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A non-iterative linear retrieval for infrared high resolution limb sounders

L. Millán^{1,*} and A. Dudhia¹

¹Atmospheric, Oceanic and Planetary Physics, Clarendon Laboratory, Department of Physics, Parks Road, Oxford, UK ^{*}now at: Jet Propulsion Laboratory, California Institute of Technology, Pasadena, California, USA

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Correspondence to: L. Millán (luis.f.millan@jpl.nasa.gov)

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Abstract

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Currently most of the high spectral resolution infrared limb sounders use subsets of the recorded spectra (microwindows) in their retrieval schemes to reduce the computing time of rerunning the radiative transfer model. A fast linear retrieval scheme is described which allows the whole spectral signature of the target molecule to be used. We determine how close the linearisation point needs to be to the solution in order to fall in the linear regime and also suggest an adjustment to the forward model and Jacobians to propagate the change in pressure and temperature on the gas concentration

retrievals. As an example, this technique is implemented for the Michelson Interferometer for Passive Atmospheric Sounding instrument, but it is applicable to any high resolution limb sounder.

1 Introduction

In an infrared limb viewing instrument, the observed radiance is a function of the wavenumber, pressure, temperature and composition along the viewing path (the at-¹⁵ mospheric profile). Hence, a set of measurements at different viewing angles (different tangent heights) may be used to infer geophysical parameters of the scene against altitude (Gille, 1971). The inversion of such atmospheric radiances into atmospheric parameters is basically the problem of fitting a theoretical model (the forward model) to observations subject to errors.

- The forward model simulates the observations using the radiative transfer equation assuming a given atmospheric state to estimate the radiance that reaches the instrument and modifies it to consider the instrumental effects such as finite field-of-view (FOV) and the instrument spectral response. The retrieval procedure is the search of the atmospheric state whose simulated radiances best fit these observations.
- A common approach used to find such an atmospheric state is the least-squares fit or a constrained version of it. The core of these approaches is to linearise the relationship



between the measurements and a given atmospheric state. However the observations do not depend linearly on the atmospheric state and hence the process needs to be iterated. At each iteration the forward model is run using the atmospheric estimate from the previous iteration as a linearisation point until the given convergence criteria are fulfilled.

High spectral resolution limb sounders, such as the Michelson Interferometer for Passive Atmospheric Sounding (MIPAS) (Endemann, 1999; Fisher, 2000) and the Atmospheric Chemistry Experiment, (ACE) (Bernath et al., 2006) provide thousands of spectral measurements each scan increasing dramatically the computing time spent
by the retrieval running the forward model. To reduce this computing time, most retrieval schemes select optimum subsets of spectra (microwindows) that contain most of the potential information (Dudhia et al., 2002). However, even using this microwindow approach, when the signal to noise (S/N) ratio is small the number of microwindows needed to process in order to obtain an useful signal is too computationally expensive to include in an on-line algorithm.

This study investigates how close the linearisation point needs to be to the solution in order to be able to perform a linear (non-iterative) retrieval scheme (i.e. to establish a linear regime for the retrievals). An inversion under these conditions can use precomputed radiative transfer calculations allowing to incorporate all the spectral profiles

²⁰ corresponding to the target parameters without worrying about the computing speed limitations associated with re-running the forward model.

We examine how closely climatological zonal mean profiles come to satisfying this linear range. Also we suggest a method for linearising the propagation of temperature and pressure effects on the gas concentration retrievals.

25 2 Inverse model

The task of the inverse problem is to estimate an atmospheric profile x that best matches the measured radiances y. A common solution used is to use a least squares



fit, in other words, to minimize the sum of the square differences between the vector y and the simulated measurements $\mathbf{F}(x_0)$ around a given linearisation point x_0 (usually called initial guess).

The iterative Twomey–Tikhonov regularised (Tikhonov, 1963; Twomey, 1963) least squares fit solution is given by (Rodgers, 2000)

$$\mathbf{x}_{i+1} = \mathbf{x}_i + \left(\mathbf{K}_i^T \mathbf{S}_y^{-1} \mathbf{K}_i + \gamma^{-1} \mathbf{R}\right)^{-1} \mathbf{K}_i^T \mathbf{S}_y^{-1} \left(\mathbf{y} - \mathbf{F}(\mathbf{x}_i)\right)$$
(1)

where \mathbf{K}_i is the Jacobian matrix ($K_{jk} = \partial y_j / \partial x_k$) evaluated at iteration *i* using the forward model around the atmospheric profile \mathbf{x}_i , \mathbf{S}_y is the measurement error covariance matrix, γ is a factor that determines the strength of the regularisation and where \mathbf{R} is a regularisation matrix which can be chosen to constrain the magnitude of a given order derivative of the retrieved profile \mathbf{x}_i . This technique estimates the uncertainties in the retrieved state vector \mathbf{x} according to:

$$\mathbf{S}_{x} = \left(\mathbf{K}^{T}\mathbf{S}_{y}^{-1}\mathbf{K} + \gamma^{-1}\mathbf{R}\right)^{-1}$$

where \mathbf{S}_{x} is the covariance matrix of the estimated atmospheric profile \mathbf{x}_{i+1} .

If the linearisation point is sufficiently close to the solution, it follows that the solution may be obtained in one iteration using x_0 and precomputed Jacobians K_0 , hence requiring no on-line forward model calculations. The first task is to establish the "distance" $x - x_0$ over which the following equation applies:

 $\boldsymbol{y} - \boldsymbol{\mathsf{F}}(\boldsymbol{x}_0) \simeq \boldsymbol{\mathsf{K}}_0(\boldsymbol{x} - \boldsymbol{x}_0)$

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3 Linear regimes

Simulations of CO₂ and CH₄ spectra were used to investigate how close to the solution an initial guess profile needs to be in order to lie within the linear regime. Then, the



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(3)

question of whether the atmospheric state can be predicted within this linear regime was investigated.

