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Key Point:

- A significant improvement in the modeled isentropic mixing in the stratosphere is obtained if the vertical resolution is improved there

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Diagnosing Mixing Properties in Model Simulations for CH₄ in the Stratosphere

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Abstract Isentropic mixing properties in the stratosphere modeled by the forward calculation of an inverse model (TM5-4DVAR) are evaluated against Michelson Interferometer for Passive Atmospheric Sounding (MIPAS) and Microwave Limb Sounder (MLS) observations. The isentropic mixing processes are separated into large-scale stirring described by the “equivalent length” and small-scale diffusion described by the diffusivity. Compared to the measurements, we find that the modeled stirring is not strong enough and that the small-scale diffusivity is too large. TM5-4DVAR produces excessive mixing-induced poleward flux for stratospheric CH₄. The flux convergence presents negative biases in the tropics and positive biases in the polar regions. The biases cannot be reduced by improving the horizontal resolution only. Modeled isentropic mixing depends on the horizontal as well as the vertical resolution of the model. An increase in vertical resolution reduces numerical diffusion of the model in the vertical. The decreased vertical diffusion leads to reduction in the modeled isentropic diffusivity. Biases in modeled total column-averaged mixing ratios of CH₄ are significant for both models with a coarse vertical resolution of 4° × 6° × 25 (and 2° × 3° × 25, 1° × 1° × 25) and an improved one of 1° × 1° × 40. They are estimated to be 7–14 and 3–7 ppb in the winter extratropics under the assumption that isentropic mixing is dominant over vertical transport on a time scale of 3 days. Correspondingly, resulted biases in inverted CH₄ surface emissions are estimated to be 0.5–1 and 0.2–0.5 mg/m²/hr, respectively, in the extratropics.

1. Introduction

Inverse modeling is widely used to infer methane (CH₄) emissions from measured concentrations. The most frequently used measurements include CH₄ mixing ratios at the surface, and total column-averaged mixing ratios from satellite-based sensors (e.g., SCIAMACHY described in Bovensmann et al., 1999, and GOSAT in Kuze et al., 2016). Compared to in situ surface measurements, satellite-based instruments provide valuable information due to their better spatial coverage. This larger coverage is particularly important for tropical regions where in situ measurements are sparse, such as in South America (Pison et al., 2013). However, inaccuracies in the models prevent correct extraction of the information on the sources by inverse models. For example, retrieved surface emissions change when applying different physical parameterization or different vertical resolutions (Locatelli et al., 2015). Specially, model errors in the stratosphere could be aliased into wrong tropospheric emissions when the inverse models use total column-averaged mixing ratio measurements. An approach to reduce the impact of model deficiencies in the stratosphere is to use satellites measurements that have small sensitivities for the stratospheric CH₄. For example, Payne et al. (2009) describe a method that extracts the tropospheric CH₄ concentrations with peak sensitivities around 500 hPa from the Tropospheric Emission Spectrometer (TES) measurements. Crevoisier et al. (2013) and Xiong et al. (2013) retrieve a middle-to-upper tropospheric CH₄ concentration from the Infrared Atmospheric Sounding Interferometer (IASI) (where the most sensitive layer is between 100–600 hPa in the tropics and 200–750 hPa in the extratropics). Worden et al. (2015) present lower tropospheric CH₄ retrieved by using a combination of TES and GOSAT sensors with an estimated accuracy of 10–30 ppb for monthly averages. However, all these CH₄ products still have a stratospheric contribution even if their highest sensitivities are in the troposphere.

CH₄ is emitted at the surface and transported into the stratosphere mainly in the tropics. In the stratosphere, CH₄ is distributed globally by the residual circulation and isentropic mixing processes induced by wave breaking events, for example, planetary wave breaking (Mcintyre & Palmer, 1983). Chemical oxidation destroys CH₄ mostly in the upper stratosphere and results in a strong decline of CH₄ concentrations with altitudes above the tropopause. However, in the stratosphere transport barriers exist, such as the subtropical barrier (Chen et al., 1994a; Neu, 2003; Trepte & Hitchman, 1992) and the polar vortex barrier (Chen et al., 1994b; Steinhorst et al., 2005). Isentropic mixing processes contribute to (1) horizontally uniform tracer distributions in the surf zone, which is a region strongly stirred by planetary wave breaking in the winter stratosphere as defined by McIntyre and Palmer (1983), and (2) strong gradients of tracer concentrations at ends of the surf zone.

In the stratosphere the transport of chemical tracers across the subtropical and polar barrier occurs as filamentary structures induced by wave breaking events (Chen, 1994; Chen et al., 1994b; Kalicinsky et al., 2013; Randel et al., 1993). However, usually Eulerian models do not represent filament structures as good as Lagrangian models (e.g., Khosrawi et al., 2005) and thus could have difficulties in capturing transport across barriers. For example, CTMs (Chemistry Transport Model) using a Lagrangian transport scheme give much better performances in representing large CH₄ gradients across these transport barriers in the stratosphere compared to the Eulerian based models (Hoppe et al., 2014; Stenke et al., 2009). Model deficiencies in the stratosphere need to be understood to improve their performance and to obtain an estimate to what degree the CH₄ emissions from inversion studies can be trusted when total column-averaged mixing ratio measurements are used.

In this work the forward mode of the inverse model TM5-4DVAR, a model frequently used for inversions of CH₄ emissions, is evaluated in the stratosphere. Precisely, the mixing properties of the model in the stratosphere are analyzed in comparison to satellite observations. The analysis is based on the modified Lagrangian-mean (MLM) approach (Allen & Nakamura, 2001; Haynes & Shuckburgh, 2000; Nakamura, 1996; Nakamura & Ma, 1997). Further, the model biases in the stratosphere are estimated. The influence of these model biases on retrieved surface emissions is briefly discussed.

