First retrieval of tropospheric aerosol profiles using MAX-DOAS and comparison with lidar and sky radiometer measurements

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Abstract. Ground-based Multi-Axis Differential Optical Absorption Spectroscopy (MAX-DOAS) measurements were performed at Tsukuba, Japan (36.1° N, 140.1° E), in November–December 2006. By analyzing the measured spectra of scattered sunlight with DOAS and optimal estimation methods, we first retrieve the aerosol optical depth (τ) and the vertical profile of the aerosol extinction coefficient (k) at 476 nm in the lower troposphere. These retrieved quantities are characterized through comparisons with coincident lidar and sky radiometer measurements. The retrieved k values for layers of 0–1 and 1–2 km agree with lidar data to within 30% and 60%, respectively, for most cases, including partly cloudy conditions. Results similar to k at 0–1 km are obtained for the retrieved τ values, demonstrating that MAX-DOAS provides a new, unique aerosol dataset in the lower troposphere.

1 Introduction

Atmospheric aerosols reflect and absorb solar radiation and modify cloud properties, altering the Earth’s radiation balance and hence climate. A major uncertainty in the prediction of future climate change arises from the difficulty of modeling the effects of atmospheric aerosols. Because aerosols impact both regional and global energy budgets, owing to their highly non-uniform spatial and temporal distributions in the troposphere, a multiple-measurement approach is highly desirable to assess their impacts on the Earth’s system (Kaufman et al., 2002).

Recently, the ground-based Multi-Axis Differential Optical Absorption Spectroscopy (MAX-DOAS) technique has been proposed as a new technique to measure aerosol optical properties in the troposphere (Wagner et al., 2004; Frieß et al., 2006). The measurements can be performed with a relatively simple setup and very low power consumption. The optical properties that are potentially measurable by MAX-DOAS include the aerosol optical depth (τ) as well as the vertical profile of the aerosol extinction coefficient (k), while the well-established Sun photometer measurement technique can only retrieve optical properties of the total atmospheric column. The MAX-DOAS technique basically utilizes the differential absorption structures of the oxygen collision complex (O2-O2 or O3) in the visible wavelength region to derive aerosol information. Because no absolute radiometric calibrations are generally needed, MAX-DOAS is suitable for conducting long-term automated measurements in a consistent manner, even at remote sites. In addition, MAX-DOAS measurements can yield significant information about several important trace gases, such as nitrogen dioxide (NO2), formaldehyde, and glyoxal (e.g., Hönninger et al., 2004; Wittrock et al., 2004; Heckel et al., 2005; Hendrick et al., 2006; Leigh et al., 2006; Sinreich et al., 2007), and thus have a variety of potential applications for understanding the Earth’s system. Necessary input parameters for trace gas retrievals are given by the derived aerosol properties. However, no MAX-DOAS aerosol measurements for the real atmosphere have been reported in the literature.

We performed ground-based MAX-DOAS measurements at Tsukuba, Japan (36.1° N, 140.1° E, 29 m a.s.l.), from 1 November to 21 December 2006. Here we present, for the first time, the retrieval algorithm for determining τ and the vertical profile of k at a wavelength (λ) of 476 nm. The wavelength of 476 nm corresponds to the O4-cross-section-weighted mean wavelength over the fitting window chosen for the spectral fitting (460–490 nm). We next show comparisons with lidar and sky radiometer measurements conducted at the same site. Although the geometry and the integration time for completing a set of MAX-DOAS measurements are different from those of the lidar and sky radiometer, we attempt to estimate the overall uncertainty of the retrieved k...
values for layers of 0–1 and 1–2 km, to which the sensitivity of MAX-DOAS measurements has been simulated to be much higher than at other altitudes (Frieß et al., 2006). Limitations in the retrieval of $\tau$ are discussed based on the comparisons with sky radiometer and lidar data.

