEE emissions are significantly higher in northern temperate regions  
27 compared with MEGAN, with a marked seasonal cycle and the largest differences between  
28 the two models occurring in summer. In July (August) in particular, calculated isoprene  
29 emissions in ORC\_CRU are about  $4 \text{ Tg C month}^{-1}$  ( $5.5 \text{ Tg C month}^{-1}$ ) higher than in



1 MEG\_CRU. In July (August), a further important contribution to the global emission peak is  
2 attributed to the northern and southern tropics, where ORCHIDEE isoprene emissions are  
3 higher, in total, by about 4 Tg C month<sup>-1</sup> (5 Tg C month<sup>-1</sup>) in comparison to MEGAN in July  
4 (August), (Fig. 3, first plot, left column).

5 MEGAN isoprene emissions are indeed dominant from the tropical regions, leading to an  
6 overall stable global emission budget throughout the year (Fig. 2). The northern and southern  
7 tropics have an opposite seasonal cycle, with isoprene emissions coming mostly from the  
8 northern tropics between March and October and from the southern tropics for the rest of the  
9 year (Fig. 3). The overall stable global emission budget is generally characteristic of the  
10 compounds for which tropical regions are strong emitters all year round, such as  
11 sesquiterpenes (Table 3 and Fig. 3). On the other hand, the global BVOC emissions for which  
12 temperate regions are strong emitters will have a more marked seasonal cycle (Fig. 2), such as  
13 for methanol and isoprene in ORCHIDEE.

14 Indeed, the two models exhibit a very different inter-seasonal variation in terms of isoprene  
15 global emissions. Sindelarova et al. (2014) compared the monthly isoprene emissions time  
16 series from different data-sets, showing, for some of them, an inter-seasonal variation similar  
17 to ORCHIDEE, and, for others, no seasonal cycle. Based on our current knowledge, we  
18 cannot establish which is the best representation because of the lack of long-term observations  
19 at the global scale. However, we can extensively investigate why the differences between the  
20 two models occur, performing sensitivity simulations and looking at the various processes  
21 modelled. This is the main purpose of the next section.

22 Additionally, Fig. 3 shows that in northern and southern temperate and northern boreal  
23 regions, the seasonal cycle is very similar between the two models, even if ORCHIDEE  
24 calculates higher emissions than MEGAN, especially for isoprene.

### 25 **3.3 Emission geographical distribution**

26 The spatial patterns of BVOC emissions in winter and summer for ORC\_CRU and  
27 MEG\_CRU simulations are presented in Figs. 5–9 for isoprene, monoterpenes, methanol,  
28 acetone and sesquiterpenes. To better assess the impact of EFs on emissions, we show the  
29 resulting emission potential for each grid cell, summing the EFs, each weighted by the cell

1 area occupied by each PFT. In MEGAN, emission potentials are already provided per grid  
 2 | cell ~~instead of EF value per PFT~~ for isoprene, monoterpenes and MBO (see Sect. 2.3).  
 3 Emission potentials per grid cell can be interpreted as the average EFs associated with the  
 4 ecosystem present in the grid cell.

5 For a particular compound, the formula to convert the ORCHIDEE EF ( $\mu\text{gC g}^{-1} \text{h}^{-1}$ ) in the  
 6 potential emission ( $\mu\text{g m}^{-2} \text{h}^{-1}$ ) consistent to those provided by MEGAN are, for emission not  
 7 depending on light (LDF = 0):

$$8 \quad EP = \sum_i EF_i \cdot M / M_{\text{Carbon}} \cdot LAI_{\text{REF}} \cdot SWL_i \cdot A_i \quad (6)$$

9 and for light-dependent emissions (LDF = 1):

$$10 \quad EP = \sum_i EF_i \cdot M / M_{\text{Carbon}} \cdot LAI_{\text{REF}} \cdot SWL_i \cdot A_i \cdot C_{\text{CE}} \quad (7)$$

11 where  $i$  is the index related to PFTs,  $M_{\text{Carbon}}$  and  $M$  are the molar mass of carbon and the  
 12 compound, respectively,  $LAI_{\text{REF}}$  equals to  $5.0 \text{ m}^2 \text{ m}^{-2}$  is the LAI in MEGAN standard  
 13 conditions,  $SWL$  is the MEGAN specific leaf weight depending on PFTs,  $A$  is the PFT grid  
 14 fraction and  $C_{\text{CE}}$  is the canopy environment coefficient, a scaling factor dependent on the  
 15 canopy radiation module, which equals 0.57 in this MEGAN configuration (Guenther et al.,  
 16 2012).

17 In general, for every compound, we observe a similar geographical distribution. High  
 18 emission areas are identified in Brazil, equatorial Africa, southern East Asia and southern  
 19 East USA for both models, with values for ORCHIDEE (MEGAN) ranging between:  
 20  $5.0\text{--}12.0 \cdot 10^{10} \text{ kg C m}^{-2} \text{ s}^{-1}$  ( $3.0\text{--}9.0 \cdot 10^{10} \text{ kg C m}^{-2} \text{ s}^{-1}$ ) for isoprene,  $0.8\text{--}2.0 \cdot 10^{10} \text{ kg C m}^{-2} \text{ s}^{-1}$   
 21 ( $0.6\text{--}1.3 \cdot 10^{10} \text{ kg C m}^{-2} \text{ s}^{-1}$ ) for monoterpenes,  $0.3\text{--}1.2 \cdot 10^{10} \text{ kg C m}^{-2} \text{ s}^{-1}$  ( $0.2\text{--}0.7 \cdot 10^{10} \text{ kg C}$   
 22  $\text{m}^{-2} \text{ s}^{-1}$ ) for methanol,  $0.2\text{--}0.5 \cdot 10^{10} \text{ kg C m}^{-2} \text{ s}^{-1}$  ( $0.1\text{--}0.3 \cdot 10^{10} \text{ kg C m}^{-2} \text{ s}^{-1}$ ) for acetone and  
 23  $0.4\text{--}0.6 \cdot 10^{10} \text{ kg C m}^{-2} \text{ s}^{-1}$  ( $0.2\text{--}0.3 \cdot 10^{10} \text{ kg C m}^{-2} \text{ s}^{-1}$ ) for sesquiterpenes, respectively. For  
 24 methanol, in summer, high emitting areas also appear in Europe and Russia, with values of  
 25  $0.3\text{--}0.5 \cdot 10^{10} \text{ kg C m}^{-2} \text{ s}^{-1}$  for ORCHIDEE and  $0.1\text{--}0.3 \cdot 10^{10} \text{ kg C m}^{-2} \text{ s}^{-1}$  for MEGAN.  
 26 Indeed, these regions are populated by temperate and boreal needleleaf evergreen trees, which  
 27 are strong methanol emitters (Table 3 and Fig. 7, last row).

1 In southeast China and south-eastern USA, for methanol, acetone and, to a lesser extent,  
2 monoterpenes, ORCHIDEE emission estimates are higher than MEGAN. This is directly  
3 linked to the larger fraction of temperate needleleaf evergreen trees (TeNeEv) in ORCHIDEE  
4 in comparison to MEGAN (not shown), which are strong emitters of these compounds. The  
5 emission potentials (last row, Figs. 6–8) show the same geographical pattern that is mainly  
6 driven by the PFT distribution in these regions.

7 Other notable differences between the two models appear in South America for isoprene,  
8 directly in relation with the EP distribution. The pattern of isoprene ~~emission in emission in~~  
9 MEGAN has higher values in western Brazil, Bolivia and northern Argentina, while in  
10 ORCHIDEE the values are more homogeneous, with higher emissions in central Brazil. The  
11 same pattern differences are detected in the emission potential (Fig. 5, last row on the right),  
12 and we therefore infer that the EP distribution drives the isoprene emission geographical  
13 distribution. The same conclusion also holds for monoterpenes, where lower emissions along  
14 the Amazonian river follow perfectly the lower EPs in this area. In general, comparing the  
15 emission geographical distribution for each compound and the corresponding emission  
16 potential, we can state that, in both models, emission spatial patterns are mostly affected by  
17 the EF and PFT distributions.

### 18 **3.4 BVOC emission sensitivity to LAI**

19 In this section, we investigate in detail the differences between the two models arising from  
20 LAI and we explore to what extent LAI can affect BVOC emission estimates.

21 Figs 4 and 10 show large differences in the geographical distribution and global average of  
22 ORCHIDEE LAI and MODIS LAI (Yuan et al., 2011). As illustrated in Fig. 10, the global  
23 monthly mean LAI calculated by ORCHIDEE is  $1.5\text{--}2\text{ m}^2\text{ m}^{-2}$  higher compared to the LAI  
24 used in MEGAN and based on MODIS data-sets. In addition the LAI peaks at different times  
25 throughout the year in ORCHIDEE and MEGAN. We investigate the contribution of different  
26 areas and we observe that, whilst in northern temperate region the MODIS LAI peaks in July  
27 and afterwards decreases quite fast, the ORCHIDEE LAI peak in both July and August.  
28 Furthermore, in the boreal region, the ORCHIDEE LAI peaks one month later (August) than  
29 the MODIS LAI (July). Therefore, the time shift observed globally is due to the greater

1 persistence of the growing season provided by ORCHIDEE in the northern temperate area  
2 and its delay in the northern boreal region compared with what is detected by MODIS.

3 Furthermore, in the tropics, the MODIS LAI exhibits quite a clear seasonal cycle, especially  
4 in Amazonia, Central Africa and Indonesia that is not simulated by ORCHIDEE (Fig. 4).

5 The differences between these LAI estimates are significant, but our current state of  
6 knowledge does not allow us to ~~establish~~ which estimate is ~~more reliable~~~~correct~~. Field and  
7 satellite data bring very useful and complementary information regarding the order of  
8 magnitude, the seasonal and the geographical variability of LAI. Nevertheless, inferring  
9 values for LAI on small or large regional scales is particularly challenging, and data available  
10 from, either field or satellite measurements also have significant uncertainties. Satellites, for  
11 instance, do not measure the *real* LAI, but the *effective* LAI obtained from indirect optical  
12 methods and strongly determined by the *a priori* assumptions necessary for the inversion  
13 procedure. Even starting from the same input reflectance, diverse retrieval methods can lead  
14 to LAI values that are highly different (Garrigues et al., 2008; Fang et al., 2013). The  
15 effective LAI can be very dissimilar to the LAI directly measured in situ and relative  
16 differences can reach 100% (Fang et al. 2012a, b).

17 The transition from *effective* to *real* LAI is possible only when additional information about  
18 the vegetation structure is available (Pinty et al. 2011), increasing the risk of inaccuracy. The  
19 sources of uncertainties are numerous (Garrigues et al., 2008). First, foliage clumping is, in  
20 general, not taken into account, leading to underestimates of LAI of up to 70% over the  
21 coniferous forest. Second, the forest understory is not systematically taken into account since  
22 the satellite LAI product is derived from a vertical integrated radiation signal. Third, in dense  
23 canopies, such as broadleaf tropical forests, the optical signal can saturate, leading to an  
24 underestimate of the effective LAI in comparison with the true value with a saturation limit of  
25  $3.0 \text{ m}^2 \text{ m}^{-2}$  (Pinty et al. 2011). Forth, the presence of ice and snow can strongly upset ~~the~~~~LAI~~  
26 retrieval, making it very difficult to estimate LAI in boreal and mountain regions.

27 Conversely, in a validation study using satellite-derived vegetation index time series,  
28 Maignan et al. (2011) pointed out some weaknesses in the ability of ORCHIDEE to correctly  
29 model the LAI seasonal cycle, especially in the equatorial forest (Amazonia, central Africa,  
30 Indonesia) where a poor correlation of model output with satellite data was demonstrated. In

1 general, quite large and comparable uncertainty is found when different LAI databases are  
2 compared. Krinner et al. (2005) found that the difference between ORCHIDEE and MODIS  
3 satellite LAI (Myneni et al., 2002) is as much as the difference between the satellite data that  
4 they used and an alternative satellite vegetation cover data set (Tucker et al., 2001). Therefore  
5 given the many existing limitations, we cannot ~~precisely estimate to conclude~~ which extent  
6 ORCHIDEE LAI estimate is more reliable. ~~(LAI obtained from MODIS satellite or calculated~~  
7 ~~by ORCHIDEE)~~. It is likely that the ORCHIDEE LAI modelling has room of  
8 improvement could be improved and a possible component to be upgraded is the allocation of  
9 the different carbon stocks, but further investigations are needed. Performing a robust  
10 evaluation of the model's ability to simulate the LAI, especially at the global scale, still  
11 remains challenging, and is beyond the scope of our study.

12 In this context, model inter-comparison and sensitivity tests give an essential insight to assess  
13 the impact of different LAI estimates and their uncertainties on BVOC emissions.

#### 14 **3.4.1 LAI seasonal cycle impact**

15 LAI has an important role in driving the seasonal cycle of emissions. To show this, we  
16 perform an extra 10-year simulation following the same configuration as in the previous runs,  
17 but forcing MEGAN with the ORCHIDEE LAI (MEG\_CRULAI simulation, Table 5) and we  
18 compare the results with MEG\_CRU and ORC\_CRU simulations.