2. Data and Method

2.1. Model and Measurements

TM5-4DVAR is a four-dimensional data assimilation system for inverse modeling of atmospheric methane emissions (Bergamaschi et al., 2013, 2015; Meirink et al., 2008). The system is based on the TM5 atmospheric transport model (Krol et al., 2005). The model is driven by ERA-Interim reanalysis data that has a resolution of $1^\circ \times 1^\circ \times 60$ (latitudinal and longitudinal resolution, and number of vertical levels). In this study, we only use the forward mode of the model to simulate spatiotemporal CH₄ fields to prescribed surface emissions. The surface emissions of CH₄ are taken from the MACC (Monitoring Atmospheric Composition and Climate) data set. The model resolutions under testing are $4^\circ \times 6^\circ \times 25$, $2^\circ \times 3^\circ \times 25$, $1^\circ \times 1^\circ \times 25$, $1^\circ \times 1^\circ \times 40$ and $1^\circ \times 1^\circ \times 63$. Pressures at interface of the model levels are tabulated in Table 1. TM5-4DVAR automatically maps the ERA-Interim meteorology data to the defined model resolution.

The resolution $4^\circ \times 6^\circ \times 25$ is a standard setup in previous studies for investigating the surface emissions of CH₄ (Bergamaschi et al., 2013, 2015). Compared to the 25 vertical layers setup, the additional layers in 40 vertical layers setup are all located above 165 hPa. The setup with 40 vertical layers has the same vertical layers as the original ERA-Interim data sets between 165 and 1.8 hPa. The 63 vertical layers setup has double vertical resolution in that pressure range. The parameterized chemical sinks of CH₄ include reactions with tropospheric hydroxyl (OH) radicals (Spivakovsky et al., 2000) and stratospheric radicals from ECHAM5 (Bergamaschi et al., 2013; Joeckel et al., 2006). The concentration fields of the radicals are prescribed monthly with no interannual changes.

The measured stratospheric CH₄ profiles during 2009–2011 are taken from MIPAS (Michelson Interferometer for Passive Atmospheric Sounding) instrument (Fischer et al., 2008). MIPAS samples the middle and upper atmosphere along polar orbits with a swath width of 3×30 km. There are 14.3 orbits per day and about 73 limb scans along one orbit. The primary geophysical parameters of interest are vertical

Table 1
Pressures at the Interface of 25, 40, and 63 Model Levels Assuming a Surface Pressure of 1,013.25 hPa

p (hPa)/63	p (hPa)/40	p (hPa)/25
0.00	0.00	0.00
0.96	0.96	0.96
1.81	1.81	
2.08		
2.35	2.35	
2.67		
2.98	2.98	2.98
3.36		
3.74	3.74	
4.19		
4.65	4.65	
5.20		
5.76	5.76	
6.44		
7.13	7.13	7.13
7.98		
8.84	8.84	
9.89		
10.95	10.95	
12.26		
13.56	13.56	
15.19		
16.81	16.81	16.81
18.81		
20.82	20.82	
23.31		
25.80	25.80	
28.88		
31.96	31.96	
35.78		
39.60	39.60	39.60
44.34		
49.07	49.07	
54.62		
60.18	60.18	60.18
66.62		
73.07	73.07	
80.40		
87.73	87.73	87.73
95.98		
104.23	104.23	
113.42		
122.61	122.61	122.61
132.76		
142.90	142.90	
153.99		
165.09	165.09	165.09
177.11		
215.02	215.02	215.02
272.04	272.04	272.04
336.01	336.01	336.01
406.08	406.08	406.08
480.72	480.72	480.72
557.88	557.88	557.88
635.18	635.18	635.18
710.12	710.12	710.12
780.17	780.17	780.17
843.07	843.07	843.07
896.90	896.90	896.90

profiles of atmospheric pressure, temperatures, and volume mixing ratios of about 25 trace constituents. The product used here is the operational V7 data processed by the European Space Agency (ESA).

For comparison of the model with the measurements, modeled 3-D fields are sampled at each measurement location and time through spatiotemporal interpolations.

2.2. Estimation of Trace Gas Fields From Point Observations

The upcoming analysis requires a global spatiotemporal trace gas concentration field along isentropic surfaces. We construct such a field by fitting a truncated spherical harmonics expansion to the observed/simulated concentrations at the measurement locations using a weighted least squares approach. Spherical harmonics are a set of Fourier-like basis functions defined on a spherical surface. The series of spherical harmonics is truncated with the R20 (rhomboidal) truncation (e.g., Daley & Bourassa, 1978), which selects the large-scale components of the expansion. The estimated CH_4 field is a globally defined approximation on the isentropic surface with a spatial resolution of about $3.6^\circ \times 6^\circ$.

This spectral approximation is performed for both the model and the measurement data. One approximation is estimated for each 10 days period at each isentrope. The MIPAS data are weighted according to the measurement uncertainties. The weights for the model are unitary. The 10 days period is required to accumulate sufficient MIPAS observations for a stable estimate of the spherical approximation at the chosen resolution, R20. Actually, the accumulated MIPAS data of 7 days suffices to stably estimate the R20 truncated approximation. However, to validate the diagnostic method with data from the MLS (Microwave Limb Sounder; see section 3.1) a longer aggregation period is required due to the larger measurement noise of MLS.

For the following analysis the concentration field is sampled on isentropic surfaces on a $1^\circ \times 1^\circ$ grid. The spectral approximation of the model data is solely based on the data at the measurement locations of the satellite instead of using the full model output. This approach allows a direct comparison between the model data and the observations. At the same time certain features of the modeled CH_4 field could be lost. A comparison of the spectral approximations from the full model simulation and from the subset of observations shows that both approaches lead to virtually identical results. This result implies that our approach to compare model data and measurements is reasonable.

2.3. Theory and Analysis Method

The meridional mixing in the stratosphere is investigated by analyzing the mixing ratio of long-lived trace gases along isentropic surfaces. The meridional transport in the stratosphere is characterized by regions of strong mixing separated by transport barriers. Zonal averaging approaches are less suited for the analysis of transport barriers, because exact positions of transport barriers are permanently disturbed by atmospheric waves. A more adequate approach is the modified Lagrangian-mean framework.

