GOME Level 1-to-2 Data Processor Version 3.0: A Major Upgrade of the GOME/ERS-2 Total Ozone Retrieval Algorithm

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Abstract

The Global Ozone Monitoring Experiment (GOME) was launched in April 1995, and the GOME Data Processor (GDP) retrieval algorithm has processed operational total ozone amounts since July 1995. GDP Level 1-to-2 is based on the two-step Differential Optical Absorption Spectroscopy (DOAS) approach, involving slant column fitting followed by Air Mass Factor (AMF) conversions to vertical column amounts. We present a major upgrade of this algorithm to Version 3.0. GDP 3.0 was implemented in July 2002, and the 9-year GOME data record from July 1995 to December 2004 has been processed using this algorithm.

The key component in GDP 3.0 is an iterative approach to AMF calculation, in which AMFs and corresponding vertical column densities are adjusted to reflect the true ozone distribution as represented by the fitted DOAS effective slant column. A neural network-ensemble is used to optimize the fast and accurate parameterization of AMFs. We describe results of a recent validation exercise for the operational version of the total ozone algorithm; in particular, seasonal and meridian errors are reduced by a factor of two. On a global basis, GDP 3.0 ozone total column results lie between –2% and +4% of ground-based values for moderate solar zenith angles lower than 70°. A larger variability of about +5% and –8% is observed for higher solar zenith angles up to 90°.

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OCIS codes

010.4950, 010.1120, 010.1310, 280.1120, 280.1310
1. Introduction

1.1 The GOME instrument

The Global Ozone Monitoring Experiment (GOME) is an across-track nadir-viewing spectrometer on board the Second European Remote Sensing Satellite (ERS-2) platform launched in April 1995. It has been operating successfully for over 9 years. The satellite is sun-synchronous and polar orbiting, with a period of about 100 minutes and a local equator crossing time of 10.30h, a semi-major axis of 7150 km, and a repeat cycle of 35 days. In normal viewing mode, there are three forward scans followed by a back scan, with forward scan footprint size of 320x40 km for a 1.5-second detector readout integration time; the maximum swath is 960 km, with nominal scan angle ±31° at the spacecraft. With this read-out strategy, global coverage is achieved at the equator within three days. There is also a polar viewing mode for improved sounding of polar latitudes during springtime.

GOME has 3584 spectral channels distributed over four serial-readout detectors; the wavelength range is 240 to 793 nm, with a moderate spectral resolution of 0.2 to 0.4 nm. In addition to the regular measurements of backscattered light from the Earth-atmosphere system, GOME has a Pt-Ne-Cr lamp for on-board wavelength calibration, and the solar irradiance spectrum is measured via a diffuser plate (this is done on a daily basis). GOME has also three broad band (band pass > 100 nm) Polarization Measurement Devices (PMDs) measuring light in a direction parallel to the slit. The PMDs' main purpose is to generate a polarization correction for the level 1 spectra (calibrated and geolocated radiances). A comprehensive description of the GOME instrument can be found in the GOME Users Manual.
The spectral resolution of GOME is fine enough to resolve the trace gas absorption signatures of chemically important atmospheric trace species. Ozone column and profile distributions make up the main mission target, but the instrument also retrieves total columns of a number of minor trace species (NO₂, HCHO, BrO, SO₂, OClO), total water vapor content, and some ancillary information on clouds and aerosols. The main operational Level 2 products are the global distributions of total vertical column amounts of ozone and nitrogen dioxide. The core operational Level 1-to-2 retrieval algorithm is based on the DOAS approach. The GOME Data Processor (GDP) at the German Aerospace Center (DLR) has been operational since August 1996 following the GOME commissioning phase, and the entire data record since July 1995 has now been re-processed following upgrades (GDP Level 1-to-2 Versions 2.4 and 2.7 and 3.0) to the original retrieval algorithm. The GOME near-real-time service was initiated in January 1997 with the installation of GDP at the Kiruna station (which receives 10 out of 14 daily ERS-2 orbits) and extended in February 2002 with additional GDP installations at the Gatineau and Maspalomas stations. A successful application of GDP V3.0 to GOME data is presented by Thomas et al.

The GDP system is implemented using FORTRAN 77 and C. It is organized in five code layers containing specialized modules for mathematical fitting routines (layer 1), databases in layer 2 (profile climatologies, other global databases, spectroscopic data), accessing routines to the lower level functionalities (layer 3), and algorithm specific modules such as DOAS, AMF, cloud retrieval (layer 4). Each major algorithm component and the operational environment have its own independent test environment (layer 5). The system was designed to run in parallel on a number of computers. The modularity was demonstrated when GDP 2.0 was ported from a
mainframe UNIX environment to become GDP 2.7 in a more cost-effective Linux cluster environment.

1.2 GDP total ozone algorithm: earlier versions

The DOAS approach to vertical column retrieval is a two-step process involving firstly the derivation of effective slant columns from a least-squares fitting based on the Beer-Lambert absorption law, and secondly the conversion to vertical column densities by means of Air Mass Factor (AMF) computations based on detailed radiative transfer modeling. The DOAS slant column fitting method was first used over 20 years ago for slant column retrieval from ground-based instruments. DOAS was selected for GOME total ozone retrieval in 1992, and the algorithm was made operational in June 1996. Technical aspects of the DOAS fitting and AMF implementation are discussed below in Sections 2 and 3 respectively. Here we give some remarks on the history of the algorithm in order to prepare the context for the newer versions.