As an example of the CO₂ and CH₄ spectra used, Fig. 1 shows CO₂ and CH₄ modelled spectra at 30, 60 and 90 km. For CO₂, these simulations were performed between the 685 to 970 cm^{-1} at a spectral resolution of 0.0625 cm^{-1} (corresponding to the MI-

PAS band A spectral range for the optimized resolution data, for more information see Sect. 7). For CH_4 These simulations were performed between 1215 and 1500 cm⁻¹ (MIPAS band B).

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tions.

In this study the forward model used is the Reference Forward Model (RFM) (Dudhia, 2012a).

To test the linearity of the forward model due to pressure and temperature changes, CO_2 spectra for midlatitude day conditions were simulated perturbing the entire pressure profile by 1, 5, 10, 20 and 50 % and by 1, 5, 10, 20 and 50 K, respectively. Then an inversion of the previously computed spectra was performed using a linear least squares fit constrained by a Twomey–Tikhonov regularisation necessary to avoid artifacts at high altitudes induced by numerical noise inside the radiative transfer calcula-

For these simulated retrievals the spectra corresponding to the perturbed profiles were used as the measurement vector (y) and the forward model and the Jacobians were computed around the midlatitude day conditions (the unperturbed scenario). Furthermore, the measurement error covariance matrix was set to the identity matrix ($S_v = I$).

As shown in Fig. 2, the linearity assumption holds up to a 20% perturbation with a 3% margin of error. In this test, the entire profile was perturbed by a constant amount

²⁵ while the real atmospheric perturbations will vary with altitude but it seems a reasonable simplification.

As displayed in Fig. 3, the linearity for temperature holds up to 10 K perturbation with a 3 K error margin. However, it should be noted that this difference is localized only



near the stratopause (i.e. where the temperature gradient changes sign) and outside this region the actual error margin is always less than 0.5 K.

To try to improve the linearity of the forward model due to changes in temperature, rather than retrieving temperature directly, the Planck function or even $B_v(T)/T$ can be retrieved instead.

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For example, to retrieve temperature through the Planck function, the temperature Jacobian needs to be multiplied by $dT/\partial B_{\bar{\nu}}(T)$ where dT corresponds to a temperature change of 1 K and $dB_{\bar{\nu}}(T)$ is the corresponding change in Planck function at a fixed wavenumber chosen to be in the middle of the spectral region of interest.

¹⁰ Figure 4 displays the retrieved and actual values for the same temperature increments as before. As can be seen the linearity of the problem improves considerably when retrieving temperature via the Planck function.

However despite this linearity improvement, this approach cannot be easily used (or the $B_{\nu}(T)/T$ approach) with real data because oscillations in the retrieval profile may cause negative Planck Function values to be retrieved.

Once the pT conditions of the scene are known the gas concentrations may be retrieved since it is the only remaining unknown in the radiative transfer model equation. Hence, the gas of interest can be retrieved selecting a spectral region where it is the principal emitter or where the emissions of the other gases are already accurately known.

Simulations of CH_4 spectra are used to determine the linear regime. First perfect knowledge of temperature and pressure will be assumed to test the linearity under ideal circumstances. Then, since the linear algorithm avoids running the forward model precomputing the CH_4 simulations with the pT conditions of a given linearisation point,

²⁵ a discussion of how to adjust these simulations to the *pT* scene conditions (previously retrieved) will follow.

To test the linearity of the forward model due to CH_4 concentration changes, CH_4 spectra for midlatitudes day time conditions were simulated perturbing the entire CH_4



profile by 1, 5, 10, 20 and 50 % and an inversion was performed in the same manner as before. The results are shown in Fig. 5.

As displayed, the linearity for CH_4 retrievals holds up to a 20 % $x - x_0$ "distance" with a 2 % margin of error at most altitudes. This linearisation range is only shown as a maximum theoretical accuracy of the linear CH_4 retrieval because in practice, if the radiative transfer model is not going to be run in the retrieval scheme, the CH_4

simulations will have to be adjusted for the difference between the pT linearisation and scene conditions.

Climatological variability

- ¹⁰ Having established the linear range for a pressure, temperature and VMR retrieval the next question is which climatological resolution is needed to ensure that the linearisation point is likely to lie within this linear range of the true state. To answer it, an analysis was performed using MIPAS middle atmospheric retrievals from MORSE (an iterative inversion algorithm, for more information see Sect. 7.1) for the days available in lanuary for 2007, 2008, and 2009. It is expected that high variability will be found in
- ¹⁵ in January for 2007, 2008 and 2009. It is expected that high variability will be found in the Northern Hemisphere during this month.

The analysis consisted in computing the pressure/temperature/CH₄ deviation from a zonal mean with respect to latitude bin mean profiles (i.e. global, 60° , 30° , etc) until the deviation at any given point was smaller than the linear regime established previ-

- ²⁰ ously. As shown in Figs. 6–8 a global mean climatology cannot represent the variability found in the atmosphere. An analysis of these figures suggests that for pressure and temperature, a linearisation point for every 30° of latitude is needed to ensure that the "distance" $x x_0$ is less than the required 20 % or 10 K proximity; while the CH₄ retrieval needs a linearisation point for every 20° of latitude to ensure that the "distance"
- $x x_0$ is less than the 20 % required proximity.



4 Forward model adjustments

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To avoid running the radiative transfer model during the linear inversion scheme, the gas simulations are calculated using the pT linearisation conditions and then adjusted to the pT scene conditions retrieved in an earlier stage of the retrieval. Two adjustments are needed: one to adjust the pT dependence of the simulated spectra and a second

one to adjust the pT dependence of the Jacobians.

