Acetylene (C$_2$H$_2$) and hydrogen cyanide (HCN) from IASI satellite observations: global distributions, validation, and comparison with model

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Abstract. We present global distributions of C$_2$H$_2$ and hydrogen cyanide (HCN) total columns derived from the Infrared Atmospheric Sounding Interferometer (IASI) for the years 2008–2010. These distributions are obtained with a fast method allowing to retrieve C$_2$H$_2$ abundance globally with a 5% precision and HCN abundance in the tropical (sub-tropical) belt with a 10% (25%) precision. IASI data are compared for validation purposes with ground-based Fourier transform infrared (FTIR) spectrometer measurements at four selected stations. We show that there is an overall agreement between the ground-based and space measurements with correlation coefficients for daily mean measurements ranging from 0.28 to 0.81, depending on the site. Global C$_2$H$_2$ and subtropical HCN abundances retrieved from IASI spectra show the expected seasonality linked to variations in the anthropogenic emissions and seasonal biomass burning activity, as well as exceptional events, and are in good agreement with previous spaceborne studies. Total columns simulated by the Model for Ozone and Related Chemical Tracers, version 4 (MOZART-4) are compared to the ground-based FTIR measurements at the four selected stations. The model is able to capture the seasonality in the two species in most of the cases, with correlation coefficients for daily mean measurements ranging from 0.50 to 0.86, depending on the site. IASI measurements are also compared to the distributions from MOZART-4. Seasonal cycles observed from satellite data are reasonably well reproduced by the model with correlation coefficients ranging from −0.31 to 0.93 for C$_2$H$_2$ daily means, and from 0.09 to 0.86 for HCN daily means, depending on the considered region. However, the anthropogenic (biomass burning) emissions used in the model seem to be overestimated (underestimated), and a negative global mean bias of 1% (16%) of the model relative to the satellite observations was found for C$_2$H$_2$ (HCN).
Table 1. Global HCN and C$_2$H$_2$ sources and sinks (Tg yr$^{-1}$).

<table>
<thead>
<tr>
<th>Sources</th>
<th>HCN</th>
<th>C$_2$H$_2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Biomass burning</td>
<td>0.1–3.18</td>
<td>1.6</td>
</tr>
<tr>
<td>Biofuels</td>
<td>0.21</td>
<td>3.3</td>
</tr>
<tr>
<td>Fossil fuel</td>
<td>0.02–0.04</td>
<td>1.7</td>
</tr>
<tr>
<td>Residential coal</td>
<td>0.2</td>
<td></td>
</tr>
<tr>
<td>Biogenic</td>
<td>0.2</td>
<td></td>
</tr>
<tr>
<td><strong>Total</strong></td>
<td>0.47–3.22</td>
<td>6.6</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Sinks</th>
<th>HCN</th>
<th>C$_2$H$_2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Ocean uptake</td>
<td>1.1–2.6</td>
<td></td>
</tr>
<tr>
<td>Reaction with OH</td>
<td>0.3</td>
<td>3.3</td>
</tr>
<tr>
<td>Photolysis</td>
<td>$0.2 \times 10^{-2}$</td>
<td></td>
</tr>
<tr>
<td>Reaction with O($^1$D)</td>
<td>$0.3 \times 10^{-3}$</td>
<td></td>
</tr>
<tr>
<td><strong>Total</strong></td>
<td>1.4–2.9</td>
<td>$\sim$ 6.6</td>
</tr>
</tbody>
</table>

1 Introduction

Hydrogen cyanide (HCN) and acetylene (or ethyne; C$_2$H$_2$) are ubiquitous atmospheric trace gases with medium lifetime, which are frequently used as indicators of combustion sources and as tracers for atmospheric transport and chemistry. Typical abundances of C$_2$H$_2$ (HCN) range from 1 to 2 (2 to 5) $\times 10^{15}$ molec cm$^{-2}$ for background levels, and from 8 to 20 (30 to 40) $\times 10^{15}$ molec cm$^{-2}$ for biomass burning plume levels (Rinsland et al., 1999, 2001, 2002; Zander et al., 1991; Zhao et al., 2002; Clarisse et al., 2011a; Vigouroux et al., 2012; Duflot et al., 2013). Table 1 summarizes the main sources and sinks for HCN and C$_2$H$_2$. For HCN, biomass burning is the primary source, followed by biofuel and fossil fuel combustions, and its primary sink is thought to be ocean uptake (Li et al., 2000, 2003). For C$_2$H$_2$, biofuel combustion is considered to be the dominant source, followed by fossil fuel combustion and biomass burning (Xiao et al., 2007). Reaction with hydroxyl radical (OH) is the main sink for C$_2$H$_2$, which may also act as a precursor of secondary organic aerosols (Volkamer et al., 2009).

With a tropospheric lifetime of 2–4 weeks for C$_2$H$_2$ (Logan et al., 1981) and 5–6 months for HCN (Li et al., 2000; Singh et al., 2003), these two species are interesting tracers for studying atmospheric transport. The study of the ratio C$_2$H$_2$ / CO (carbon monoxide) can also help to estimate the age of emitted plumes (Xiao et al., 2007).

Long-term local measurements of HCN and C$_2$H$_2$ are sparse and mainly performed from ground-based Fourier transform infrared (FTIR) spectrometer at selected stations of the Network for the Detection of Atmospheric Composition Change (NDACC; http://www.ndacc.org) (Vigouroux et al., 2012, and references therein). Global distributions of HCN and C$_2$H$_2$ may thus help to reduce the uncertainties remaining with regard to the magnitude of their sources and sinks, as well as to their spatial distribution and seasonality in the atmosphere (Li et al., 2009; Parker et al., 2011).

Satellite sounders have provided considerable new information in the past years, with measurements from the Atmospheric Chemistry Experiment (ACE-FTIR) (Lupu et al., 2009; González Abad et al., 2011), the Michelson Interferometer for Passive Atmospheric Sounding (MIPAS) (Park et al., 2011; Wiegele et al., 2012; Glathor et al., 2015) and the microwave limb sounder (MLS) (Pumphrey et al., 2011). These measurements were all made in limb geometry and consequently mostly in the upper troposphere or higher; also the spatial sampling from these instruments is limited, making it less well-suited when studying dynamical events on short timescales.