All GDP versions up to 3.0 have used a single contiguous fitting window from 325 nm to 335 nm covering part of the ozone Huggins bands absorption features. The initial GDP total ozone algorithm (GDP 2.0, July 1996) used an ozone cross-section reference spectrum taken from literature\textsuperscript{7,8}, but in version 3.0, the GOME-measured flight model cross-sections have been used\textsuperscript{9}. The latter were derived from laboratory measurements made with the in-flight GOME instrument during the calibration phase prior to launch. Versions 2.7 and earlier also assumed a single effective temperature input to be used for the temperature dependency of the ozone cross-sections. A more recent approach\textsuperscript{10} uses two ozone reference cross-sections to retrieve an effective temperature in addition to the slant column, and this technique has been incorporated in version 3.0 (section 2.2).
To deal with wavelength grid registration problems, it is necessary to re-sample spectra on new wavelength grids established through shift and squeeze parameters. Shifting and squeezing of the reference spectra (including the solar spectrum) is therefore implemented. In the UV, GOME actually samples slightly below the Nyquist criterion, and in 1998 it was found that this effect could be largely compensated by the introduction of an undersampling correction\(^1\). A reference spectrum was prepared to handle this effect, and is now standard for version 3.0.

Interference from the Ring effect (inelastic rotational Raman scattering producing “filling-in” of Fraunhofer and absorption features) has been considered from the outset. This effect is treated as a pseudo-absorber in DOAS fitting, with the addition of one or two fitting parameters associated with Ring reference spectra. Up to version 2.7 a single Ring spectrum derived from pre-launch GOME zenith sky measurements; this was found to have limited accuracy. Version 3.0 uses a theoretical Ring Fraunhofer spectrum derived from a folding of Raman cross-sections with a high-resolution solar spectrum\(^2\). Attempts have been made to introduce a second Ring spectrum to deal with telluric (absorption) interference effects.

AMF computations in GDP have gone through some changes. “On-the-fly” radiative transfer (RT) computations have been generally avoided, ostensibly for performance reasons and the need to achieve data turnover in real time. In versions up to 2.7, the single-scatter AMF was computed from scratch, and multiple scatter correction factors were interpolated from a large look-up table (LUT) classified according to scenario geometry and various atmospheric parameters (time-latitude dependent trace gas profiles, aerosol loading, surface albedo and topography. The GOMETRAN radiative transfer model\(^3\) including the so-called pseudo-spherical approximation was used to create the multiple scatter LUTs. Surface Lambertian albedos were derived from a 1°x1° data set that contains information about the surface type\(^4\) but

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reduced to five surface types only (water, sand, bare soil, vegetation, snow), while the spectral dependency is taken from Bowker et al.\textsuperscript{15} for a number of different surface types. Topography is extracted from the ETOP05 data set provided by The National Snow and Ice Data Center (NDISC). A substantial new determination of the AMF has been implemented in GDP 3.0; this is one of the cornerstones of the new version, and is described in detail in Section 3.

In addition to the basic Level 1 input of calibrated geolocated earthshine measurements (radiances), the GDP total ozone algorithm requires inputs from a pre-processing cloud property algorithm in order to deal with partially cloudy scenes in the independent pixel approximation. GDP ingests two cloud properties: the fractional cover that is retrieved by the Initial Cloud Fitting Algorithm (ICFA)\textsuperscript{16} based on fitting of reflectance measurements in and around the O\textsubscript{2}-A band at 760 nm, and the cloud-top pressure supplied from the International Satellite Cloud Climatology Project (ISCCP) data base\textsuperscript{17}.

The development of operational retrieval algorithms such as GDP is an ongoing and iterative process requiring accurate validation of the resulting geophysical products after each significant modification of the operational processing chain. During the commissioning phase of the ERS-2 satellite in 1995, an intensive validation campaign\textsuperscript{18} was conducted to assess the quality of the developmental versions of GDP, using well-controlled ground-based measurements from SAOZ/DOAS UV-visible spectrometers, Brewer and Dobson spectrophotometers, UV filter radiometers and ozonesondes. In addition to comparisons with correlative data from coincident ground stations or balloon experiments, GDP ozone results were compared on a global basis with data from the Total Ozone Mapping Spectrometer (TOMS) instruments onboard Earth Probe and ADEOS, and with retrievals from independent DOAS algorithms.
Validation results for version 2.0 (which generated the first public release of GDP data) revealed some biases and regional discrepancies, in particular at high latitudes and high solar zenith angles, and for situations with high or low ozone content\textsuperscript{19,20}. A major validation of version 2.4 was performed in 1998 after the accumulation of three years of data; this showed clear improvements in both the ozone and NO\textsubscript{2} column retrievals\textsuperscript{21}. Improvements introduced in version 2.7 were mainly confined to NO\textsubscript{2} column retrieval; a “delta” validation of the expected improvement was performed in 1999 on this version\textsuperscript{22}. The entire data record was reprocessed after these validations. Results from these validation programmes as well as a preliminary delta validation of GDP V3.0 are presented in Lambert et al.\textsuperscript{23}. Further comparison of GOME results with ground-based measurements, space-based measurements and results of alternative retrieval methods is presented e.g. in Corlett et al.\textsuperscript{24} and Bramstedt et al.\textsuperscript{25}, the latter also showing first results for GDP V3.0.

1.3 New GDP versions: scope of the paper

The present work concerns the major upgrade to GDP Level 1-to-2 version 3.0 carried out over the 2-year period 2001-2002. Version 3.0 represents a new departure for GDP total ozone using the DOAS approach. The DOAS algorithm (summarized in section 2.1) has been upgraded to include most of the latest research improvements; in particular the use of two ozone reference spectra to derive two pieces of information (effective slant column and temperature) from the fit. These aspects are treated in Section 2.2.

The most significant change concerns the AMF implementation, and this is the heart of the new GDP Version. It was shown\textsuperscript{26} that the AMF (and by association the vertical column density) for ozone could be adjusted to reflect the actual ozone content as expressed through the fitted slant column. This iterative-AMF approach has been implemented now in GDP 3.0; it is
described in greater detail in section 3.1. The approach requires column-classified ozone climatology; this is summarized in section 3.2, along with the profile-column mapping that allows the AMF iteration to be linked with vertical column density adjustment.