2 Measurement and Retrieval Algorithm

2.1 MAX-DOAS

Our MAX-DOAS instrument consists of two main parts: a telescope unit placed outdoors and a miniaturized UV/visible spectrometer (B&W TEK Inc., BTC111) indoors. These are connected via a 3-m fiber-optic bundle cable consisting of seven cores, each of which having a diameter of 100 $\mu$m. The telescope unit has two viewing quartz windows, one for the zenith and one for off-axis directions. The line of sight for off-axis geometries was directed northwest. A movable mirror on a rotary actuator was installed inside the telescope unit, allowing a sequential measurement of scattered sunlight at six different elevation angles (ELs) of 3°, 5°, 10°, 20°, 30°, and 90° every 30 min. A plano-convex lens (focal length=40 mm and diameter=25 mm) focuses the received sunlight onto the fiber. The field of view (FOV) was estimated to be <1°. The spectrometer employs a 10-$\mu$m width slit, a Crossed Czerny-Turner optical layout, a plane-ruled grating with a groove frequency of 1800 lines mm$^{-1}$ at a blaze wavelength of 250 nm, a thermoelectrically cooled linear charge-coupled device (CCD) array (ILX511, Sony) with 2048 elements covering a wavelength region of 283–566 nm, and a built-in 16-bit digitizer with a high-speed USB interface.

For the measured spectra, the wavelength calibration was performed by fitting a high-resolution solar spectrum (Kurucz et al., 1984) to the spectra measured with our instrument. The instrument slit function was assumed to be a Gaussian shape, based on measurements of mercury emission lines around 404.7, 407.8, and 435.8 nm. The wavelength length calibration was made daily to take into account temporal changes of wavelength shift and spectral resolution. For the wavelength region analyzed (460–490 nm), the spectral resolution (Full Width at Half Maximum (FWHM)) was about 0.55 nm.

The retrieval algorithm for deriving aerosol information was a 2-step procedure. First, the measured spectra of scattered sunlight at 460–490 nm were analyzed based on the so-called DOAS technique (Platt, 1994) using the nonlinear least-squares spectral fitting method (Levenberg-Marquardt method Rodgers, 2000) to retrieve the differential slant column density ($\Delta$SCD) of O$_3$. This spectral fitting was performed with the logarithm of the intensity ($\ln I(\lambda)$):

$$
\ln I(\lambda) = \ln (I_0(\lambda) - c(\lambda)) - \sum_{i=1}^{n} \sigma_i (\lambda) \Delta$${SCD}_i - p(\lambda),
$$
(1)
$$

where the $\Delta$SCD is defined as the difference between the slant column density along the path of sunlight for an off-axis measurement (EL<90°) and that for a zenith-sky measurement (EL=90°). $I_0(\lambda)$ is the reference spectrum at time $t$, which was derived by interpolating two spectra measured at EL=90° within 30 min before and after the off-axis measurement was made at $t$. Over 460–490 nm, a wavelength-dependent offset, $c(\lambda)$, and a slowly-changing structure mainly due to Rayleigh and Mie scattering, $p(\lambda)$, were taken into account by 2nd- and 3rd-order polynomials, respectively. The gas species considered were O$_3$, NO$_2$, O$_3$, and H$_2$O, for which we adopted absorption cross section ($\sigma$) data of Greenblatt et al. (1990) (with manual adjustment (A. Richter, personal communication)), Vandeaele et al. (1998); Bogumil et al. (2003), and the year 2000 edition of the High-Resolution Transmission (HITRAN) database (Rothman et al., 2003), respectively. In addition, we took into account the Ring effect (Grainger and Ring, 1962) and the undersampling effect using the model of Chance and Spurr (1997) and the method of Chance et al. (2005),

**Fig. 1.** Example of nonlinear least-squares spectral fitting results for O$_3$. This is for SZA=53° and EL=10° on 8 November 2006. The red line shows the cross section scaled to the measured spectrum (black) by the DOAS technique. The spectra are plotted as a differential optical density ($\Delta\tau$) from the reference spectrum (EL=90°).

