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Extending the satellite data record of tropospheric ozone profiles from Aura-TES to MetOp-IASI: characterisation of optimal estimation retrievals

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Abstract. We apply the Tropospheric Emission Spectrometer (TES) ozone retrieval algorithm to Infrared Atmospheric Sounding Instrument (IASI) radiances and characterise the uncertainties and information content of the retrieved ozone profiles. This study focuses on mid-latitudes for the year 2008. We validate our results by comparing the IASI ozone profiles to ozone sondes. In the sonde comparisons, we find a negative bias (1–10 %) in the IASI profiles in the lower to mid-troposphere and a positive bias (up to 14 %) in the upper troposphere/lower stratosphere (UTLS) region. For the described cases, the degrees of freedom for signal are on average 3.2, 0.3, 0.8, and 0.9 for the columns 0 km – top of atmosphere, (0–6), (0–11), and (8–16) km, respectively. We find that our biases with respect to sondes and our degrees of freedom for signal for ozone are comparable to previously published results from other IASI ozone algorithms. In addition to evaluating biases, we validate the retrieval errors by comparing predicted errors to the sample covariance matrix of the IASI observations themselves. For the predicted versus empirical error comparison, we find that these errors are consistent and that the measurement noise and the interference of temperature and water vapour on the retrieval together mostly explain the empirically derived random errors. In general, the precision of the IASI ozone profiles is better than 20 %.

1 Introduction

Ozone acts as a toxic pollutant in the lower troposphere, a greenhouse gas in the upper troposphere and a protective shield against harmful ultraviolet (UV) radiation in the stratosphere. Information on the vertical distribution of ozone is therefore critical to understanding its impact on air quality, chemical composition and climate. Rapidly increasing Asian emissions of ozone precursors, land surface changes from burning and decreasing surface emissions in Europe and North America are resulting in ongoing changes to the spatial distribution of tropospheric ozone, which have yet to be well quantified (e.g. Wild and Akimoto, 2001; Fry et al., 2012). Satellite-borne measurements provide the means for global and continuous monitoring of this important trace gas.

Spectrally resolved measurements in the submillimetre, thermal infrared (TIR, emitted from the earth’s surface and the atmosphere itself) and UV regions can provide ozone profile information. However, the strong pressure and temperature dependence of the spectral lines in the TIR ozone band (around 9.6 μm) makes this region particularly sensitive to the vertical distribution of ozone. Recent studies have demonstrated the value of combining information in different spectral regions (e.g. Natraj et al., 2011; Cuesta et al., 2013;
Fu et al., 2013). However, in this work, we focus on ozone retrievals from the TIR.

Tropospheric Emission Spectrometer (TES), a Fourier transform spectrometer (FTS) on the Aura satellite, flying since 2004, was specifically designed with a focus on mapping the global distribution of tropospheric ozone (Beer, 2006). The extremely high spectral resolution (0.1 cm\(^{-1}\)) of the TES instrument enables profiling of tropospheric ozone. Aura-TES is in a sun-synchronous polar orbit with an equator overpass local time of \(\sim 01:45\) and 13:45. It makes nadir observations with a spatial resolution of 5.3 km. Successive orbit tracks are separated by about 22\(^\circ\) longitude. In the nominal TES observation mode, observations are separated by 182 km along the flight track. The TES retrieval algorithm uses an optimal estimation approach (Bowman et al., 2006), which allows straightforward characterisation of the retrieval errors and the vertical sensitivity of the retrievals. The ozone product has been subject to ongoing improvements over the years. A number of validation studies have demonstrated the quality of the TES radiances (Shephard et al., 2008; Connor et al., 2011), ozone profile retrievals (Worden et al., 2007; Nassar et al., 2008; Verstraeten et al., 2013) and ozone retrieval error estimates (Boxe et al., 2010). The Boxe et al. (2010) study utilised special observations from TES (“stare” mode) in order to obtain multiple observations of the same air mass, from which a covariance matrix could be constructed.

The capability to retrieve ozone profile information has also been demonstrated using radiance measurements from other TIR nadir sounders, such as the Interferometric Measurement of Greenhouse gases (IMG, e.g. Coheur et al., 2005) on ADEOS, the Atmospheric InfraRed Sounder (AIRS, e.g. Bian et al., 2007) on Aqua, the Infrared Atmospheric Sounding Instruments (IASI, e.g. Dufour et al., 2012) on the MetOp-A and -B satellites, the TANSO-FTS (Ohyama et al., 2012) on GOSAT, and the Cross-track Infrared Sounder (CrIS) on Suomi NPP (Han et al., 2013). Combining data sets with well-characterised sensitivity and uncertainty estimates can potentially provide the means to generate a consistent long-term record of tropospheric ozone that could be used in chemical composition and climate applications.

In addition to the IASI instruments on MetOp-A and -B, an identical instrument will fly on the MetOp-C satellite, due to launch in 2017/18. Furthermore, EUMETSAT (European Organisation for the Exploitation of Meteorological Satellites) is currently preparing the next polar-orbiting programme with the Metop Second Generation (SG) satellite series that should be launched around 2020. In this framework, studies are underway for the concept of a new instrument, the IASI-New Generation (IASI-NG), characterised by an improvement of both spectral resolution and radiometric characteristics as compared to IASI (Clerbaux and Crevoisier, 2013). The IASI/IASI-NG data record is therefore of particular interest in the context of quantifying long-term changes in tropospheric ozone. In addition, the IASI instruments, unlike TES, provide swath coverage, allowing near-coincident observations of the atmospheric state and enabling comparison between predicted and actual retrieval errors.