In version 3.0, the LIDORT scattering code\(^{27,28}\) is now used for simulations of backscatter intensities required for AMF evaluations. We summarize the implementation of LIDORT in the context of the ozone AMF calculation in section 3.3. The initial implementation of version 3.0 used very large Look-up tables (LUTs) of pre-calculated AMFs, and employed a number of extraction and interpolation steps to determine appropriate AMFs for a given scenario. This time-consuming task can be replaced by a much faster AMF parameterization based on the application of neural network techniques to calculate AMFs analytically, as suggested by Loyola\(^{29}\). Section 3.4 provides a description of the AMF parameterization with neural networks. We conclude section 3 with some remarks on the implementation and performance of GDP in the operational environment.

Following the operational implementation of the new AMF formulations and the DOAS improvements in GDP 3.0, the delta validation campaign was executed in 2002\(^{23}\). A set of 2250 previously validated orbits was used (sampled over a 4-year time period). Validation tools and ground-based data already set up for previous delta-validations were used again. These aspects are summarized in section 4.1, which deals with the validation methodology. GDP 3.0 total ozone retrievals were compared not only with GDP 2.7 results (section 4.2), but also with ground-based and balloon data used in previous validation campaigns (section 4.3). Section 4.4 contains the validation summary.
The paper also has some concluding remarks on the transition to GDP versions beyond 3.0. The complete GOME total ozone data set from July 1995 as re-processed with GDP Version 3.0 is publicly available.

2. Upgrades to DOAS Slant Column Fitting

2.1 DOAS fitting of slant columns

In the fitting step, the forward model is reduced in essentials to the Beer-Lambert extinction law for trace gas absorbers. This is a reasonable assumption to make for optically thin trace gas absorption in the atmosphere, where the cross-sections are weakly dependent on pressure, and strongly linearly dependent on temperature. In DOAS, it is standard practice to use logarithms of the sun-normalized intensities, and then add a low-order polynomial (typically of degree 2) to account for broadband molecular and aerosol scattering and also for reflection from the Earth’s surface. The fitting model is then:

\[
\ln \left[ \frac{I_\lambda (\Omega)}{I_\lambda^0 (\Omega)} \right] = - \sum_g E_g (\Omega) \sigma_g (\lambda) - \alpha_0 - \alpha_1 (\lambda - \lambda^*) - \alpha_2 (\lambda - \lambda^*)^2. \tag{1}
\]

Here, \(I_\lambda\) is the earthshine spectrum at wavelength \(\lambda\), and for GOME, \(I_\lambda^0\) is usually taken to be the extraterrestrial solar spectrum. This model applies to trace gas absorption through the whole atmosphere, where \(E_g (\Omega)\) is the effective slant column density of gas \(g\) appropriate to an atmospheric path characterized by directional variable \(\Omega\), \(\sigma_g (\lambda)\) is the trace gas absorption cross section, \(\alpha_0, \alpha_1\) and \(\alpha_2\) are the low-order (quadratic) polynomial fitting coefficients, and \(\lambda^*\) is a reference wavelength for this polynomial (usually the fitting window mid-point).
Leaving aside issues of wavelength registration, the above expression is linear in the fitting variables \( E_g(\Omega) \) and \( \{ \alpha_k \} \). Additional reference spectra may be considered in the fitting, to compensate for instrumental effects such as undersampling, and atmospheric effects due to inelastic rotational Raman scattering (the Ring effect). Wavelength mismatching is often dealt with by applying fitted shift and squeeze parameters to the wavelength grids of the reference spectra, and also the solar spectrum. This makes the fitting iteratively non-linear (and thus much slower), and it is desirable to use pre-established shifts and squeezes and pre-shifted reference spectra in order to maintain linearity and reduce the possibility of numerical instability.

Over the lifetime of GOME, a number of improvements have been made to the original DOAS algorithm following essential research by a number of groups associated with the instrument. Many of these improvements have been incorporated in the newer GDP algorithms, and some of these were noted in section 1.2 for earlier versions. Here, we deal first with fitting issues and choice of reference spectra as they apply to DOAS in GDP 3.0.

### 2.2 Reference Spectra in GDP 3.0

Version 3.0 uses the latest release of the GOME Flight Model cross-sections (\( \text{O}_3 \) and \( \text{NO}_2 \)), the so-called GOME FM98 data\(^{30,9}\). These cross-sections are taken from cell measurements taken with the GOME instrument prior to the April 1995 launch; measurements were taken at five temperatures (202 K, 221 K, 241 K, 273 K and 293 K). In addition, the interfering species \( \text{NO}_2 \) is also considered in the fit, with cross-sections at 241 K for this trace gas also taken from the GOME FM98 data set. In GDP 3.0, trace gas cross-sections have been subjected to a pre-shift re-sampling (0.012 nm towards longer wavelengths is the wavelength shift for ozone cross sections in the customary 325-335 nm GOME DOAS window); shift and squeeze fitting for these
reference spectra has been disabled (this avoids occasional numerical instabilities found in the earlier GDP versions).