**Fig. 2.** Time series of O$_3$ΔSCD values at EL=3° (red), 5° (green), 10° (blue), 20° (gray), and 30° (black) on 6–9 November 2006.
respectively. The so-called I₀ effect, arising due to different conditions (spectral resolution and light source) under which laboratory experiments of absorption cross sections and atmospheric measurements were made, was considered for all the absorption cross section data, according to the method of Aliwell et al. (2002).

An example of the DOAS spectral fitting results is shown in Fig. 1, where two spectra of the O₃ differential optical density (∆τ), which correspond to the scaled cross section (red), σΔSCD, and the sum of the scaled cross section and the residual (black), respectively, are plotted. The median residual of ∆τ for all the measurements around noon (solar zenith angle (SZA) = 50°±5°) was as small as 6×10⁻⁴. The corresponding O₃ ΔSCD error was ~1×10⁻⁴ molecules² cm⁻⁵, according to the two types of estimates using the measurement covariance matrix constructed based on the statistics of dark counts in nighttime and that constructed from the residuals. For 6–9 November 2006, the retrieved O₃ ΔSCD values are shown in Fig. 2. On 6 November, cloud bases were persistently detected at altitudes of 3–6 km by lidar over Tsukuba, whereas on 7–9 November the sky was relatively clear. Consistent with the fact that aerosols/clouds lower ΔSCD by shortening the effective path of sunlight reaching the telescope (Wagner et al., 2004), ΔSCD data on 6 November show smaller values and weaker dependences on EL. These indicate that ΔSCD values were retrieved precisely from our instrumentation and analyses. From 6 to 9 November, a significant temporal variation of aerosols in the lower troposphere is expected to have occurred as ΔSCD values varied over the time period (Fig. 2).

We next used the following methods to convert the O₃ ΔSCD values into τ and the vertical profile of k at λ=476 nm. We defined the measurement vector (y; representing quantities to be fitted) and the state vector (x; representing quantities to be retrieved) as

\[ y = (O₃ ΔSCD(Ω₁) \cdots O₃ ΔSCD(Ωₙ))^T \]

\[ x = (τ F₁ F₂ F₃)^T, \]

respectively, where n is the number of measurements in a 30-min interval, which corresponds to a complete sequence of ELs, and Ω is the observation geometry vector consisting of three components: the solar zenith angle (SZA), the relative azimuth angle (RAA), and EL. RAA is the azimuth angle between the telescope direction and the Sun. F values are the parameters determining the shape of a vertical profile and are defined to range between 0 and 1. Thereby, partial τ values for 0–1, 1–2, and 2–3 km can be described as τ F₁, τ(1-F₁) F₂, and τ(1-F₁)(1-F₂) F₃, respectively, and the partial τ above 3 km as τ(1-F₁)(1-F₂)(1-F₃).

From the given partial τ above 3 km, we determined the profile of k for a layer from 3 to 100 km assuming a k value at the top of the layer (100 km) and an exponential profile shape. Similarly, we determined profiles for layers of 2–3, 1–2, and 0–1 km, completing the k vertical profile from the surface to 100 km. To represent a k profile with the above method, the prerequisite k value at 100 km was obtained by extrapolating Stratospheric Aerosol and Gas Experiment III (SAGE III) k data (λ=448 and 521 nm) taken at altitudes of 15–40 km. Although this estimate might be crude, this assumption for high altitudes does not significantly influence the retrieval for the lower troposphere (Frieß et al., 2006). In support of this, a sensitivity test conducted using a tenfold k value for 100 km showed that its impact on the retrievals of τ and k below 2 km was less than 3% (Table 1).

Although a k profile is represented by only four parameters, we can obtain various continuous profiles, as shown in Fig. 3. Moreover, the parameterization has an advantage that the retrieval can be made without a priori knowledge of the absolute value of k; Frieß et al. (2006) have argued that using inappropriate a priori constrains on k can easily cause unrealistic or strongly biased results. We instead used a priori information of the profile shape, which is parameterized by the F values, because the relative variability of the profile shape, in terms of 1-km averages, was much less than that of the absolute k value for the period and time of the measurements presented here, as mentioned below. There are, however, disadvantages in that it is not easy to directly derive the vertical resolution and the measurement sensitivity. We will attempt to derive them from the comparisons with lidar and sky radiometer measurements below.