Having a twice daily global coverage and a 12 km diameter footprint at nadir, the Infrared Atmospheric Sounding Interferometer (IASI) infrared sounder (Clerbaux et al., 2009) aboard the MetOp-A satellite has the potential for providing measurements for these two species globally, and with higher spatial resolution and temporal sampling than what has been obtained up to now.

Previous studies have demonstrated that HCN and C$_2$H$_2$ can be observed with the IASI infrared nadir-looking hyperspectral sounder, e.g. in a specific biomass burning plume (Clarisse et al., 2011), as well as in an anthropogenic pollution plume uplifted in the free troposphere (Clarisse et al., 2011b). More recently, Duflot et al. (2013) have shown that HCN and C$_2$H$_2$ columns can be routinely retrieved from IASI spectra, even in the absence of exceptional columns or uplift mechanisms, when CO$_2$ line mixing is accounted for in the inversion scheme. These previous works were based on an optimal estimation method (OEM) developed and formalized by Rodgers (2000).

In this paper, we first present a fast scheme for the global detection and quantification of HCN and C$_2$H$_2$ total columns from IASI spectra. We describe 2008–2010 time series and analyze the seasonality of the columns of these two species above four NDACC sites in comparison with ground-based FTIR measurements. We finally present the global distributions for the years 2008 to 2010 that we compare with model outputs for these two species.

2 Instrument and method

2.1 IASI

IASI is on board the MetOp-A platform launched in a Sun-synchronous orbit around the Earth at the end of 2006. The overpass times are 09:30 and 21:30 MLT (mean local time). Combining the satellite track with a swath of 2200 km, IASI provides global coverage of the Earth twice a day with a footprint of 12 km at nadir. IASI is a Fourier transform spectrometer that measures the thermal infrared radi-
ation emitted by the Earth’s surface and atmosphere in the 645–2760 cm\(^{-1}\) spectral range with a spectral resolution of 0.5 cm\(^{-1}\) apodized and a radiometric noise below 0.2 K between 645 and 950 cm\(^{-1}\) at 280 K (Clerbaux et al., 2009).

The IASI spectra used in this study are calibrated radiance spectra provided by EUMETCast near-real-time service.

2.2 Retrieval strategy

Up to now, 24 trace gases have been detected from IASI radiance spectra, including HCN and C\(_2\)H\(_2\) (see Clarisse et al., 2011a, for the list of detected species), with an OEM (Rodgers, 2000) implemented in a line by line radiative transfer model called Atmostophit (Coheur et al., 2005). In the cases of HCN and C\(_2\)H\(_2\), the accuracy of the retrievals has been recently improved by taking into consideration the CO\(_2\) line mixing in the radiative transfer model (Duflot et al., 2013).

This retrieval method, relying on spectral fitting, needs a high computational power and is time-consuming, especially when a large number of spectra has to be analyzed and fitted. This is therefore not suitable for providing global-scale concentration distributions of these trace gases in a reasonable time.

One of the commonly used methods for the fast detection of trace gases is the brightness temperature difference (BTD) between a small number of channels, some being sensitive to the target species, some being not. Such a method has been used from IASI spectra for sulfur dioxide (SO\(_2\)) (Clarisse et al., 2008) and ammonia (NH\(_3\)) (Clarisse et al., 2009). It is of particular interest in operational applications (quick alerts) or when large amounts of data need to be processed. However, relying on a cautious selection of channels to avoid the contamination with other trace gases, the BTD method does not fully exploit all the information contained in hyperspectral measurements. Especially, low concentrations of the target species may not be detected with such a method.

Walker et al. (2011) presented a fast and reliable method for the detection of atmospheric trace gases that fully exploits the spectral range and spectral resolution of hyperspectral instruments in a single retrieval step. They used it to retrieve SO\(_2\) total column from a volcanic plume and NH\(_3\) total column above India. More recently, Van Damme et al. (2014) presented a retrieval scheme to retrieve NH\(_3\) from IASI spectra based on the work of Walker et al. (2011), and introduced a metric called hyperspectral range index (HRI). We use in the present study a similar approach.

2.2.1 Hyperspectral range index

The method used in this study is a non-iterative pseudo retrieval method of a single physical variable or target species x expressed following the formalism developed by Rodgers (2000):

\[
\hat{x} = x_0 + (K^T S_{tot}^{-1} K)^{-1} K^T S_{tot}^{-1} (y - F(x_0)),
\]

where y is the spectral measurements, x\(_0\) is the linearization point, F is the forward model (FM), S\(_{tot}\) is the covariance of the total error (random + systematic), and the Jacobian K is the derivative of the FM to the target species in a fixed atmosphere.

S\(_{tot}\) can be estimated considering an appropriate ensemble of \(N\) measured spectra, which can be used to build up the total measurement error covariance S\(_{obs}\):

\[
S_{tot} \simeq \frac{1}{N - 1} \sum_{j=1}^{N} (y_j - \bar{y})(y_j - \bar{y})^T = S_{obs},
\]

where \(\bar{y}\) is the calculated mean spectrum for the ensemble.

To generate S\(_{obs}\), we randomly chose one million cloud free spectra observed by IASI all over the world, above both land and sea, during the year 2009. Then, we applied a BTD test to remove the spectra contaminated by the target species. For HCN (C\(_2\)H\(_2\)), the wave numbers 716.5 and 732 cm\(^{-1}\) (712.25 and 737.75 cm\(^{-1}\)) were used as reference channels and 712.5 cm\(^{-1}\) (730 cm\(^{-1}\)) was used as test channel (Fig. 1, middle panel). Given the medium lifetimes of the target species (few weeks for C\(_2\)H\(_2\) to few months for HCN), and the limited accuracy of the BTD test due to the weak spectral signatures of the target species, it is likely that such randomly chosen and filtered spectra still contain a small amount of the target species whose signal may come out from the noise. This limitation decreases the sensitivity of the method, which is discussed in Sect. 2.2.3.

The spectral ranges considered to compute the S\(_{obs}\) matrices are 645–800 cm\(^{-1}\) for HCN and 645–845 cm\(^{-1}\) for C\(_2\)H\(_2\) (Fig. 1, top panel). These ranges were chosen as they include parts of the spectrum which have a relatively strong signal from the target species but also from the main interfering species (CO\(_2\), H\(_2\)O and O\(_3\); Fig. 1, bottom panel) in order to maximize the contrast with the spectral background.