Ozone cross sections in the Huggins bands are temperature dependent, and this must be treated in the fitting. Earlier versions of GDP used a single ozone cross-section as the trace gas reference, and the Huggins band temperature dependence was dealt with by assuming a single effective temperature \( T_{\text{eff}} \) corresponding to the maximum number density in the appropriate climatological ozone profile (temperatures are selected from climatology). However, two ozone cross-sections at different temperatures may be used in the fitting to retrieve \( T_{\text{eff}} \) in addition to the effective slant column \(^{10}\); this avoids the potentially large source of uncertainty associated with an external choice of effective temperature. Specifically, we use the ozone cross-section at temperature \( T_1 \), and an ozone difference cross-section (between temperatures \( T_1 \) and \( T_2 \)), and the temperature dependence of the complete cross-section is assumed to be:

\[
\sigma(T) = \sigma(T_1) + (T - T_1) \frac{\partial \sigma}{\partial T} \approx \sigma(T_1) + (T - T_1) \frac{\sigma(T_1) - \sigma(T_2)}{(T_1 - T_2)}. \quad (2)
\]

The dependence is linear if we assume the temperature derivative is constant. This assumption is a good one for the limited range of stratospheric temperatures\(^8\). Taking \( T_1 = 221 \text{ K} \) and \( T_2 = 241 \text{ K} \), so that \( \Delta T = T_1 - T_2 = 20 \text{ K} \), we use \( \sigma(T_1) \) and \( \Delta \sigma_{12} = \sigma(T_1) - \sigma(T_2) \) as the reference spectra in the fitting. The total slant column optical depth is \( E \sigma(T) \), for effective slant column density \( E \), which is then the fitting parameter corresponding to \( \sigma(T_1) \). If \( D \) is the fitting parameter corresponding to the difference cross sections \( \Delta \sigma_{12} \), then we can define an effective temperature through the relation:

\[
T_{\text{eff}} = T_1 + (T_1 - T_2) \frac{D}{E}. \quad (3)
\]
Effective slant columns are now independent of any assumed temperature climatology. The fitted temperature varies with latitude and season and values can be compared against climatological data and analysis data from forecast models for diagnostic purposes (see also section 5.1).

In Figure 1, we illustrate these changes for a spring orbit. RMS has improved by a factor of two for all cases. In the right-hand panel, fitted effective temperatures vary from 222 K to 240 K in the Northern high latitudes. These values are generally lower than ozone-maximum temperatures selected from climatology, as used in version 2.7; the validation exercise on GDP 3.0 has proved that the use of this fitted-temperature DOAS formalism gives consistently better total ozone column results (see section 4.2).

In subsequent studies after the GDP 3.0 algorithm was defined, ECMWF temperature fields were averaged over atmospheric height using climatology ozone profiles as weighting factors, and derived effective temperatures obtained. It was found that there was a positive bias in GDP 3.0 effective temperatures (10-20 K, depending on orbit). The source of this bias is in part due to the wavelength registration, and improved pre-shift parameters for ozone cross-sections have gone some way to eliminating this bias. Uncertainties in weather-model temperature fields are also a factor. However this is a complex issue that requires a more detailed assessment that is beyond the scope of this paper.

In all GDP versions up to and including 3.0, the earthshine spectrum wavelength grid is used as the standard; for each fitting, the sun spectrum is re-sampled on this reference grid by application of a shift parameter to the sun spectrum wavelength. This particular wavelength mismatch is due mainly to the solar spectrum Doppler shift (an average shift value is 0.008 nm), and it will vary across an orbit due to changes in the instrument temperature. As in Version 2.7,
the static undersampling correction\textsuperscript{11} has been incorporated to deal with slit function sampling below the Nyquist criterion.

3. Iterative AMFs and Neural Network Parameterization

3.1 Iterative AMF adjustment

The AMF definition is

\[ A = \frac{\log(I_{\text{nog}} / I_g)}{\tau_{\text{vert}}}, \]  

for which two calculations of backscatter intensity are required, one \((I_g)\) for an atmosphere including ozone, the other \((I_{\text{nog}})\) for an atmosphere excluding ozone, and \(\tau_{\text{vert}}\) is the vertical optical depth of ozone for the whole atmosphere. All GDP versions have used this definition for the ozone AMF, though other definitions have been used for minor trace species [see for example Palmer et al.\textsuperscript{31}]. AMFs for total ozone column are calculated at 325 nm at the lower end of the DOAS fitting window (325-335 nm). For a clear sky scenario, the final vertical column amount is then defined as the effective slant column divided by a single AMF computed from this chosen ozone profile. For partially cloudy GOME scenarios, the conversion from slant column to vertical column density proceeds via the relation:

\[ V = \frac{E + \phi G A_{\text{cloud}}}{(1 - \phi) A_{\text{clear}} + \phi A_{\text{cloud}}}, \]  

where \(E\) is the effective slant column, with AMFs \(A_{\text{clear}}\) and \(A_{\text{cloud}}\) for clear and cloudy scenarios respectively. In GDP, the factor \(\phi\) is just the cloud fraction \(f\). An alternative definition is \(\phi = f I_{\text{cloud}} / I_{\text{total}}\), where \(I_{\text{total}} = (1-f) I_{\text{clear}} + f I_{\text{cloud}}\). In the above equation, the cloud fraction comes from
the ICFA pre-processing step. The “ghost column” $G$ is the quantity of ozone below cloud-top; it must be computed from ozone profile climatology.

In traditional DOAS retrievals, slant column fitting and AMF calculation steps are decoupled; for a given trace species, the AMF radiative transfer computations are based on \textit{a priori} climatological profile inputs that may have no real connection to the true profile. This is the case with GDP versions 2.7 and earlier, where a suitable ozone profile has traditionally been interpolated (by time and latitude) from a zonal-mean monthly climatology of profiles. If the profile shape is a function of the total column (as it is for ozone), this approach is logically inconsistent - in order to retrieve the total vertical column, it is necessary to know the profile first in order to compute the correct AMF.