4.1 VMR forward model *pT* adjustment

The simulated spectra can simply be modified using the temperature and pressure Jacobian of the corresponding spectral range, i.e.

¹⁰
$$\mathbf{F}^{*}(\mathbf{x}) = \mathbf{F}(\mathbf{x}_{0}) + \mathbf{K}_{T_{0}}(\mathbf{T} - \mathbf{T}_{0}) + \mathbf{K}_{\rho_{0}}(\boldsymbol{\rho} - \boldsymbol{\rho}_{0})$$

where $\mathbf{F}^*(\mathbf{x})$ is the corrected spectrum for the actual pT scene conditions (p, **T**), $\mathbf{F}(\mathbf{x}_0)$ is the linearisation point simulated spectrum and \mathbf{K}_{T_0} and \mathbf{K}_{p_0} are the linearisation point temperature and pressure Jacobians, respectively.

Figure 9 displays the CH_4 linear retrievals for different changes in temperature. To isolate the effects of the term " $K_{T_0}(T - T_0)$ " in Eq. (4) the CH_4 Jacobians used were computed with the temperature of the scene (ie. the true temperature) with the CH_4 concentration remaining constant (only the temperature varies). As shown, for a 10 K perturbation the errors induced are less than 20 % practically at all altitudes. For a 20 K perturbation the retrieved values are nonrealistic, for this reason, the retrieved CH_4 corresponding to a 50 K increment is not shown.

Figure 10 is the equivalent figure for pressure increments (testing only the " $\mathbf{K}_{p_0}(\boldsymbol{p} - \boldsymbol{p}_0)$ " term of Eq. (4). As portrayed, the adjustment works up to a 20 % perturbation with less than 2 % error margin at most of the altitudes.

Figures 9 and 10 suggest that the VMR forward model simulations pT adjustment given by Eq. (4) may be valid as long as the $pT x - x_0$ "distances" are not greater than 10 K and 20 % accordingly.



(4)

4.2 VMR Jacobians *pT* adjustment

As the forward model has been adjusted to the temperature and pressure of the scene, in a similar manner, the VMR Jacobians can be adjusted using,

$$\mathbf{K}_{\nu}^{*} = \mathbf{K}_{0\nu} + \mathbf{H}_{\tau}(\mathbf{T} - \mathbf{T}_{0}) + \mathbf{H}_{\rho}(\boldsymbol{\rho} - \boldsymbol{\rho}_{0})$$

where \mathbf{K}_{ν}^{*} is the adjusted VMR Jacobian for the actual scene pT conditions (p, T), $\mathbf{K}_{0\nu}$ is the VMR Jacobian computed with the linearisation point conditions (p_0 , T₀) and where H corresponds to the changes of the VMR Jacobian with respect to either temperature or pressure, that is to say,

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$$H_{Tijk} = \frac{\partial K_{v_{ij}}}{\partial T_k} = \frac{\partial}{\partial T_k} \left(\frac{\partial L_i}{\partial v_j} \right) = \frac{\partial^2 L_i}{\partial T_k \partial v_j}$$
(6)
$$H_{pijk} = \frac{\partial K_{v_{ij}}}{\partial p_k} = \frac{\partial}{\partial p_k} \left(\frac{\partial L_i}{\partial v_j} \right) = \frac{\partial^2 L_i}{\partial p_k \partial v_j}$$
(7)

where *i* corresponds to the measurement tangent height while *j* and *k* correspond to perturbation heights. Theoretically, these matrices are dimensionally $m \times n \times n$ (where *m* and *n* are the number of elements in the measurement vector *y* and the atmospheric profile *x*, respectively) but it should be noted that for optically thin cases the atmospheric limb monochromatic radiance, L_v , leaving an stratified atmosphere, is given by (see appendix),

$$L_{v} \simeq \sum_{i} B_{v}(T_{i}) \frac{p_{i}}{RT_{i}} I_{i} \left(\sum_{j} v_{ij} \sigma_{v(ij)}\right)$$

where *i* refers to the *i*th stratified level, B(T) is the Planck function, *p*, *T* and *l* are the pressure, temperature and length along the given path, respectively, and where *j*



(5)

(8)

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corresponds to the *j*th absorber with absorption coefficient σ and volume mixing ratio v. Equations (6) and (7) become,

$$\mathbf{K}_{v} = \frac{\partial L_{i}}{\partial v_{j}} \simeq B(T_{j}) \frac{\rho_{j}}{RT_{j}} I_{j} \sigma_{j}$$

5 and hence

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$$\frac{\partial^{2} L_{i}}{\partial T_{k} \partial v_{j}}, \frac{\partial^{2} L_{i}}{\partial p_{k} \partial v_{j}} = 0 \quad \text{if} \quad k \neq j$$

$$\frac{\partial^{2} L_{i}}{\partial T_{k} \partial v_{i}}, \frac{\partial^{2} L_{i}}{\partial p_{k} \partial v_{j}} \neq 0 \quad \text{if} \quad k = j$$
(10)
(11)

which reduces the size of **H** to $m \times n$.