Table 1
Continued

p (hPa)/63	p (hPa)/40	p (hPa)/25
940.32	940.32	940.32
972.75	972.75	972.75
994.50	994.50	994.50
1,007.02	1,007.02	1,007.02
1,013.00	1,013.00	1,013.00

Consider a two-dimensional distribution of a trace gas mixing ratio $q = q(x, y, t)$. The mixing ratio evolves according to the advection-diffusion equation, that is (assuming constant air density),

$$\frac{\partial q}{\partial t} + \nabla \cdot (\vec{v}q - k\nabla q) = s, \quad (1)$$

where $\vec{v}q$ is the advective flux and $-k\nabla q$ is the diffusive flux with constant diffusion coefficient and s is term describing sources and sinks.

The modified Lagrangian-mean approach observes the physical system in a different variable. Following Nakamura (1996), the areal integral of a function f over the area enclosed by the contour line of some trace gas mixing ratio q by

$$\langle f \rangle_q = \int_{q'(x,y) \leq q} f ds.$$

Note that the area enclosed by this contour line is thus given by $\langle q \rangle = \langle 1 \rangle_q$. In this way $A(q)$ is a function that maps the mixing ratio q to the area enclosed by the contour q . Generally, the inverse relationship $q = q(A)$ can be established as well. A snapshot of the atmospheric mixing ratio q and the corresponding function $q(A)$ are illustrated in Figure 1. In the absence of diffusion and sources/sinks the function $q(A)$ is not changed by advection. Advection can deform or even split the contours of the trace gas field but it does not change the function $q(A)$. This is due to the fact the advection conserves the mixing ratio and the areal coordinate A is of Lagrangian type; that is, it follows the flow of the atmosphere if the wind is divergence-free and hence area preserving. These conditions nearly apply to the stratosphere. The system observed on the areal coordinate is only changed by nonconservative processes such as diffusion and sources/sinks.

The advection-diffusion equation expressed with respect to A has the form (cp. Nakamura, 1996)

$$\frac{\partial q}{\partial t}(A, t) = \frac{\partial}{\partial A} \left(k L_e^2(A, t) \frac{\partial q}{\partial A}(A, t) \right) + \hat{s}, \quad (2)$$

where $\hat{s}$ is source/sink averaged along the contour line and $L_e(A, t)$ is called equivalent length given by

$$L_e^2(A, t) = \frac{\partial \langle |\nabla q|^2 \rangle_{q(A)} / \partial A}{(\partial q / \partial A)^2}.$$

In Equation 2, large values of equivalent length increase the diffusive flux and eventually lead to flat gradients. Small values act as a barrier and gradients can form (cp. Figure 1). The equivalent length is approximately the length of the contour line and proportional to the gradient of the mixing ratio across the contour. Atmospheric wave disturbances can enhance L_e through enlarging the length and strengthening the gradients. This action of wave disturbances is referred to as stirring (Haynes, 2004). So equivalent length acts as an indicator of the strength of stirring.

For spherical surfaces it can be convenient to express the areal coordinate A in terms of equivalent latitude ϕ_e . The corresponding equivalent latitude to the area A is the latitudinal circle centered at the pole whose enclosed area is A , that is, $\phi_e = \arcsin(1 - A/2\pi r^2)$, where r is the radius of the sphere. In the spherical coordinate the MLM formalism has the form (cp. Nakamura & Ma, 1997)

$$\frac{\partial q(\phi_e, \theta, t)}{\partial t} = \frac{1}{\cos \phi_e} \frac{\partial}{\partial \phi_e} \left(\cos \phi_e \frac{k_h L_e^2}{(2\pi r \cos \phi_e)^2} \frac{\partial q}{\partial \phi_e} \right) + \text{others}, \quad (3)$$

where k_h is the small-scale isentropic diffusion coefficient. Equivalent length L_e represents the strength of stirring. Both coefficients together are also referred to as the effective diffusivity $k_{eff} =$

Figure 1. Snapshot of the Northern Hemispheric CH_4 distribution on an isentropic surface (top). The same distribution observed in the areal variable A and in equivalent latitude (bottom). The marked points correspond to the highlighted contours in the upper panel.

$k_h L_e^2 / L_0^2$, where $L_0 = 2\pi r \cos \phi_e$ is the length of the circle on the spherical surface at latitude ϕ_e . In Equation 3 the first term on the right-hand side describes mixing-induced flux convergence of the trace gas and is evaluated in this study. The “others” term includes contributions by diabatic motion and sources/sinks and is not concerned.

It should be noted that large-scale stirring has different properties than small-scale diffusion. Two processes cannot compensate for each other even though they are described by the same term in the MLM theory. Large-scale distortions stretch tracer contours but do not smooth the tracer gradients. Small-scale diffusion degrades the tracer gradient during the transport by decreasing high tracer concentrations and increasing lower tracer concentrations. On the isentropes the small-scale diffusion transports tracers more isotropically and uniformly than the large-scale disturbances. For example, wave disturbances can take air masses near the polar vortex boundary to lower latitudes where the tracer concentrations are very different from the polar regions in just a few days. Such transport exchanges only small amounts of trace gas with latitudes passed by the air mass. This effect cannot be achieved by a diffusion process. The L_e is tightly related to wave activities and enlarged by wave breaking events as will be shown in section 3. At the same time, the small-scale diffusion decreases L_e through smoothing tracer distribution.

A normalization as suggested in Nakamura and Ma (1997) is applied to the equivalent length, that is, $L_e = \ln(L_e^2 / L_0^2)$. In this study, the normalized equivalent length is used as a measure for the strength of stirring processes that are resolved by the R20 truncated approximation. Hereafter the L_e is just referred to as L_e for convenience. For CH₄ with maximum mixing ratio in the tropical regions and generally decreasing values toward the poles we evaluate each hemisphere separately.