The shape and total content of the selected ozone profile may bear little resemblance to the true profile, and (particularly for scenarios with high ozone content and/or high solar zenith angles) the AMF may be significantly in error due to a poor choice of input profile. The motivation behind the iterative AMF approach is to circumvent this uncertainty by using information about the true profile to establish the AMF more accurately. The only relevant profile information available to us in the DOAS context is the fitted slant column, and the iterative AMF algorithm uses the slant column result $E$ to make an adjustment to the AMF (and by extension, the vertical column density) that reflects the trace gas content as expressed in the value of $E$. This adjustment depends on the use of a \textit{column-classified} climatology of ozone profiles; the choice of profile is uniquely determined by a specified vertical column amount. Such climatology was developed some years ago for the TOMS retrievals\textsuperscript{32}, and this is summarized in Section 3.2.
The iteration process is straightforward to describe; we treat first the clear-sky case. An initial AMF $A_0$ is computed given an initial choice $V_0$ of vertical column; the required profile is drawn from the column-classified ensemble. The initial choice $V_0$ may be taken from the TOMS zonal mean column climatology, or (in the GOME operational context) from a previously retrieved result from an adjacent footprint. $V_0$ is the first guess for the iterated vertical column density. Given a slant column $E$ from the DOAS fitting, an updated vertical column $V_1$ is calculated through the relation $V_1 = E / A_0$. From the climatology, $V_1$ determines a new choice of profile, which is in turn used as input to a new AMF calculation, with result $A_1$. A new guess $V_2$ for the vertical column follows from $V_2 = E / A_1$. This process is repeated until convergence has been reached (the relative difference between iterations of $V$ is less than some small number).

For the partially cloudy footprint, the iteration proceeds via:

$$V^{(n+1)} = \frac{E + fG^{(n)}A_{\text{cloud}}^{(n)}}{(1-f)A_{\text{clear}}^{(n)} + fA_{\text{cloud}}^{(n)}},$$

where the $(n)$ superscript indicates the iteration number. In addition to the AMF results, the ghost column is also updated at each step. In this way, the profile, the vertical column density and the AMF have all been adjusted to fit the "true-situation" constraint imposed by the effect slant column. For the great majority of scenarios, convergence for ozone columns is rapid (3 or 4 iterations for a relative change of 0.1% in the vertical column).

A simple flow diagram of the AMF and vertical column iteration is shown in .

### 3.2 Use of a column-classified ozone climatology

We require a way to assign a profile for a given choice of total column. GDP 3.0 uses the TOMS Version 7 ("TV7") column-classified ozone profile climatology$^{32}$ for this task. In the TV7
climatology, there are 10 high- and mid-latitude profiles with total columns of 125 DU to 575 DU at intervals of 50 DU, and 6 low latitude profiles from 225 to 475 DU, also with a 50 DU increment. There are 11 layers in total with pressure levels defined using scale heights; each partial column is given in DU. Cross-latitude jump artifacts are avoided by mixing profiles from adjacent zones using a distance-based weighting scheme.

To define a unique correspondence between profile and column, we proceed as follows. If the profile is represented as a set \( \{ U_j \} \) of partial columns, then the total column is \( V = \sum_j U_j \). For two adjacent TV7 profiles \( \{ U_j^{(1)} \} \) and \( \{ U_j^{(2)} \} \) with total columns \( V^{(1)} \) and \( V^{(2)} \) we define an intermediate profile with column amount \( V \) according to:

\[
U_j(V) = \left( \frac{V - V^{(1)}}{V^{(2)} - V^{(1)}} \right) U_j^{(2)} + \left( \frac{V^{(2)} - V}{V^{(2)} - V^{(1)}} \right) U_j^{(1)}.
\] (7)

This defines a linear profile-column map. This map allows us to interpolate smoothly between profile entries in the climatology; the shape will vary continuously. We are drawing on an ensemble of possible profiles of which the climatology is a sample. In Fig. 3, we illustrate the application of this map for the set of 10 high-latitude TV7 profiles. The AMF and ghost column iterations are based on this linear profile-column correspondence, and the choice of profile is restricted to the ensemble represented by the climatology. There are some circumstances where the climatology may be inappropriate; these are discussed below when we look at the validation.

It is worth noting here that the simplest profile-column map is a scaling \( U_j(V) = U_j^{(0)} V / V_0 \) in terms of a fixed profile \( U_j^{(0)} \) with column \( V_0 \). This mapping is implicit in traditional single-AMF DOAS retrieval algorithms. In this case, any profile associated with a retrieved total column will preserve the shape of the fixed profile (all layer partial columns are
scaled equally). The assumption of a fixed ozone profile shape (with its typical stratospheric bulge) may generate sizeable AMF errors in (for instance) an ozone hole scenario.

The TV7 climatology has some limitations in its classification system, particularly in the tropics, where the fixed vertical distributions of ozone in the lowest two Umkehr layers are unrepresentative. The newly-released TOMS Version 8 climatology has a wider column classification and a more extensive latitudinal distribution, and variations in the Umkehr columns in the first two layers of the TOMS atmospheres are better accounted for in the tropical profile climatology. [The Version 8 climatology was not available at the time of the GDP 3.0 implementation].

Uncertainty in the tropospheric ozone burden is one of the main sources of error in the AMF. Using the LIDORT model (section 3.3) to calculate AMFs, we found that changing the burden in second-lowest layer (9 DU in the TV7 climatology) has little effect on the AMF at 325 nm (less than 0.5%). On the other hand, uncertainty in ozone in the lowest layer (set at 15 DU in the TV7 data) is more important. For a solar zenith angle of 25°, and a surface albedo of 10%, we find that, for a total atmospheric column of ozone of 250 DU, doubling the lowest layer burden to 30 DU induces a decrease of 3% in the AMF, and reducing this burden to zero results in a 3% increase. These values are representative for tropical scenarios, for which the tropospheric ozone burden is largest.