For different sets of aerosol profiles and observation geometries, we created a lookup table (LUT) of the box-airmass-factor (A_box) vertical profile using our radiative transfer model, the Monte Carlo Atmospheric Radiative Transfer
For off-axis measurement geometries, we also created a LUT for profiles of \( \Delta A_{\text{box}} \), which was calculated by subtracting the corresponding zenith-sky \( A_{\text{box}} \) value. Instead of \( A_{\text{box}} \), the \( \Delta A_{\text{box}} \) value was used in the retrieval, as it is more directly linked to \( \Delta SCD \).

Under the above setup conditions, the \( \tau \) value and \( k \) profile were retrieved by applying the optimal estimation method (Rodgers, 2000). The nonlinear inversion problem was solved using the iteration equation:

\[
x_{i+1} = x_i + \left( S_{a}^{-1} + K_i^T S_{s}^{-1} K_i + \gamma_i D \right)^{-1} \left( K_i^T S_{s}^{-1} [y - F(x_i)] - S_{a}^{-1} [x_i - x_a] \right),
\]

where \( x_{i+1} \) and \( x_i \) are the current and previous state vectors, respectively. \( S_{a} \) is the measurement error covariance matrix, \( K_i \) is the weighting function matrix, and \( F \) represents the forward model converting a given aerosol profile to \( \text{O}_4 \Delta SCD \) values. \( D \) is the diagonal scaling matrix and \( \gamma_i \) is the parameter updated in each iteration to optimize the retrieval.

The a priori state vector \( (x_a) \) and the a priori covariance matrix \( (S_{a}) \) were constructed based on averages and standard deviations (s.d.) of two months of lidar data, mentioned below.

The a priori values (± error) used were \( \tau = 0.21 \pm 0.06 \), \( F_1 = 0.70 \pm 0.05 \), \( F_2 = 0.60 \pm 0.05 \), and \( F_3 = 0.60 \pm 0.05 \), yielding a \( \tau \) of 0.21, \( k \) (0–1 km), defined as the mean for \( k \) of 0–1 km, of 0.15 km\(^{-1}\), \( k \) (1–2 km) of 0.04 km\(^{-1}\), and \( k \) (2–3 km) of 0.02 km\(^{-1}\). The errors are \( \pm 1.0 \), \( +0.76/-0.66 \) km\(^{-1}\), \( +0.23/-0.15 \) km\(^{-1}\), and \( +0.11/-0.05 \) km\(^{-1}\), respectively.

The relative error (variation) ranges of the \( F \) values are much smaller than those of the absolute \( k \) values. The above a priori errors correspond to 20% of the s.d. of \( F \) values from the lidar data, but three times that of \( \tau \). This modification was made to stabilize the retrieval and allow for a wide range of values. Non-diagonal elements of the a priori covariance matrix were set to zero. Results from the sensitivity studies of the a priori values and their errors are summarized in Table 1.

In the tested ranges, the impacts on \( \tau \), \( k \) (0–1 km), and \( k \) (1–2 km) values were less than 6%, 5%, and 31%, respectively.

Figure 4 shows the mean averaging kernels for all the MAX-DOAS aerosol retrievals. Except for \( F_3 \), the peaks of the averaging kernels are located at the corresponding variable. \( \tau \) indicates the largest value, followed by \( F_1 \) and \( F_2 \), suggesting a higher sensitivity of MAX-DOAS measurements to lower-altitude aerosols. The resulting degrees of freedom for signal (Rodgers, 2000) ranged from 1.0 to 2.7 for the whole period. The area (Rodgers, 2000), which provides a rough measure of the fraction of the retrieval that comes from the measurements, was calculated as the sum of all elements in the averaging kernel profile weighted by the a priori error (Liu et al., 2005). The areas were 1.0, 0.2, 0.1, and 0.1 for \( \tau \), \( F_1 \), \( F_2 \), and \( F_3 \), respectively, indicating that the retrieval was done by scaling the given a priori profile first, followed by changing the profile shape. The \( \text{O}_4 \Delta SCD \) values modeled under these conditions (Fig. 5) agreed well
Table 1. Median Absolute (Relative) Errors Estimated From Sensitivity Tests and the Retrieval Covariance Matrix.