Having calculated S\(_{obs}\) and , the HRI of a measured spectrum y can be defined as

\[
\text{HRI} = G(y - \bar{y})
\]

with G the measurement contribution function

\[
G = (K^T S_{obs}^{-1} K)^{-1} K^T S_{obs}^{-1}
\]

The HRI is a dimensionless scalar similar, other than units, to the apparent column retrieved in Walker et al. (2011). Unlike the optimal estimation method, no information about the vertical sensitivity can be extracted. Note also that the use of a fixed Jacobian to calculate HRI does not allow for generating meaningful averaging kernels.

2.2.2 Conversion of HRI into total columns

Having calculated the matrices G for HCN and C\(_2\)H\(_2\), each observed spectrum can be associated through Eq. (3) with

\[
\text{HRI} = G(y - \bar{y})
\]

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To determine these coefficients linking the HRIs to total column amounts, HCN and C$_2$H$_2$ profiles have been constructed, with enhanced concentrations of the species located in a 1 km thick layer, whose altitude is varied from the ground up to 30 km for HCN and up to 20 km for C$_2$H$_2$ (the choice of these maximum altitudes are made with respect to the Jacobians of the FM that are shown in Fig. 3 and commented on in Sect. 2.2.3). Each of the constructed profile has been associated with a spectrum through the FM of Atmosphit considering standard absorption profiles from modelled atmospheres for the other species. The associated values of HRI$_{\text{HCN}}$ and HRI$_{\text{C}_2\text{H}_2}$ have then been computed for each of the simulated spectra. Figure 2 shows the look up tables (LUTs) of HRI$_{\text{HCN}}$ (top) and HRI$_{\text{C}_2\text{H}_2}$ (bottom) as a function of the abundance of the target molecule and of the altitude of the polluted layer in a standard tropical modelled atmosphere (Anderson et al., 1986). Similar LUTs have been computed for standard temperate (US standard atmosphere) and polar (Anderson et al., 1986) atmospheres (data not shown). The satellite viewing angles were taken into account in the HRI calculation similarly to Van Damme et al. (2014). One can see that, for a given atmosphere and for a given altitude of the polluted layer, the abundances of both species linearly depend on the HRI value, which validates Eq. (5). For a given atm and a given species $X$, the different values of $B$ with respect to the altitude $z$ of the polluted layer will be noted $b_{X_{\text{atm}}}(z)$ and $b_{X_{\text{atm}}}(z_{\text{atm}})$ (in molec cm$^{-2}$) in the following.

Figure 3 shows the normalized Jacobians of the FM for HCN and C$_2$H$_2$ averaged over the spectral ranges given in Sect. 2.2.1 (645–800 cm$^{-1}$ for HCN and 645–845 cm$^{-1}$ for C$_2$H$_2$) and for each of the three standard modelled atmospheres. These Jacobians express the sensitivity of the FM, i.e. both the radiative transfer model and IASI (through its instrumental function), to the target species abundance $X$ in a fixed atm:

$$K_{X_{\text{atm}}} = \left[ \frac{\partial F_{\text{atm}}}{\partial X}(z_1) \ldots \frac{\partial F_{\text{atm}}}{\partial X}(z_n) \right]$$

$$= [k_{X_{\text{atm}}}(z_1) \ldots k_{X_{\text{atm}}}(z_n)].$$

We then obtain the coefficients $B_{X_{\text{atm}}}$ by multiplying the $b_{X_{\text{atm}}}(z)$ by the value of the Jacobian at the altitude $z$:

$$B_{X_{\text{atm}}} = \sum_{i=1}^{n} (b_{X_{\text{atm}}}(z_i) \times k_{X_{\text{atm}}}(z_i))$$

with $\sum_{i=1}^{n} k_{X_{\text{atm}}}(z_i) = 1$. (7)

Applying this method to the three standard modelled atmospheres (tropical, temperate and polar), we get a $B_X$ value for each, which we have associated with the corresponding range of latitude ($\pm20^\circ$, $\pm45^\circ$ : $\pm60^\circ$, $\pm75^\circ$ : $\pm90^\circ$, respectively), and linearly interpolated between. Figure 4 gives the resulting values of $B_{\text{HCN}}$ (blue) and $B_{\text{C}_2\text{H}_2}$ (green) in a function of the latitude.

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Figure 1. (Top) Simulated spectra in the region of the HCN $\nu_2$ band and C$_2$H$_2$ $\nu_3$ band. The green (brown) double sided arrow gives the spectral range used to compute the $S_i$ matrices for HCN (C$_2$H$_2$). (Middle) contributions of climatological background levels of HCN and C$_2$H$_2$. (Bottom) contribution of CO$_2$ (red line), O$_3$ (green line) and H$_2$O (blue line) to a simulated spectrum for background concentrations. Calculations have been made for the US standard atmosphere (Standard Atmosphere: National Atmospheric and Oceanic Administration, 1976) with CO$_2$ concentrations scaled to 390 ppmv.

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A value of HRI for HCN (HRI$_{\text{HCN}}$) and C$_2$H$_2$ (HRI$_{\text{C}_2\text{H}_2}$). These HRIs are only metrics for determining whether levels of the gas are enhanced with respect to the climatological background over the vertical levels where the instrument is sensitive. For a given atmosphere (atm), the main challenge is then to link the HRI to a column amount of the target molecule, i.e. to find $B_{\text{HCN}_{\text{atm}}}$ and $B_{\text{C}_2\text{H}_2_{\text{atm}}}$ (in molec cm$^{-2}$) such as

$$[X] = B_{X_{\text{atm}}} \times \text{HRI}_X.$$

$[X]$ being the species abundance in molec cm$^{-2}$.
2.2.3 Sensitivity and stability of the method

The sensitivity of the method can be assessed from the Jacobians presented in Fig. 3. For HCN, one can see that there is no sensitivity at the surface and above ~30 km, and the altitude of the sensitivity peak is located close to the tropopause at ~9 km, ~11 km and ~14 km for the polar, temperate and tropical atmospheres, respectively. For C$_2$H$_2$, there is no sensitivity above ~20 km, and the maximum sensitivity is reached at ~8, ~10 and ~11 km for the polar, temperate and tropical atmospheres, respectively. The vertical distribution in a standard temperate atmosphere (US standard atmosphere) is also shown for both species in Fig. 3. These standard distributions agree reasonably well (in shape and value) with observed profiles exhibited in previous studies (e.g. Li et al., 2003; Xiao et al., 2007). The C$_2$H$_2$ Jacobians match quiet well with the standard distribution of the molecule (i.e. from the ground up to ~20 km), with somehow a lack of sensitivity close to the ground where C$_2$H$_2$ is the most abundant. For HCN, which shows a nearly “flat” vertical distribution up to 30 km, the Jacobians show a lack of sensitivity close to the ground and above 30 km, where HCN is still present (HCN distribution decreases down to 60 pptv at 60 km – data not shown).