Although these matrices can be computed by perturbing height by height the temperature or pressure profiles and then subtracting the original VMR Jacobians as computed by the forward model, here a simpler technique was implemented. Noting that, while the VMR Jacobian at the scene conditions will be given by Eq. (9), the linearisation profile VMR Jacobian will be given by,

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$$\mathbf{K}_{0\nu} = \frac{\partial L_{0i}}{\partial \nu_i} \simeq B(T_{0j}) \frac{\rho_{0j}}{RT_{0i}} I_j \sigma_j$$
(12)

Hence, the linearisation point VMR Jacobian can be modified to the pT scene conditions by,

$$\mathbf{K}_{\nu}^{*} = \mathbf{K}_{0\nu} \left[\frac{B(T_{j})\rho_{j}}{B(T_{0j})\rho_{0j}} \frac{T_{0}}{T_{j}} \right]$$

The effect of this simple adjustment can be appreciated in Fig. 11. Note that the only parameter that it is varying in these simulations is the temperature dependence of the



(9)

(13)

VMR Jacobian to isolate its effects in the retrieval. As shown when the Jacobians are not adjusted, the error induced in the retrievals for a 10 K perturbation are up to 30 % while when the Jacobians are adjusted the errors induced are significantly reduced (less than 10 % for a 20 K perturbation).

⁵ Figure 12 is the equivalent figure for pressure increments. As can be seen the errors induced to the retrieval when the VMR Jacobians are adjusted to the pT conditions of the scene (right) are much smaller than when the VMR Jacobians are not adjusted (left).

5 Linear regime summary

- ¹⁰ The results shown so far suggest that the retrievals of pressure and temperature can be treated linearly up to a $x - x_0$ "distance" of 20 % for a 3 % error margin and up to a 10 K difference for a 3 K error margin near the stratopause and less than 0.5 K elsewhere. The CH₄ simulated retrievals shown suggest that, for simulations with perfect pT knowledge, the CH₄ retrievals can be treated linearly up to a 20 % CH₄ $x - x_0$ "dis-
- tance" with a margin error of 2 %. However, since in the linear algorithm the radiative transfer model is not run, the adjustments for the pT difference between the linearisation point used and the scene conditions worsen this linear error estimate. These CH₄ pT adjustments hold up to a 10 K and 20 % (pressure) $x x_0$ "distance" with an error margin of 20 % and 10 %, respectively.
- An analysis of the pressure, temperature and CH_4 variability revealed that to ensure the proximity required between the linearisation and the true profile a climatology with a latitudinal resolution of at least 20° is needed.

It should be noted that so far p, T and v have been treated independently although Eq. (8) predicts that they multiply each other and that these simulations have been performed perturbing the entire profile rather than level by level and hence these linear regimes are only a guidance. Furthermore, signal to noise limitations, overlapping emission by other molecules and nonLTE effect have been ignored. Before the linear



algorithm can produce useful results in a practical application, the influence of these factors needs to be properly taken into account.

6 Error analysis

The total error in the linear retrieval will be given by the sum of the random, systematic, and the linear approximations errors. The random errors are determined by the random noise in the measurements and are given by the diagonal elements of the covariance matrix of the retrieved state (Eq. 2). The systematic errors arise from forward model uncertainties, instrumental issues, and retrieval approximations and will have to be included in a full error analysis, not pursued here.

¹⁰ The linear approximation errors are a function of the deviation between the retrieved profile and the linearisation point. For pressure and temperature, the following equation is suggested,

$$\xi = \frac{\lambda}{\Lambda} (x - x_0) \tag{14}$$

where ξ are the pressure or temperature linear approximation errors either in % or in K, λ is the pressure or temperature error margin (also in % or in K) established in Sect. 3 for a given perturbation, Λ .

For VMR, the linear approximation errors are given by,

$$\xi_{\nu} = \left[\left(\frac{\lambda_{\nu}}{\Lambda_{\nu}} \left(\nu - \nu_{0} \right) \right)^{2} + \left(\frac{\lambda_{\nu T}}{\Lambda_{\nu T}} \left(\mathbf{T} - \mathbf{T}_{0} \right) \right)^{2} + \left(\frac{\lambda_{\nu \rho}}{\Lambda_{\nu \rho}} \left(\frac{\boldsymbol{\rho} - \boldsymbol{\rho}_{0}}{\boldsymbol{\rho}_{0}} \right) \right)^{2} \right]^{1/2}$$
(15)

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where ξ_{ν} is in percentage, λ_{ν} , $\lambda_{\nu T}$ and $\lambda_{\nu p}$ are error margins in percent due to perturbations in VMR, temperature or pressure, Λ_{ν} , $\Lambda_{\nu T}$ and $\Lambda_{\nu p}$.

Figure 13 displays the random error estimates of the linear pT and VMR retrievals, as well as the sum of the random errors and several linear approximation errors computed



for different deviations between the retrieved profile and the linearisation point. The random errors were computed using the same spectral regions as before and assuming the characteristic noise present in the MIPAS radiances (see Sect. 7 for more details). The linear errors were computed using the error margins and perturbations discussed

⁵ in Sect. 5 (for instance, $\lambda_T = 3$ K for a $\Lambda_T = 10$ K, $\lambda_p = 2\%$ for a $\Lambda_p = 20\%$, etc.). Note that the deviations from the linearisation points were assumed constant throughout the profile and as such should be considered only as worst case scenarios.

As shown in Fig. 13, the error in the linear pT or VMR retrieval, at least when the deviation between the retrieved profile equals the maximum deviation allowed, might be too high for scientific results. In an operational retrieval a decision will need to be

be too high for scientific results. In an operational retrieval a decision will need to be made profile by profile to decide whether the linear approximation error was to big, in those cases, the linear retrieval could be used as a first iteration of an iterative scheme, presumably reducing its computing time considerably.

7 Implementation for MIPAS

- As a practical implementation, a *pT* and CH₄ linear retrieval was implemented using data from the Michelson Interferometer for Passive Atmospheric Sounding (MIPAS) instrument (Endemann, 1999; Fisher, 2000) on board the Envisat satellite. MIPAS was a Fourier transform spectrometer conceived to detect limb emission spectra on a global scale. MIPAS covered the infrared spectral region between 685 cm⁻¹ and 2410 cm⁻¹
- ²⁰ (divided into five bands listed in Table 1) where emissions features from many atmospheric species are present.