In this work, we apply the TES ozone retrieval algorithm to radiances from IASI on MetOp-A as a first step towards the goal of creating a consistent long-term record of tropospheric ozone from multiple TIR instruments (while this study is concerned with TES and IASI, instruments such as AIRS and CrIS also have potential to contribute information to such a record). This study concentrates on mid-latitudes in 2008 in order to facilitate comparison of our results with other IASI ozone retrievals as presented in the study by Dufour et al. (2012). The focus of this study is the characterisation and validation of the error estimates on the ozone retrievals produced with this algorithm.

Section 2 gives the technical details of the IASI instrument. In Sect. 3, a summary of existing IASI ozone retrievals is given, while Sect. 4 describes the retrieval of ozone profiles with optimal estimation theory and the theoretical tools for the characterisation of the profiles used in this study. In Sect. 5, an overview of the different reference data sets for validation is provided, followed by the results in Sect. 6.

2 The IASI instrument on MetOp-A

The IASI instrument is an FTS based on the Michelson interferometer. 8461 channels cover a spectral range between 645 and 2760 cm\(^{-1}\) with a resolution of 0.5 cm\(^{-1}\) (after apodisation). The spectral sampling interval is 0.25 cm\(^{-1}\). IASI was designed by the Centre National d’Études Spatiales (CNES) (Cayla, 1993; Blumstein et al., 2004) and launched in October 2006 onboard the MetOp-A satellite. The mission is operated by EUMETSAT. Operational measurements have been performed since June 2007.

IASI flies on MetOp-A at an altitude of 817 km in a polar sun-synchronous orbit. The local overpass times at the equator are 09:30 and 21:30. MetOp-A completes slightly over 14 orbits a day. IASI is a nadir-viewing instrument and scans across the track within ±48.3\(^\circ\) in a step and stare mode. There are 30 scans per swath. These 30 individual effective fields of view (EFOVs) are made up of four instantaneous fields of view (IFOVs) arranged in a \(2 \times 2\) pixel matrix resulting in 120 measurements per scan line. The surface footprint of a nadir IFOV is circular, of 12 km diameter. Towards the edge of the swaths, the IFOVs are elliptically elongated to a footprint of 20 km \(\times\) 39 km. A swath spans 2200 km in width. Global coverage is achieved twice daily.

IASI data are processed operationally by EUMETSAT. We obtained IASI Level 1c radiances (geolocated, calibrated, and apodised radiance spectra) via the National Oceanographic and Atmospheric Administration (NOAA) Comprehensive Large Array-data Stewardship System (CLASS). EUMETSAT produces operational retrievals of column in-
tegrated ozone, carbon monoxide, nitrous oxide, methane, carbon dioxide, three partial columns of ozone (between the surface and 6, 12, and 16 km altitude), surface temperature, cloud information, and profiles for temperature and water vapour. More technical details can be found in the EUMETSAT IASI Level 2 Product Guide (http://oisswww.eumetsat.org/WEBOPS/eps-pg/IASI-L2/IASIL2-PG-0TOC.htm) and in Clerbaux et al. (2009). In this work, we have utilised EUMETSAT temperature, water vapour and surface temperature products in the retrieval input (Sect. 4.2). In addition, we have made use of the cloud fraction that is reported in the Level 2 product (Sect. 4.3).

3 Existing IASI ozone data products

A number of different groups have previously developed ozone retrieval algorithms for IASI. Operational processing of IASI ozone data for the Level 2 product is done by EUMETSAT with a neural network based approach (Schlüssel et al., 2005; August et al., 2012) and by the Center for Satellite Applications and Research (STAR), part of the NOAA Satellite and Information Service (NESDIS), with an iterative regularised least squares minimisation algorithm (Bian et al., 2007; Pittman et al., 2009). At LATMOS/ULB (Laboratoire Atmosphères, Milieux, Observations Spatiales/Université Libre de Bruxelles), the FORLI (Fast Operational Retrievals on Layers for IASI) algorithm uses an optimal estimation approach (Hurtmans et al., 2012). In the future FORLI ozone profiles will be distributed by EUMETSAT with the Level 2 product. We compare our retrievals to FORLI results (see Sects. 5.3 and 6.5).

Other non-operational ozone data products have been derived at LISA (Laboratoire Inter-universitaire des Systèmes Atmosphériques, Universités Paris-Est Créteil et Paris Diderot, CNRS/INSU) applying an altitude-dependent Tikhonov-Phillips regularisation (Eremenko et al., 2008; Dufour et al., 2010) and at LA (Laboratoire d’Aérologie/OMP) utilising an optimal estimation approach using a priori constraints (Barret et al., 2011).

Validation with and comparison to independent measurements of other IASI products have been performed (Boynard et al., 2009; Keim et al., 2009; Pittman et al., 2009; Viatte et al., 2011; Dufour et al., 2012; Pommier et al., 2012; Scanell et al., 2012; Gazeaux et al., 2013) and also comparison to models (Parrington et al., 2012). In general, these studies focus on the determination of the bias and the correlation with respect to independent measurements and on the determination of the information content. Only Keim et al. (2009) and Dufour et al. (2012) also studied the precision of the retrievals.