3.3 Radiative transfer implementation

GDP Version 3.0 uses the LIDORT discrete ordinate scattering code\textsuperscript{27,28} for off-line AMF computation. This model incorporates full multiple scattering effects in a stratified atmosphere, with the solar beam as the driving source of scattered light. Azimuthal dependence of the intensity is expressed in terms of the cosine of the relative azimuth between the solar and line of
sight directions. The phase function for scattering is expanded in terms of a Legendre polynomial series in the cosine of the scattering angle. Integration over the multiply scattered diffuse field is approximated by a Gauss-Legendre quadrature sum over a set of polar directions (the discrete ordinates or streams). The resulting coupled linear differential equations of radiative transfer are solved by standard means, and the discrete ordinate solution for the radiation field found by application of boundary-value conditions. Finally, the radiation field at arbitrary viewing angles and optical depth may be obtained by substitution of the discrete ordinate solution in the original RTE and subsequent integration of the source function (this is the “post-processing” step). Polarization is not taken into account.

LIDORT uses the so-called pseudo-spherical (P-S) approximation, in which the solar beam attenuation is treated in a spherical shell atmosphere, but all scattering processes are still plane parallel. This approximation is sufficiently accurate for solar zenith angles up to 90° and for line-of-sight viewing up to 30-35° from the nadir. A number of studies have shown that AMFs calculated with this assumption are sufficiently accurate for converting trace gas slant columns into vertical columns (see for example Sarkissian et al.33).

LIDORT requires as input the set of layer optical thickness values, total single scatter albedos and total phase function Legendre expansion coefficients. These optical property inputs must be prepared from knowledge of atmospheric profile distributions of pressure and temperature, trace gas absorbers, and aerosols, in addition to optical properties of molecules and aerosols. Rayleigh scattering properties (cross-sections, depolarization ratios) are taken from Bodhaine et al.35. Ozone and nitrogen dioxide cross sections are taken from the GOME FM98 database. The ground and clouds are treated as reflecting lower boundaries with Lambertian reflection properties.
The TV7 climatology is specified on a fixed pressure grid. It also possesses an auxiliary temperature data set for use with the ozone profiles (in particular for assigning the temperature dependence of the Huggins band cross-sections), and this temperature data set has been retained in the AMF computations for GDP 3.0. In the RT calculations, we apply a finer altitude grid with spacing at 1 km\(^3\), and ozone profiles on the higher resolution grid are spline-interpolated from cumulative column amounts in the TV7 data sets.

Aerosol burdens and optical properties are extracted and interpolated from the LOWTRAN database\(^3^6\). Maritime and rural continental boundary layer aerosol types are taken for footprints over sea and land surfaces respectively in the lowest 2 km of the atmosphere, while a standard aerosol loading is used in other tropospheric and stratospheric layers. Choice of aerosol regime is important for AMF simulations. For minor trace species such as HCHO and NO\(_2\), AMF errors due to poor choices of boundary layer aerosols can be very substantial\(^3^7\). Since most of the atmospheric ozone is in the stratosphere, we would expect ozone AMFs to be less dependent on tropospheric aerosol burdens, but there are situations such as biomass burning where errors may be more than 5\%. For non-absorbing aerosols (single scattering albedo 0.95 or higher), changing the aerosol extinction optical thickness in the lowest atmospheric layer (thickness about 5.5 km) from regular values (~0.3) to high values (~2.0) has very little effect on the AMF (less than 0.5\%). The situation is different for absorbing aerosols. For a solar zenith angle of 25°, a total ozone burden of 250 DU, and with a boundary layer aerosol optical thickness of 1.0, we find a 4.5% decrease in the AMF when the boundary layer aerosol single scattering albedo is reduced from 0.95 to 0.60. The latter is representative for biomass burning scenarios. Similar figures pertain for desert dust aerosols. Errors are generally lower for higher solar zenith angles and they increase slowly with aerosol optical thickness. A worst-case scenario with a
coal-burning thick sooty boundary layer aerosol (optical thickness 2.5, single scatter albedo 0.3) and low solar zenith angle generates a difference of 7.5% compared with a default scenario with scattering aerosols and low optical thickness.

3.4 AMF parameterization with neural networks

Earlier GDP versions used one very large AMF look-up table (LUT) covering the range of viewing geometries and atmospheric scenarios appropriate to the global monitoring of ozone. These algorithms required the determination of one AMF for each clear-sky and cloudy scenario, and the LUT extraction was done by multiple interpolations through the data set. The use of LUTs conveys a considerable performance advantage over direct “on-the-fly” RT calculations. However, the iterative AMF approach requires repeated extraction of AMFs from LUTs, and it is here that the use of a neural network algorithm can really enhance the performance. Before summarizing the AMF parameterization with neural networks, we first classify the LUTs.

In GDP 3.0, there are 12 separate LUTs, classified according to three latitude bands (tropics, mid-latitude and sub-arctic, with divisions every 30° latitude), two aerosol types (maritime and rural planetary boundary layers, as noted above), and two viewing modes (GOME normal, and GOME polar). For each table we have a subdivision into 6 surface albedo classes (0.01, 0.1, 0.3, 0.5, 0.75, 0.98), 7 lower boundary height levels (0.0, 2.0, 4.0, 6.0, 8.0, 10.0, 12.0 km), and a wide range of viewing geometries, namely 14 solar zenith angles from 15° to 89.99°, 7 relative azimuths from 0° to 180° in steps of 30°, and 13 viewing zenith angles from 0° (direct nadir view) to 60° (extreme polar view). With cloud-tops treated as highly reflecting Lambertian boundaries, the range of albedos and lower boundary height levels will encompass any AMF cloudy-sky radiative transfer requirements.
A novel approach for the AMF parameterization using neural networks is implemented in GDP 3.0. The AMF parameterization can be seen as the old problem in approximation theory that tries to reproduce a given function, either exactly or approximately, by evaluating a given set of primitive functions. Neural networks can be used as universal function approximators when the function to be approximated is specified implicitly through a representative set of input and output examples. In our case, the AMF LUTs are the input and output examples. Each of the 12 AMF tables was divided into training, test and validation data sets. Perturbed training sets were generated using the bagging technique. A neural network was trained for each table and the resulting weights stored for use in the operational processing, and AMFs are then generated analytically from the stored neural network weights.