The HRIs presented here above are sensitive to the abundance of the target species – this is what they are made for – and to their vertical distribution. However, once cloudy spectra have been discarded, the measured column amount may also depend on (1) the proper suppression of the spectral
background, (2) the conditions of thermal contrast (TC) with the surface, and (3) the accuracy of the FM to simulate the spectra used to build up the LUTs. The latter was discussed already by Duflot et al. (2013). In order to test the impact of the two first factors (spectral background suppression and TC) on the retrieved column amount, HCN and C$_2$H$_2$ profiles have been constructed with varying TC and concentrations of the interfering and target species. In general thermal contrast can be defined as the temperature difference between the surface and the air temperature at some altitude of interest. We consider here the same definition for the TC as in Van Damme et al. (2014): the TC is defined here as the difference between the skin (surface) temperature and that of the air at an altitude of 1.5 km. These variations in interfering species abundances and TC were considered to be independent and were taken within the range ±2% for CO$_2$ and ±20% for H$_2$O and O$_3$, and in the range ±10 K for the TC. For a fixed column amount of the target species, the HRIs were compared one by one to a HRI corresponding to a standard spectrum (i.e. with background concentrations of the interfering species and a TC equal to zero) and if the difference between the two HRIs was lower than 10%, then this fixed abundance of the target species was tagged as independently detectable from the listed parameters.

The TC was found to be the major source of HRI variation for both target species, and a serious cause of limitation only for HCN. Figure 5 shows the variation of HRI$_{HCN}$ caused by a TC equal to ±10 K. One can see that the HCN column amount can be detected with a variation due to the TC below 10% when its abundance is higher than 0.28, 1.2 and 1.6 $\times$ 10$^{16}$ molec cm$^{-2}$ for the tropical, temperate and polar atmospheres, respectively. This gives the stability thresholds above which HCN column amount can be measured with a 10% confidence in the independence of the retrieval method to the atmospheric parameters. Consequently, as the stability thresholds of the method for HCN in temperate and polar atmospheres are too high (1.2 and 1.6 $\times$ 10$^{16}$ molec cm$^{-2}$, respectively) to allow for the detection of HCN background abundances as compared to usual background column of typically 0.35 $\times$ 10$^{16}$ molec cm$^{-2}$ (Vigouroux et al., 2012; Duflot et al., 2013), IASI HCN measurements have to be rejected in these two types of atmosphere, and considered in the tropical belt for values above 0.28 $\times$ 10$^{16}$ molec cm$^{-2}$. In order to broaden the exploitable latitude range, we performed the same study for subtropical latitudes (considering a mix of tropical and temperate atmospheres), and we found a 25% confidence in the independence of the retrieval method to the atmospheric parameters (data not shown). Consequently, in the following no IASI C$_2$H$_2$ measurements are rejected.

3 Results

The goal of this section is to describe and evaluate the C$_2$H$_2$ and HCN total columns as measured by IASI. We first compare HCN and C$_2$H$_2$ total columns retrieved from IASI spectra and from ground-based FTIR spectra. We then depict the C$_2$H$_2$ and HCN total columns at global and regional scales. IASI global and regional distributions are finally compared with output from the Model for Ozone and Related Chemical Tracers, version 4 (MOZART-4) in order to evaluate the agreement between the model and the IASI distributions.

3.1 Comparison with ground-based observations

We compare in this section HCN and C$_2$H$_2$ total columns retrieved from IASI spectra and from ground-based FTIR spectra for the years 2008–2010 for four selected ground-based FTIR observation sites (i.e. wherever observations for these two species were available during the period of study): Wollongong (34°S; 151°E; 30 m above mean sea level, a.m.s.l.), Reunion Island (21°S; 55°E; 50 m a.m.s.l.), Izaña (28°N; 16°W; 2367 m a.m.s.l.) and Jungfraujoch (46°N; 8°E; 3580 m a.m.s.l.) (Fig. 6). IASI cloudy spectra were removed from the data set using a 10% contamination threshold on the cloud fraction in the pixel. As exposed in Sect. 2.2.3, errors in retrieved species abundances from IASI spectra due to variations in atmospheric parameters are 10% at tropical latitudes ($\pm$20°) and 25% at subtropical latitudes ($\pm$35°: $\pm$20°) for HCN and 5% for C$_2$H$_2$, and comparison with ground-based HCN measurements are only performed for tropical and subtropical sites (Reunion Island, Wollongong and Izaña).
The total error for ground-based measurements at Reunion Island is 17% for both species, total error for HCN ground-based measurements at Wollongong is 15%, total error for HCN ground-based measurements at Izaña is 10%, and total error for C$_2$H$_2$ ground-based measurements at Jungfraujoch is 7%. Detailed description of ground-based FTIR data set, retrieval method and error budget can be found in Vigouroux et al. (2012) for Reunion Island and in Mahieu et al. (2008) for Jungfraujoch. However, at Reunion Island, the retrieval strategies have been slightly improved from Vigouroux et al. (2012), mainly concerning the treatment of the interfering species, but the same spectral signatures are used. Izaña data set and error budget were obtained from the NDACC database (ftp://ftp.cpc.ncep.noaa.gov/ndacc/station/izana/). The Wollongong data set and error budget were calculated by N. Jones from the University of Wollongong, personal communication, 2014.

Figure 7 shows the mean total column averaging kernels for the ground-based FTIR at each of the four sites. Similar to IASI (Fig. 3), information content from ground-based instruments measurements is mostly in the middle-high troposphere for both species. The main difference can be observed for tropical C$_2$H$_2$: while IASI Jacobian peaks at 10 km for C$_2$H$_2$ in a tropical atmosphere, ground-based FTIR averaging kernel peaks at 15 km for C$_2$H$_2$ at Reunion Island.