Originally, between July 2002 and March 2004 MIPAS operated with an spectral sampling of 0.025 cm⁻¹ (referred to as "full resolution" data), however the instrument developed problems with the interferometer mirror slide mechanism and the instrument

²⁵ was switched off. In January 2005 operations were resumed restricting the mirror movement to 40 % of is original maximum displacement corresponding to a reduced spectral sampling (referred to as "optimized resolution" data) of 0.0625 cm⁻¹. MIPAS took



quasi-continuous measurements until April 2012 when the European Space Agency (ESA) lost contact with Envisat.

In the optimized resolution operation, the MIPAS nominal (NOM) measurement mode consists of 27 tangent heights from 6 km to 70 km but MIPAS altitude coverage was

- ⁵ extended well into the thermosphere using special measurements modes (Fig. 14) such as the Middle Atmosphere (MA) mode with 29 tangent heights from 18 km to 102 km and the Upper Atmosphere (UA) mode with 35 tangent heights from 42 km to 172 km. Since April 2007, MIPAS operated in MA mode for one day in every 10, and similarly in the UA mode.
- ¹⁰ Here we use the spectra taken in the MA mode from band A where most of the CO_2 lines occur and from band B where most of the CH_4 lines occur. This viewing mode was selected to apply the algorithm here described because it is in this mode where this algorithm has the more potential.

7.1 MIPAS retrieval algorithms

- ¹⁵ Several algorithms have been developed to retrieve altitude profiles of temperature, pressure and composition from the MIPAS spectra (Ridolfi et al., 2000; von Clarmann et al., 2003; Hoffmann et al., 2005; Carlotti et al., 2006; Dudhia, 2012b). These retrievals exploit the redundancy in MIPAS measurements which allows the selection of subsets of the spectra, known as microwindows. These microwindows maximize the information content as well as minimize the contribution of other parameters that in-
- fluence the spectral lines selected, such as spectrally overlapping species, nonLTE emissions, etc.

The use of microwindows instead of broader spectral intervals helps to reduce the size of the matrices involved in the inverse problem as well as the computing cost of

the forward model. Nevertheless when retrieving at mesospheric heights (using special viewing modes) the number of additional microwindows needed to counteract the low S/N ratio increases the computing time of the retrieval considerably.



In this study the linear retrievals are compared against the MORSE (MIPAS Oxford Retrieval using Sequential Estimation) results. The MORSE inversion strategy is based in a optimal estimation technique (Rodgers, 2000). It was developed at Oxford University to replicate and if possible improve the results of the standard ESA processor. It has been upgraded to retrieve more constituents as well as to retrieve MIPAS data acquired using different viewing modes (MA mode and UA mode) (Dudhia, 2012b).

7.2 *pT* Practical considerations

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7.2.1 Emission from other gases and nonLTE emissions

Due to the high resolution of MIPAS the signal contributions of interfering species can be masked out of the measurements by simply removing the measurement points influenced by their emission. The spectral profiles assumed to be due to "pure" CO₂ emissions were selected using,

$$0.9 \le \frac{L_{\rm CO_2}}{L_{\rm TOT}} \le 1.1$$
 (16)

where L_{CO_2} is a CO₂-only simulated spectrum and L_{TOT} is a simulated spectrum including other gases.

In a similar manner the CO_2 spectral points affected by departures from LTE conditions can be masked out from the measurements. These spectral masks can be constructed using,

$$0.9 \le \frac{L_{\text{nonLTE}}}{L_{\text{LTE}}} \le 1.1 \tag{17}$$

where L_{nonLTE} and L_{LTE} are CO₂-only simulated spectra either in nonLTE or in LTE, respectively.

7.2.2 Hydrostatic constraint

The MIPAS pT retrievals can be constrained using the engineering data defining the line of sight (the pointing information at each tangent height) as an indirect measurement of the state vector. Assuming hydrostatic balance, these engineering altitudes can be used to constrain the temperature and pressure via the hydrostatic equation (Ridolfi et al., 2000),

$$z_{i+1} - z_i = \left(\frac{R}{g}\right) \frac{T_i + T_{i+i}}{2} \log\left(\frac{p_i}{p_{i+1}}\right)$$

where *z* correspond to two tangent profile altitudes (*i* and *i* + 1) with associated tem-¹⁰ peratures *T* and pressures p, *R* is the specific gas constant for dry air and *g* is gravity (as a function of height and latitude).

This constraint is implemented by incorporating the engineering altitudes as part of the measurement vector (with their corresponding uncertainties) and by constructing pT Jacobians differentiating Eq. (18) with respect to either pressure or temperature. This scheme is also used by the operational ESA level 2 retrieval.

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7.2.3 Measurement noise

One problem of the linear retrieval scheme is that, since it uses entire emission bands rather than just microwindows, the analysis of a single tangent height involves several thousand spectral points. Thus, inverting the measurement error covariance matrix (which in the case of MIPAS is not diagonal due to the correlation of adjacent spectral points due to the apodization of the spectra) becomes a time consuming process.

This can be aided exploiting the almost negligible MIPAS measurement error variation through the day. That is to say, the error covariance matrix can be inverted just once each day allowing for the minor changes in each scan by



(18)

$$\mathbf{S}_{y}^{-1} = \mathbf{S}_{y_{0}}^{-1} \left(\frac{\epsilon}{\epsilon_{0}}\right)^{-1}$$

(20)

(21)

where \mathbf{S}_{y} is the updated error covariance matrix, \mathbf{S}_{y_0} is the error covariance matrix constructed with the mean error of the first scan (e_0), and e is the mean error associated with the current scan.