19 First of all, we observe that, for the MEG\_CRU simulation, the isoprene emission seasonal  
20 cycle in the tropics (particularly in the South) is more marked than for ORC\_CRU simulation  
21 (Fig. ~~4 and~~ 11). This behaviour is principally related to the differences in seasonal variation  
22 between the MODIS and the ORCHIDEE LAI (Fig. 4), since the ORCHIDEE LAI presents  
23 smaller variations between winter and summer in tropical regions, in particular in Amazonia,  
24 (Fig. 4, left column) in comparison with MODIS LAI (Fig. 4, right column). Whereas, the  
25 two models have a similar inter-seasonal variability when they are driven by the same LAI  
26 (MEG\_CRULAI and ORC\_CRU). Moreover, MEG\_CRULAI simulation gives a lower peak  
27 in the northern tropics April and May emission than MEG\_CRU (Fig. 11), being more similar  
28 to ORC\_CRU.

1 Generally, for every compound, we observe a better agreement between the MEG\_CRULAI  
2 and the ORC\_CRU simulations than between MEG\_CRU and ORC\_CRU, especially in the  
3 tropical regions.

#### 4 **3.4.2 LAI rangesize impact**

5 The global and zonal emission budgets (Table 7) in the MEG\_CRULAI simulation are not  
6 significantly different than those determined in MEG\_CRU, even if the ORCHIDEE LAI is  
7 significantly higher than MODIS LAI, suggesting a low sensitivity of MEGAN to LAI size  
8 Indeed, at the regional scale, in boreal and temperate regions, the MEG\_CRULAI emissions  
9 are slightly higher than those in MEG\_CRU, and in the tropics they are even slightly lower  
10 for some compounds. As proposed by Sindelarova et al. (2014), a possible reason for the  
11 emission decrease calculated in the tropics by MEGAN is to the strengthened effect of leaf  
12 self-shading caused by an increase in LAI in locations characterized by a dense vegetation  
13 (e.g. in central Africa or Amazonia). This effect can be predominant for compounds for which  
14 biogenic emissions are strongly dependent on light, such as isoprene or methanol.

15 Indeed, for the other compounds the MEG\_CRU and MEG\_CRULAI emission budgets are  
16 very similar. We could foresee that these results are linked to the leaf self-shading effect on  
17 leaf temperature. In contrast to ORCHIDEE, where the air temperature is used, in MEGAN  
18 the leaf temperature is calculated for shaded and sunlit leaves. If the leaf self-shading effect  
19 was crucial even for light-independent compounds, we would expect a much higher leaf  
20 temperature for sunlit leaves than for shaded leaves. Calculating the difference in hourly leaf  
21 temperature between sunlit and shaded leaves in the case of dense vegetation (TrBrEv and  
22 TrBrDe), we estimate differences of about 1–1.5 °C, which would unlikely be high enough to  
23 explain such differences in emissions. Lathière et al. (2006), for instance, found that an  
24 increase in the global surface temperature by 1°C leads to an increase of isoprene emissions  
25 of at most 11%. We therefore doubt that the only mechanism behind the static BVOC  
26 emissions for light-independent compounds is leaf self-shading.

27 We, therefore, investigate in more detail whether models show the same response to a  
28 particular change in LAI. We perform two extra simulations for each model, using the  
29 ORCHIDEE LAI multiplied by a factor of 0.5 or 1.5. The scaling factor considered are  
30 consistent with the LAI uncertainties (see the begging of Sect. 3.4). Fig. 12 shows the four

1 simulations: MEG\_LAI05, ORC\_LAI05 (ORCHIDEE LAI multiplied by 0.5) and  
2 MEG\_LAI15 and ORC\_LAI15 (ORCHIDEE LAI multiplied by 1.5), for the year 2006  
3 (details in Table 5). Only the zonal average for the tropics and southern and northern  
4 temperate areas, for isoprene and monoterpenes are displayed, but they are also representative  
5 of other regions.

6 Regarding isoprene, we observe that ORCHIDEE and MEGAN present a similar response to  
7 LAI variation. When the LAI is multiplied by a factor of 0.5 (1.5), change in emissions  
8 compared to the reference runs (MEG\_CRULAI, ORC\_CRU) reaches  $-18\%$  ( $+12\%$ ) for  
9 MEGAN and  $-21\%$  ( $+8\%$ ) for ORCHIDEE in the southern tropics, and reaches  $-34\%$   
10 ( $+21\%$ ) for MEGAN and  $-32\%$  ( $+16\%$ ) for ORCHIDEE in northern temperate areas. In the  
11 tropics especially, the emissions calculated by the two models are little sensitive to the LAI  
12 increase. Indeed isoprene is a light-dependent compound thus, beyond a given LAI threshold,  
13 the contribution of the highest LAI layers is very low, as there is no more or very little direct  
14 light available. We observe that MEGAN is less sensitive than ORCHIDEE to an LAI  
15 increase, which is likely due to the different parameterisation of CTLD factor in the two  
16 models as described Sect. 2.5. In more details, as LAI increases, the growth of sunlit leaves  
17 fraction is dumped by an exponential factor as in Spitter et al. (1986b), implying lower  
18 contribution of sunlit leaves with respect to shaded leaves for high LAI values. In MEGAN,  
19 for equal incoming radiation, the relative contribution of sunlit leaves, with respect to shaded  
20 leaves, is roughly twice than in ORCHIDEE. This explains the different sensitivity of the two  
21 models.

22 Monoterpene emissions show a different response in terms of sensitivity to LAI. In the  
23 southern tropics, the relative difference in monoterpene emission budget between  
24 ORC\_LAI05 (ORC\_LAI15) and ORC\_CRU is  $-43\%$  ( $+40\%$ ), and  $-9\%$  ( $+3\%$ ) between  
25 MEG\_LAI05 (MEG\_LAI15) and MEG\_CRULAI. In northern temperate regions, the relative  
26 difference in the monoterpene emission budget between ORC\_LAI05 (ORC\_LAI15) and  
27 ORC\_CRU is  $-44\%$  ( $+40\%$ ), and  $-14\%$  ( $+6\%$ ) between MEG\_LAI05 (MEG\_LAI15) and  
28 MEG\_CRULAI. These simulations confirm a much smaller emission impact of LAI variation  
29 on emissions in MEGAN, even for compounds not fully dependent on light, such as  
30 monoterpenes (LDF=0.6).

1 Table 8 shows the total emission budget calculated for MEG\_LAI05, ORC\_LAI05,  
2 MEG\_LAI15 and ORC\_LAI15 simulations for every compound. In general in ORCHIDEE,  
3 the lower the light dependence, the higher the sensitivity to LAI, while for MEGAN, the  
4 sensitivity to LAI does not significantly change with LDF. The explanation for this difference  
5 in emission response lies in the different formulation for light independent emissions in the  
6 two models. Such differences are detailed in point 6 of Sect. 2.5. In particular, in  
7 ORCHIDEE, the light independent emission linearly depends on LAI whereas, in MEGAN it  
8 is determined by the  $\gamma_{\text{LAI}}$  factor and it varies almost linearly for low LAI ( $< 2 \text{ m}^2 \text{ m}^{-2}$ ) and  
9 then more and more slowly up to become almost constant for LAI higher than  $5 \text{ m}^2 \text{ m}^{-2}$ . The  
10 light-independent emission descriptions in the two models therefore respond differently to  
11 LAI variation, with differences more striking when LAI is greater than  $2 \text{ m}^2 \text{ m}^{-2}$ , while the  
12 ORCHIDEE emissions keep increasing linearly with LAI, the MEGAN emissions do not  
13 increase as strongly anymore. In this case, the different modelling choices bring significant  
14 discrepancies in emission sensitivity between the two models.

### 15 **3.4.3 MODIS LAI**

16 Considering the high sensitivity of BVOC emissions to the LAI and the high differences  
17 between ORCHIDEE and MODIS LAI, we perform an additional simulation, forcing  
18 ORCHIDEE with the LAI provided by MODIS (ORC\_CRUMOD) for the year 2006. Details  
19 of ORC\_CRUMOD are provided in Table 5. In Fig. 13, we present the differences between  
20 the seasonal averages of ORC\_CRUMOD and ORC\_CRU for monoterpenes and isoprene  
21 emissions. In ORC\_CRUMOD, isoprene emissions significantly decrease in the tropics, up to  
22  $3\text{--}6 \cdot 10^{-10} \text{ kgC m}^{-2} \text{ s}^{-1}$  in Brazil, in the African savanna, India and Northern Australia, while  
23 they increase up to  $0.75\text{--}1 \cdot 10^{-10} \text{ kgC m}^{-2} \text{ s}^{-1}$  in some areas of South America, Australia, and  
24 Africa and up to  $1\text{--}3 \cdot 10^{-10} \text{ kgC m}^{-2} \text{ s}^{-1}$  in equatorial Africa. The monoterpene emissions  
25 decrease almost everywhere, especially in many tropical and equatorial areas and northern  
26 temperate and boreal areas (up to  $0.5 \cdot 10^{-10} \text{ kgC m}^{-2} \text{ s}^{-1}$ ).

27 Fig. 13 also illustrates the seasonal variation for both isoprene and monoterpene emissions in  
28 the tropics, and clearly shows that the use of MODIS LAI implies a seasonality in tropical and  
29 equatorial emissions which is almost not present in ORC\_CRU simulation. Confirming the  
30 results presented in section 3.4.2, monoterpene emissions show higher sensitivity to LAI



1 variations than isoprene, with the monoterpene annual global budget for ORC CRUMOD  
2 being 32% lower than for ORC\_CRU, while for isoprene, the annual global budget is 6%  
3 lower. Considering the other species, the impact of using the MODIS LAI is stronger for  
4 species with a lower LDF. The relative difference between ORC CRUMOD and ORC\_CRU  
5 is -4% for methanol, -30% for acetaldehyde, formaldehyde, acetic acid and formic acid,  
6 -36% for acetone and -44% for MBO.

### 8 **3.5 BVOC emission sensitivity to LDF**

9 As described in Sect. 2.2, the LDF parameter sets the light-dependent fraction of emissions  
10 for each compound. Many experimental studies point out for several plant species that, if  
11 emissions can be totally light-independent for some BVOCs, the emissions of most of them  
12 are actually light-dependent to a degree that depends on the compound (Jacob et al. 2002,  
13 2005; Hansen and Seufert, 2003; Dindorf et al., 2006; Holzke et al., 2006; Harley et al., 2007;  
14 Millet et al., 2008, 2010; Hu, et al., 2011; Wells et al., 2014). Since the results of these studies  
15 are highly heterogeneous, assigning a single LDF value to each compound is as difficult as  
16 assigning the EFs to each PFT (Sect. 2.2). Hence, the LDF uncertainty could be even higher  
17 than the uncertainties associated with EFs, as there have been fewer less quantitative studies  
18 on this subject published to date.

19 The objective of this section is to quantify, for both ORCHIDEE and MEGAN, the relative  
20 contribution of the light-dependent and light-independent part to the total emissions, and  
21 consequently to determine the impact of LDF-attributed values on emission estimates, giving  
22 clues to better understand the different behaviours of the two models.

23 For the fully light dependent (isoprene: LDF=1) or largely light dependent compounds  
24 (methanol: LDF=0.8) (Fig. 5 and Fig. 7), we observe that a higher EP in ORCHIDEE than in  
25 MEGAN does not necessarily lead to higher emissions in ORCHIDEE. In the case of a LDF  
26 close to 1, even when the same EP value is used in both models, the emissions calculated by  
27 MEGAN are higher compared to ORCHIDEE, suggesting a different emissions response to  
28 light. Indeed, this effect is less important for compounds which are less dependent on light  
29 such as monoterpenes (LDF = 0.5) (Fig. 6) and sesquiterpenes (LDF = 0.6) (Fig. 9), and

1 indeed are even negligible for acetone (LDF = 0.2) (Fig. 8). It therefore seems that the choice  
2 of LDF parameter can be crucial in the emission estimate and in the sensitivity to EF  
3 variation.

4 To isolate the signal related to the LDF, we investigate the hourly variation of two “test  
5 compounds”, the first defined as light independent (LDF = 0) and the second defined as  
6 totally light-dependent (LDF = 1). All EFs are set to  $1 \mu\text{C g}^{-1} \text{ h}^{-1}$  for each PFT. The other  
7 settings are specified as in the reference run and are the same for the two test compounds (for  
8 further details see Table 5). We refer in the text to the first compound as *orcldf0* if it is  
9 calculated by ORCHIDEE and as *megldf0* if it is calculated by MEGAN, while we refer to the  
10 second compounds as *orcldf1* and *megldf1*, respectively.

11 In order to quantify the contribution of the light-dependent part in comparison to the light-  
12 independent one, we use the LDF index, which we define as the ratio between the light-  
13 dependent and the light-independent test compound, multiplied by 100 (*orcldf1/orcldf0*·100,  
14 *megldf1/megldf0*·100). Using the LDF index we can easily compare the behaviour of the two  
15 models, avoiding the complication arising from the mismatch between the two land covers.  
16 Indeed, the direct comparison of the absolute values of *orcldf* and *megldf* compounds could  
17 be affected by the differences between the PFT distributions in the two models, and the signal  
18 due to LDF change could therefore not be well isolated.

19 In Fig. ~~1413~~ the daily profile averaged over each month of the LDF index is presented to  
20 investigate the daily and annual variations. At the global scale (left panel), we observe that the  
21 LDF index associated with MEGAN is much higher (up to 20%), than the index associated  
22 with ORCHIDEE. At the regional scale, in the southern tropics for example (second panel)  
23 the index reaches up to 70% and is the twice as large the index calculated for ORCHIDEE.  
24 The light-dependent part of the emissions in MEGAN is therefore more important than  
25 ORCHIDEE, with important impacts on emission estimates. Firstly, we show that based on  
26 the same EF value, the MEGAN emissions are higher than in ORCHIDEE for compounds  
27 associated with high LDF, as expected from Sect. 3.3.