4 TOE – TES optimal estimation

We apply an optimal estimation (OE) approach (Rodgers, 2000) following the TES algorithm to IASI Level 1c radiances. The optimal estimation approach minimises the cost function:

\[ C = \| y - L(x, b) \|^2_{S_e^{-1}} + \| z_{\text{apriori}} - z \|^2_{S_a^{-1}} \]  

(1)

in a non-linear Levenberg–Marquardt iterative scheme (Bowman et al., 2006). Here, \( y \) is the measured radiance, a discrete vector, related to the true state \( L^{\text{true}} \) by an additive noise model \( \varepsilon \):

\[ y = L^{\text{true}} + \varepsilon. \]  

(2)

\( L \) can also be interpreted as an operator describing the radiative transfer dependent on the atmospheric state. Hence \( L(x, b) \) is the forward model of a specific state vector \( x \) and parameters \( b \) held constant in the retrieval. The second term of Eq. (1) describes the difference between the a priori profile \( z_{\text{apriori}} \) and the retrieved state \( z \). The retrieval is constrained with the measurement noise covariance matrix \( S_e \) and the covariance matrices corresponding to the a priori profiles \( S_a \).

4.1 OE framework for the characterisation of the retrievals

Several diagnostics can be derived to describe the quality and uncertainty of the retrieved atmospheric state: the Jacobian matrix \( K \) is an output of the radiative transfer model and represents the sensitivity of the forward model towards changes in the retrieved state:

\[ K = \frac{\partial L(z)}{\partial z}. \]  

(3)

The gain matrix \( G \) describes the sensitivity of the retrieved state towards changes in the measured radiances and can be calculated from

\[ G = \left( K^T S_e^{-1} + S_a^{-1} \right)^{-1} K^T S_e^{-1}. \]  

(4)

The averaging kernel matrix \( A_{zz} \) can be calculated from the gain matrix and the Jacobian:

\[ A_{zz} = G K. \]  

(5)

The averaging kernels describe the sensitivity of the retrieval to the true state. Example averaging kernels for the IASI-TOE ozone retrieval are shown in Fig. 1. The trace of the averaging kernel matrix gives the degrees of freedom for signal (DOFS) of the retrieval.

Various factors contribute to the overall uncertainty of a retrieved ozone profile, i.e. the smoothing by the retrieval/instrument due to the limited information content of the measurement, the measurement noise of the instrument,
Figure 1. Averaging kernels and errors for selected locations. The angle in the individual panels refers to the viewing angle of the IASI instrument and DOF to the degrees of freedom for signal.

coupling or cross-correlation between simultaneously retrieved parameters, and the uncertainties associated with parameters that are not included in the retrieval state vector (see Sect. 4.3 for more details). Mathematically, the error covariance can be described as a sum of four terms:

\[
S_\tilde{z} = (A_{zz} - I) S_z (A_{zz} - I)^T + GS_z G^T + \sum G K_b S_b (G K_b)^T + \sum A_{xs} S_{b}^{ret} (A_{xs})^T + \text{res}
\]  

(6)

with \( I \) the identity matrix. Systematic errors originate from parameters \( b \), which are held constant in the retrieval, with \( K_b \) and \( S_b \) their respective Jacobians and error covariance matrices. The cross-state errors are caused by the simultaneous retrieval of other atmospheric parameters that are not the target. They can be derived from the cross-state part of the averaging kernel matrix \( A_{xs} \) and the corresponding error covariance matrix \( S_{b}^{ret} \). The residual term \( \text{res} \) includes all uncertainties not considered or unknown. The square root of the diagonal of Eq. (6) yields the uncertainty on the ozone profile.

4.2 Retrieval input

The TES forward model is described in Clough et al. (2006). Calculations of radiances and Jacobians are performed using a fixed pressure grid of 66 levels. The absorption parameters used in the forward model are pre-calculated using the Line-By-Line Radiative Transfer Model (LBLRTM) (Clough et al., 2005) and the TES_v1.4 line parameter database (http://rtweb.aer.com).

As mentioned above, the IASI Level 1c radiances are obtained from NOAA CLASS. The measurement noise covariance matrix \( S_\varepsilon \) is a diagonal matrix with the diagonal elements set according to the noise-equivalent spectral radiance (NESR). The NESR in the TIR ozone band around 9.6 \( \mu \)m has been estimated at 20 nW (cm\(^2\) sr cm\(^{-1}\))\(^{-1}\) from a set of representative spectra measured in orbit (Clerbaux et al., 2009).

Ozone profiles are retrieved simultaneously with water vapour profiles in the spectral windows 990–1031, 1040–1049, and 1069–1072 cm\(^{-1}\). Figure 2 shows these three spectral windows in relation to the TIR ozone band around 9.6 \( \mu \)m.

In the TOE retrievals, the initial guess profiles are the same as the a priori profiles and they are the same as those used for the TES V05 ozone retrievals. These profiles were generated by merging the climatological monthly mean tropospheric and lower stratospheric ozone field from a 1997–2004 simulation from the Model for Ozone And Related chemical Tracers, MOZART-4 (Emmons et al., 2010) with the climato-
logical monthly mean stratospheric and mesospheric ozone field from a 2005–2010 simulation from the Whole Atmosphere Chemistry–Climate Model (WACCM) model (Kinnison et al., 2007). The a priori profiles for ozone are binned by months and by region in 10° latitude and 60° longitude steps. The a priori water vapour profiles are taken from the EUMETSAT operational Level 2 product.