7.3 Multiple linearisation points

Despite the linear regime summarized in Sect. 5, there will be always regions where a single linearisation point will be unlikely to be adequate (like the polar vortex). In such cases, several linearisation points may be used to cover more atmospheric variability and the one that minimizes a χ^2 function (i.e. the square of the residuals between the measurements and the expected spectrum for the linearly adjusted profile) is selected

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as the solution. The χ^2 function suggested for this purpose is,

$$\chi^2 = \frac{\left[\Delta \boldsymbol{y}\right]^T \left[\Delta \boldsymbol{y}\right]}{\boldsymbol{y}^T \boldsymbol{y}}$$

where Δy is given by

 $\Delta \boldsymbol{y} = \boldsymbol{y} - \left[\mathbf{F}(\boldsymbol{x}_0) + \mathbf{K}(\boldsymbol{x} - \boldsymbol{x}_0) \right]$

and the $y^T y$ term is needed to weight the residuals according to each tangent height radiance strength.

Figure 15 shows the linearisation points selected for the 7 June 2007 using Eq. (20). This linearisation point ensemble consists of five profiles corresponding to equator, midlatitude day, midlatitude night, polar summer and polar winter conditions.

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As can be seen, this criterion is selecting for the south pole mostly the polar winter profile while for the north pole mostly the polar summer profile which are the expected choices. For equatorial and midlatitude conditions this criterion is selecting a mixture of the day, night midlatitude and the equatorial profiles. This ensemble of linearisation points will be used to gain further insight into the performance of a linear retrieval.

7.4 Results

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To gain an idea of the performance of the linear retrieval Fig. 16 shows a zonal mean comparison for the 7 June 2007 between the linear retrieval, MORSE and Microwave Limb Sounder (MLS) (Waters et al., 1999, 2006) dataset. It also displays the COSPAR International Reference Atmosphere (CIRA, 1988) climatology for that period.

MLS is an instrument that measures thermal microwave limb emission in five spectral regions from 118 GHz to 2.5 THz. It was launched in July 2004 on board the Aura spacecraft. MLS scans the limb from the ground to about 95 km 3500 times per day. It covers between 82° S and 82° N providing near global coverage. The vertical resolution of this data is about 3 km (Waters et al., 1999, 2006). Its temperature retrievals have been extensively validated (Schwartz et al., 2008).

The linear retrieval results display the structure shown in the climatology data, in MORSE and in MLS with cold temperatures in the polar summer mesopause (> 50° N for this day) and a higher stratopause in the polar winter (<50° S). However, as seen

- in the absolute difference subplot, compared to MORSE and MLS the linear retrieval seems to overestimate the temperatures for pressure between 0.1 and 0.03 hPa at all latitudes and underestimate the temperature for smaller pressures. For pressures greater than 0.1 hPa there seems to be no significant difference (less than 3 K) between the linear, MORSE and the MLS results.
- Figure 17 shows a profile by profile seasonal comparison between the linear retrieval, MORSE and MLS data. For the linear–MLS comparisons, two profiles were considered to be coincident if they were within 500 km radius and 6 h. The seasonal bins compared correspond to latitude bins between 90° S–50° S, 20° S–20° N and 50° N–90° N



and sorted into winter, equatorial and summer according to the date of the profile. These bins were selected to cover most of the atmospheric variability.

As shown in Fig. 17, there seems to be an overestimation of the temperature by the linear retrieval for pressures lower than around 0.1 hPa. This overestimation varies with

5 season and it is around 3–5 K for the MORSE comparisons and up to around 5–10 K for MLS ones. For greater pressures the bias found is less than 3 K.

In addition to the mean bias, this Figure also show the standard deviation of the profiles compared to gain an idea of the variability of the data. For pressures greater than 0.5 hPa, the variability found in most seasonal bins (for the MORSE and MLS comparisons) is approximately 3 K which is the expected error margin for a 10 K $x - x_0$ "distance".

These comparisons might suggest that the linear retrieval is not as precise as needed for pressures lower than 0.01 hPa. However, it should be pointed out that the performance of the linear retrieval will improve when a proper linearisation point climatology is used. Even if when using this climatology the retrievals are not as precise

as needed, the linear results are presumably going to be closer to the true state than the climatology, so that they may be used as a first iteration in the iterative schemes potentially saving computing time.

7.5 Further VMR practical considerations

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²⁰ As for the *pT* retrievals, the VMR retrievals will be influenced by overlapping emissions of other gases and by nonLTE processes. These influences can be reduced using spectral masks as the ones described in Sect. 7.2.1.

The linearly retrieved VMR altitude range was truncated above the altitude where the estimated random error was greater than 30%. Above these altitudes, the retrieved

values are dominated by noise. Although coaddition could be used to retrieve gases concentrations at upper mesospheric altitudes, this approach is not pursued here.

Currently, for each pT linearisation point there is one CH₄ profile associated with it. However, it should be noted that, tentatively, for each pT linearisation point, different



CH₄ scenarios can be simulated. At each scan, a retrieval will be performed for each scenario and the one that best minimizes the χ^2 suggested in Sect. 7.3 (Eq. 20) used as solution.

7.6 VMR results

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⁵ Figure 18 displays the zonal mean CH₄ comparison between the linear retrieval and MORSE for the 7 June 2007. It also displays the IG climatology (Remedios et al., 2007) for that period. As shown, the linear retrieval displays the structure found in the climatology and in MORSE with higher values around the equator, however, as seen in the percentage difference subplot, the linear retrieval seems to overestimate the CH₄ concentration by around 20 % for pressures between 10 and 1 hPa.