Figure 8 shows the comparison between the IASI and the ground-based measurements. IASI retrieved total columns were averaged on a daily basis and on a 1° × 1° area around the observation sites. HCN retrieved abundances below 2.8 × 10$^{15}$ molec cm$^{-2}$ have been removed from both ground-based and space measurements to allow for comparison of both data sets (cf. Sect. 2.2.3). One can see that there is an overall agreement between the IASI and the ground-based FTIR measurements considering the error bars. An important result from this study is that IASI seems to capture the seasonality in the two species in most of the cases. This is best seen by looking at the IASI monthly mean retrieved total columns (black circles and lines). The scatter of the IASI daily mean measurements (red dots) are due to the averaging on a 1° × 1° area around the observation sites.
and HCN peaks are higher in 2010. As South American biomass burning plumes are known to impact trace gases abundance above Reunion Island (Edwards et al., 2006a, b; Duflot et al., 2010), these 2010 higher peaks are probably influenced by the 2010 great Amazonian fires (Lewis et al., 2011).

At Wollongong HCN peaks also in October–November due to the Southern Hemisphere biomass burning season (Paton-Walsh et al., 2010). We find maxima of around $11 \times 10^{15}$ molec cm$^{-2}$ in October 2010 for both space and ground-based instruments, which is, similar to Reunion Island, very likely to be a signature of the great Amazonian fires as South American biomass burning plumes are known to impact trace gases abundance above Australia (Edwards et al., 2006a, b). The seasonality and interannual variability matches well with that of the ground-based FTIR measurements (correlation coefficient of 0.55 for the entire daily mean data set, and of 0.83 for the monthly mean data set), with the IASI columns being biased low by $0.48 \times 10^{15}$ molec cm$^{-2}$ (10%).

At Izaña HCN peaks in May–July due to the biomass burning activity occurring in northern America and Europe (Sancho et al., 1992). We find maxima of around $8 \times 10^{15}$ molec cm$^{-2}$ in the IASI (ground-based FTIR) data set. The seasonality and interannual variability matches poorly with that of the ground-based FTIR measurements (correlation coefficient of 0.28 for the entire daily mean data set, and of 0.64 for the monthly mean data set), with the IASI columns being biased high by $0.45 \times 10^{15}$ molec cm$^{-2}$ (11%). One can notice that HCN total columns as measured by ground-based FTIR are below the HCN stability threshold in boreal winter, which may result in erroneous IASI measurements (because unstable) and explain this poor match between the two data sets.

For $C_2H_2$ at the Jungfraujoch site, the agreement between IASI and the ground-based retrieved columns is good (correlation coefficient of 0.70 for the entire daily mean data set, and of 0.85 for the monthly mean data set), with the IASI columns being biased low by $0.15 \times 10^{15}$ molec cm$^{-2}$ (12%), opposite to the observations at Reunion. The larger columns observed in late winter are caused by the increased $C_2H_2$ lifetime in that season (caused by the seasonal change in OH abundance) (Zander et al., 1991), and we find corresponding maxima of up to $3 \times 10^{15}$ molec cm$^{-2}$ in the IASI (ground-based FTIR) data set.

### 3.2 IASI Global distributions

We focus in this section on the description of the $C_2H_2$ and HCN distributions retrieved from IASI spectra. For practical reasons, the figures used in this section also show simulated distributions that will be analyzed afterwards.

The left panels of Figs. 9 and 10 provide the seasonal global and subtropical distributions of $C_2H_2$ and HCN total columns, respectively, as measured by IASI and averaged over the years 2008 to 2010.

---

**Figure 8.** Time series of HCN (left panel) and $C_2H_2$ (right panel) measurements for Reunion Island (HCN and $C_2H_2$), Wollongong (HCN only), Izaña (HCN only), and Jungfraujoch ($C_2H_2$ only). IASI measurements are shown as daily and $1^\circ \times 1^\circ$ means (red dots) with associated standard deviations (light red lines), and as monthly and $1^\circ \times 1^\circ$ means (black circles and line) with associated standard deviation (vertical black lines). Ground-based FTIR measurements are shown as daily means with associated total error by green crosses and lines. Correlation coefficients are given on each plot for daily means in red and for monthly means in black.

---

**Table 1.** Seasonal and interannual variability of HCN and $C_2H_2$ columns observed at Izaña, Reunion Island, and Jungfraujoch.

<table>
<thead>
<tr>
<th>Location</th>
<th>HCN columns</th>
<th>$C_2H_2$ columns</th>
</tr>
</thead>
<tbody>
<tr>
<td>Izaña</td>
<td>$0.98 \times 10^{15}$</td>
<td>$0.70 \times 10^{15}$</td>
</tr>
<tr>
<td>Reunion Island</td>
<td>$0.98 \times 10^{15}$</td>
<td>$0.72 \times 10^{15}$</td>
</tr>
<tr>
<td>Jungfraujoch</td>
<td>$0.64 \times 10^{15}$</td>
<td>$0.28 \times 10^{15}$</td>
</tr>
</tbody>
</table>
Looking at IASI measurements (Figs. 9 and 10 – left panels), one can notice the following main persisting features for both \( \text{C}_2\text{H}_2 \) and HCN:

- the hot spots mainly due to the biomass burning activity occurring in Africa and moving southward along the year (Sauvage et al., 2005; van der Werf et al., 2006);

- the hot spot located in Southeast Asia being likely a combination of biomass burning and anthropogenic activities;

- the transatlantic transport pathway linking the African west coast to the South American east coast and moving southward along the year (Edwards et al., 2003, 2006a, b; Glatthor et al., 2015).

The following seasonal features can also be observed:

- the transpacific transport pathway linking eastern Asia to western North America, especially in March–April–May (MAM) (Yienger et al., 2000);
– the transport pathway from southern Africa to Australia in June–July–August (JJA) and September–October–November (SON) (Annegarn et al., 2002; Edwards et al., 2006a, b);
– the transport pathway linking South America (especially Amazonia) to southern Africa and Australia during the SON period (Edwards et al., 2006a, b; Glatthor et al., 2015);
– the transport of the northern African plume over southern Asia to as far as the eastern Pacific by the northern subtropical jet during the MAM period (Glatthor et al., 2015);
– the Asian monsoon anticyclone (AMA), which is the dominant circulation feature in the Indian–Asian upper troposphere–lower stratosphere (UTLS) region during the Asian summer monsoon, spanning Southeast Asia to the Middle East and flanked by the equatorial and subtropical jets (Hoskins and Rodwell, 1995). The AMA is a known region of persistent enhanced pollution in the upper troposphere, linked to rapid vertical transport of surface air from Asia, India, and Indonesia in deep convection, and confinement by the strong anticyclonic circulation (Randel et al., 2010). The enhanced abundance of C$_2$H$_2$ and HCN within the AMA in JJA observed by IASI is in accordance with previous studies (Park et al., 2008; Randel et al., 2010; Parker et al., 2011; Glatthor et al., 2015); however, one should keep in mind that this enhanced abundance measured by IASI is likely due to the combination of this pollution uplift and confinement with the higher sensitivity of the method in the upper troposphere (Fig. 2).