The retrieval for ozone is performed on 26 levels and for water vapour on 18 levels. Those retrieval grids are strongly linked with the constraints. The a priori constraint matrices are altitude-dependent Tikhonov constraints (Kulawik et al., 2006). In this technique, the constraint is optimised for a specified a priori covariance, for which MOZART-3 was used (Brasseur et al., 1998). The ozone constraint matrices are binned into five latitude bands and the water vapour constraint matrix is the same for all locations on the globe.

Atmospheric temperature profiles and skin temperature are not retrieved. These are set to the IASI EUMETSAT operational Level 2 values. Other parameters that are not retrieved include the CO₂ profiles created from a 2004 MATCH (Model of Atmospheric Transport and Chemistry) model run (Nevison et al., 2008) and scaled by 1.0055 per year, the surface emissivity over land from the Zhou et al. (2011) monthly climatology and the emissivity over water from Wu and Smith (1997). The surface pressure is taken from the operational Level 2 data.

4.3 Sources of error in the TOE retrievals and quality control

Many factors contribute to the overall uncertainty of the retrieved ozone profile. Uncertainties arise from parameters fixed to climatological values or auxiliary data sets. These include the temperature profile, trace gas profiles for interfering species, surface temperature, surface pressure, and emissivity climatology. In theory, the contribution of these errors to the overall uncertainty can be calculated with the systematic term of Eq. (6). Uncertainties associated with the atmospheric state are rather difficult to quantify since the true state and consequently the corresponding error covariance are unknown. The only other parameter retrieved besides the ozone profile is the water vapour profile, and we can calculate the error on the ozone profile caused by cross-correlation for this species with Eq. (6).

Further, possible contamination of an IASI scene with clouds can lead to errors if not considered in the retrieval. The approach taken to account for the influence of cloud in the TES products is described in Kulawik et al. (2006). However, in this study we have chosen to screen for clouds. IASI radiances were selected for cloud fractions smaller than 13 % from EUMETSAT’s Level 2 product (following Clerbaux et al., 2009) and then the retrievals were performed on individual IASI IFOVs assuming a cloud-free scene. This is different from the approach applied for the operational TES products where cloud parameters are retrieved and then included in the radiative transfer simulations for the trace gas retrievals. We might expect that the presence of thin clouds could have some influence on the result. For example, Wassmann et al. (2011) have shown that neglecting the presence of a cirrus cloud with an IR optical thickness of 0.1 and a cloud top at 10 km in the ozone retrievals can lead to errors in the ozone profile of up to 25 % in the troposphere.

An overview of the IASI instrument calibration is given in Hilton et al. (2012): the spectral calibration accuracy is $\delta \nu / \nu = 2 \times 10^{-6}$ where $\nu$ is the frequency. The absolute calibration in brightness temperature is better than 0.35 K and it was shown that AIRS and IASI agree within 0.2 K. The NESR for IASI was estimated from the measured radiances themselves. The NESR should be a good representation of the random errors although over time instrument issues could change this value.

In Eq. (1), the radiance term of the cost function is solely constrained by the measurement noise covariance (see Sect. 4.1). However, differences between modelled and measured radiances can also originate from uncertainties in the forward model with the covariance $S_q$. This adds an additional term of $G f G^T$ to Eq. (6). $S_q$ contains contributions from the discretisation and interpolation of atmospheric profiles, and spectroscopic errors. The vertical grid in the TES forward model has been chosen to be fine enough to make discretisation errors negligible, and the difference between radiances from the TES forward model and LBLRTM are less than 0.1 % (Clough et al., 2006). The dominant errors in the modelling of clear-sky radiances arise from uncertainties in the spectroscopic parameters used as input to the line-by-line calculations (see, for example, Alvarado et al., 2013).

In order to remove unphysical results, such as oscillations in the profile, we remove profiles with $\chi^2$ larger than 1.3, which is calculated by

$$\chi^2 = \frac{1}{n} \left( y - L (x, b) \right) S_e^{-1} \left( y - L (x, b) \right)^T,$$

with $n$ being the number of elements in the radiance vectors $y$ and $L(x, b)$. This is the only additional quality screening besides the cloud filtering.

5 Validation data and selection criteria

We have validated the IASI-TOE ozone retrievals against independent measurements from ozone sondes for mid-latitudes in 2008. In addition to quantifying the overall bias relative to the sondes, we have focused here on validating the error estimates obtained from the optimal estimation framework.

There are three key empirical methods helping to quantify the actual errors of a retrieval: (a) determining the bias of the retrieved profiles with respect to independent measurements, here ozone sondes, (b) determining the root-mean-square (rms) deviation of the IASI profiles with respect to the...
sonde profiles, and (c) determining the range of the deviation from a mean of an ensemble of quasi-coinciding retrieved profiles, i.e. calculating the so-called sample covariance matrix. In the sample covariance approach, we assume that the true atmospheric ozone field is relatively constant over some limited spatial domain. For the IASI-TOE retrieval, which is primarily sensitive to ozone in the free troposphere (see Fig. 3), we assume that this spatial domain can extend over up to 10 IFOVs or about 220 km.