<table>
<thead>
<tr>
<th>Error Source</th>
<th>Test</th>
<th>( \tau )</th>
<th>( k(0-1 \text{ km}) )</th>
<th>( k(1-2 \text{ km}) )</th>
</tr>
</thead>
<tbody>
<tr>
<td>Fixed ( k ) at 100 km</td>
<td>Tenfold</td>
<td>&lt;0.01 (3%)</td>
<td>&lt;0.01 (3%)</td>
<td>&lt;0.01 (2%)</td>
</tr>
<tr>
<td>Fixed asymmetry para.</td>
<td>±0.05</td>
<td>0.01 (6%)</td>
<td>0.01 (7%)</td>
<td>&lt;0.01 (8%)</td>
</tr>
<tr>
<td>Fixed single scat. albedo</td>
<td>±0.05</td>
<td>&lt;0.01 (&lt;1%)</td>
<td>&lt;0.01 (&lt;1%)</td>
<td>&lt;0.01 (&lt;1%)</td>
</tr>
<tr>
<td>Fixed surface albedo</td>
<td>±0.05</td>
<td>&lt;0.01 (2%)</td>
<td>&lt;0.01 (2%)</td>
<td>&lt;0.01 (1%)</td>
</tr>
<tr>
<td>A priori ( \tau )</td>
<td>±0.1</td>
<td>&lt;0.01 (&lt;1%)</td>
<td>&lt;0.01 (&lt;1%)</td>
<td>&lt;0.01 (&lt;1%)</td>
</tr>
<tr>
<td>A priori ( F_1 )</td>
<td>±0.1</td>
<td>0.01 (4%)</td>
<td>&lt;0.01 (5%)</td>
<td>0.01 (31%)</td>
</tr>
<tr>
<td>A priori ( F_2 )</td>
<td>±0.1</td>
<td>0.02 (6%)</td>
<td>&lt;0.01 (5%)</td>
<td>&lt;0.01 (5%)</td>
</tr>
<tr>
<td>A priori ( F_3 )</td>
<td>±0.1</td>
<td>&lt;0.01 (&lt;1%)</td>
<td>&lt;0.01 (&lt;1%)</td>
<td>&lt;0.01 (&lt;1%)</td>
</tr>
<tr>
<td>A priori ( \tau ) error</td>
<td>±0.1</td>
<td>&lt;0.01 (&lt;1%)</td>
<td>&lt;0.01 (&lt;1%)</td>
<td>&lt;0.01 (&lt;1%)</td>
</tr>
<tr>
<td>A priori ( F_1 ) error</td>
<td>±0.01</td>
<td>&lt;0.01 (&lt;1%)</td>
<td>&lt;0.01 (&lt;1%)</td>
<td>&lt;0.01 (&lt;1%)</td>
</tr>
<tr>
<td>A priori ( F_2 ) error</td>
<td>±0.01</td>
<td>&lt;0.01 (&lt;1%)</td>
<td>&lt;0.01 (&lt;1%)</td>
<td>&lt;0.01 (&lt;1%)</td>
</tr>
<tr>
<td>A priori ( F_3 ) error</td>
<td>±0.01</td>
<td>&lt;0.01 (&lt;1%)</td>
<td>&lt;0.01 (&lt;1%)</td>
<td>&lt;0.01 (&lt;1%)</td>
</tr>
<tr>
<td>Smoothing and retrieval noise errors</td>
<td>—</td>
<td>0.01 (7%)</td>
<td>0.02 (14%)</td>
<td>0.01 (30%)</td>
</tr>
<tr>
<td>Overall error</td>
<td>—</td>
<td>0.03 (12%)</td>
<td>0.02 (18%)</td>
<td>0.01 (44%)</td>
</tr>
</tbody>
</table>

with those measured (Fig. 2). For each 30-min interval, the relative residual, (residual of \( O_4 \Delta SCD \))/(mean measured \( O_4 \Delta SCD \)), was calculated. The median of the relative residuals for all the retrievals was 9%.