In order to quantify CH₄ biases induced by mixing processes the small-scale isentropic diffusivity k_h must be estimated. For the observational data k_h includes contributions from eddy motions unresolved by R20 truncated approximation, turbulences and the molecular diffusion. The k_h in the model data includes eddies resolved by the original model resolution but unresolved by R20 truncated approximation, parameterized subgrid mixing, and the numerical diffusion. Previous studies estimate the average diffusivity from the evolution of the globally averaged tracer variance (Abernathy et al., 2010; Allen & Nakamura, 2001; Leibensperger & Plumb, 2014; Marshall et al., 2006). When restricting the averaging to the stratospheric region between the isentropic layers 500 and 1200 K, the spatially averaged variance equation for that region reads (see Appendix A for a detailed derivation)

$$\begin{aligned} \frac{1}{2} \frac{d\overline{q^2}}{dt} - \frac{1}{2} \int_{\theta=500K} \sigma \dot{\theta} q^2 \frac{ds}{M} + \frac{1}{2} \int_{\theta=1200K} \sigma \dot{\theta} q^2 \frac{ds}{M} = -k_\theta \left(\frac{1}{2} \int_{\theta=500K} \sigma \frac{\partial q^2}{\partial \theta} \frac{ds}{M} - \frac{1}{2} \int_{\theta=1200K} \sigma \frac{\partial q^2}{\partial \theta} \frac{ds}{M} + \overline{\left(\frac{\partial q}{\partial \theta} \right)^2} \right) \\ - k_h |\nabla_h q|^2 - \frac{\overline{q^2}}{\tau_{chem}}. \end{aligned} \quad (4)$$

In this equation q is the trace gas mixing ratio, $\sigma = -g^{-1} \partial p / \partial \theta$ is the potential temperature coordinate density, $\dot{\theta}$ denotes the diabatic heating rate, M is the air mass between the two isentropes, ∇_h denotes the components of the gradient operator within the isentropic surface, k_θ is the small-scale diabatic diffusivity, and τ_{chem} is the averaged chemical lifetime of the trace gas in this region. The overbar represents the mass-weighted mean over the volume enclosed by the isentropic layers 500 and 1200 K. The tracer variance equation lists the processes that change the mean variance of the tracer in the defined region. Advective and diffusive fluxes over the vertical boundaries can transport variance in and out of the region. Within the region the variance is reduced by horizontal and vertical diffusion as well as by chemical oxidation.

For the modeled CH₄ the boundary conditions and chemical sinks are known and can be calculated explicitly. Then the tracer variance equation is of the form $f(t) = k_\theta a(t) + k_h b(t)$. The coefficient of k_θ has a much larger variation range than that of k_h . So first, k_θ is estimated by linear regression of k_θ and a constant c in $f(t) = k_\theta a(t) + c$. In a second step k_h is estimated through linear regression against $b(t)$ with fixed k_θ .

The estimation for the MIPAS data follows a different procedure. Time series of the retrieval have discontinuities due to changes in the gain of the detector for B-band of MIPAS. Evaluation of the time derivative of the averaged CH₄ variance is interfered significantly. To reduce influences of the discontinuities, Equation 4

is normalized through being divided by $\overline{q^2}$. The diffusivities are estimated from the difference of the normalized tracer variance equations for CH₄ and N₂O. With $q_1 = \overline{q_{CH_4}^2}$ and $q_2 = \overline{q_{N_2O}^2}$ the difference equation include the time derivative of the variance in a form $\frac{d}{dt} \ln\left(\frac{q_1}{q_2}\right)$. Since both trace gases are retrieved from the B-band the fraction q_1/q_2 reduces the impact of the discontinuities significantly. The diffusivities can now be estimated using the above mentioned linear regression.

Diabatic heating rates $\dot{\theta}$ from two different reanalysis data sets, ERA-Interim and MERRA-2, are applied for the MIPAS data. It is revealed that the estimated isentropic and diabatic diffusivities are $(1.48 \times 10^5 \text{ m}^2/\text{s}, 0.0001 \text{ K}^2/\text{s})$ when applying the ERA-Interim diabatic heating rates and $(2.17 \times 10^5 \text{ m}^2/\text{s}, 0.0003 \text{ K}^2/\text{s})$ when applying those of the MERRA-2. For the following analysis, the value $k_h = 2.17 \times 10^5 \text{ m}^2/\text{s}$ is used because they are closer to model results. In this way, we try to estimate lower limits in the models biases.

3. Validation of the L_e as a Measure for Stirring Strength

3.1. L_e Magnitude Validation

The L_e is determined by the structure of the tracer contour for a specific moment. For inert tracers the tracer contours are in equilibrium with the flow and share similar geometries (Haynes & Shuckburgh, 2000). Consequently, the L_e derived from different inert tracers are the same. As long as the local lifetimes of tracers having a sink in the stratosphere are much longer than the timescale of isentropic mixing processes, L_e derived from their distribution does not depend on the tracer and their destruction details. In the stratosphere both CH₄ and N₂O have long lifetimes and their contours at a certain altitude are mainly the results of air motions. So the L_e derived from these tracers is a measure of stirring strength.