5.1 Ozone sondes

Ozone sonde profiles have been obtained from the World Ozone and Ultraviolet Data Centre (WOUDC, www.woudc.org) and from the Global Monitoring Division of NOAA (www.esrl.noaa.gov/gmd). The theoretical uncertainty of electrochemical concentration cell (ECC) sondes for a typical mid-latitude ozone profile peaks at the ozone minimum (about 13 km) at about 9% (WMO, 2011). However, the absolute uncertainty depends on various factors, e.g. the concentration of the potassium iodide solution and the sonde preparation. Through laboratory studies for precision of the sondes as well as comparisons to UV-photometer ozone measurements, the accuracy is estimated to be 10% or better for ECC sondes and 13% or better for Brewer-Mast type sondes (only at Hohenpeissenberg) for the altitude range considered in the IASI-TOE comparison (WMO, 2011). In some cases, the WOUDC distributes a correction factor together with the sonde profiles. This correction factor is calculated from the ratio of a Brewer/Dobson instrument coincident total ozone measurement to the integrated ozone profile from the sonde plus a climatology above the burst height of the sounding balloon. Three different scenarios for soundings used in this study were encountered: (a) a correction factor was not given, (b) a correction factor was stated and applied to the sonde profile, and (c) a correction factor was stated, but not applied. For cases a and b we used the ozone profiles as given, but for case c we applied the factor to consolidate the measurements. Sondes with correction factors larger than 15% were excluded from this study.

IASI scenes were selected to be within ±7 h and within a circle of 110 km (equivalent to 1° latitude) radius around the sounding site following Dufour et al. (2012). These sites are listed in Table 1 together with the geolocation, elevation above sea level (a.s.l.) and the numbers of the soundings and of the IASI scenes.

In order to estimate the IASI-TOE retrieval’s limited vertical sensitivity, the averaging kernel matrix \( A_{zz} \) together with the a priori profile \( z_{apriori} \) is applied to the sonde profile \( z_{sonde} \) (Rodgers and Connor, 2003) to obtain a new profile \( \hat{z} \) mimicking the IASI measurement:

\[
\hat{z} = z_{apriori} + A_{zz} (z_{sonde} - z_{apriori}).
\] (8)

We determine the mean bias of the IASI-TOE ozone profiles \( z_{i,TOE} \) with respect to the ozone sonde profiles \( z_{i,sonde} \):

\[
\bar{z}_{sondebias} = \frac{1}{N} \sum_{i} \left( z_{i,TOE} - z_{i,sonde} \right) \tag{9}
\]

Note that \( z_{i,sonde} \) is not always unique if there is more than one IASI scene fulfilling the coincidence criteria for the same sonde. The mean bias is a measure of the absolute accuracy of the retrievals, while the corresponding standard deviation \( \sigma \) is a measure of the precision of the retrievals. Note that the fact that the sonde and the satellite are not viewing exactly the same air mass could make the standard deviation slightly larger than the true precision. The corrected sample variance \( \sigma^2, \text{corrected} \) referring to the degrees of freedom of \((N - 1)\), is

\[
\sigma^2 = \frac{1}{N - 1} \sum_{i} \left( z_{i,sondebias} - \bar{z}_{sondebias} \right)^2. \tag{10}
\]

The rms deviation (\( RMSD \)) can be calculated from

\[
RMSD = \sqrt{\frac{1}{N} \sum_{i} \left( z_{i,TOE} - z_{i,sonde} \right)^2} \tag{11}
\]

and hence can be understood as a measure for the mean bias combined with the precision: \( RMSD^2 = \text{precision}^2 + \text{bias}^2 \).
### 5.2 Sample covariance matrix

The sample covariance matrix provides an estimate of the error covariance matrix: we have \( n \) observations \( z_{1,n} \) of the random vector \( \mathbf{Z} \). Then the sample covariance matrix is given by

\[
\mathbf{Q} = \frac{1}{n-1} \sum_{i=1}^{n} (\mathbf{z}_i - \bar{\mathbf{z}}) (\mathbf{z}_i - \bar{\mathbf{z}})^T,
\]

where \( \bar{\mathbf{z}} \) is the sample mean. The degrees of freedom used here are \((n-1)\) because \( \bar{\mathbf{z}} \) does not equal \( \mathbf{Z} \). Here, the measured quantities \( z_{1,n} \), i.e. the retrieved ozone profiles, are not random with respect to the smoothing error because they all depend on the same a priori profile. Also other systematic errors cancel out since only relative differences are calculated. Consequently, the square root of the diagonal of \( \mathbf{Q} \) is a measure for the precision only. Note that Eq. (12) is equivalent to the variance of Eq. (10) if only one ozone sonde profile for the bias determination is used.

It is assumed that several retrieved ozone profiles of adjacent and concurrent IASI scenes are representative sample measurements of the true ozone profile. This assumption is valid for ozone in the free troposphere and lower stratosphere, the region of the atmosphere IASI is mostly sensitive to (see Fig. 3), and for a sufficiently large number of the sample size \( n \). The same co-location criterion as above is applied, i.e. measurements within a circle with a radius of 110 km. In the best case, this results in about 64 scenes that could fulfil the co-location criterion. In addition to this, only IASI-TOE profiles from the same swath are used, guaranteeing the measurements to be within less than 38 s of each other. Here, the cloud fraction was limited to smaller than 6 %.