The error of the retrieved state vector is quantified by the retrieval covariance matrix,

\[
\hat{S} = \left( \mathbf{K}^T \mathbf{S}_e^{-1} \mathbf{K} + \mathbf{S}_n^{-1} \right),
\]

which is defined to represent the sum of the smoothing error and the retrieval noise error (Rodgers, 2000). For this error estimate, the measurement error covariance matrix \( \mathbf{S}_e \) was constructed from the residual of \( O_4 \Delta SCD \), because it was much larger than the \( O_4 \Delta SCD \) errors. For all the retrievals, medians of the errors estimated from each of the sensitivity tests conducted above and the retrieval covariance matrix are summarized in Table 1. Overall errors, calculated as the root-sum-squares of these errors, were 0.03 (12%), 0.02 km\(^{-1}\) (18%), and 0.01 km\(^{-1}\) (44%) for \( \tau \), 0–1 km, and 1–2 km, respectively (Table 1).

It should be noted that outputs from the retrieval are available only for retrieved \( \tau \) less than 3, the largest value in the LUTs. This excludes considerably large optical depth cases, most of which should be due to optically thick clouds. Further data screening was made using the relative residual, which was a good measure for screening retrievals that yielded a \( k \) profile too far from the true profile. This deviation likely happened when the above-mentioned method constructing a profile was too simple to represent the true profile, particularly with a steep vertical gradient of \( k \) due to clouds.

From all the retrievals, the threshold for this data screening was determined statistically to be 20%, which corresponded to (the mode) plus (the one-sigma width) in a histogram of the relative residuals on a logarithm basis.

2.2 Lidar

The lidar system used here was a compact Mie-scattering system utilizing the second harmonics of a flashlamp-pumped Nd:YAG laser (532 nm) as a light source (Shimizu et al., 2004). One aerosol profile comprising 3000 shots was obtained in 5 min followed by a 10-min rest to extend the laser life. The lidar equation was solved using the inversion method described by Fernald (1984) to derive vertical profiles of aerosol optical properties, such as \( k \), with a vertical resolution of 30 m. In the method used here, a constant...
extinction-to-backscattering ratio ($S$) of 50 sr was assumed. According to the estimates by previous studies for Aerosol Robotic Network (AERONET) locations and Tsukuba (Catrall et al., 2005; Tatarov et al., 2006), the $S$ ratio can vary by more than 30%, with resulting errors in $k$ due to the use of a fixed $S$ value occasionally exceeding 30%. Only $k$ data for which no clouds at 0–6 km were judged by the method of Shimizu et al. (2004) were used below. For reference, vertical profiles of $k$ at 532 nm measured by the lidar are shown in Fig. 6. We converted the lidar $k$ values at 532 nm to the values at 476 nm using Ångström exponent data obtained from the sky radiometer measurements described below.

### 2.3 Sky radiometer

The sky radiometer measured the direct solar irradiance and the distribution of radiances in the aureole region (Aoki and Fujiyoshi, 2003). In the present study, we analyzed a set of measurements of the direct solar irradiance and the solar aureole radiance distributions within 160$^\circ$ of the center of the Sun. The measurements were made in 30 s to 2 min, depending on SZA, and were repeated every 5 min. The $\tau$ values at 400, 500, 675, 870, and 1020 nm were derived using the inversion algorithm (SKYRAD.pack, version 4.2) developed by Nakajima et al. (1996). The error in $\tau$ was generally as small as $\sim$0.01 (Kim et al., 2005), comparable to Sun photometer measurements implemented in AERONET. Only 30% of the data are used below to minimize the influence by cloud contamination (Aoki and Fujiyoshi, 2003). The Ångström exponent was calculated from the above five $\tau$ values and was used to derive $\tau$ at 476 nm.