The 12 individual neural networks are combined with the Mixture-of-experts model using an overall distribution in which the mixing weights and component distributions are dependent on the input $x$. The “experts” are neural networks appropriated in different regions of the AMF input space. Expert network $i$ maps its input $x$ to an output $y_i$. A special module, referred as a gating network, identifies the expert or blend of experts whose output is most likely to approximate the corresponding response $y$. The output of the gating network for an input $x$ is a vector $g$ of scalars $g_i$ that weight the contributions of the various experts according to the input $x$. The final output of the model is a combination of $K$ expert outputs calculated as

$$y = \sum_{i=1}^{K} g_i y_i, \quad \text{ (8)}$$

where for each possible input $x$, the gating weights $g_i$ are greater than or equal to zero and their sum is equal to one.
It should be emphasized that in this scheme, the operational processor does not contain any AMF LUTs, or indeed any auxiliary data sets required for interpolation from such tables. We need only the neural network weights, and knowledge of the footprint coordinates (center point), GOME viewing mode (polar, normal) and a land/sea mask (for aerosol discrimination) is sufficient to select and combine sets of neural network results. Differences between AMF values calculated with the neural network ensemble and computed directly with the LIDORT radiative transfer model are well below 1%. The errors are normally distributed with a mean of the order of $10^{-3}$ and a standard deviation of $\sim 10^{-2}$. Fig. 4 shows the excellent multidimensional interpolation and extrapolation capability of the neural network.

Fig. 5 (left) shows ratios of clear-sky AMFs (GDP 3.0 to 2.7), for the springtime orbit used in previous figures. The largest changes occur at mid and high northern latitudes, where 3.0 AMFs are consistently higher, in agreement with the normally high ozone content at this time of year. In GDP 2.7 AMFs were too low because of inappropriate profile climatology choice. Jump artifacts at 20°N and 30°N are present in this algorithm due to the zonal classification used in the ozone climatology – such artifacts are absent in GDP 3.0. The corresponding vertical column densities are shown in the right-hand panel of Fig. 5; southern mid-latitude differences are due mainly to the improved slant column fit (effective temperature fitting).

The use of neural networks for AMF parameterization doubles the speed of the algorithm, compared with the use of AMF look-up tables with conventional extraction and multidimensional interpolation. This is notable also, because the iterative AMF scheme requires several AMFs to be retrieved for a given footprint. This performance improvement is necessary to establish quick turn-over for near real time GDP implementations, and it enables reprocessing of the entire GOME data record (a little over 9 years of data as of July 2004) to be completed in
an efficient and timely manner. With current hardware capability the entire GOME record can be reprocessed on a Linux-cluster of 20 CPU’s in a few days.

4. Validation of GDP 3.0 total ozone

4.1 Validation principles and methodology

Validation of operational official GDP products has been co-ordinated at the Institut d'Aéronomie Spatiale de Belgique (BIRA-IASB) under the aegis of the European Space Agency. Correlative studies relying on independent ground-based observations from the Network for the Detection of Stratospheric Change (NDSC) have proved to be an efficient validation technique for most of satellite data products, including GOME and TOMS total ozone\textsuperscript{20,22}. The complementary instrumentation involved in the NDSC (DOAS UV-visible and Fourier Transform IR spectrometers, Dobson and Brewer spectrophotometers, millimetre wave radiometers, lidars, and balloon-borne electrochemical ozonesondes) is well maintained and thoroughly documented. Most of these instruments contribute to World Meteorological Organization’s Global Atmospheric Watch (GAW) programme through participation in the NDSC and/or the World Ozone and Ultraviolet Data Centre (WOUDC).

Comparison-based studies must deal with a variety of problems arising from the remote sensing nature of the measurement and the geophysical nature of the observed ozone field. These include differences in spatial and temporal resolution, differences in measurement time (mid-morning GOME against twilight ground-based UV-visible spectrometers), geophysical variability, inhomogeneity of atmospheric parameters along the line of sight, etc. Other problems are related to the preparation and integration of massive amounts of correlative data acquired by a network of ground-based instruments of different types. BIRA-IASB has developed dedicated
comparison and interpretation tools to address these issues. These include radiative transfer tools for modelling the actual geolocation of the sampled air mass and retrieval tools for calculating AMF and temperature effects. The database available for this project started well before the GOME launch in 1995, and is upgraded regularly to reflect newly acquired data and to accommodate latest algorithm versions.

4.2 Ground-based characterization of GDP 2.7 errors

Numerous validation studies of GDP version 2.7 have been carried out since the delta validation with GDP version 2.4 in 1999. These studies have highlighted several sources of errors that could impact the geophysical interpretation of the data. Among the most significant errors are those that generate fictitious cycles and geographical patterns in the GOME data product. Driven by variations in parameters to which the measurement and the retrieval are sensitive (solar illumination, atmospheric profiles of ozone, temperature and pressure, fractional cloud cover etc.), these artificial features are superimposed on the real geophysical variations of ozone. Fictitious seasonal cycles and meridian structures generated by GDP 2.7 appear clearly in Fig. 6, where GDP 2.7 total ozone is compared to ground-based network total ozone data. On a monthly average basis, GDP 2.7 reports systematically lower values by 3% in the tropics. At higher latitudes this annual systematic bias vanishes but a seasonal cycle appears with amplitude increasing from ±3% at middle altitudes to ±6% at polar latitudes. Further ground-based validation has highlighted the direct dependence of GOME/ground intercomparisons on the solar zenith angle, the ozone column, and stratospheric temperatures. Upgrades implemented in GDP 3.0 – especially the improved calculation of AMF and the spectral derivation of the effective temperature – were expected to reduce the amplitude of those errors.
4.3 Delta-validation from GDP 2.7 to 3.0