Figure 19 displays a profile by profile comparison between the linear retrieval and MORSE for the same days as in Sect. 7.4. It seems that the linear retrieval is always overestimating the CH_4 concentrations. This overestimation may be due to the lack of proper linearisation points, especially considering the temperature requirements to fall within the linear regime (see Sect. 4).

Note that although the current linear VMR retrieval is probably not as good as the MORSE results, its results should improve once a more reasonable linearisation point climatology is used.

Since CH₄ does not possess pure rotational bands and hence no microwave signature in the spectrum, a comparison against MLS is not possible. A comparison against other instruments was consider unnecessary at this stage due to the clear overestimation (considering the differences against the climatology and MORSE) of the linear CH₄ retrievals.



8 Retrieval speed

An important feature of the linear retrieval is its speed. Since there is no need to run the radiative transfer model, once the error covariance matrices are inverted (only needed once a day), the main time consuming processes are ingesting the data into the re-⁵ trieval and inverting the multiple linearisation points (five in the current configuration of the scheme). Using the same computer to run MORSE and the Linear retrieval, it has been determined that the linear retrieval is faster than MORSE at least by a factor of 20 (reducing the processing time of a day of middle atmosphere data from days to a couple of hours on a single CPU) when retrieving *pT* and CH₄. Furthermore, it should be noted that speed of the linear retrieval can be improved by retrieving only some of the linearisation points selecting them according to the time and latitude of the scene.

9 Conclusions

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An alternative algorithm to retrieve temperature, pressure and composition from limb viewing infrared spectrometers has been introduced. This algorithm exploits the linear properties of an optically thin path making it possible to perform the inversion without re-running the radiative transfer model as long as the linearisation point is close enough to the scene conditions. This new algorithm can use the whole spectral band of the molecule rather than a small subset of it (the microwindows) and therefore, theoretically, leads to a higher precision.

²⁰ It was shown that the retrievals of pressure and temperature can be treated linearly up to a 20 % $x - x_0$ "pressure distance" for a 3 % error margin and up to 10 K "distance" for a 3 K error margin near the stratopause and less than 0.5 K elsewhere.

Assuming perfect pT knowledge, it was shown that the CH₄ concentrations can be treated linearly up to a $x - x_0$ "distance" of 20 % CH₄ concentration for a 2 % error margin. However, since the linear algorithm requires precomputed simulated spectra and Jacobians for the pT and CH₄ conditions of the linearisation points, adjustments



for the pT dependence of the simulated spectra and Jacobians have been proposed. These CH₄ pT adjustments hold up to a 10 K and 20 % pressure $x - x_0$ "distance" with an error margin of 20 % and 10 %, respectively.

In order to fulfill all these restrictions (i.e. the required proximity between the linearisation point and the true profile) a pT and CH_4 climatology with a latitudinal resolution of at least 20° of latitude is needed. Furthermore it was also shown how to select the best retrievals from different linearisation points for regions where a single linearisation point is unlikely to be adequate.

This retrieval scheme was implemented using MIPAS radiances and compared to the results of the MIPAS iterative scheme MORSE and the results of MLS.

The pT comparisons revealed that linear retrieval is within a 3K range difference most of the time for pressures greater than 0.01 hPa while for lower pressures there seems to be an overestimation of the temperature by the linear scheme up to 5K. The CH₄ comparisons showed that the linear retrieval is overestimating the CH₄ concentrations. This overestimation may be due to the lack of proper linearisation points climatology, especially considering the CH₄ retrieval temperature requirements to fall within the linear regime.

Even if the linear retrieval is not as precise as required, it has proven its potential to be so close to the actual scene conditions that they may be used as the first iteration of the iterative schemes, thereby reducing the processing time.

Appendix A

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Linearising the forward model

The radiance I_{ν} measured by an instrument is given by

$$I_{v} = \iint_{v \,\omega} F_{v} \mathrm{d}v * \Phi \mathrm{d}\omega * L_{v}$$

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(A1)

where v is wavenumber, ω is solid angle, F_v is the instrument spectral response, Φ is the field of view and L_v is the radiance that reaches the satellite given by the radiative transfer equation

$$L_{\nu} = L_{0\nu}\tau_{\nu}(x) + \int_{0}^{x} B_{\nu}(T) \frac{d}{dx'}(\tau_{\nu}(x')) dx'$$

where x is distance along the path, $L_{0\nu}$ represents the background emission at the end of the path (negligible for the infrared region of the spectrum), $B_{\nu}(T)$ is the Planck function, which describes the thermal emission of the atmosphere (assuming Local Thermodynamic Equilibrium and neglecting scattering) and τ_{ν} is the transmittance of the path from the point x' to the spacecraft (x = 0). This quantity is defined by

$$\tau_{\nu}(x) = \exp\left(-\chi(x_{\nu})\right) \tag{A3}$$
$$\chi_{\nu}(x) = \int_{0}^{x} \sum_{j} \nu(x')\sigma_{\nu}(x')\rho(x')dx' \tag{A4}$$

where χ_v is the optical thickness of the path, *j* corresponds to the *j*th absorber with absorption coefficient σ_v , volume mixing ratio *v* and density ρ .

Assuming a stratified atmosphere, the numerical integration of Eqs. (A2) and (A3) can be formulated as

$$L_{\nu} = L_{0\nu}\tau_{0\nu} + \sum_{i} B_{\nu}(T_{i})[\tau_{\nu(i-1)} - \tau_{\nu(i)}]$$

$$\tau_{\nu} = \exp\left(-\sum_{i} \sum_{j} \rho_{i} I_{i} v_{ij} \sigma_{ij}\right)$$
(A5)
(A6)

where *i* refers to the *i*th stratified level, τ_i corresponds to the transmittance from the satellite up the given stratified level and I_i is the length along the path in that level.