28 Secondly, the variable *orcldf0* (*megldf0*) represents the emissions when LDF is zero while  
29 *orcldf1* (*megldf1*) represents the emissions when LDF is one; thus, they define the interval  
30 spanned by emissions as LDF varies. Therefore, a low LDF index is associated with a greater

1 variability of emissions for equal light-independent emissions. Consequently, ORCHIDEE  
2 results more sensitive to LDF variation than MEGAN, as the ORCHIDEE LDF index is lower  
3 than the MEGAN index. Furthermore, the LDF index provides an evaluation of error due to a  
4 diverse choice of LDF values. The LDF index is always less than 100, meaning that the light-  
5 independent component of the emission is always bigger than the light-dependent part.  
6 Therefore, if LDF in the model is greater than it should be, emissions will be underestimated,  
7 while if it is less, emissions will be overestimated. At regional scale, tropical areas, that are  
8 associated to high LDF index, will be less sensitive to LDF variation than other regions.

9

#### 10 **4 Conclusions**

11 The main objectives of this study were to (i) present the new version of the BVOC emission  
12 module embedded in the ORCHIDEE model, (ii) provide BVOC emission estimates for the  
13 2000–2009 period for a large diversity of compounds, (iii) compare the ORCHIDEE model  
14 results to emissions calculated by MEGAN in terms of global, regional and seasonal patterns,  
15 and (iv) investigate how the uncertainty linked to some key variables or parameters such as  
16 the LAI and the LDF could affect the BVOC emission estimate in the two models.

17 The new ORCHIDEE emission module now considers many speciated monoterpenes and  
18 bulk sesquiterpenes, that have been shown to be important regarding SOA formation, uses  
19 updated EFs and includes ~~development~~development in the physical processes related to  
20 BVOC formation, such as the emission dependence on light for almost all compounds, ~~and~~ a  
21 multi-layer calculation of diffuse and direct radiation, and shaded and sunlit leaves over LAI  
22 layers-.

23 The ORCHIDEE emission estimates are within the range of the published emission budgets.  
24 The ORCHIDEE global budgets averaged over the period investigated (2000–2009) are 465  
25 Tg C yr<sup>-1</sup> for isoprene, 108 Tg C yr<sup>-1</sup> for monoterpenes, 38 Tg C yr<sup>-1</sup> for methanol, 25 Tg C  
26 yr<sup>-1</sup> for acetone and 24 Tg C yr<sup>-1</sup> for sesquiterpenes. The global emission budgets are, in  
27 general, in good agreement between the two models, with the ORCHIDEE emissions being  
28 8% higher for isoprene, 8% lower for methanol, 17% higher for acetone, 18% higher for  
29 monoterpenes and 39% higher for sesquiterpenes compared to the MEGAN results. At the  
30 regional scale, the largest differences in terms of spatial emission distribution between

1 ORCHIDEE and MEGAN occur in the northern temperate region for isoprene. This different  
2 behaviour is directly linked to differences in the EF and PFT distribution in this area.

3 More generally, considering the ~~emissions~~-geographical distribution of emissions for each  
4 compound and the corresponding emission potential, we show that, in both models, EF and  
5 PFT ~~distributions~~distribution are the main drivers of the geographical emission pattern. In  
6 terms of seasonal variation, the differences between the two models in the tropics are mostly  
7 due to the different seasonal cycles of LAI between MODIS and ORCHIDEE, while the large  
8 discrepancy in northern temperate regions is attributed to differences in the EF distribution.

9 The LAI calculated by ORCHIDEE is  $1.5\text{--}2\text{ m}^2\text{ m}^{-2}$  higher than the LAI retrieved by MODIS.

10 We examined ~~how~~what these discrepancies can impact ~~on~~ the BVOC estimates. Sensitivity  
11 tests ~~are then~~ performed forcing both models with the ORCHIDEE LAI multiplied by a factor  
12 of 0.5 and 1.5 showed that, for isoprene, ORCHIDEE and MEGAN emissions present a  
13 similar response to these LAI variations. Conversely, for monoterpenes, ORCHIDEE is much  
14 more sensitive to LAI variations; in comparison to MEGAN. These discrepancies are due to  
15 differences in the light-independent emission formulation between the two models. In

16 ORCHIDEE, the dependence of emissions on LAI is linear, while in MEGAN it is quasi-  
17 linear for LAI up to  $2\text{ m}^2\text{ m}^{-2}$  ~~is quasi-linear~~, then the increase is progressively  
18 ~~reduced~~reducing the increase up to become nearly constant for LAI greater than  $5\text{ m}^2\text{ m}^{-2}$ .  
19 The sensitivity test performed forcing ORCHIDEE with MODIS LAI, confirmed that in  
20 tropical areas the principal differences between ORCHIDEE and MEGAN BVOC estimation  
21 come from the LAI, and that compounds with lower LDF show a higher sensitivity to LAI  
22 variation.

23 We ~~investigated~~investigate the contribution of the light-dependent and light-independent part  
24 of emissions and consequently the impact that a different choice of LDF can have on  
25 emissions. In MEGAN, the light-independent part of emissions is more important than in  
26 ORCHIDEE, reaching a factor of two in the southern tropics. We find that ~~ORCHIDEE~~  
27 estimates are more sensitive to LDF variation than MEGAN. Moreover, we ~~showed~~show that  
28 overestimation (underestimation) of their LDF value leads to emission underestimation  
29 (overestimation).

30

## 1 **5 Future directions**

2 Model inter-comparison and sensitivity tests are extremely useful to define which  
3 parameters/variables mainly affect BVOC emissions, ~~what~~~~which~~ is the cause of this  
4 sensitivity, and how estimates can be improved. Previous works have already investigated the  
5 impact of different experimental set-ups (climate forcing and vegetation distribution) (Arneeth  
6 et al., 2011), differences in the canopy structure description (Keenan et al., 2011) and land  
7 cover classification (Oderbolz et al., 2013) on emissions.

8 In the present work, we focused on the impact of LAI, LDF, EFs and PFT distribution. Our  
9 results underline that the high uncertainties in the involved variables/parameters, and the  
10 different choices in modelling processes, result in a high variability of BVOC emission  
11 estimates. The outcome of this analysis provides some guidelines for future developments of  
12 BVOC emission models at the global scale. In particular, the following issues should be  
13 carefully addressed:

- 14 - LAI uncertainties are still extremely high and have a considerable impact on  
15 emissions. Improvements in LAI ~~modelling~~~~modelisation~~ or estimation at the global  
16 scale are essential;
- 17 - EF allocation is a big concern because of its high variability. A proper way to assign  
18 statistically robust values at a global scale has not yet been found. Significant  
19 improvement can be achieved only by increasing the observation data coverage of  
20 many regions and performing long-term measurements;
- 21 - ~~model~~-LDF parameterisation is still oversimplified and has a significant impact on  
22 emissions. Future developments should, therefore, improve LDF parameterization  
23 accuracy, ~~for~~~~For~~ example, by including PFT dependency. As for EFs, more reliable  
24 results can be achieved only by increasing observation coverage;
- 25 - the rather low number of PFTs in global models is a limiting factor in ~~an~~-accurate  
26 emission estimates;

27 Further analysis will certainly be needed ~~in order~~ to include other important  
28 parameters/variables in the investigation, for example, leaf temperature versus air

1 | temperature usage, leaf age classes, parameters in the Guenther formulation, and the soil  
2 | moisture activity factor.

3 | Finally, it is worth mentioning that, besides model inter-comparison, there is a strong need to  
4 | evaluate model results against emission observations. This has already been done in other  
5 | domains, for example in atmospheric chemistry modelling (Mann et al., 2014; Tsigaridis et  
6 | al., 2014). In the case of BVOC, however, observational data are very challenging to acquire,  
7 | especially on the long-term scale. Therefore, for BVOC emission modelling, a robust  
8 | validation of model results against observations, is still lacking.

9

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19

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1 Table 1. Plant Functional Types in ORCHIDEE and MEGAN and corresponding occupied  
 2 surfaces in  $10^{12}$  m<sup>2</sup>.

PFT acronym		PFT full name		PFT surface	
ORCHIDEE	MEGAN	ORCHIDEE	MEGAN	ORCHIDEE	MEGAN
	BaSo		Bare soil	21.43	40.30
	TrBrEv		Tropical broadleaf evergreen tree	12.84	11.40
	TrBrDe		Tropical broadleaf deciduous tree	7.49	5.82
	TeNeEv		Temperate needleleaf evergreen tree	4.50	3.43
	TeBrEv		Temperate broadleaf evergreen tree	4.04	1.81
	TeBrDe		Temperate broadleaf deciduous tree	5.79	4.45
	BoNeEv		Boreal needleleaf evergreen tree	5.74	9.71
	BoBrDe		Boreal broadleaf deciduous tree	5.14	1.68
	BoNeDe		Boreal needleleaf deciduous tree	1.98	1.47
C3Gr	C3GrCold C3GrCool	C3 Grass	C3 Grass Cold C3 Grass Cool	37.00	4.20 12.55
	C4Gr		C4 Grass	14.89	11.025
C3Ag C4Ag	Crop	C3 Agriculture C4 Agriculture	Crop	10.19 4.88	14.58
-	TeSbEv	-	Temperate shrub evergreen	-	0.074
-	TeSbDe	-	Temperate shrub deciduous	-	5.39
-	BoSbD	-	Boreal shrub deciduous	-	8.02

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1 Table 2. Comparison between the old and new versions of the biogenic emission module in  
 2 ORCHIDEE: list of emitted compounds, principal parameters for emission equations,  
 3 radiation model type and compounds for which the leaf emission activity is activated.

	Output Species	Light (LDF) and temperature dependence (Beta) function		Radiation model type	Species with leaf age activation	
		Species	LDF Beta			
<b>ORCHIDEE new version</b>		isoprene, MBO	1.0 -			
		acetaldehyde, formaldehyde, acetic acid, formic acid	0.8 0.10			
		methanol, acetone, acetaldehyde, formaldehyde, acetic acid, formic acid, total monoterpene, $\alpha$ -pinene, $\beta$ -pinene, limonene, myrcene, sabinene, camphene, 3-carene, t- $\beta$ -ocimene, other monoterpenes, sesquiterpene, MBO, Other VOCs	0.2 0.10		Light multilayer vertical profile to calculate radiation extinction inside the canopy for both sunlit and shaded leaves	isoprene methanol
		acetone	0.8 0.8			
		methanol	0.6 0.6			
		total monoterpene, $\alpha$ -pinene, $\beta$ -pinene, limonene, myrcene, sabinene, camphene 3-carene, t- $\beta$ -ocimene, other monoterpenes	0.5 0.17			
	total sesquiterpene					
<b>ORCHIDEE old version</b>	methanol, acetone, acetaldehyde, formaldehyde, acetic acid, formic acid, total monoterpene, MBO, other VOCs	isoprene, MBO	1.0 0.9	One layer	isoprene methanol	

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1 Table 3. Emission Factors (EFs,  $\mu\text{gC g}^{-1} \text{h}^{-1}$ ) for each PFT for the main compounds emitted, in the previous (first line) and actual (second line,  
 2 in bold) version of the ORCHIDEE emission module. The list of references used to set the new values is provided in the last column.