One can also notice the very good agreement between the seasonal HCN distributions shown in our Fig. 10 and the ones published recently in Glatthor et al. (2015, Fig. 3).

Figures 11 and 12 show the C$_2$H$_2$ and HCN total columns time series, respectively, as measured by IASI (red dots) with the associated standard deviation (light red lines) for each of the zones defined in Fig. 6.

In northern America, Europe and boreal Central Asia (Fig. 11 – zones NAM, EUR and BCA), C$_2$H$_2$ peaks in late boreal winter due to the increased C$_2$H$_2$ lifetime as already noticed over Jungfraujoch (Fig. 8). The boreal summer 2008 California wildfires event (Gyawali et al., 2009) is clearly visible in the NAM plot, as well as the August 2009 Russian
wildfires in the NAM, EUR and BCA plots (Parrington et al., 2012; R’Honi et al., 2013).

In north-central America (Fig. 12 – zone NCA), the annual HCN peak in April–June is driven by local fire activity (van der Werf et al., 2010).

In South America, southern Africa and Australia (Figs. 11 and 12 – zones SAM, SAF and AUS), the Southern Hemisphere biomass burning season clearly drives the C$_2$H$_2$ and HCN peaks in September–November each year. The signature of the great 2010 Amazonian fires (Lewis et al., 2011) is visible on each of the three zones, South American fire plumes being known to impact southern Africa and Australia (Edwards et al., 2003, 2006a, b). The February 2009 Australian bush fires (Glatthor et al., 2013) are also noticeable on zone AUS for both species.

In northern Africa (Figs. 11 and 12 – zone NAF), C$_2$H$_2$ and HCN peak in boreal winter because of the biomass burning activity occurring in the zone, and peak also in boreal summer because of the European and southern Mediterranean fires (van der Werf et al., 2010).

In Southeast Asia (Figs. 11 and 12 – zone SEA), the observed C$_2$H$_2$ and HCN peaks in July–September and January–March are due to local fire activity (Fortems-Cheiney et al., 2011; Magi et al., 2012). Additionally, the July–September peaks are also likely due to the combination of the pollution uplift and confinement within the AMA with the higher sensitivity of the method in the upper troposphere.

In equatorial Asia (Figs. 11 and 12 – zone EQA), local fire activity is visible in July–October, as well as the Southeast Asian fire activity in January–March (Fortems-Cheiney et al., 2011; Magi et al., 2012). The high biomass burning activity occurring in Indonesia from July to December 2009 (Yulianti et al., 2013; Hyer et al., 2013) is also clearly noticeable.

C$_2$H$_2$ and HCN sharing important common sources (cf. Introduction), the same annual and seasonal features are observed for both species. However, biomass burning being the major source for HCN (while it is biofuel and fossil fuel combustions for C$_2$H$_2$), one can notice the especially high increase in HCN abundance (up to $13 \times 10^{15}$ molec cm$^{-2}$) in the Southern Hemisphere during the austral biomass burning season (September to November). These observations are in accordance with previous studies (Lupu et al., 2009; Glatthor et al., 2009; Wiegele et al., 2012).

### 3.3 Comparison with model

In order to further evaluate the HCN and C$_2$H$_2$ distributions retrieved from IASI spectra, they are compared in this section to the output of MOZART-4 for the years 2008–2010. We first describe the simulation set-up before comparing simu-
MOZART-4 simulation set-up

The model simulations presented here are performed with the MOZART-4 global 3-D chemical transport model (Emmons et al., 2010a), which is driven by assimilated meteorological fields from the NASA Global Modeling and Assimilation Office (GMAO) Goddard Earth Observing System (GEOS). MOZART-4 was run with a horizontal resolution of 1.875° latitude × 2.5° longitude, with 56 levels in the vertical and with its standard chemical mechanism (see Emmons et al. (2010a) and Lamarque et al. (2012), for details). The model simulations have been initialized by simulations starting in July 2007 to avoid contamination by the spin-up in the model results. MOZART-4 simulations of numerous species (CO, O₃ and related tracers including C₂H₂ and HCN) have been previously compared to in situ and satellite observations and used to track the intercontinental transport of pollution (e.g. Emmons et al., 2010b; Pfister et al., 2006, 2008, 2011; Tilmes et al., 2011; Clarisse et al., 2011b; Wespes et al., 2012; Viatte et al., 2015).

The surface anthropogenic (including fossil fuel and biofuel) emissions used here were taken from the inventory provided by D. Streets and University of Iowa and created for the Arctic Research of the Composition of the Troposphere from Aircraft and Satellites (ARCTAS) campaign (see http://bio.cgrer.uiowa.edu/arctas/emission.html for more information). This inventory was developed in the frame of the POLARCAT Model Intercomparison Program (POLMIP) and is a composite data set of regional emissions as representative of current emissions as possible; it is built upon the INTEX-B Asia inventory (Zhang et al., 2009) with the US NEI (National Emission Inventory) 2002 and CAC 2005 for North America and the EMEP (European Monitoring and Evaluation Programme) 2006 for Europe inventory to make up NH emissions (see Emmons et al. (2015) for an evaluation of POLMIP models). Emissions from EDGAR (Emissions Database for Global Atmospheric Research) were used for missing regions and species. Since only total volatile organic compounds (VOCs) were provided with this POLMIP inventory, the VOC speciation based on the RETRO emissions inventory as in Lamarque et al. (2010) was used. The anthropogenic emissions are constant in time with no monthly variations.