Figure 2 gives L_e at 600 and 1100 K during the period 2009–2011 derived from CH₄ and N₂O mixing ratios measured by MIPAS and MLS. The MIPAS measured CH₄ and N₂O give very similar but not completely the same values for L_e . The reason is that the MIPAS measurement errors are different for the two gases. Validations of MIPAS retrievals with in situ measurements (Wetzel et al., 2016) show a positive bias for CH₄ of 5% at 20 km ($\approx 450 \text{ K}$), which slowly increases to 10% at 40 km ($\approx 1,300 \text{ K}$) with about 5% standard deviation. For N₂O the validation shows a positive bias of 15% between 20 and 32 km ($\approx 900 \text{ K}$). Above 32 km the bias of the N₂O largely oscillates from 15% to 35% with standard deviation as large as 40%. As L_e depends on tracer contours the larger standard deviation of 40% in the validation of N₂O retrievals indicates that the L_e derived from MIPAS measured N₂O mixing ratios suffer from more measurement noise. The L_e distributions calculated with CLaMS model data as described in Pommrich et al. (2014) for 450–1500 K between CH₄ and N₂O are very similar (not shown). The MLS measured N₂O gives similar L_e as MIPAS but the differences between the instruments are larger than between CH₄ and N₂O from MIPAS data. Comparisons of MIPAS and MLS measured N₂O with ACE-FTS measured N₂O (Sheese et al., 2017) reveal that there is no mean bias between these instruments from 20 to 45 km but the MLS measurements present larger variabilities. The 1- σ relative differences are 10% at 20 km, 30% at 38 km, and 60% at 45 km for MIPAS and 10% at 20 km and 100% at 38 km for MLS. The correlations between MIPAS and ACE-FTS stay near 0.9 in the range 20–45 km, but those between MLS and ACE-FTS decrease with respect to altitudes above 30 km and become about 0.2 at 45 km. In this study the L_e derived from MIPAS CH₄ are used to evaluate the modeled one since ACE-FTS measurements are usually considered to be highly accurate.

3.2. Relation Between L_e and Wave Activities

In Eulerian (zonal) mean formulation zonal inhomogeneity in any quantity is referred as wave disturbance. Large wave disturbances usually are associated with wave breaking events in the extratropical stratosphere. Here, *Eliassen-Palm flux* (E-P flux) (Holton, 2004, p325) is calculated to reveal connection between the L_e and wave activities. The quasi-geostrophic E-P flux is given by

$$\vec{F} = (F_\phi, F_z) = r \cos \phi e^{-z/H} \left(\frac{-\overline{u'v'}, f\overline{v'\theta'}}{\overline{\theta}_z} \right),$$

where r is the Earth's radius and $z = H \ln(p_0/p)$ is the log-pressure coordinate and H is a constant atmospheric scale height, p_0 and p are constant reference pressure and atmospheric pressure, u and v are

Figure 2. Temporal evolutions of L_e at 600 K (left) and 1100 K (right) for the period 2009–2011 derived from MIPAS measured CH_4 (top) and N_2O (middle) mixing ratios, and from MLS measured N_2O (lower) mixing ratios.

zonal and meridional disturbance velocities, ϕ is the latitude, and f is the Coriolis parameter; the overbar means zonal average, and the primes indicate departure from zonal mean. The E-P flux vector represents the transport flux of wave activities.

The divergence of the E-P flux, $\nabla \cdot \vec{F} / (r \cos \phi e^{-z/H})$, represents the acceleration of the zonally averaged zonal wind by wave disturbances and drives the residual circulation. According to the noninteraction theorem (Andrews et al., 1987; Edmon et al., 1980) for adiabatic conditions steady waves cannot change the zonal mean state of the atmosphere because the divergence of the E-P flux equals to 0. Only waves under development or dissipation produce nonzero E-P flux divergence, which occurs in wave breaking events. For quasi-geostrophic flow the divergence of the E-P flux equals to the meridional eddy flux of potential vorticity $\overline{v'q'}$ (Edmon, 1980; Palmer, 1982). When scaled by the meridional gradient of the zonal mean potential vorticity the eddy diffusivity in Eulerian mean formulation $K_{yy} = -\overline{v'q'} / \frac{\partial \bar{q}}{r \partial \phi}$ is produced. K_{yy} is comparable to L_e that composes the effective diffusivity in the tracer coordinate (Abalos et al., 2016).

Figure 3 compares MIPAS measured L_e to the temporal evolutions of eddy diffusivity and the vertical component of the E-P flux (scaled by H) calculated using ERA-Interim reanalysis data. The resolution of the reanalysis data is $2.5^\circ \times 3.75^\circ \times 60$. All quantities are averaged over the latitudes 30° – 50°S . The similarity between their seasonal cycles indicates that L_e is enlarged by wave disturbances. In order to know how much wave activities can be attributed to planetary scale waves the eddy diffusivity and E-P flux are calculated for both total disturbances and those with zonal wavenumber ≤ 4 . The results indicate that almost all wave activities result from planetary scale disturbances at the altitudes 450–1200 K (approximate 20–38 km).

In Abalos et al. (2016) the eddy diffusivity and L_e for a passive tracer is compared in an altitude range 5–30 km and found high similarity between them. They also recognized a high similarity between eddy fluxes of the potential vorticity and the tracer. As a summary, L_e is closely related to the wave disturbances enlarged by wave breaking events.

4. Results and Discussions

4.1. Evaluation of Modeled Mixing Strength in the Stratosphere

As explained in section 2 L_e calculated from the R20 truncated CH_4 fields gives the strength of stirring processes which are resolvable under the R20 resolution. Large L_e indicates that the CH_4 contours on the isentropic surface are stretched and disturbed strongly. Figure 4 shows measured and modeled temporal evolutions of L_e at 600 and 1100 K in both the Northern and Southern Hemispheres for the period 2009–2011. In the stratosphere one of the important dynamical structures is the surf zone, which

Figure 3. Temporal evolutions of the eddy diffusivity in Eulerian mean formulation K_{yy} (colors) and the vertical component of the scaled E-P flux vector (black contours, in unit m^2/s^2) of total disturbances (top panel). Disturbances with wavenumber ≤ 4 averaged over the latitudes 30–50°S (middle panel) and the temporal evolution of MIPAS measured L_e averaged over the region 30–50°S (lower panel).

corresponds to the region with large L_e in the extratropics. The extratropical L_e shows seasonal cycles with maxima in the winter and minima in the summer. This seasonal variation is due to the seasonal cycle of the amount of the tropospheric wave propagating up into the stratosphere. Planetary waves can propagate upwards only in the westerlies and the zonal winds of the stratosphere change seasonally. The westerlies are dominant in the winter and easterlies in the summer (Andrews et al., 1987). During the transition from the westerlies to the easterlies the large L_e region extends into the polar region due to strong disturbances associated with the breakup of the polar vortex (Allen & Nakamura, 2001; Haynes & Shuckburgh, 2000).