	TrBrEv	TrBrDe	TeNeEv	TeBrEv	TeBrDe	BoNeEv	BoBrDe	BoNeDe	C3Gr	C4Gr	C3Ag	C4Ag	References
<b>Isoprene</b>	24.0 <b>24.0</b>	24.0 <b>24.0</b>	8.0 <b>8.0</b>	16.0 <b>16.0</b>	45.0 <b>45.0</b>	8.0 <b>8.0</b>	8.0 <b>18.0</b>	8.0 <b>0.5</b>	16.0 <b>12.0</b>	24.0 <b>18.0</b>	5.0 <b>5.0</b>	5.0 <b>5.0</b>	He et al., 2000; Klinger et al., 2002; Levis et al., 2003; Stewart et al., 2003; Padhy and Varshney, 2005; Bai et al., 2006; Geron et al., 2006; Guenther et al., 2006, 2012; Smiatek and Steinbrecher, 2006; Karl et al., 2007, 2009; Steinbrecher et al. 2009, 2013; Tsui et al., 2009; Lathière et al., 2006, 2010; Bracho-Nunez, Arneeth et al., 2011; Fu and Liao, 2012; Oderbolz et al., 2013.
<b>Monoterp.</b>	0.800 <b>2.000</b>	0.800 <b>2.000</b>	2.400 <b>1.800</b>	1.200 <b>1.400</b>	0.800 <b>1.600</b>	2.400 <b>1.800</b>	2.400 <b>1.400</b>	2.400 <b>1.800</b>	0.800 <b>0.800</b>	1.200 <b>0.800</b>	0.200 <b>0.220</b>	0.200 <b>0.220</b>	Janson et al., 1999; He et al., 2000; Janson and De Serves, 2001; Stewart et al., 2003; Hayward et al., 2004; Karl et al., 2004, 2007, 2009; Spirig et al., 2005; Tarvainen et al., 2005; Bai et al., 2006; Geron et al., 2006; Guenther et al., 2006, 2012; Hakola et al., 2006; Lathière et al., 2006; Smiatek and Steinbrecher, 2006; Helmig et al., 2007; Ortega et al., 2008; Steinbrecher et al. 2009; Kim et al., 2010; Bracho-Nunez et al., 2011; Fares et al., 2011; Guenther et al., 2012.
<b>Sesquiterp.</b>	- <b>0.450</b>	- <b>0.450</b>	- <b>0.130</b>	- <b>0.300</b>	- <b>0.360</b>	- <b>0.150</b>	- <b>0.300</b>	- <b>0.250</b>	- <b>0.600</b>	- <b>0.600</b>	- <b>0.080</b>	- <b>0.080</b>	LathièreGuenther et al., 2006; Helmig et al., 2007, 2012; Duhl et al., 2008; Matsunaga et al., 2009; Steinbrecher et al. 2009; Karl et al., 2009; Ortega et al., 2008; Bracho-Nunez et al., 2011; Hakola et al., 2006; Kim et al., 2010; Fares et al., 2011; Guenther et al., 2012.
<b>Methanol</b>	0.600 <b>0.800</b>	0.600 <b>0.800</b>	1.800 <b>1.800</b>	0.900 <b>0.900</b>	0.600 <b>1.900</b>	1.800 <b>1.800</b>	1.800 <b>1.800</b>	1.800 <b>1.800</b>	0.600 <b>0.700</b>	0.900 <b>0.900</b>	2.000 <b>2.000</b>	2.000 <b>2.000</b>	Schade and Goldstein, 2001; Geron et al., 2002; Karl et al., 2004, 2005, 2009; Hayward et al., 2004; LathièreGuenther et al., 2006, 2012; Smiatek and Steinbrecher, 2006; Harley et al., 2007; Chang et al., 2009; Steinbrecher et al. 2009; Bracho-Nunez et al., 2011; Fares et al., 2011; Guenther et al., 2012.
<b>Acetone</b>	0.290 <b>0.250</b>	0.290 <b>0.250</b>	0.870 <b>0.300</b>	0.430 <b>0.200</b>	0.290 <b>0.300</b>	0.870 <b>0.300</b>	0.870 <b>0.250</b>	0.870 <b>0.250</b>	0.290 <b>0.200</b>	0.430 <b>0.200</b>	0.070 <b>0.080</b>	0.070 <b>0.080</b>	Janson et al., 1999; Janson and De Serves 2001; Schade and Goldstein, 2001; Karl et al., 2004, 2005, 2009; Villanueva-Fierro et al., 2004; LathièreGuenther et al., 2006, 2012; Smiatek and Steinbrecher, 2006; Chang et al., 2009; Steinbrecher et al. 2009; Bracho-Nunez et al., 2011; Fares et al., 2011; Guenther et al., 2012.
<b>Acetaldehy.</b>	0.100 <b>0.200</b>	0.100 <b>0.200</b>	0.300 <b>0.200</b>	0.150 <b>0.200</b>	0.100 <b>0.250</b>	0.300 <b>0.250</b>	0.300 <b>0.160</b>	0.300 <b>0.160</b>	0.100 <b>0.120</b>	0.150 <b>0.120</b>	0.025 <b>0.035</b>	0.025 <b>0.022</b>	Kesselmeier et al., 1997; Janson et al., 1999; Martin et al., 1999; Schade and Goldstein, 2001; Hayward et al., 2004; Karl et al., 2004, 2005, 2009; Villanueva-Fierro et al., 2004; LathièreGuenther et al., 2006, 2012; Smiatek and Steinbrecher, 2006; Chang et al., 2009; Steinbrecher et al. 2009; Fares et al., 2011; Guenther et al., 2012.
<b>Formaldehy.</b>	0.070 <b>0.040</b>	0.070 <b>0.040</b>	0.200 <b>0.080</b>	0.100 <b>0.040</b>	0.070 <b>0.040</b>	0.200 <b>0.040</b>	0.200 <b>0.040</b>	0.200 <b>0.040</b>	0.070 <b>0.025</b>	0.100 <b>0.025</b>	0.017 <b>0.013</b>	0.017 <b>0.013</b>	Kesselmeier et al., 1997; Janson et al., 1999; Martin et al., 1999; Villanueva-Fierro et al., 2004; LathièreGuenther et al., 2006, 2012; Smiatek and Steinbrecher, 2006; Chang et al., 2009; Karl et al., 2009; Steinbrecher et al. 2009; Guenther et al., 2012.
<b>Acetic acid</b>	0.002 <b>0.025</b>	0.002 <b>0.025</b>	0.006 <b>0.025</b>	0.003 <b>0.022</b>	0.002 <b>0.080</b>	0.006 <b>0.025</b>	0.006 <b>0.022</b>	0.006 <b>0.013</b>	0.002 <b>0.012</b>	0.003 <b>0.012</b>	0.001 <b>0.008</b>	0.001 <b>0.008</b>	Kesselmeier et al., 1997, 1998; Martin et al., 1999; Staudt and Kesselmeier, et al., 2000; Villanueva-Fierro et al., 2004; LathièreGuenther et al., 2006, 2012; Smiatek and Steinbrecher, 2006; Chang et al., 2009; Karl et al., 2009; Steinbrecher et al. 2009; Guenther et al., 2012.
<b>Formic Acid</b>	0.010 <b>0.015</b>	0.010 <b>0.015</b>	0.030 <b>0.020</b>	0.015 <b>0.020</b>	0.010 <b>0.025</b>	0.030 <b>0.015</b>	0.030 <b>0.015</b>	0.030 <b>0.015</b>	0.010 <b>0.010</b>	0.0150 <b>0.010</b>	0.0025 <b>0.008</b>	0.0025 <b>0.008</b>	Kesselmeier et al. 1997, 1998; Martin et al., 1999; Staudt and Kesselmeier, et al., 2000; Villanueva-Fierro et al., 2004; LathièreGuenther et al., 2006, 2012; Smiatek and

														Steinbrecher, 2006; Chang et al., 2009; Karl et al., 2009; Steinbrecher et al. 2009; <a href="#">Guenther et al., 2012.</a>
<b>MBO</b>	0.000	0.000	20.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000	<a href="#">Baker et al., 1999;</a> <a href="#">Schade and Goldstein, 2001;</a> <a href="#">Tarvainen et al., 2005;</a> <a href="#">Guenther et al., 2012;</a> <a href="#">Hakola et al., 2006;</a> <a href="#">Lathière et al., 2006;</a> <a href="#">Chang et al., 2009;</a> <a href="#">Kim et al., 2010;</a> <a href="#">Guenther et al., 2012.</a>
	<b>0.00002</b>	<b>0.00002</b>	<b>1.4</b>	<b>0.00002</b>	<b>0.00002</b>	<b>0.14</b>	<b>0.00002</b>	<b>0.00002</b>	<b>0.00002</b>	<b>0.00002</b>	<b>0.00002</b>	<b>0.00002</b>	<b>0.00002</b>	

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1 Table 4. Percentage of speciated monoterpene EFs with respect to the PFT bulk monoterpene EF (forth line, in bold the Table 3) in the new  
 2 version of the ORCHIDEE emission module.

	TrBrEv	TrBrDe	TeNeEv	TeBrEv	TeBrDe	BoNeEv	BoBrDe	BoNeDe	C3Gr	C4Gr	C3Ag	C4Ag	
<b><math>\alpha</math>-Pinene</b>	<b>39.5</b>	<b>39.5</b>	<b>35.4</b>	<b>46.3</b>	<b>32.6</b>	<b>35.4</b>	<b>31.6</b>	<b>66.2</b>	<b>23.1</b>	<b>20.0</b>	<b>27.7</b>	<b>27.7</b>	Janson et al., 1999; He et al., 2000; Janson and De Serves 2001; Schade and Goldstein, 2001; Greenberg et al. 2004; Villanueva-Fierro et al., 2004; Tarvainen et al 2005; Geron et al., 2006; Ortega et al., 2008; Smiatek and Steinbrecher, 2006; Dominguez-Taylor et al., 2007; Karl et al., 2007, 2009; Steinbrecher et al. 2009; Guenther et al., 2012
<b><math>\beta</math>-Pinene</b>	<b>11</b>	<b>11</b>	<b>14.6</b>	<b>12.2</b>	<b>8.7</b>	<b>14.6</b>	<b>6.3</b>	<b>15.0</b>	<b>12.3</b>	<b>8.0</b>	<b>15.4</b>	<b>15.4</b>	Janson et al., 1999; He et al., 2000; Janson and De Serves 2001; Villanueva-Fierro et al., 2004; Tarvainen et al 2005; Geron et al., 2006; Smiatek and Steinbrecher, 2006; Dominguez-Taylor et al., 2007; Karl et al., 2007, 2009; Ortega et al., 2008; Steinbrecher et al. 2009; Guenther et al., 2012
<b>Limonene</b>	<b>9.2</b>	<b>9.2</b>	<b>8.3</b>	<b>12.2</b>	<b>6.1</b>	<b>8.3</b>	<b>7.1</b>	<b>3.7</b>	<b>14.6</b>	<b>28.0</b>	<b>9.2</b>	<b>9.2</b>	Janson et al., 1999; He et al., 2000; Janson and De Serves 2001; Villanueva-Fierro et al., 2004; Bai et al., 2006; Geron et al., 2006; Smiatek and Steinbrecher, 2006; Dominguez-Taylor et al., 2007; Karl et al., 2007, 2009; Ortega et al., 2008; Steinbrecher et al. 2009; Guenther et al., 2012
<b>Myrcene</b>	<b>7.3</b>	<b>7.3</b>	<b>5.0</b>	<b>5.4</b>	<b>2.8</b>	<b>5.0</b>	<b>1.9</b>	<b>2.5</b>	<b>6.2</b>	<b>5.7</b>	<b>4.6</b>	<b>4.6</b>	Janson et al., 1999; He et al., 2000; Janson and De Serves 2001; Villanueva-Fierro et al., 2004; Geron et al., 2006; Smiatek and Steinbrecher, 2006; Karl et al., 2007, 2009; Ortega et al., 2008; Steinbrecher et al. 2009; Guenther et al., 2012
<b>Sabinene</b>	<b>7.3</b>	<b>7.3</b>	<b>5.0</b>	<b>8.3</b>	<b>0.4</b>	<b>5.0</b>	<b>26.3</b>	<b>3.0</b>	<b>6.5</b>	<b>5.0</b>	<b>6.2</b>	<b>6.2</b>	He et al., 2000; Tarvainen et al 2005; Smiatek and Steinbrecher, 2006; Karl et al., 2007, 2009; Ortega et al., 2008; Steinbrecher et al. 2009; Guenther et al., 2012
<b>Camphene</b>	<b>5.5</b>	<b>5.5</b>	<b>4.2</b>	<b>4.9</b>	<b>0.4</b>	<b>4.2</b>	<b>0.5</b>	<b>2.3</b>	<b>5.4</b>	<b>5.3</b>	<b>3.1</b>	<b>3.1</b>	Janson et al., 1999; He et al., 2000; Janson and De Serves 2001; Tarvainen et al 2005; Bai et al., 2006; Geron et al., 2006; Smiatek and Steinbrecher, 2006; Karl et al., 2007, 2009; Ortega et al., 2008; Steinbrecher et al. 2009; Guenther et al., 2012;
<b>3-Carene</b>	<b>4.8</b>	<b>4.8</b>	<b>17.5</b>	<b>1.0</b>	<b>2.4</b>	<b>17.5</b>	<b>1.3</b>	<b>4.2</b>	<b>6.5</b>	<b>5.7</b>	<b>20.0</b>	<b>20.0</b>	Janson et al., 1999; He et al., 2000; Janson and De Serves 2001; Villanueva-Fierro et al., 2004; Tarvainen et al 2005; Bai et al., 2006; Hakola et al., 2006; Smiatek and Steinbrecher, 2006; Dominguez-Taylor et al., 2007; Karl et al., 2007, 2009; Steinbrecher et al. 2009; Ortega et al., 2008; Guenther et al., 2012;
<b>t-<math>\beta</math>-Ocimene</b>	<b>9.2</b>	<b>9.2</b>	<b>5.4</b>	<b>4.4</b>	<b>11.3</b>	<b>5.4</b>	<b>10.5</b>	<b>2.8</b>	<b>13.8</b>	<b>12.0</b>	<b>3.1</b>	<b>3.1</b>	Steinbrecher et al. 2009; Karl et al., 2009; Ortega et al., 2008; Guenther et al., 2012
<b>Other Monoterpene</b>	<b>6.2</b>	<b>6.2</b>	<b>4.6</b>	<b>5.3</b>	<b>5.3</b>	<b>4.6</b>	<b>14.5</b>	<b>0.3</b>	<b>11.6</b>	<b>10.3</b>	<b>10.7</b>	<b>10.7</b>	Janson et al., 1999; He et al., 2000; Janson and De Serves, 2001; Stewart et al., 2003; Hayward et al., 2004; Karl et al., 2004, 2007, 2009; Spirig et al., 2005; Tarvainen et al., 2005; Bai et al., 2006; Geron et al., 2006; <del>Guenther et al., 2006, 2012</del> ; Hakola et al., 2006; Smiatek and Steinbrecher, 2006; Helmig et al., 2007; Ortega et al., 2008; Steinbrecher et al. 2009; Kim et al., 2010; Bracho-Nunez et al., 2011; Fares et al., 2011; <del>Guenther et al., 2012</del> ;

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1 Table 5. Configuration of simulations performed by ORCHIDEE and by MEGAN.