### 5.3 FORLI

A detailed description of the FORLI algorithm can be found in Hurtmans et al. (2012). Briefly, ozone profiles are retrieved on a constant height grid with 1 km thick layers from the surface to 40 km altitude. The absorbance cross-sections are pre-calculated from the HITRAN database (Rothman et al., 2005). The spectral range is 1025–1075 cm\(^{-1}\). Only one global ozone a priori profile and corresponding covariance matrix are used for all seasons. Just like for TOE, FORLI utilises the operational Level 2 temperatures and the water vapour profiles as input as well as the emissivity climatology by Zhou et al. (2011). Also, TOE and FORLI use the same value for the diagonal of the measurement noise covariance matrix, i.e. 20 nW (cm\(^{-2}\) sr cm\(^{-1}\))\(^{-1}\) (Clerbaux et al., 2009).

IASI-FORLI ozone profiles were selected around the ozone sounding location with the same selection criteria as for IASI-TOE (although actual ozone sonde measurements were not included in this comparison). In general, since the retrieved ozone profiles undergo different quality screening, TOE and FORLI processing result in different subsets of successful profile retrievals for the same IASI scenes. For the comparison, we use a common subset. Since the two retrievals use different a priors profiles, the IASI-TOE a priori \( z_{\text{apriori}}(\text{TOE}) \) are swapped out with the IASI-FORLI a priori \( z_{\text{apriori}}(\text{FORLI}) \) (Rodgers and Connor, 2003):

\[
\begin{align*}
\mathbf{z}_{\text{TOE}} & \quad \Rightarrow \quad \mathbf{z}_{\text{FORLI}} \\
\mathbf{z}_{\text{TOE}} + (\mathbf{A}_{zz} - \mathbf{I}) (\mathbf{z}_{\text{apriori}}(\text{TOE}) - \mathbf{z}_{\text{apriori}}(\text{FORLI}))
\end{align*}
\]

\( \mathbf{A}_{zz} \) is the error covariance matrix, \( \mathbf{I} \) the identity matrix. This way, the bias can be determined.

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### Table 1. Locations with geoinformation and numbers of ozone sondes and IASI-TOE profiles, and mean DOFS.

<table>
<thead>
<tr>
<th>Station (elevation above sea level)</th>
<th>Latitude</th>
<th>Longitude</th>
<th># of TOE scenes</th>
<th># of sondes</th>
<th>DOFS</th>
</tr>
</thead>
<tbody>
<tr>
<td>Goose Bay (36 m)</td>
<td>53.3° N</td>
<td>60.4° W</td>
<td>363</td>
<td>30</td>
<td>2.84</td>
</tr>
<tr>
<td>Legionowo (96 m)</td>
<td>52.4° N</td>
<td>21.0° E</td>
<td>158</td>
<td>19</td>
<td>3.31</td>
</tr>
<tr>
<td>Lindenberg (112 m)</td>
<td>52.2° N</td>
<td>14.1° E</td>
<td>238</td>
<td>29</td>
<td>3.24</td>
</tr>
<tr>
<td>Valentia Island (14 m)</td>
<td>51.9° N</td>
<td>10.3° W</td>
<td>367</td>
<td>32</td>
<td>3.15</td>
</tr>
<tr>
<td>Bratt’s Lake (580 m)</td>
<td>50.2° N</td>
<td>104.7° W</td>
<td>524</td>
<td>41</td>
<td>3.09</td>
</tr>
<tr>
<td>Prague (304 m)</td>
<td>50.0° N</td>
<td>14.4° E</td>
<td>191</td>
<td>27</td>
<td>3.03</td>
</tr>
<tr>
<td>Kelowna (456 m)</td>
<td>49.9° N</td>
<td>119.4° W</td>
<td>1240</td>
<td>62</td>
<td>3.10</td>
</tr>
<tr>
<td>Hohenpeißenberg (976 m)</td>
<td>47.8° N</td>
<td>11.0° E</td>
<td>688</td>
<td>80</td>
<td>3.17</td>
</tr>
<tr>
<td>Payerne (491 m)</td>
<td>46.5° N</td>
<td>6.6° E</td>
<td>851</td>
<td>93</td>
<td>3.20</td>
</tr>
<tr>
<td>Trinidad Head (20 m)</td>
<td>40.8° N</td>
<td>124.2° W</td>
<td>753</td>
<td>60</td>
<td>3.27</td>
</tr>
<tr>
<td>Madrid (631 m)</td>
<td>40.5° N</td>
<td>3.6° W</td>
<td>261</td>
<td>34</td>
<td>3.43</td>
</tr>
<tr>
<td>Boulder (1743 m)</td>
<td>40.0° N</td>
<td>105.3° W</td>
<td>230</td>
<td>38</td>
<td>3.32</td>
</tr>
<tr>
<td>Ankara (891 m)</td>
<td>40.0° N</td>
<td>32.9° E</td>
<td>214</td>
<td>19</td>
<td>3.21</td>
</tr>
<tr>
<td>Wallops Island (13 m)</td>
<td>37.9° N</td>
<td>75.5° W</td>
<td>513</td>
<td>36</td>
<td>3.38</td>
</tr>
<tr>
<td>Macquarie Island (6 m)</td>
<td>54.5° S</td>
<td>138.9° E</td>
<td>204</td>
<td>20</td>
<td>2.74</td>
</tr>
<tr>
<td>Ushuaia (17 m)</td>
<td>54.9° S</td>
<td>68.3° W</td>
<td>166</td>
<td>19</td>
<td>2.76</td>
</tr>
</tbody>
</table>

All – – 6961 639 3.16
6 Results

Table 1 gives an overview of the number of IASI scenes that are used in this study. Overall, there are 6961 TOE retrievals that passed the various selection criteria (see Sects. 4.3 and 5.1) corresponding to 639 ozone soundings at 16 locations, two of which are in the Southern Hemisphere.