Daily biomass burning emissions were taken from the global Fire INventory from NCAR (FINN) version 1 (Wiedinmyer et al., 2011). The fire emissions for individual fires, based on daily MODIS fire counts, were calculated and then gridded to the simulation resolution (Wiedinmyer et al., 2006, 2011). The oceanic emissions are taken from the POET emissions data set (Granier et al., 2005) and the biogenic emissions from MEGANv2 (Model of Emissions of Gases and Aerosols from Nature) data set inventory (Guenther et al., 2006).

Model emissions for HCN and C₂H₂ used in this study are summarized in Table 2 and presented in Fig. 13. The majority of the emissions for both C₂H₂ and HCN are from anthropogenic source (about 80 and 55 % of the global source of C₂H₂ and HCN, respectively; see Table 2). Averaged over the period 2008–2010, the highest HCN and C₂H₂ anthropogenic surface emissions are observed over China, with elevated emissions over India, Europe and USA, due to in-
tense industrialization, where values larger than $4 \times 10^{-12}$ kg(C$_2$H$_2$) m$^{-2}$ s$^{-1}$ are entered in the model. The most intense HCN and C$_2$H$_2$ emissions due to biomass burning are observed over Southeast Asia, equatorial and southern Africa, South America, Siberia and Canada.

Figure 13. C$_2$H$_2$ and HCN surface emission fluxes (kg m$^{-2}$ s$^{-1}$) averaged over the period 2008–2010 from the anthropogenic and fire emissions inventories used in MOZART-4.

Figure 14. Time series of HCN (left panel) and C$_2$H$_2$ (right panel) measurements and simulations for Reunion Island (HCN and C$_2$H$_2$), Wollongong (HCN only), Izaña (HCN only), and Jungfraujoch (C$_2$H$_2$ only). MOZART-4 total columns are shown as daily and 1.875° x 2.5° means (blue crosses). Ground-based FTIR measurements are shown as daily means with associated total error by green circles and lines. Correlation coefficient and bias are given on each plot for daily means.

3.3.2 Model vs. ground-based FTIR observations

MOZART-4 simulations can be first evaluated by comparing them to the ground-based measurements at the four sites studied previously (Sect. 3.1). To perform this comparison, we use the FTIR averaging kernels and a priori to priorize to degrade the model vertical profile to the FTIR vertical resolution, in order to obtain the model “smoothed” total column, which represents what the FTIR would measure if the model profile was the true state (see Eq. (25) of Rodgers and Connor (2003), and Vigouroux et al. (2012) for an example). Figure 14 shows the comparison between the simulated “smoothed” total columns and the ground-based measurements. Model outputs are given in a 1.875° latitude x 2.5° longitude box (corresponding to the horizontal resolution of the model) over the ground-based measurement points. One can see that there is an overall agreement between the ground-based instruments and the model, the latter being obviously able to capture the seasonality in the two species in most of the cases.

For HCN, the model seasonality and interannual variability matches the ground-based FTIR measurements at Reunion Island very well (correlation coefficient of 0.86) and Wollongong (correlation coefficient of 0.85), and reasonably well at Izaña (correlation coefficient of 0.59). For this last site, one can see that the model correctly captures the abun-
dance peak occurring in May–July (cf. Sect. 3.1), but sets another peak around October. This second yearly peak for HCN at Izaña in the model simulations could be due to an overestimation of the southern African contribution to the northern African loading; this hypothesis will be analyzed in the next section. For HCN, the simulated total columns are biased low at Reunion Island by $1.29 \times 10^{15}$ molec cm$^{-2}$ (28 %) and at Wollongong by $1.8 \times 10^{15}$ molec cm$^{-2}$ (38 %). One can notice that the model does not capture the 2010 great Amazonian fires exceptional event visible on ground-based measurements at Reunion Island and Wollongong (cf. Sect. 3.1), which could explain the biases between the model and ground-based data sets at these two sites. At Izaña, the model is biased high by $0.46 \times 10^{15}$ molec cm$^{-2}$ (13 %), which seems to be caused by the second yearly peak simulated by the model.

For C$_2$H$_2$, the model seasonality and interannual variability matches very well that of the ground-based FTIR measurements at Jungfraujoch (correlation coefficient of 0.86) and reasonably well at Reunion Island (correlation coefficient of 0.50). For this last site, one can see that the model captures correctly the yearly peak occurring in October–November (cf. Sect. 3.1), but struggles to simulate the large day-to-day variations observed by the ground-based FTIR. This is illustrated by the very good correlation coefficient between the two data sets when dealing with monthly mean observations: 0.87 (data not shown). For C$_2$H$_2$, the simulated total columns are biased low for every site: $0.20 \times 10^{15}$ molec cm$^{-2}$ (21 %) at Reunion Island, and $0.27 \times 10^{15}$ molec cm$^{-2}$ (20 %) at Jungfraujoch. Similar to HCN, one can notice that the model does not capture the 2010 great Amazonian fires exceptional event visible on ground-based measurements at Reunion Island, which could explain the bias between the model and ground-based data sets at this site.

### 3.3.3 IASI vs. model global distributions

Figures 9 and 10 provide the seasonal global and subtropical distributions of C$_2$H$_2$ and HCN total columns, respectively, as measured by IASI and as simulated by MOZART-4 averaged over the years 2008 to 2010. For comparisons with IASI, the hourly output from MOZART-4 was interpolated to the overpass time of IASI. In addition, the high-resolution modelled layers were smoothed by applying each of the MOZART-4 simulated profiles, the Jacobians of the used forward model (cf. Sects. 2.2.3 and Fig. 3), to take into account the sensitivity of both the radiative transfer model and IASI. This has been done instead of applying the averaging kernels since our retrieval scheme does not provide such information. Note that here again HCN abundances below $2.8 \times 10^{15}$ molec cm$^{-2}$ have been removed from both space measurements and simulated columns to allow for comparison of both data sets (cf. Sect. 2.2.3).