Simulation Name	Model	Climate Forcing	EFs	LDF	LAI	T	Period	Output frequency
ORC_CRU	ORCHIDEE	CRU	Standard version	Standard version	ORCHIDEE LAI	T air	2000-2009	1 month
MEG_CRU	MEGAN	CRU	Standard version	Standard version	MODIS LAI	T leaf	2000-2009	1 month
MEG_CRULAI	MEGAN	CRU	Standard version	Standard version	ORCHIDEE LAI	T leaf	2000-2009	1 month
ORC_LAI05	ORCHIDEE	CRU	Standard version	Standard version	ORCHIDEE LAI multiplied by 0.5	T air	2006	1 month
ORC_LAI15	ORCHIDEE	CRU	Standard version	Standard version	ORCHIDEE LAI multiplied by 1.5	T air	2006	1 month
MEG_LAI05	MEGAN	CRU	Standard version	Standard version	ORCHIDEE LAI multiplied by 0.5	T leaf	2006	1 month
MEG_LAI15	MEGAN	CRU	Standard version	Standard version	ORCHIDEE LAI multiplied by 1.5	T leaf	2006	1 month
<u>ORC_CRUMOD</u>	<u>ORCHIDEE</u>	<u>CRU</u>	<u>Standard version</u>	<u>Standard version</u>	<u>MODIS LAI</u>	<u>T air</u>	<u>2006</u>	<u>1 month</u>
ORC_LDF	ORCHIDEE	CRU	EFs = 1	LDF = 1 and 0	ORCHIDEE LAI	T air	2006	1 hour
MEG_LDF	MEGAN	CRU	EFs = 1	LDF = 1 and 0	ORCHIDEE LAI	T leaf	2006	1 hour

1 Table 6. Emission budget (Tg C yr<sup>-1</sup>) averaged over the 2000–2009 period for the ORC\_CRU (gray lines) and MEG\_CRU simulations at the  
 2 global scale, for northern and southern tropics, northern and southern temperate areas and northern boreal regions.

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Model	Area	Isoprene	Methanol	Acetone	Acetald	Formald	Acetic Acid	Formic Acid	MBO	Sesqiter.	Monoter	$\alpha$ -Pinene	$\beta$ -Pinene	Limonen	Myrcene	Sabinene	3-Carene	T- $\beta$ -Ocimene
ORCHIDEE	Global	464.6	37.8	24.6	8.6	1.9	1.1	0.7	1.3	24.3	91.3	40.9	12.2	10.7	7.2	8.19	6.5	9.3
MEGAN	Global	427.6	40.9	20.5	8.7	1.6	1.2	0.8	1.0	14.9	74.4	24.6	13.1	6.9	2.1	5.5	4.8	17.4
ORCHIDEE	Tro North	176.3	12.9	8.6	2.9	0.6	0.4	0.2	0.1	9.6	32.8	14.8	4.3	4.0	2.7	2.9	2.0	3.5
MEGAN	Tro North	1685	15.1	7.4	3.2	0.6	0.5	0.3	0.2	5.7	28.4	9.5	5.1	2.6	0.7	2.2	1.9	6.5
ORCHIDEE	Tro South	217.7	13.0	10.9	3.8	0.8	0.5	0.3	0.0	12.3	42.7	19.6	5.5	5.1	3.6	3.7	2.4	4.6
MEGAN	Tro South	209.6	15.1	9.1	4.2	0.7	0.5	0.3	0.1	7.5	32.5	10.5	5.5	2.9	0.8	2.6	1.6	8.5
ORCHIDEE	Tem North	51.6	9.1	3.6	1.3	0.3	0.2	0.1	1.2	1.6	10.9	4.3	1.6	1.1	0.6	1.2	1.6	0.8
MEGAN	Tem North	30.7	7.9	2.6	0.8	0.2	0.1	0.1	0.6	1.0	9.2	3.1	1.8	1.0	0.5	0.4	1.0	1.4
ORCHIDEE	Tem South	5.4	0.6	0.3	0.1	0.03	0.01	0.01	0.0	0.2	1.0	0.5	0.1	0.1	0.06	0.1	0.04	0.1
MEGAN	Tem South	9.1	1.0	0.4	0.1	0.03	0.02	0.01	0.01	0.2	1.2	0.5	0.2	0.14	0.03	0.06	0.06	0.25
ORCHIDEE	Boreal	4.4	1.5	0.6	0.2	0.05	0.03	0.02	0.03	0.2	2.0	0.9	0.3	0.2	0.1	0.2	0.3	0.15
MEGAN	Boreal	2.2	1.1	0.4	0.1	0.02	0.01	0.01	0.02	0.1	1.6	0.5	0.3	0.19	0.07	0.07	0.18	0.23

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1 Table 7. Mean emission budgets (Tg C yr<sup>-1</sup>) for the 2000–2009 period estimated in MEG\_CRULAI simulation.

MEGAN	Isoprene	Methanol	Acetone	Acetald	Formald	Acetic Acid	Formic Acid	MBO	Sesquiterp	Monoterp	$\alpha$ -Pinene	$\beta$ -Pinene	Limonen	Myrcene	Sabinene	3-Carene	T- $\beta$ -Ocimene
Global	422.7	41.1	20.2	8.5	1.5	1.2	0.8	1.1	14.5	74.0	24.5	13.0	6.9	2.1	5.7	4.8	17.0
Tro North	162.5	14.8	7.2	3.2	0.6	0.4	0.3	0.2	5.5	28.0	9.3	5.0	2.5	0.7	2.1	1.8	6.4
Tro South	210.1	15.0	8.9	4.1	0.7	0.5	0.3	0.1	7.3	32.0	10.5	5.4	2.8	0.7	2.5	1.5	8.2
Tem North	30.9	8.2	2.8	0.8	0.2	0.1	0.1	0.7	1.1	9.6	3.2	1.8	1.0	0.5	0.4	1.1	1.5
Tem South	9.2	1.1	0.4	0.1	0.03	0.02	0.01	0.01	0.2	1.3	0.5	0.2	0.15	0.03	0.07	0.06	0.26
Boreal	2.4	1.3	0.5	0.1	0.02	0.01	0.01	0.02	0.15	1.8	0.6	0.3	0.21	0.08	0.08	0.20	0.27

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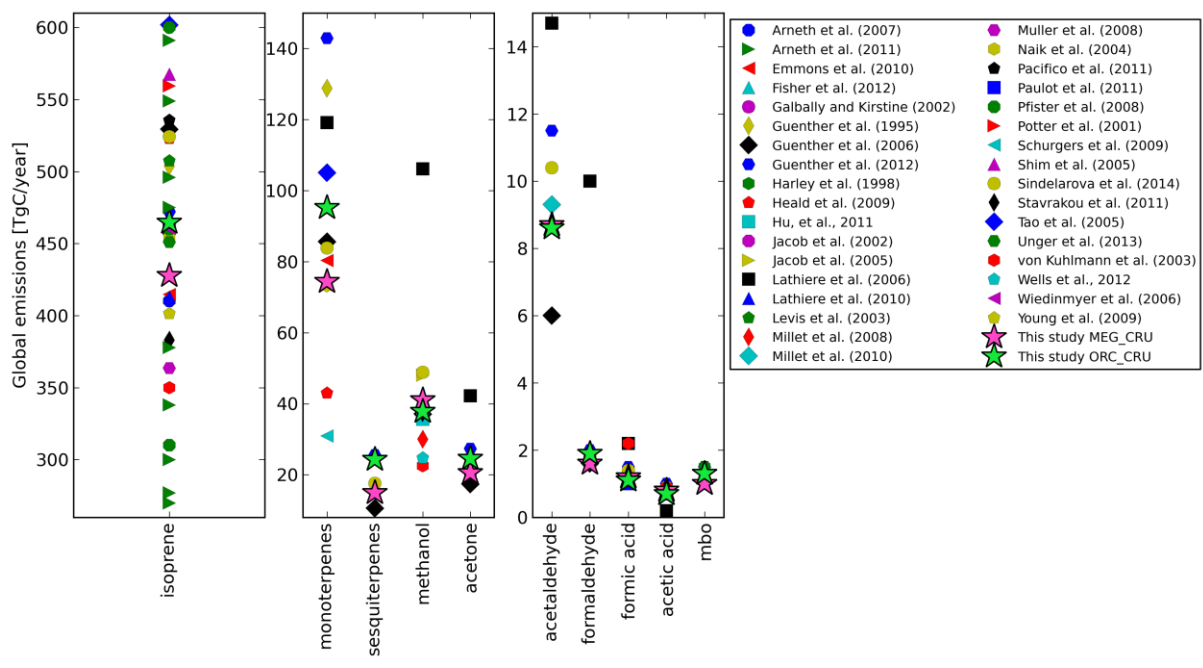
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1 Table 8. Annual emission budgets (Tg C yr<sup>-1</sup>) for the year 2006 in ORC\_CRU,  
 2 MEG\_CRULAI (taken as reference) and in the LAI sensitivity tests (ORC\_LAI05,  
 3 ORC\_LAI15, MEG\_LAI05 and MEG\_LAI15).

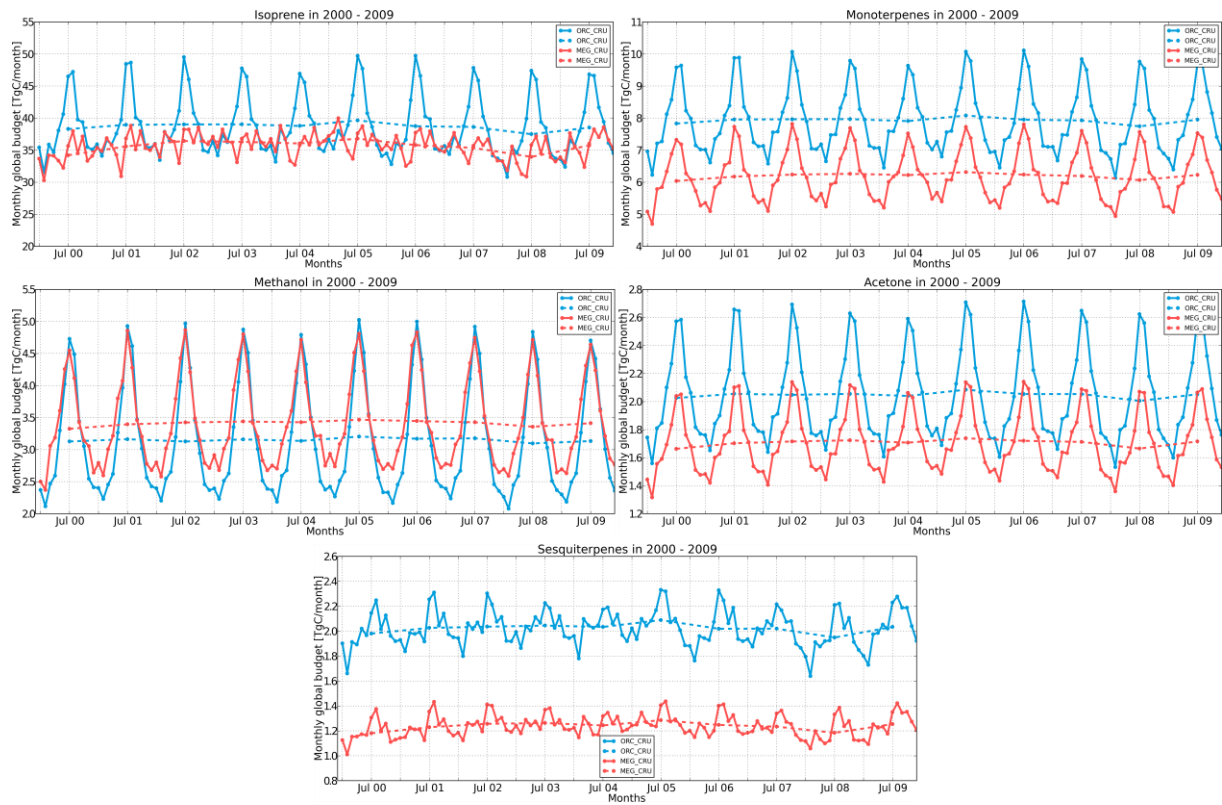
Simulation	Isoprene	Methanol	Acetone	Acetald	Formald	Acetic Acid	Formic Acid	MBO	Sesquiter	Monoter	$\alpha$ -Pinene	$\beta$ -Pinene	Other Monoter
ORC_CRU	464.9	38.0	24.6	8.6	1.9	1.1	0.7	1.4	24.2	95.4	41.0	12.2	42.3
ORC_LAI05	365.3	23.3	12.7	5.3	1.1	0.7	0.4	0.7	13.5	54.1	23.2	6.9	23.9
ORC_LAI15	501.1	50.4	36.5	11.4	2.5	1.5	1.0	2.1	34.2	133.8	57.4	17.1	59.2
MEG_CRULAI	422.5	41.4	20.3	8.6	1.6	1.2	0.8	1.1	14.5	74.2	24.6	13.1	36.5
MEG_LAI05	360.9	34.4	18.3	7.6	1.4	1.0	0.7	1.0	13.5	66.4	21.5	11.7	33.2
MEG_LAI15	450.2	45.1	20.7	8.9	1.6	1.2	0.8	1.2	14.6	76.8	25.8	13.5	37.5