### 3 Results and discussion

Figure 7a shows the time series of all the MAX-DOAS $k$(0–1 km) values at 476 nm from 1 November to 21 December 2006. The value of $k$(0–1 km) represents the mean value of $k$ in the layer from 0 to 1 km, as defined earlier. In Figs. 7b and 8, only coincident pairs of MAX-DOAS (red) and lidar (black) data are plotted. Note that the coincident pairs have been selected using cloud-free lidar data, providing favorable comparison conditions. However, this does not necessarily mean that no clouds occurred at the MAX-DOAS viewing directions. It can readily be seen from Figs. 7b and 8 that the MAX-DOAS $k$(0–1 km) values show temporal variations very similar to the lidar data. The correlations between MAX-DOAS and lidar $k$(0–1 km) values are compact (slope=1.01 and correlation coefficient $R^2=0.85$), and in most cases the differences are less than 30% (Fig. 9a).

As mentioned above, the lidar data have an uncertainty of more than 30%, due to the use of a fixed $S$ ratio. In addition, MAX-DOAS and lidar measurements were made with...
different geometries (the combination of zenith-sky and off-axis versus zenith-sky only, respectively) and different integration times for completing a set of measurements (30 versus 5 min, respectively). Because these factors should explain at least part of the differences seen in Fig. 9a, the overall intrinsic uncertainties of MAX-DOAS k(0–1 km) are smaller than 30%. Similarly, the uncertainty of the k(1–2 km) values was estimated from the comparisons with lidar data (Fig. 9b) to be less than 60%. These agreements may indicate that the vertical resolution at altitudes below 2 km was roughly 1 km, corresponding to the vertical spacing chosen in the parameterization of k, consistent with the simulation done by Frieß et al. (2006). MAX-DOAS and lidar k(2–3 km) values were both about 0.02 km$^{-1}$. However, it was difficult to assess the MAX-DOAS k(2–3 km) values, because the values were very close to the a priori value used in the MAX-DOAS retrieval.

It is interesting to note that these comparisons have been made without detailed cloud screening methods for the MAX-DOAS data, and it is very likely that some clouds were occasionally near the FOVs of the MAX-DOAS system, as mentioned above. In the MAX-DOAS/lidar comparisons of k for the 0–1 km (1–2 km) layer shown in Fig. 9, however, no significant differences exceeding 30% (60%) are generally seen, suggesting that uncertainties due to such cloud influences were less than 30% (60%).

We next compare $\tau$ values from MAX-DOAS and the sky radiometer (Fig. 10). Also plotted in Figs. 10b and 11 are the mean values of the Moderate Resolution Imaging Spectroradiometer (MODIS) data (collection 5) obtained within 0.3° latitude and longitude of the measurement site. Two MODIS datasets from different satellites (Terra and Aqua) have been averaged separately but are plotted with the same symbols for simplicity. The MODIS data with a cloud fraction less than 0.4 were used. The original MODIS $\tau$ values at 470 nm have been converted to those at 476 nm using MODIS Ångström exponent data. Good general agreement between the three datasets can be seen in Figs. 10b and 11.

The correlations between MAX-DOAS and sky radiometer data are shown in Fig. 12. For most cases the agreement is within 30% (Fig. 12). However, MAX-DOAS values tend to be smaller than sky radiometer values at sky radiometer $\tau$ values greater than 0.4 (Fig. 12).

As the sky radiometer tracked the Sun, it was oriented mainly toward the east-south-west sky, whereas the
MAX-DOAS measured the sky in a different direction (northwest). In addition, measurements by the sky radiometer were made for an integration time of 30 s to 2 min, which was different than that of MAX-DOAS (30 min). Therefore, the \( \tau \) values in air masses measured by MAX-DOAS and the sky radiometer should have occasionally been different.