<table>
<thead>
<tr>
<th>Zones</th>
<th>C$_2$H$_2$ Bias (%)</th>
<th>HCN Bias (%)</th>
</tr>
</thead>
<tbody>
<tr>
<td>NAM</td>
<td>0.93</td>
<td>47</td>
</tr>
<tr>
<td>NCA</td>
<td></td>
<td>0.44</td>
</tr>
<tr>
<td>SAM</td>
<td>0.54</td>
<td>-50 0.76</td>
</tr>
<tr>
<td>EUR</td>
<td>0.88</td>
<td>115</td>
</tr>
<tr>
<td>NAF</td>
<td>0.66</td>
<td>-35 0.07</td>
</tr>
<tr>
<td>SAF</td>
<td>0.69</td>
<td>-67 0.86</td>
</tr>
<tr>
<td>BCA</td>
<td>0.82</td>
<td>105</td>
</tr>
<tr>
<td>SEA</td>
<td>-0.31</td>
<td>-23 0.57</td>
</tr>
<tr>
<td>EQA</td>
<td>0.45</td>
<td>-51 0.09</td>
</tr>
<tr>
<td>AUS</td>
<td>0.65</td>
<td>-61 0.77</td>
</tr>
<tr>
<td>Global</td>
<td>0.72</td>
<td>-1 0.69</td>
</tr>
</tbody>
</table>

MOZART-4 simulations can be evaluated by looking at Figs. 9 and 11 for C$_2$H$_2$, and Figs. 10 and 12 for HCN. Figures 11 and 12 show the simulated C$_2$H$_2$ and HCN total columns time series, respectively, for each of the zones defined in Fig. 6 superimposed to IASI observations. Table 3 summarizes the biases and correlation coefficients resulting from the comparison between model and observations. Looking at these tables and figures, the following conclusions can be drawn:

- Seasonal cycles observed from satellite data are reasonably well reproduced by the model.
- The African, South American, Asian and Indonesian hot spots are clearly visible in the model.
- Exceptional events that are captured by IASI (cf. Sect. 3.2) are not simulated by MOZART-4.
- The model is more negatively biased in the Southern Hemisphere (bias = -61 % for C$_2$H$_2$ and bias = -25 % for HCN) than in the Northern Hemisphere (bias = 40 % for C$_2$H$_2$ and bias = -3 % for HCN), suggesting that anthropogenic (biomass burning) emissions are likely overestimated (underestimated) in the model. Note that in the MOZART-4 simulation presented here, the fire emissions are injected at the surface, which might result in an underestimation of concentrations at higher altitudes where IASI shows an increased sensitivity.
- The model reasonably reproduces the main transport pathways identified on IASI observations (cf. Sect. 3.2). However, the low background concentrations in the Southern Hemisphere as simulated by the model, especially for southern Africa and Australia (Figs. 11 and 12 – zones SAF and AUS and previous section at Reunion
Island and Wollongong), is possibly due to a mix of uncertainties introduced by the coarse grid of the model producing too much diffusion, problems in the transportation scheme for fine-scale plumes, the fire injection set at the surface and uncertainties in the emissions. The fact that only three representative Jacobians are used to perform the global comparison might also play a role.

In Table 3, for C$_2$H$_2$, the correlation coefficients are good ($\geq 0.6$) to very good ($\geq 0.9$) except for the zones SAM (South America), SEA (Southeast Asia) and EQA (equatorial Asia). For HCN, the correlation coefficients are good ($\geq 0.6$) except for the zones NCA (north-central America), NAF (northern Africa), SEA and EQA.

For South America (zone SAM), correlation coefficient is not as good for C$_2$H$_2$ ($R = 0.54$) due to a backward shift of the species abundance peaks in years 2008 and 2009: in the model, this increase occurs from July to October, while observations (and previous studies, e.g. van der Werf et al., 2010) show an increase from August to December. This backward shift is also visible for HCN (Fig. 12), but to a lesser extent.

For Southeast Asia and equatorial Asia (zones SEA and EQA), the low correlation coefficients (cf. Table 3) can be attributed to the difficulty of locating precisely with the model the intercontinental convergence zone (ITCZ) which drives the long-range transport of C$_2$H$_2$ and HCN-loaded plumes into the zone. Moreover, for Southeast Asia, the very low correlation coefficient for C$_2$H$_2$ ($\approx -0.31$) may be caused by (i) the model setting the high abundance peaks in DJF and MAM (Fig. 9, especially visible over China coast, and Fig. 11), which may be due to an overestimation of the Asian anthropogenic emissions, and (ii) IASI observations exhibiting July–September peaks, which are likely due to the combination of the pollution uplift and confinement within the AMA with the higher sensitivity of the method in the upper troposphere. Additionally, for equatorial Asia, the too low fire emissions considered in the model for Indonesia from July to December 2009 may also be a cause for these low correlation coefficients.

For HCN in northern Africa (zone NAF), correlation coefficient is very low ($R = 0.07$) because the model sets the abundance peaks around August while observations show peaks occurring around December, which is in accordance with previous studies (van der Werf et al., 2010). This inadequate timing for HCN in the model simulations could be due to an overestimation of the southern African contribution to the northern African loading and is visible on Fig. 10 (JJA).

4 Conclusions

We have presented a fast method to retrieve HCN and C$_2$H$_2$ total columns from IASI spectra. The sensitivity of this method to the two species is mostly in the mid-upper troposphere. With this method, C$_2$H$_2$ total columns can be retrieved globally with 5 % precision, while HCN abundances can be retrieved for abundances greater than $0.28 \times 10^{16}$ molec cm$^{-2}$ with 10 % precision in the $\pm 20^\circ$ latitudinal band and with 25 % precision in the $[\pm 35^\circ : \pm 20^\circ]$ latitudinal band.

Total columns have been retrieved globally for a 3-year period and compared to routine FTIR measurements available at Reunion Island (HCN and C$_2$H$_2$), Wollongong (HCN), Jungfraujoch (C$_2$H$_2$), and Izaña (HCN). The comparison between IASI and FTIR retrieved total columns demonstrates the capabilities of IASI to capture the seasonality in HCN and C$_2$H$_2$ in most cases.

Global seasonal distributions, as well as regional time series of the total columns, have been shown for the two species. IASI is able to capture persistent, seasonal and exceptional features for both species, and the observed patterns are in a general good agreement with previous spaceborne studies (ACE-FTIR and MIPAS).

The comparison between MOZART-4 simulations and the ground-based FTIR measurements, together with the comparison between the IASI observations and MOZART-4 simulations leads to the following conclusions: (i) the model is able to capture most of the hot spots and seasonal cycles, but not the exceptional events; (ii) the anthropogenic (biomass burning) emissions used in the model seem to be overestimated (underestimated) for both species; (iii) discrepancies between simulations and observations are most probably due to the coarse grid of the model producing too much diffusion and problems in the transportation scheme for fine-scale plumes, as well as to the fire injection set at the surface.

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