Delta validation of GDP total ozone from version 2.7 to 3.0 is based on cross-correlation studies of GDP ozone data and on pole-to-pole comparisons with NDSC ground-based measurements. It is clear that there is a general improvement of the total ozone data product. Fig. 7 displays the remaining seasonal cycles and meridian structures of GDP 3.0 total ozone with respect to ground-based data records. Although seasonal and meridian dependencies persist with this new version of GDP, their amplitudes have decreased almost everywhere by about 30-50%. A similar improvement is noted with the well-known dependencies on solar zenith angle and ozone column. Fig. 8 illustrates the improvement with respect to the solar zenith angle dependence in Northern summer: instead of the strong GDP 2.7 underestimation of ozone values by 10% and more at large SZA, the agreement of GDP 3.0 with ground-based data ranges now between +3% and −5% at all latitudes. Similarly, the strong 10%-15% overestimation in 2.7 of the very low ozone columns observed during the Antarctic springtime ozone depletion has now been reduced to the 5% level. It is important to note that the improvements vary with latitude and time: while it is obvious that the situation has improved for Northern summer and Antarctic springtime, the SZA dependence during Northern winter and spring has not changed significantly and still varies from month to month. GDP improvements are also larger and consistent in the Southern than in the Northern Hemisphere.

An end-to-end study\textsuperscript{23} of the impact of the changes on the comparison results indicates that the improvement is mainly driven by changes in the air mass factor calculation. In particular, the use of more appropriate ozone profile climatology reduces the amplitude of the seasonal cycles and meridian structures. Fitted cross-sections temperatures contribute to an additional change in the seasonal cycles of a few percent at high latitudes only. Changes in the fractional
cloud cover have no significant impact. The new climatology has no dramatic effect on the calculation of the ghost column, except in polar springtime where it is reduced by up to 6% from GDP 2.7 to GDP 3.0. In polar areas, especially during springtime, these effects can have comparable magnitudes, sometimes leading to an improvement in the total ozone; on other occasions there is no such improvement.

5. Concluding Remarks

GDP 3.0 is a major upgrade to the operational DOAS-AMF retrieval of total ozone from the GOME instrument. A number of improvements have been incorporated in the DOAS slant column fitting, but the major new departure is the use of an iterative AMF solution scheme based on column-classified ozone climatology together with a neural network parameterization for fast and accurate reproduction of AMF results. Validation results have shown a marked improvement in retrieved total ozone values in several regions and seasons, with the largest improvements in accuracy at high latitudes, northern mid-latitudes and in general for higher solar zenith angles. The seasonal pattern is clearly reduced and results for ozone-hole conditions are also much better in GDP 3.0 than in previous versions.

There have been a number of studies on the degradation of the GOME instrument (see for example Tanzi et al.\textsuperscript{39}), and results from this work have been included in GDP 3.0 processing. Although there has been significant degradation particularly in the UV channel since the year 2000, we have found that the total ozone retrieval is remarkably stable, with no noticeable bias over the 9-year GDP 3.0 processing period (1995-2004). In this regard, the use of spectral ratios in the DOAS algorithm is an advantage, since this lowers dependence on absolute radiometric
calibration, and degradation effects can be partly subsumed in the polynomial closure that is a feature of the slant column fitting.

The main retrieval problem with Version 3.0 is the lack of a correction for molecular Ring effect (filling-in of Ozone absorption features due to inelastic scattering by air molecules). This is responsible for a persistent solar zenith angle dependent underestimation of ozone slant columns that has been present in all versions up to and including 3.0. A new version of GDP (4.0) has very recently been implemented and will replace GDP 3.0 in 2005. In addition to a new molecular Ring correction implemented in GDP 4.0, the cloud pre-processing algorithms have been upgraded, wavelength registration has been improved, and ozone AMFs are now calculated at 325.5 nm to further reduce bias. This algorithm was validated at the end of 2004, and publication of this work is currently in preparation.

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References


FIGURE CAPTIONS

**Fig. 1.** (Left) Improvement from version GDP 2.7 to GDP 3.0 in the fitting RMS for ozone for a single orbit (15621, 16 April 1998) of GOME data. (Right) Fitted effective temperatures for the same orbit.

**Fig. 2.** Functional diagram of the iterative solution scheme for ozone vertical column densities.

**Fig. 3.** (Left) TOMS high latitude ozone profiles (partial columns in [DU]). (Right) Intermediate profiles (dotted) for total columns 10 DU intervals between the climatology values.

**Fig. 4.** Multidimensional interpolation and extrapolation capability of the neural network ensemble for AMF calculations. The figure in the left panel shows the AMF as function of ozone profile and sun zenith angle; the right panel shows AMF as function of surface height and sun zenith angle. The red surface corresponds to the neural network results, while the green stars are the AMF values computed directly with the LIDORT radiative transfer model.

**Fig. 5.** (Left) Ratios of clear sky AMFs from GDP 3.0 and 2.7 for a springtime GOME orbit. (Right) Vertical column densities.

**Fig. 6:** Latitude/month cross-section of the percent relative difference between GOME GDP 2.7 total ozone and ground-based correlative measurements ([GOME-ground]/ground) over the period 1996-1999. Shaded areas highlight positive deviations of GOME from ground-based data.
**Fig. 7:** Latitude/month cross-section of the percent relative difference between GOME GDP 3.0 total ozone and ground-based correlative measurements ([GOME-ground]/ground) for the same period 1996-1999. Shaded areas highlight positive deviations of GOME from ground-based data.

**Fig. 8:** Solar zenith angle dependence of the mean percent relative difference between GOME GDP and ground-based NDSC total ozone data in Northern summer. Left: GDP 2.7; right: GDP 3.0. Data are averaged in 5°-SZA bins and in 4 latitude belts (Tropics 0°-23°N; middle latitudes 23°N-63°N; Polar Circle 67°N±3°; high latitudes >70°N).
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