Compared to the measurements the model with the standard resolution $4^\circ \times 6^\circ \times 25$ underestimates stirring strength, especially in the extratropics. In the model the L_e in the surf zone shows a smaller latitudinal coverage and a shorter temporal extent. Compared to Figure 2 the L_e differences between the model and the measurements are larger than the differences in the different measured tracers. This result indicates that the L_e differences between model and measurements result from model deficiencies. Since the equivalent length is

Figure 4. Temporal evolutions of L_e at 600 K (left column) and 1100 K (right column) from the MIPAS measurements and TM5-4DVAR simulations with resolutions $4^\circ \times 6^\circ \times 25$, $2^\circ \times 3^\circ \times 25$, $1^\circ \times 1^\circ \times 25$, $1^\circ \times 1^\circ \times 40$, and $1^\circ \times 1^\circ \times 63$. The top panel shows results for the Northern Hemisphere and the lower panel for the Southern Hemisphere.

enlarged by wave disturbances and dissipated by the small-scale diffusion the differences between measurements and model output suggest that the deficiencies in the modeled L_e should be attributed to dynamics and numerical diffusion of the transport scheme adopted by the model. The numerical diffusion mainly depends on the resolution of the model for a specific transport scheme. Figure 4 shows that increasing the horizontal resolution improves the representation for L_e only to a minor extent. However, increasing the vertical layers from 25 to 40 at the same time results in a better agreement for L_e between

Table 2

Estimates of Isentropic and Diabatic Diffusivity, Globally Averaged L_e^2/L_0^2 , and Effective Diffusivity k_{eff} , the Pe Number Assuming a Eddy Speed of 10–100 m/s and a Typical Eddy Size of 1,000 km for TM5-4DVAR and MIPAS Measurements

	k_h (m ² /s)	k_θ (K ² /s)	L_e^2/L_0^2	k_{eff} (m ² /s)	Pe
TM5-4DVAR 4° × 6° × 25	19.72×10^5	0.0008	2.71	53.35×10^5	5–50
TM5-4DVAR 2° × 3° × 25	15.58×10^5	0.0007	2.87	44.68×10^5	6–64
TM5-4DVAR 1° × 1° × 25	17.11×10^5	0.0009	3.03	51.90×10^5	6–58
TM5-4DVAR 1° × 1° × 40	6.64×10^5	0.0002	3.94	26.19×10^5	15–151
TM5-4DVAR 1° × 1° × 63	5.51×10^5	0.0001	4.20	23.12×10^5	18–181
MIPAS with ERA-Interim diabatic heating rate	1.48×10^5	0.0001	5.34	7.90×10^5	68–676
MIPAS with MERRA-2 diabatic heating rate	2.17×10^5	0.0003	5.34	11.58×10^5	46–460

the model output and the measurements, especially in the northern hemisphere. The improvement is more significant at higher altitudes in both of the hemispheres.

For the small-scale diffusivity displayed in Table 2 the modeled isentropic diffusivity k_h (5.51 – 19.72×10^5 m²/s) is higher than the measured one (2.17×10^5 m²/s). As mentioned above, the modeled isentropic diffusivity results from small-scale dynamics with scales between the model and the R20 resolution, and the horizontal numerical diffusion. The modeled k_h can be attributed to the horizontal numerical diffusion of the transport scheme and small-scale dynamics. The k_h stays more or less the same for all the models with 25 vertical layers and decreases only when the number of the vertical layers is increased to 40. This result indicates that an improved horizontal resolution does not necessarily decrease the horizontal numerical diffusion proportionally when it is determined by the vertical diffusion set by the low vertical resolution of the model. According to Haynes and Anglade (1997), the combination of vertical shear and horizontal strain leads to a large number of thin, sloping structures in stratospheric tracer fields. These structures have a large aspect ratio (ratio of horizontal to vertical length scale) of about 250 in the lower stratosphere. Consequently, horizontal diffusivity is enhanced by an amount $k_z \alpha^2$ where α is the aspect ratio and k_z is vertical diffusivity. The theory of Haynes and Anglade (1997) is consistent with the result we obtained; an increase of vertical resolution reduces horizontal numerical diffusion and the estimated isentropic diffusivity k_h . When the number of vertical layers is increased to 63 the vertical diffusivity decreases to half that of 40 vertical layers but the isentropic diffusivity changes only to a minor extent. This indicates that isentropic diffusivity is not dominated by vertical diffusion for models with at least 40 layers. The remaining higher isentropic diffusivity in the 40-layer model compared to that of the measurements can be attributed to horizontal numerical diffusion.

The magnitude of the equivalent length L_e depends on the dynamics and the small-scale diffusivity. The dynamics applied in the model are created from the 1° × 1° × 60 ERA-Interim data. With a very limited vertical resolution in the stratosphere in the model with only 25 vertical layers the 3-D cascade of tracer variance is not adequately resolved. Severe truncation of tracer variance leads to biases in the modeled L_e . Impact of the small-scale diffusivity on L_e can be explained with a theoretical argument according to Shuckburgh and Haynes (2003) and Marshall et al. (2006). The relationship between L_e and k_h depends on the relative magnitude of the advective time scale (L/V) and the diffusive time scale (L^2/k_h), summarized as the Peclet number (Pe) VL/k_h (V , L are typical scales for eddy speed and size). In the large Pe limit, the enhancement of the length of tracer contours by eddy disturbances dominates over the small-scale diffusion. In this case k_{eff} is independent of k_h and L_e^2 is proportional to $1/k_h$ (here L_e^2 refers to the original definition in Equation 2, not to the normalized one). The k_{eff} scales like $k_h^{1/2}$ when Pe is small and approaches k_h as Pe becomes even smaller. Consequently, L_e^2 scales like $1/k_h^{1/2}$ and approaches L_0^2 , respectively. Pe ranges about 10–100 if assuming an eddy speed of 10–100 m/s, a k_h of 1.0×10^6 m²/s and a typical eddy size of 1,000 km. For these values of Pe L_e^2 scales like k_h^α with $-1\%3C \alpha \%3C -0.5$ according to Marshall et al. (2006), so the equivalent length strongly depends on the magnitude of the small-scale diffusivity. The theoretical expectation is consistent with the results here; the models underestimate L_e and overestimate k_h ; the modeled L_e increases significantly only when k_h decreases significantly.