6.1 Vertical sensitivity of the IASI-TOE ozone retrievals

Figure 3 shows the vertical distribution of the sum of the averaging kernel matrix rows split by season. Those are the means at the individual pressure levels and the corresponding 1σ standard deviation. Only between 300 and 700 hPa the values differ slightly with the season. Highest values are observed for summer (JJA) and lowest for winter (DJF). Also shown are the statistics for the DOFS for the different seasons for selected partial columns, and those follow the same seasonal trend (see Fig. 4). The overall mean DOFS for the total column are 3.2 (see Table 1). The DOFS for the partial columns in the troposphere (0–11 km) and in the upper troposphere/lower stratosphere (UTLS; 8–16 km) are slightly lower than 1 in summer. The whiskers in Fig. 4 represent the maximum and minimum values, respectively. Interestingly, the largest maximum values can be found for the winter partial columns (0–6, 0–11, 8–16 km) adding more than half a DOFS to the tropospheric partial columns in the extreme case. This indicates that not only the thermal contrast determines the DOFS, but also the low humidity in winter plays a role in the information content of the retrieval.

6.2 Theoretical error estimates

Figure 1 shows examples of IASI-TOE theoretical ozone retrieval errors and averaging kernels. These are examples for single ozone retrievals and not averages. The individual errors are calculated with the different terms of Eq. (6), smoothing and noise as labelled and water with the cross-state term. The temperature error in Fig. 1 is calculated somewhat differently: we estimate how the temperature error propagates into the ozone profile with a temperature error covariance matrix derived from an ensemble of EUMETSAT Level 2 temperature profiles of quasi-coincident IASI scenes for these individual cases (Eq. 6), systematic term). See also Sects. 5.2 and 6.4. The sample size used for the calculation of the covariance is given in the label as \( n \). The uncertainties are dominated by the smoothing error and peak at about 30% in the lowest retrieval level. This reflects the limited sensitivity of the measurements as can also be seen in the averaging kernels. In the UTLS, the overall error is between 10 and 12%. In the case of Bratt’s Lake, the averaging kernels do not decrease as rapidly towards the surface as in the other three examples due to a stronger thermal contrast (14 K vs. 4, 1, and 3 K for Valentia Island, Legionowo, and Wallops Island, respectively). Consequently, although the uncertainty at the surface is the same as in the other cases, the uncertainty decreases more rapidly with height. This results in more information on the tropospheric profile in this case, as indicated by the rather high number of 3.7 DOFS.

6.3 Biases relative to ozone sondes

The bias of IASI-TOE with respect to the ozone sonde profiles is shown in Fig. 5 for the 16 individual locations as listed in Table 1. A clear picture emerges for all locations: a positive bias can be found in the UTLS region. Towards the surface the bias vanishes because of the limited sensitivity of IASI there. From this figure it can be concluded that the biases and standard deviations of the TOE retrievals with respect to the sondes are similar for different locations, even though different kinds of ozone sondes, as well as different a priori profiles/constraints, were applied in the retrievals. Consequently in the following, the bulk quantities for all locations together can be analysed.

Figure 6, left panel shows the bias in TOE ozone retrievals with respect to the sonde with the averaging kernels and a priori being applied to the sonde profile (see Eq. 8). The mean bias is largest between 200 and 70 hPa. The maximum mean bias is 14%. This positive mean bias in this region is consistent with previously reported values for IASI (e.g. Dupont et al., 2012) and for TES (e.g. Verstraeten et al., 2013). Since this bias is platform independent, it seems likely that it originates from uncertainties in spectroscopic parameters. The right panel of Fig. 6 shows the bias with respect to the
Figure 5. Bias of the IASI-TOE profiles with respect to ozone sonde profiles separated by location. The numbers of profiles included are indicated in the legend. The averaging kernels and ozone a priori profiles have been applied to the ozone sonde profiles. The coloured lines show the individual biases, the black solid lines the means, and the dashed lines the 1σ standard deviations of the averaging. No significant differences in the individual biases or standard deviations can be observed for the different locations.

raw ozone sonde data. As expected, the standard deviation is larger in this case since the variability of the ozone sonde measurements will be reduced due to the application of the averaging kernels (see also Fig. 7). In Fig. 7, the standard deviation is plotted together with the rms deviation (see Eq. 11). The rms deviation is a measure for the precision and the bias error; the standard deviation is a measure for the precision only. As mentioned above, the bias error can be calculated from these quantities. When looking at the absolute numbers in Fig. 7, the following two points have to be kept in mind: (a) the estimates also include the precision of the ozone sonde (~5 % for ECC sondes; WMO, 2011) and (b) the strength of the constraint of the retrieval is reflected in the precision, i.e. a strong constraint will pull the retrieved profile towards the a priori profile, resulting in less scatter in the results.