These influences of the incomplete matching of the measured air masses might explain the differences between MAX-DOAS and sky radiometer data. It is unlikely, however, that these influences are the dominant factor causing the systematic differences seen at sky radiometer \( \tau \) values greater than 0.4, which are expected to show random features in the differences.

A simulation of MAX-DOAS aerosol measurements by Frieß et al. (2006) has shown that the sensitivity of the measurements decreases with increasing altitude in the troposphere. This is supported by the averaging kernels for our MAX-DOAS aerosol retrievals (Fig. 4). As discussed above, the profiles of \( k \) from MAX-DOAS agreed well with those from the lidar for altitudes below 3 km. Therefore, an underestimate of the MAX-DOAS \( \tau \) values might have occurred, owing to optically thick aerosols above 3 km. This will be confirmed in the future by investigating other sites and seasons, where and when optically thick aerosols occur more often at high altitudes.

Another possible cause is the cloud influence. To quantify the maximum influence of clouds on the MAX-DOAS retrievals, an additional sensitivity test was performed...
assuming that all extinction occurred by clouds with $g=0.86$ and $s=1.00$. The sensitivity test showed that MAX-DOAS $\tau$ values ($k$ below 2 km) retrieved assuming typical aerosols ($g=0.65$, $s=0.95$) were systematically smaller than those retrieved assuming clouds by about 30% (40%). This result suggests that for more realistic situations containing both aerosols and clouds, the $\tau$ value ($k$ below 2 km) retrieved assuming typical aerosols includes the 30%–(40%) reduced contribution of the cloud optical depth, possibly explaining part of the systematic differences seen in the comparisons between MAX-DOAS and sky radiometer $\tau$ data. In addition, this non-negligible cloud influence qualitatively explains why MAX-DOAS data showed occasional enhancements of $\tau$ values when no coincidences with sky radiometer measurements under cloud-free conditions occurred (Fig. 10).

Considering the above results suggesting the reduced cloud contribution, the simulation performed by Frieß et al. (2006), and the averaging kernels for our retrieval method, MAX-DOAS measurements would be very insensitive to high-altitude clouds, suggesting that $k$ profiles in the lower troposphere can be measured even when clouds are present at high altitudes. While MAX-DOAS has these advantages, the development of a proper cloud screening method will improve the performance of aerosol measurements by MAX-DOAS, as well as other remote sensing techniques for aerosol measurements.

4 Conclusions

We have presented, for the first time, a new aerosol retrieval algorithm applicable to MAX-DOAS measurements made in the real atmosphere. To characterize the retrieval algorithm for deriving $\tau$ and the vertical profile of $k$ at 476 nm in the lower troposphere, we performed MAX-DOAS measurements at Tsukuba, Japan, from 1 November to 21 December 2006, and compared them with lidar and sky radiometer measurements. The comparisons between $k$ (0–1 km) values from MAX-DOAS and the lidar showed good agreement, within 30% for most cases, although the cases compared probably included partly cloudy conditions. Considering that errors in lidar $k$ data could occasionally exceed 30%, it is likely that the comparisons yielded an upper limit to the overall uncertainty of MAX-DOAS $k$ (0–1 km) data. Similarly, the overall uncertainty of MAX-DOAS $k$ (1–2 km) was estimated to be less than 60%. The MAX-DOAS $\tau$ values generally agreed with sky radiometer data to within 30%. Thus, by applying our aerosol retrieval algorithm, MAX-DOAS provides a new, reliable dataset of aerosols in the troposphere. Moreover, the derived aerosol properties can be used as reliable input parameters for the vertical profile retrieval of trace gases detected by MAX-DOAS. In some cases, however, the MAX-DOAS $\tau$ values tended to be smaller than sky radiometer data. This underestimation likely occurred when optically thick aerosols were present at high altitudes and/or clouds occurred in the lower troposphere. It was suggested that $k$ profiles can be measured by MAX-DOAS even when clouds are present at high altitudes. However, the development of a proper cloud screening method is highly desirable for more accurate aerosol measurements by MAX-DOAS.

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