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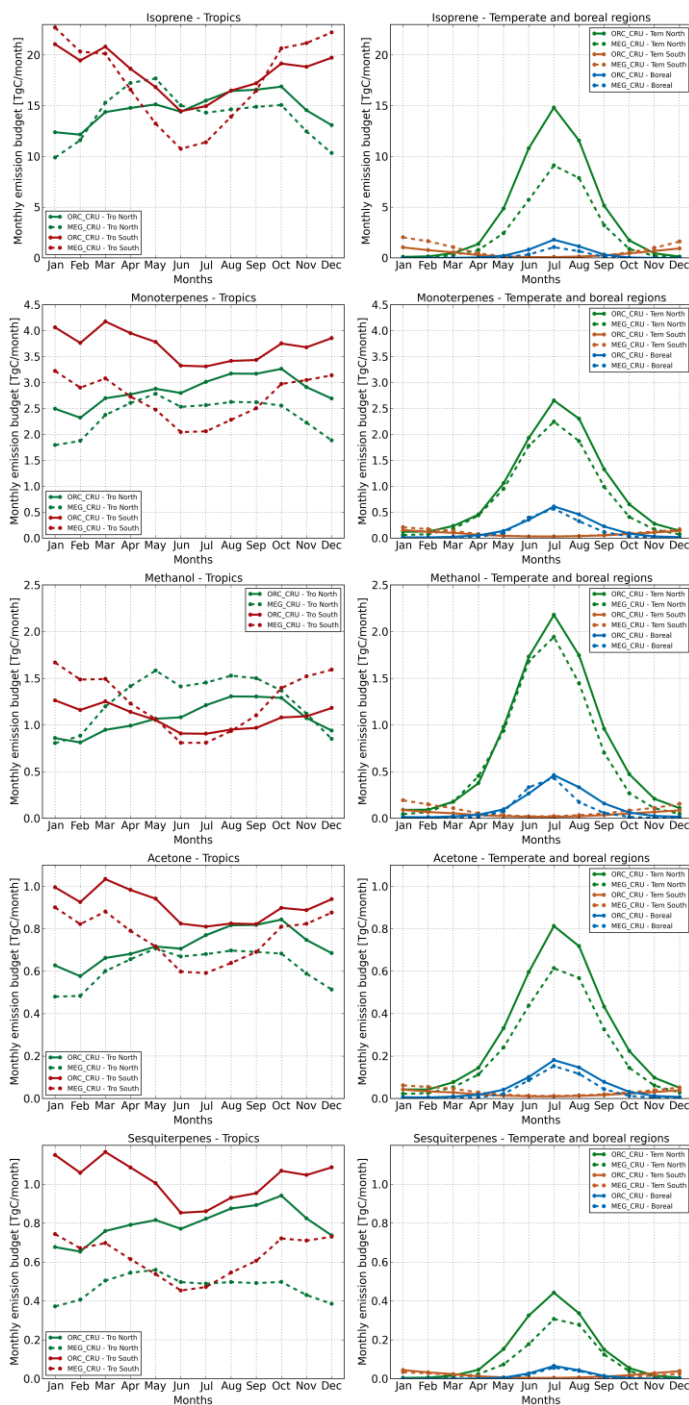


1  
 2 Figure 1. Global emission budgets ( $\text{Tg C yr}^{-1}$ ) calculated by ORCHIDEE (ORC\_CRU  
 3 simulation, green stars) and MEGAN (MEG\_CRU simulation, pink stars), compared with  
 4 published estimates for the main BVOCs presented in this work. Note that the vertical axes  
 5 have different scales in the three plots.  
 6



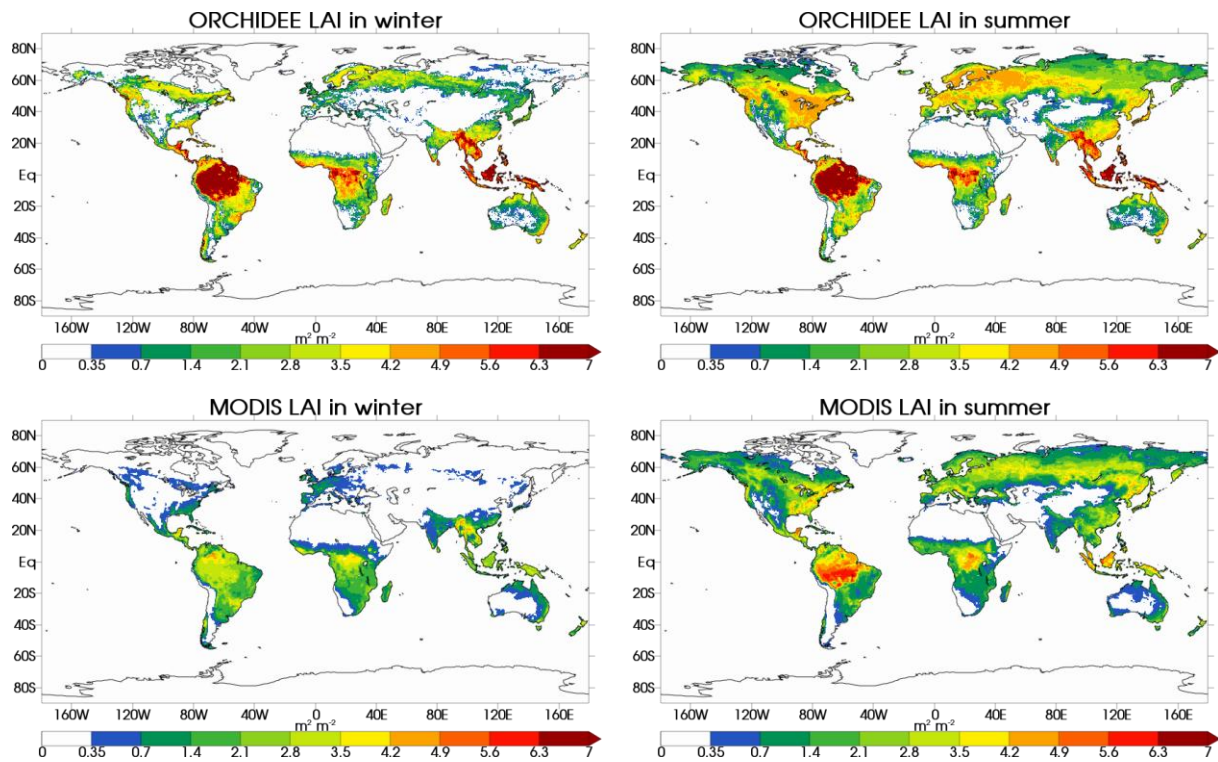
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Figure 2. Monthly global (solid lines) and yearly averaged (dashed lines) emission budgets in  $\text{Tg C month}^{-1}$  for ORC\_CRU and MEG\_CRU simulations for isoprene, monoterpenes, methanol, acetone and sesquiterpenes.

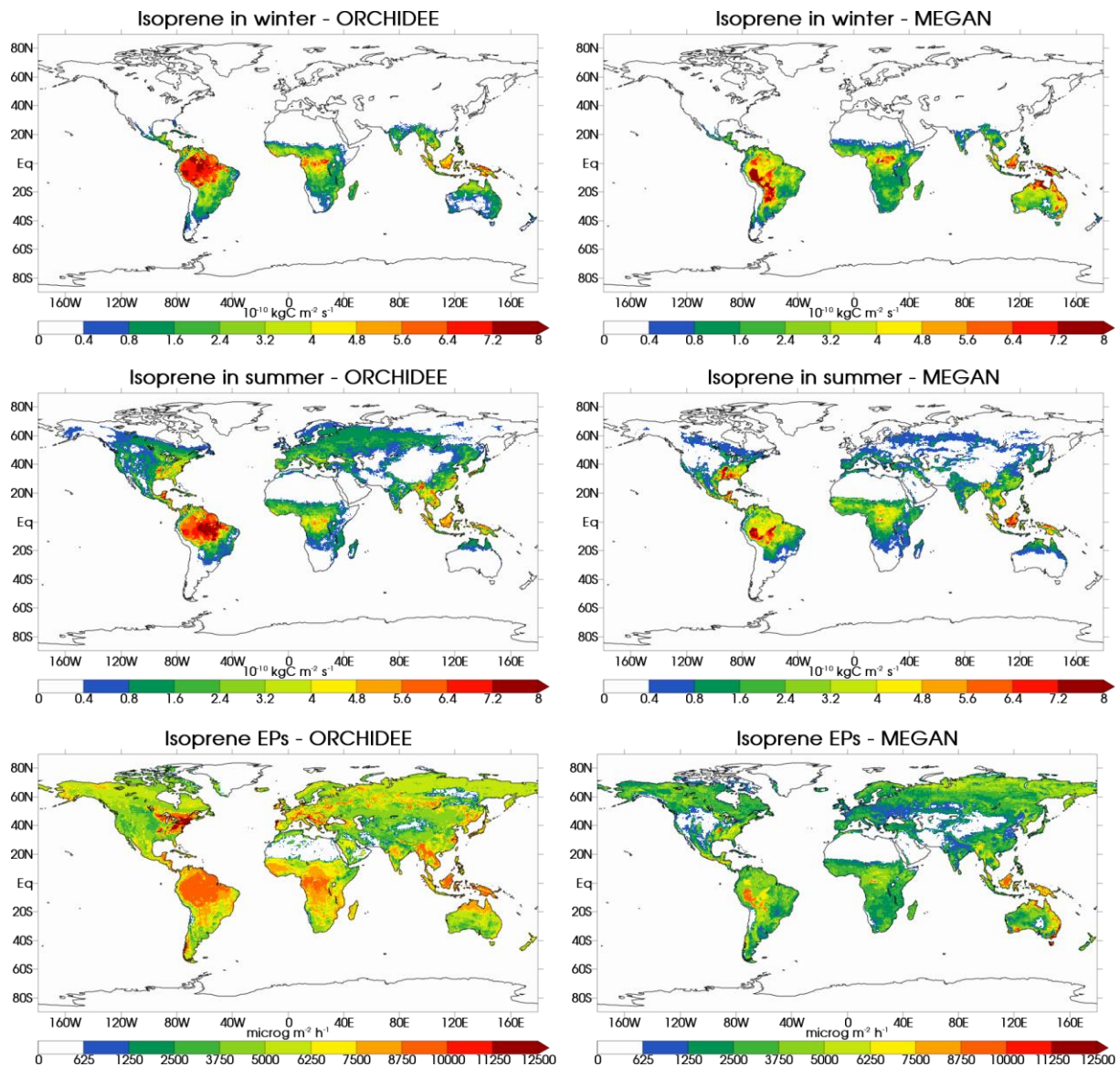


1  
 2 Figure 3. Zonal mean for northern and southern tropics (left column), northern and southern  
 3 temperate and northern boreal latitudes (right column) of the monthly emission budget (Tg C  
 4 month<sup>-1</sup>) averaged over the simulation period (2000–2009) in ORC\_CRU and MEG\_CRU  
 5 runs for isoprene, monoterpenes, methanol, acetone and sesquiterpenes, respectively.

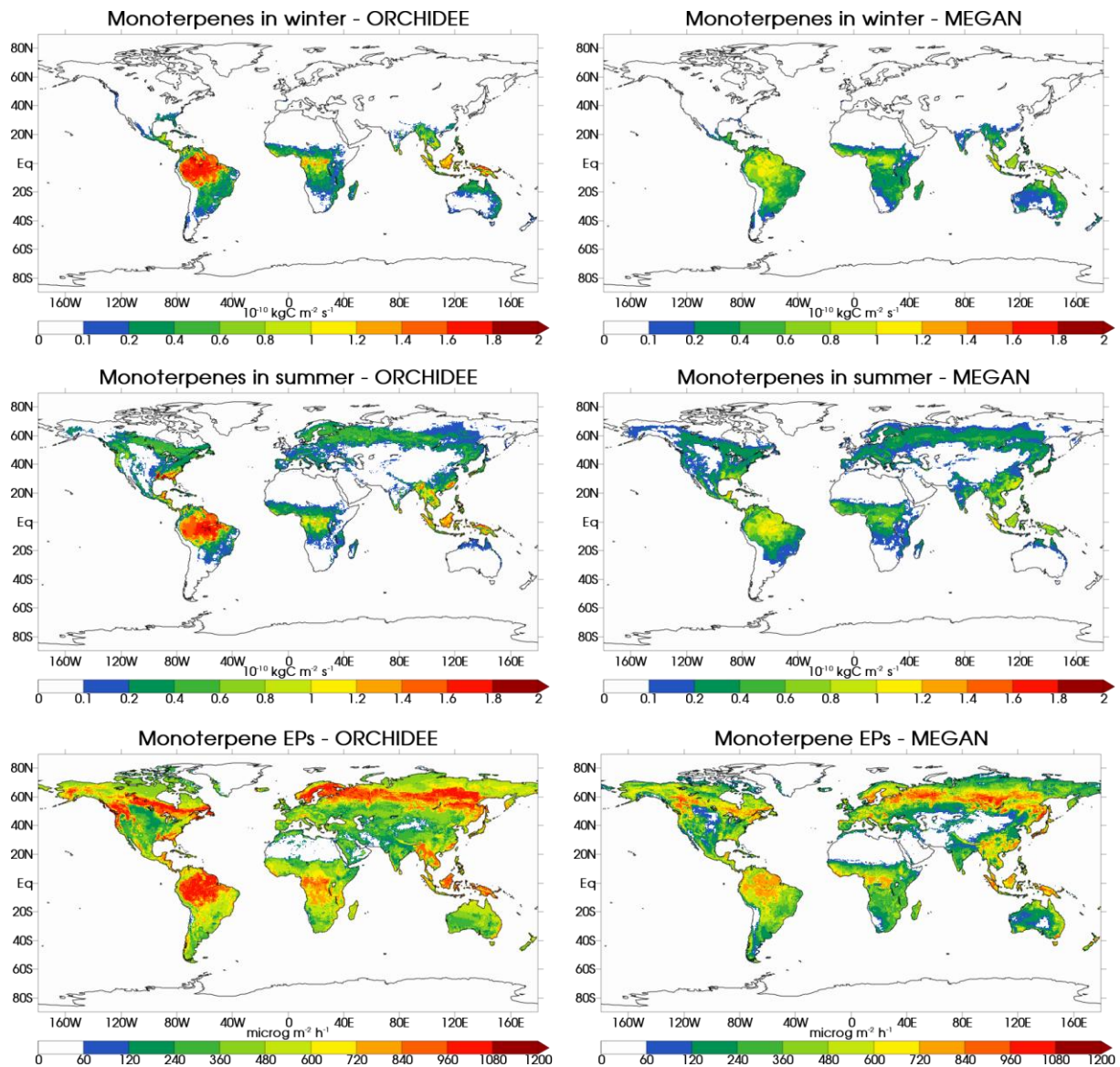
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1  
 2 Figure 4. Leaf area index (LAI) considered for BVOC emission estimates in ORCHIDEE  
 3 (LAI calculated on line) and in MEGAN (MODIS retrieval) in summer (June, July, August)  
 4 and winter (December, January, February), averaged over the 2000–2009 period ( $\text{m}^2 \text{m}^{-2}$ ).  
 5