6.4 Empirical vs. theoretical random errors

Examples for the comparison between empirical and theoretical errors are shown in Fig. 8. The theoretical errors are equivalent to the ones in Fig. 1, described in Sect. 6.2 but without the smoothing error. The ozone profiles themselves are not used in this comparison. Here, we assume that for a concurrent group of IASI IFOVs associated with a given site at a given time, the instrument is effectively viewing the same atmosphere. The comparison is between the observed
variation in the retrieved ozone within the concurrent set of IFOVs (the empirical error) and the theoretical error calculated using Eq. (6). Twelve of the 16 different sites had 15 or more concurrent ozone profiles and one example per site is shown in Fig. 8. The black lines are the square root of the diagonal of the sample covariance matrix calculated with Eq. (12) hereafter referred to as empirical error. The magnitude of the empirical error is similar to the theoretical error (brown line). In general, (although not in all cases) the empirical errors are larger than the theoretical errors. Absolute differences are less than 11 %. The profile shape of the empirical errors and the theoretical errors are quite variable in the examples. In some cases they look similar (e.g. Macquarie Island) in other cases they do not (e.g. Legionowo). For Legionowo, only 15 scenes were available of the theoretically possible more than 60 (see Sect. 5.2). Hence in this case, the majority of the IASI scenes did not pass the quality screening (see Sect. 4.3). We investigated the distribution of
The individual IASI scenes within the coincidence area, but no outliers were detected. We also tested the standard deviation of the individual theoretical errors for all the shown cases and it is smaller than 1 \% for water vapour, smaller than 2 \% for noise, and smaller than 2.5 \% for the smoothing error. Note that the remaining differences could reflect actual atmospheric variability or a too small sample size to obtain a representative sample covariance matrix.

For the presented cases, the empirical errors usually peak in the lower troposphere and the profile shape at sites with a large sample size (e.g. Kelowna and Ankara) is very similar to the profile shape of the mean standard deviation with respect to the ozone sondes as presented in Fig. 7. The empirical errors are smaller than 18 \% and can be as low as 6 \% in the UTLS region. These values also compare to the standard deviation of Fig. 7. Overall, it can be concluded that the observed variability is consistent with theoretical error budgets.

### 6.5 Comparison with IASI-FORLI

The comparisons between IASI-FORLI and IASI-TOE ozone profiles are shown in Fig. 9. The two retrievals agree within the 1σ-level when the IASI-TOE a priori profile has been modified to be the same as the IASI-FORLI a priori profile using Eq. (13) (Fig. 9, left panel). The particular shape of the bias between TOE and FORLI can be explained in terms of the biases of the respective retrieval profiles in comparison to ozone sondes. The FORLI positive bias towards ozone sondes (see Dufour et al., 2012) is positioned slightly lower than TOE’s. The differences between the two retrievals can be attributed to the differences in the ozone covariance matrices: FORLI uses a weaker constraint, but a larger correlation length for the retrieval levels in comparison to TOE/TES. In terms of the raw differences, the TOE retrieval results in larger ozone concentrations than FORLI at most levels (Fig. 9, right panel).

### 7 Summary of results and discussion

In this section the main findings are summarised and put into context.

- On average, the overall DOFS are 3.2 and the DOFS for the partial columns are 0.3, 0.8, and 0.9 for (0–6), (0–11), and (8–16) km, respectively. These numbers are similar to the numbers reported in Dufour et al. (2012) for three different IASI ozone retrievals, which include the FORLI retrieval.

- In general, the lower atmospheric DOFS as well as the overall DOFS are higher for warmer seasons (see Fig. 4).

- The theoretical error of the ozone profile is dominated by the smoothing error. The water vapour error plays only a minor role. The temperature profile error values and vertical distributions are quite variable and uncertainties can be as low as 3 \%, but up to 20 \% in some cases (see Figs. 1, 8). Overall, the error is about 30 \% at the surface and decreases to about 10 \% in the UTLS.

- The empirical random errors are broadly consistent with the theoretical random errors considering that some actual variations of the ozone can occur over the ensemble of IASI scenes. The precision of the IASI-TOE retrievals is better than 20 \% at all levels. Since the random errors are mainly made up by the noise and the temperature profile errors (Figs. 1, 8), the precision can possibly improve if the temperature profile is retrieved from the measured IASI radiances in a step prior to the ozone retrievals as is done for the operational TES retrievals. The magnitude of the theoretical random errors for IASI-TOE ozone retrievals is broadly consistent.
with theoretical errors for TES ozone retrievals (Boxe et al., 2010).

– The IASI-TOE ozone profiles show a positive bias in comparison to the ozone sondes profiles. The bias is greatest between 200 and 70 hPa when the averaging kernels are applied to the sondes profiles (Fig. 6, left panel). The maximum bias is 14%. A positive bias has been observed before for IASI ozone profiles of three different retrievals (Dufour et al., 2012). This positive bias is found for TES as well (e.g. Verstraeten et al., 2013), but also TANSO-FTS (Ohyama et al., 2012) and consequently the bias might be caused by incorrect spectroscopic parameters since it is independent from the observing system.

– FORLI and TOE ozone profiles are consistent with each other when the a priori profiles have been consolidated for the two retrievals (Fig. 9, left panel).

– When looking at the raw differences, the TOE retrieval results in larger ozone concentrations than FORLI at most levels (Fig. 9, right panel) as a result of the choice of a priori ozone values for FORLI versus TOE.

It can be concluded that the IASI-TOE profile errors are consistent with other retrievals for IASI and TES and that the IASI-TOE ozone profiles are consistent with the IASI-FORLI profiles. Dedicated comparisons between IASI-TOE and TES ozone will be the topic of future work.

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