One of the uncertainties in the validation of the modeled isentropic diffusivity is that diabatic heating rates in the real atmosphere are unknown and have to be replaced with those from reanalysis datasets for the

Figure 5. CH_4 mixing ratios (the top panels), meridional fluxes of CH_4 due to isentropic mixing processes $-k_{\text{eff}} \frac{\partial q_{\text{CH}_4}}{r \partial \phi_e}$ (the middle panels), and the flux convergence $\frac{1}{\cos \phi_e} \frac{\partial}{\partial \phi_e} \left(\cos \phi_e k_{\text{eff}} \frac{\partial q_{\text{CH}_4}}{r \partial \phi_e} \right)$ (the lowermost panels) averaged over 2009–2011. The left, middle, and right columns show results on 500, 800, and 1100 K isentropic surfaces, respectively. Different lines indicate results for MIPAS (black solid), TM5-4DVAR with a resolution of $4^\circ \times 6^\circ \times 25$ (black dash), $2^\circ \times 3^\circ \times 25$ (black dotted line), $1^\circ \times 1^\circ \times 25$ (gray solid), $1^\circ \times 1^\circ \times 40$ (gray dash), and $1^\circ \times 1^\circ \times 63$ (gray dotted line).

derivation of k_h from measurements. The diabatic heating contributes flows of tracer variance into the region bounded by 500 and 1200 K surfaces at the lower boundary (500 K) and out of the region at the upper boundary (1200 K) (see Equation 4). Because of the exponential decrease of air density with respect to altitude, the total contribution by the diabatic heating tends to increase the tracer variance in this region. As a result, k_h derived from the measurements is underestimated if the residual circulation in the reanalysis data is too slow and overestimated if it is too fast.

4.2. Estimation of CH_4 Biases Due to Isentropic Mixing

Figure 5 shows measured and TM5-4DVAR modeled CH_4 mixing ratios, the meridional flux, and the flux convergence of CH_4 due to isentropic mixing averaged over 2009–2011 at 500, 800, and 1100 K. Compared to the MIPAS measurements the model underestimates CH_4 mixing ratios in the tropics and overestimates them in the polar regions. The systematic underestimates of CH_4 mixing ratios in the tropics by the models at 800 and 1100 K could be due to underestimates of vertical advection associated with the residual circulation. The poleward fluxes of CH_4 due to isentropic mixing are higher in the model than in the measurements. The modeled excessive poleward fluxes transport tropical CH_4 to high latitudes too fast. The flux convergence of CH_4 from isentropic mixing is negative in the tropics and positive in the extratropics. One exception is around 60°S on the isentropic surface of 500 K surface where the modeled flux convergence derivative changes significantly from negative values to positive values. This anomaly can be explained with the polar transport barrier located there. The tropical side shows a large negative flux convergence due to a fast increase of tracer fluxes along latitudes and the polar side shows a large positive flux convergence due to a decrease of tracer fluxes. The models give more negative mixing-induced flux convergence in the tropics and more positive mixing-induced flux convergence in the extratropics. When the number of vertical layers is fixed to 25, the magnitudes of these biases change to a minor extent for models with various horizontal resolutions ($4^\circ \times 6^\circ$ to $1^\circ \times 1^\circ$). Significant reduction in the biases of the modeled fluxes and flux convergence is obtained by improving the vertical resolution.

For yearly averages the modeled temporal tendencies of CH_4 mixing ratios due to isentropic mixing are similar to the measured ones as the number of vertical levels of the model increases from 25 to 40 as shown in

Figure 6. Temporal evolution of mixing-induced flux convergence of CH_4 at 800 K in the measurement and models biases. The top panel gives the flux convergence for MIPAS and the following panels show the biases in the modeled flux convergence under the resolutions $4^\circ \times 6^\circ \times 25$, $2^\circ \times 3^\circ \times 25$, $1^\circ \times 1^\circ \times 25$, $1^\circ \times 1^\circ \times 40$, and $1^\circ \times 1^\circ \times 63$.

Figure 5. However, isentropic mixing is closely related to wave activities that present seasonal variations. The winter hemisphere is disturbed strongly by wave activities but the summer hemisphere is calm. Correspondingly, isentropic mixing is stronger in the winter hemisphere than in the summer hemisphere. To see the seasonal variation of the model errors, Figure 6 shows the evolution of the flux convergence derived from the measurements and the model biases in the flux convergence at 800 K for a period of 3 years. It is clear that the flux convergence is significant in the winter only in the northern extratropics. For the model with a resolution of $1^\circ \times 1^\circ \times 40$ biases in the flux convergence are as large as 0.5 ppb/hr compared to about 0.2 ppb/hr for the yearly averages in Figure 5. This difference is important since the time series of measurements is an input to inverse models.