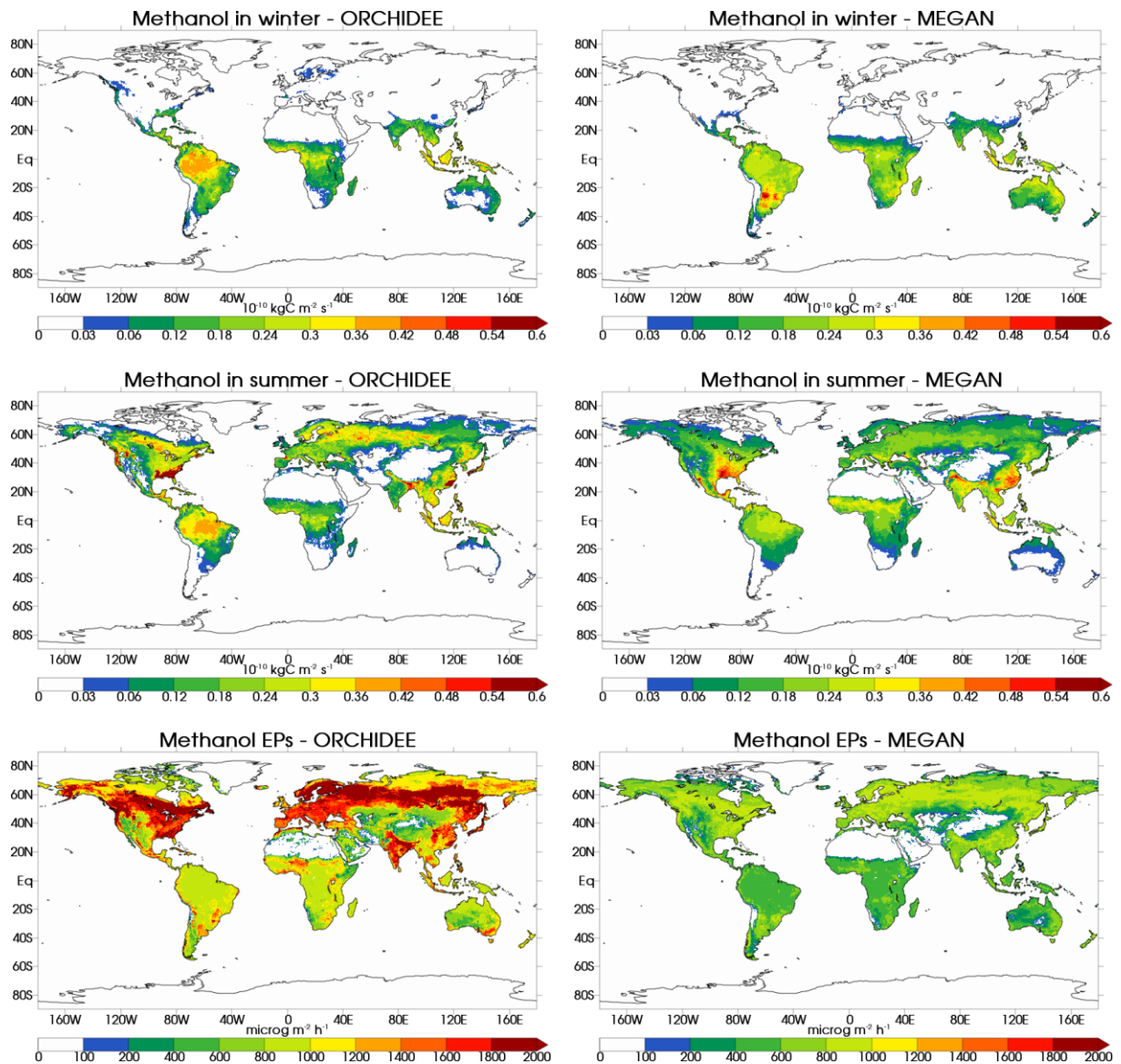


1  
 2 Figure 5. Emissions in winter (first row) and summer (second row) in  $10^{-10} \text{ kg C m}^{-2} \text{ s}^{-1}$  and  
 3 emission potentials (EPs) (third row) in  $\mu\text{g m}^{-2} \text{ h}^{-1}$  for ORCHIDEE (ORC\_CRU, left column)  
 4 and MEGAN (MEG\_CRU, right column) for isoprene.  
 5

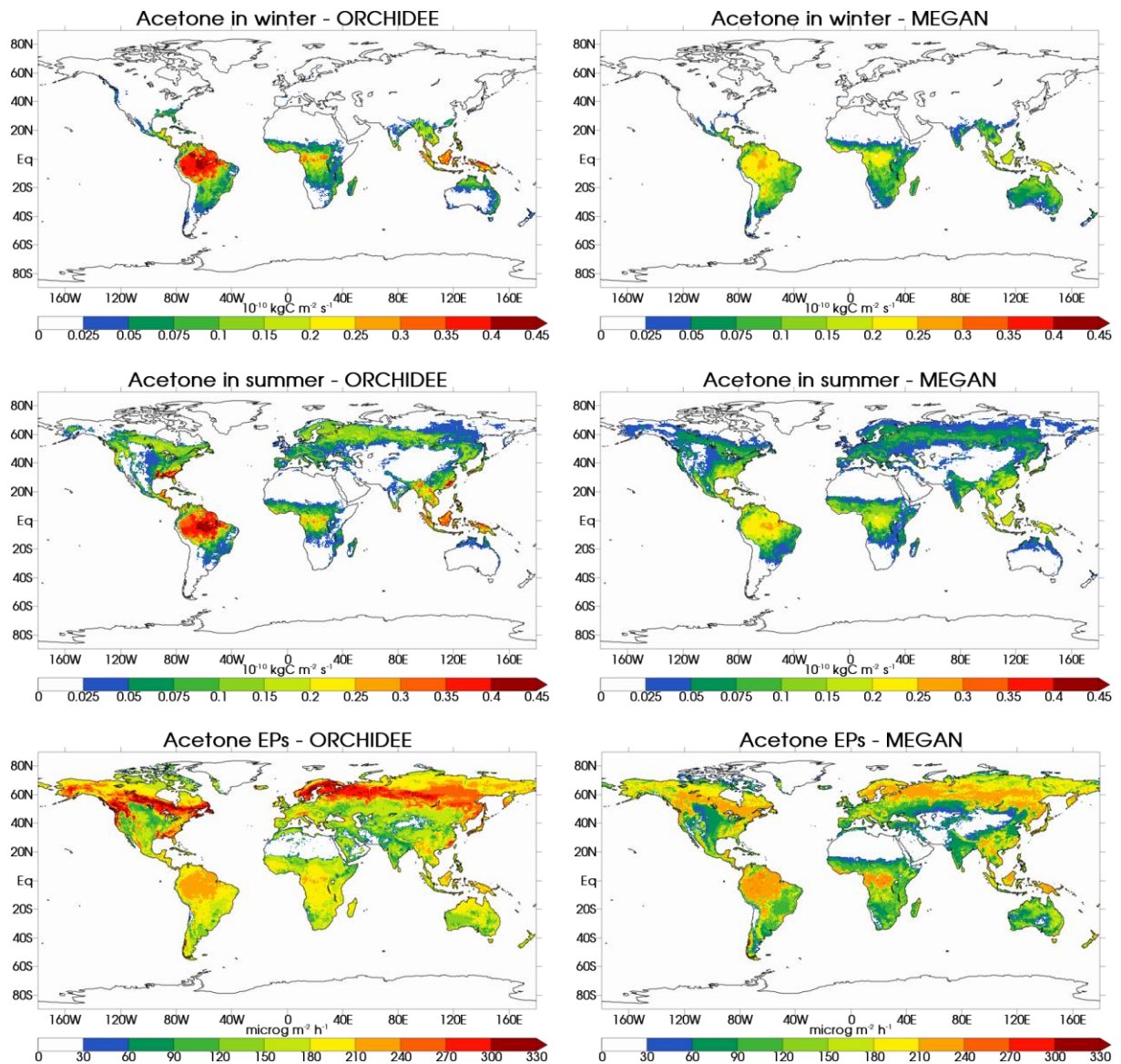


1  
 2 Figure 6. The same as Fig. 5, but for monoterpenes.  
 3

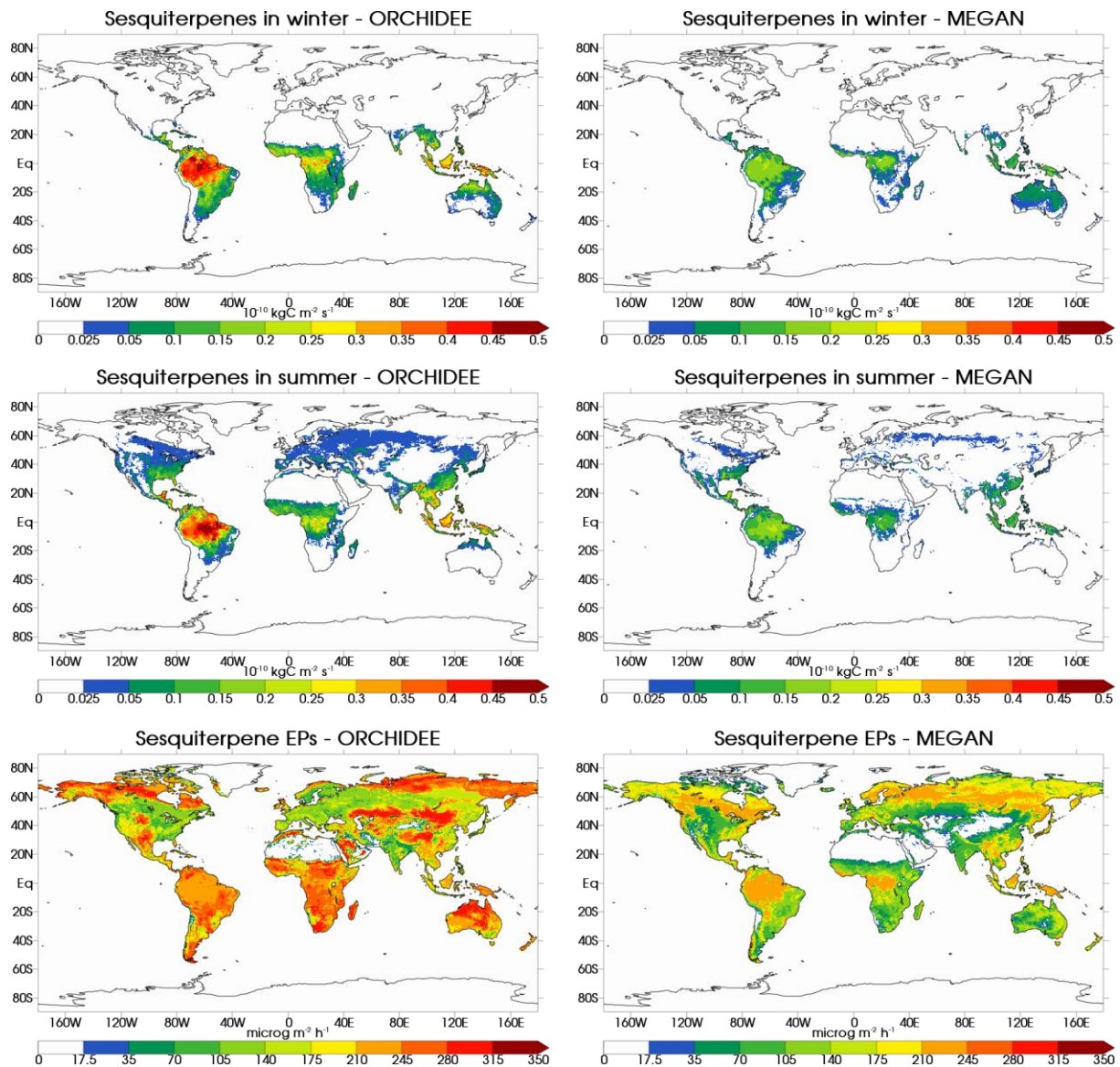




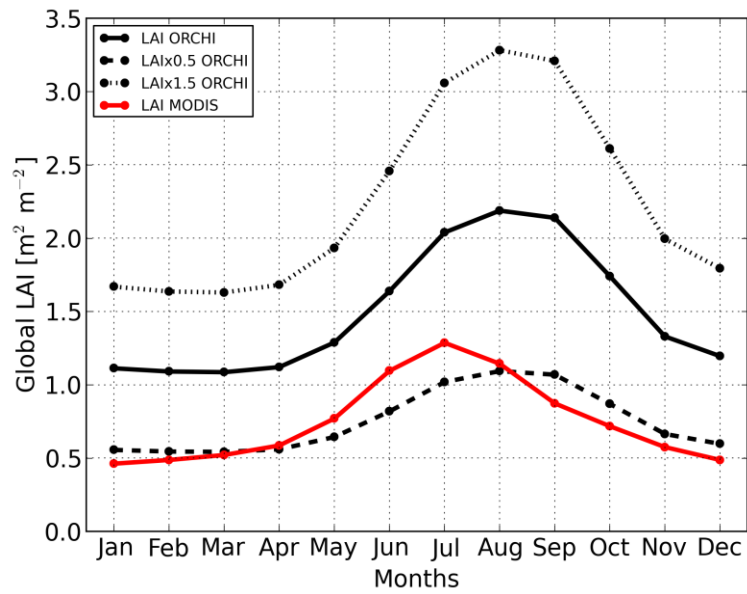
1  
 2 Figure 7. The same as Fig. 5, but for methanol.  
 3