(A2)

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Assuming that the path is optically thin ($\chi \ll 1$), it can be shown that Eq. (A5) can be approximated by,

$$L_{\nu} \simeq \sum_{i} B_{\nu}(T_{i}) \rho_{i} I_{i} \left(\sum_{j} v_{ij} \sigma_{\nu(ij)} \right)$$

5 and using the ideal gas law by

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$$L_{v} \simeq \sum_{i} B_{v}(T_{i}) \frac{p_{i}}{RT_{i}} I_{i} \left(\sum_{j} v_{ij} \sigma_{v(ij)} \right)$$

which is the "most" linear scenario of the radiative transfer (except in the microwave region of the spectra where $B_v(T) \simeq T$ hence much of the temperature dependence cancels) and still involves the product of atmospheric variations depending on *T*, *p*, *v* and σ_v .

However, assuming that the spectral sampling of the instrument ($\simeq 0.5 \text{ cm}^{-1}$) is coarse relative to the actual line width ($\simeq 0.001 \text{ cm}^{-1}$ in the upper stratosphere) we can assume σ_v independent of p, T or v. Under this scenario, Eq. (8) implies that when the path is optically thin, the forward model is linear with respect to p, v and $B_v(T)/T$.

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(A7)

(A8)

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| MIPAS | S Wavenumber range | NESR* $[nW (cm^2 sr cm^{-1})^{-1}]$ | Principal atmospheric |
|-------|---------------------|-------------------------------------|---|
| band | [cm ⁻¹] | | emitting species |
| A | 685–970 | 32 | $\begin{array}{c} {\rm CO}_2,{\rm O}_3,{\rm H}_2{\rm O},{\rm HNO}_3\\ {\rm CO}_2,{\rm O}_3,{\rm H}_2{\rm O},{\rm N}_2{\rm O}\\ {\rm H}_2{\rm O},{\rm CH}_4,{\rm N}_2{\rm O}\\ {\rm NO}_2,{\rm H}_2{\rm O}\\ {\rm CO}_2,{\rm O}_3,{\rm H}_2{\rm O},{\rm N}_2{\rm O},{\rm NO},{\rm CO}\\ \end{array}$ |
| AB | 1020–1170 | 25 | |
| B | 1215–1500 | 13 | |
| C | 1570–1750 | 4 | |
| D | 1820–2410 | 2.6 | |

* Noise equivalent spectral radiance (NESR) for MIPAS operating at the "optimized resolution" (Dudhia, 2007).





Fig. 1. (Top) CO_2 spectra for midlatitude conditions. (Bottom) CH_4 spectra for midlatitude conditions. The dashed black line shows the typical MIPAS noise equivalent spectral radiance (NESR) value.





Fig. 2. Linearly retrieved pressure for different increments of the entire pressure profile (left) and percentage difference between the actual and retrieved values (right).





Fig. 3. Linearly retrieved temperature for different increments of the entire temperature profile (left) and absolute difference between the actual and retrieved values (right).





Fig. 4. Linearly retrieved temperature via the Planck function for different increments of the entire temperature profile (left) and absolute difference between the actual and retrieved values (right).













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Fig. 7. Temperature variation about a global (left), 60° (middle) and 30° (right) latitude bin profiles for June, July and August for 2007, 2008 and 2009.









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Fig. 9. Linearly retrieved CH_4 concentrations (after adjusting the forward model) for different temperature increments of the entire profile (left) and percentage difference between the actual and retrieved values (right).

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Fig. 10. Linearly retrieved CH_4 concentrations (after adjusting the forward model) for different pressure increments of the entire profile (left) and percentage difference between the actual and retrieved values (right).





different increments of the entire temperature profile when using the linearisation point Jacobians (left) and when adjusting them (using Eq. 13) to the new temperature conditions (right).

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Fig. 12. Percentage difference between the actual and linearly retrieved CH_4 concentrations for different increments of the entire pressure profile when using the linearisation point Jacobians (left) and when adjusting them (using Eq. 13) to the new pressure conditions (right).

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Fig. 13. Error estimates for pressure, temperature and CH_4 concentration. Aqua lines show the random errors while orange, purple and blue lines show the sum of the random errors and the linear approximation errors for different deviations between the retrieved profile and the linearization point. The random errors correspond to estimates for midlatitude conditions but other conditions are similar.





Fig. 14. MIPAS optimized resolution operation measurement modes.





Fig. 15. Linearisation points selected within an ensemble (equatorial, midlatitude day, midlatitude night, polar summer and polar winter profiles) using Eq. (20) for the retrieval of 7 June 2007.





Fig. 16. Temperature zonal mean for the 7 June 2007 for climatology, MORSE, MLS and the MIPAS linear retrieval. The top row displays climatology, MORSE and MLS while the bottom row displays the MIPAS linear retrieval results as well as the absolute difference between the retrievals (linear–MORSE and linear–MLS). Note that the absolute difference contours are not equally spaced.





Fig. 17. Mean temperature biases (left) and scatter (right) between the MIPAS linear retrieval, MORSE (solid) and MLS (dashed). The gray dashed line shows a 3 K difference. The number of profiles compared in each bin for the MLS comparison were 40, 2918 and 3159 for the equatorial, winter and summer case, respectively.





Fig. 18. CH_4 zonal mean for the 7 June 2007 for climatology, MORSE and the MIPAS linear retrieval. The top row displays climatology and MORSE while the bottom row displays the MI-PAS linear retrieval results as well as the percentage difference between the two retrievals (linear-MORSE).





Fig. 19. Mean CH_4 bias (left) and scatter (right) between the MIPAS linear retrieval and MORSE. The gray dotted line shows the expected equatorial CH_4 variability. The gray dashed line shows a 3K difference.