1  
 2 Figure 8. The same as Fig. 5, but for acetone.  
 3



1  
 2 Figure 9. The same as Fig. 5, but for sesquiterpenes.  
 3

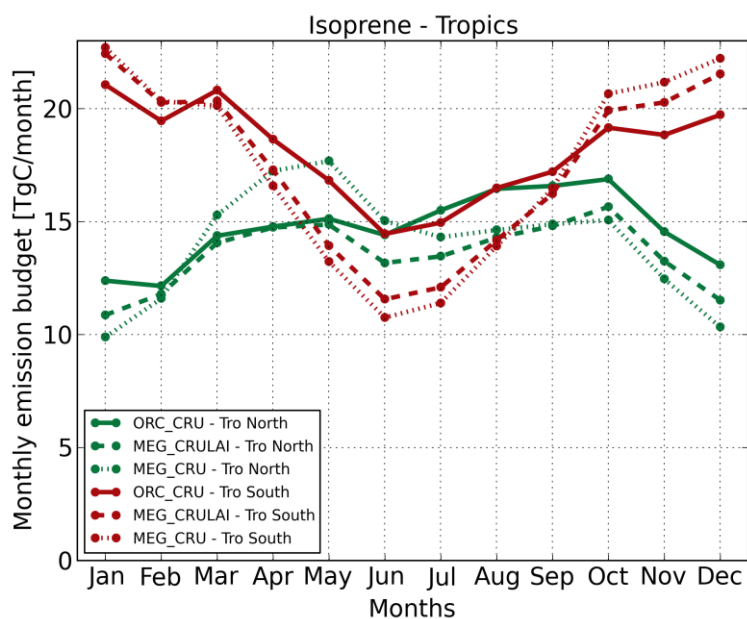


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2 Figure 10. Global monthly mean LAI ( $\text{m}^2 \text{m}^{-2}$ ) calculated by ORCHIDEE (solid black line)  
 3 and retrieved from MODIS measurements (red line). The thick and thin dashed lines represent  
 4 the LAI from ORCHIDEE multiplied by a factor 0.5 and 1.5, respectively.

5

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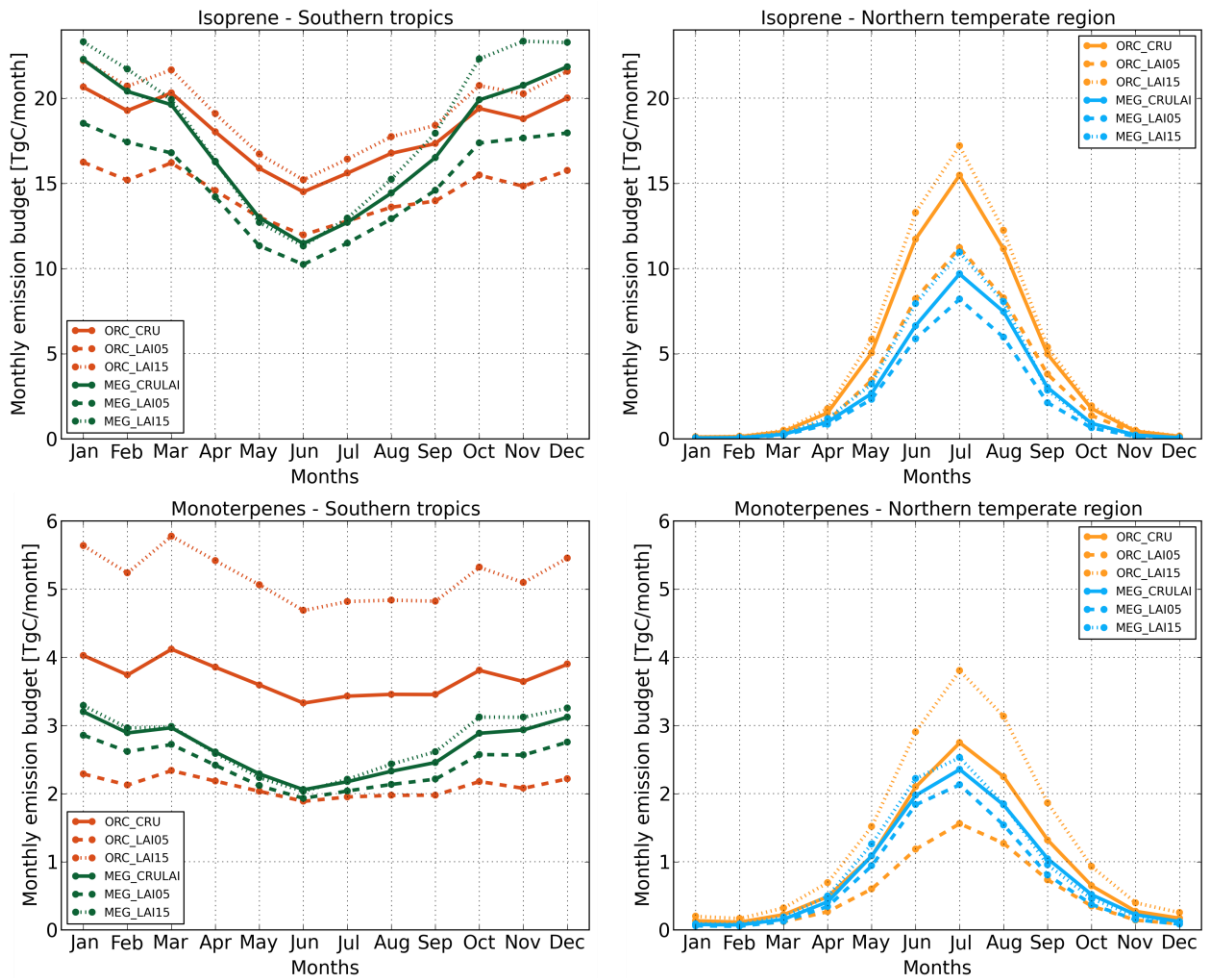


2

3 Figure 11. Zonal mean of monthly emission budgets ( $\text{Tg C month}^{-1}$ ), averaged over the  
4 simulation period (2000–2009) for the northern and southern tropics, in ORC\_CRU (solid  
5 line), MEG CRULAI (thick dashed line) and MEG\_CRU (thin CRULAI (dashed line)  
6 simulations for isoprene.

7

8

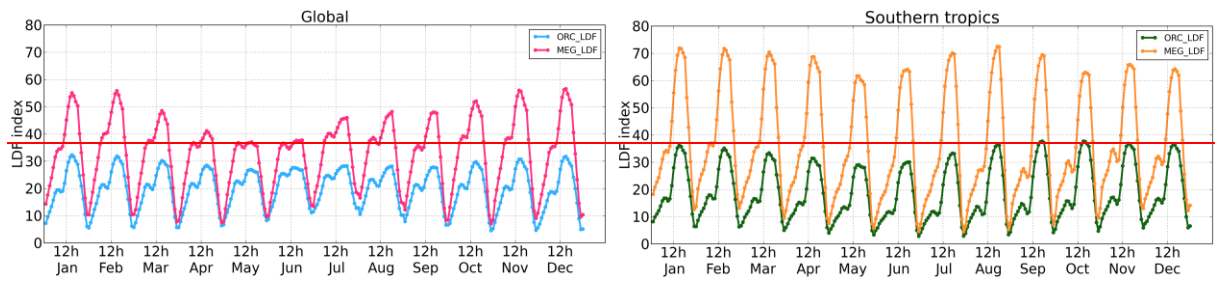


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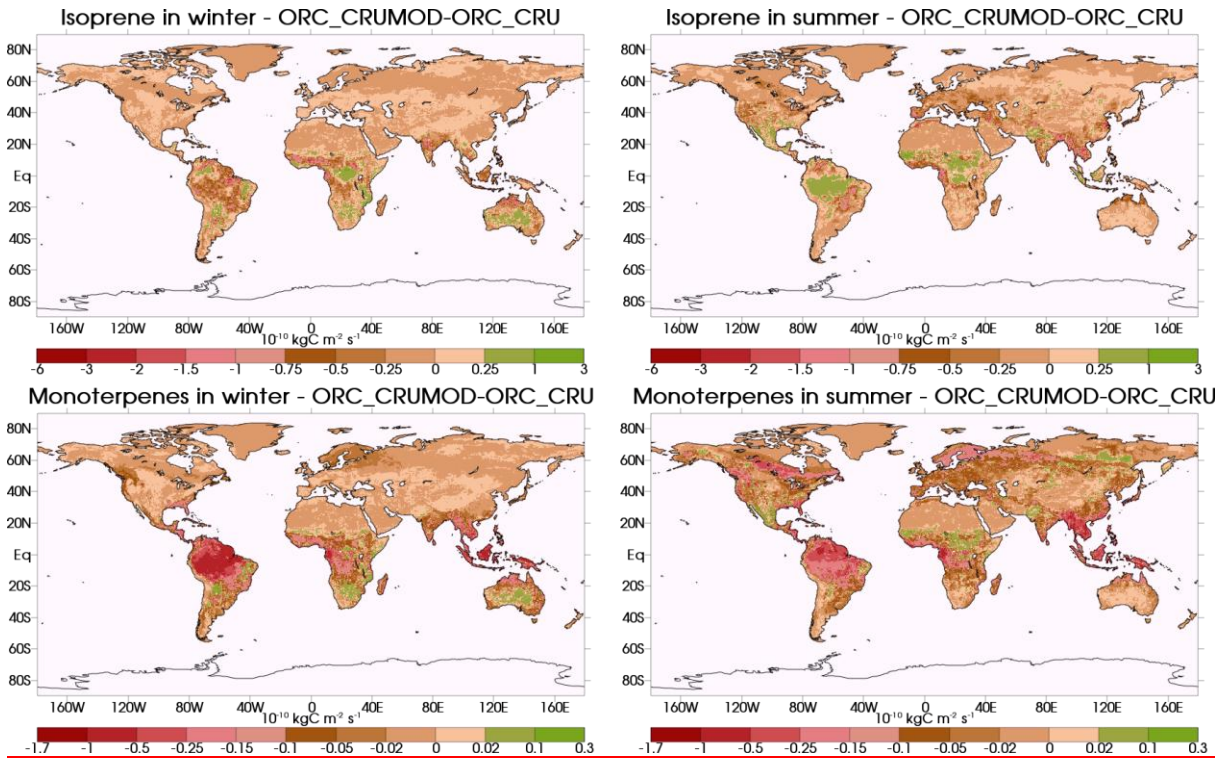
2 Figure 12. Zonal average of changed emissions in the different LAI sensitivity tests:  
 3 ORC\_CRU and MEG\_CRULAI using ORCHIDEE LAI (solid line), ORC\_LAI05 and  
 4 MEG\_LAI05 using ORCHIDEE LAI-0.5 (thick dashed line) and ORC\_LAI15 and  
 5 MEG\_LAI15 using ORCHIDEE LAI-1.5 (thin dashed line) in the year 2006, for the  
 6 southern tropical (left column) and northern temperate regions (right column) for isoprene and  
 7 monoterpenes. Emissions are given in  $\text{Tg C month}^{-1}$ .

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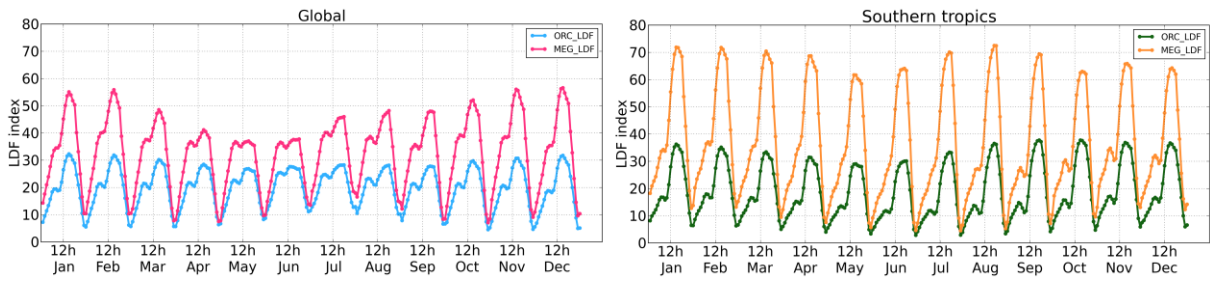
4

Figure 13. Differences between the ORC\_CRUMOD and ORC\_CRU simulation for isoprene and monoterpenes emissions in summer and winter for 2006.

6

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3 Figure 14, Figure 13. Global (left plot) and southern tropical (right plot) average of the LDF  
4 index for ORCHIDEE and MEGAN. The LDF index is provided as the hourly daily profile  
5 averaged over each month.

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