Isentropic mixing enhances CH_4 mixing ratios in the extratropics and diabatic sinking moves the added CH_4 downward into the troposphere. The diabatic sinking is slow compared to the isentropic mixing. Accordingly, accumulation of model biases due to isentropic mixing can take place only on a short time scale, that is, a few days. Assuming isentropic mixing is dominant over the vertical transport on a time scale of 3 days and model biases in the flux convergence of CH_4 are 0.5–1.0 ppb/hr for the model of $4^\circ \times 6^\circ \times 25$ (and $2^\circ \times 3^\circ \times 25$, $1^\circ \times 1^\circ \times 25$) and 0.2–0.5 ppb/hr for the model of $1^\circ \times 1^\circ \times 40$ in the extratropical stratosphere. The corresponding biases in the total column-averaged mixing ratio of CH_4 are 7–14 and 3–7 ppb, respectively, assuming that the stratosphere takes one fifth of the mass of the atmosphere. These biases in the total column-averaged mixing ratio are important for an inverse model that infers CH_4 emissions from measurements of total column-averaged CH_4 mixing ratio. For example, latitudinal/monthly bias correction of a few tens of ppb is derived during inverse modeling of global CH_4 emissions using SCIAMACHY satellite retrievals (Bergamaschi et al., 2009). The diagnosed deficiencies in the atmospheric transport scheme of TM5-4DVAR can lead to substantial biases in the retrieved trace gas emissions. We estimate emission biases by assuming that biases in the column-integrated flux convergence need to be compensated by surface emissions. The biases in the modeled flux convergence can then be translated to biases in inverted CH_4 emissions by

the factor $(1/5) \cdot (16/29) \cdot (10^5/9.8) \cdot (10^{-9}) \cdot (10^6)$ in the extratropics. The factor transforms the stratospheric flux convergence in ppb/h to surface emissions in $\text{mg/m}^2/\text{hr}$. In this factor, $1/5$ is the approximate mass fraction of the stratosphere in the extratropics, $16/29$ is ratio of molecular mass of CH_4 and air, 10^5 Pa is surface pressure, 9.8 m/s^2 is the gravitational constant, 10^{-9} is ppb, and 10^6 transform coefficient from kg to mg. Estimated biases in the surface emission are 0.5–1 $\text{mg/m}^2/\text{hr}$ for the model of $4^\circ \times 6^\circ \times 25$ (and $2^\circ \times 3^\circ \times 25$, $1^\circ \times 1^\circ \times 25$) and 0.2–0.5 $\text{mg/m}^2/\text{hr}$ for the model of $1^\circ \times 1^\circ \times 40$ in the extratropics. The estimated biases in CH_4 surface emissions are large compared to CH_4 emission rates of 0–2 $\text{mg/m}^2/\text{hr}$ in the northern extratropics and those below 0.4 $\text{mg/m}^2/\text{hr}$ in the southern extratropics (Bergamaschi et al., 2009). In the above estimations we assume a vertically uniform sampling weight for satellite measurements. For SCIAMACHY and GOSAT the sampling weights are larger in the troposphere than in the stratosphere. In addition, the inverse model estimates surface emissions based on measurements while taking a priori emissions into account. It means that only part of the information in the inverted emissions is based on measurements. The fraction of the information from measurements depends on measurement noise and the a priori information used by the inverse model. When taking sampling weights and a priori emissions into account the estimated biases in the surface emissions will decrease.

5. Conclusions

In this work the TM5-4DVAR, an inverse model for CH₄, is evaluated against MIPAS and MLS in terms of mixing properties in the stratosphere. For our studies we used only the forward part of the model. The modified Lagrangian-mean formalism is used as the analysis framework. As a result, the mixing processes are separated into large-scale stirring measured by the equivalent length L_e and small-scale diffusion measured by the isentropic diffusivity k_h . A spherical harmonics expansion with the R20 truncation is applied to extract the large-scale information from the measurements and the model outputs.

Comparisons of the measured L_e to the simulations reveal that MIPAS and MLS measured CH₄ and N₂O give similar L_e . Generally, the model underestimates L_e and overestimates k_h compared to the measurements. The magnitude of the bias in the modeled mixing parameters depends on the model resolution. Compared to the model with the frequently applied coarse resolution $4^\circ \times 6^\circ \times 25$ (latitude $\times$ longitude $\times$ vertical layers) improving the horizontal resolution alone does not reduce the model biases in the L_e and the k_h . When both, the horizontal and vertical resolutions are increased, the bias in the modeled k_h decreases from tenfold to threefold. The bias on the modeled L_e decreases significantly associated with the reduced k_h . It is argued that the reduction of the bias in the modeled k_h is due to reduced numerical diffusion in the vertical component of the applied transport scheme.

With these derived mixing parameters at hand, isentropic mixing-induced meridional fluxes and the flux convergence of CH₄ are estimated for both the measurements and the models. It is revealed that the TM5-4DVAR simulates an excessive poleward flux and, consequently, too negative flux convergence in the tropics and too positive one in the extratropics. For TM5-4DVAR with resolutions of $4^\circ \times 6^\circ \times 25$, $2^\circ \times 3^\circ \times 25$ and $1^\circ \times 1^\circ \times 25$ (latitude $\times$ longitude $\times$ vertical layers) the biases do not reduce substantially. For TM5-4DVAR with resolution of $1^\circ \times 1^\circ \times 40$ (additional vertical layers locate within the stratosphere and the vertical resolution is the same as in ERA-Interim there) the biases reduce globally. A further increase of the vertical model resolution in the stratosphere does not further reduce the biases. Biases in CH₄ surface emissions caused by the unrealistic isentropic mixing in the model are estimated to be 0.5–1 mg/m²/hr for the model of $4^\circ \times 6^\circ \times 25$ (and $2^\circ \times 3^\circ \times 25$, $1^\circ \times 1^\circ \times 25$) and 0.2–0.5 mg/m²/hr for the model of $1^\circ \times 1^\circ \times 40$ in the extratropics.

For upcoming versions of the transport model, we recommend to improve the parameterization of small-scale diffusion in the stratosphere and to take steps toward reducing numerical diffusion.

Appendix A: Derivation of the Tracer Variance Equation

The tracer variance equation (Equation 4) is derived here. Starting from the advection-diffusion equation for the mixing ratio q of some trace gas, that is,

$$\frac{\partial \sigma q}{\partial t} + \nabla \cdot (\sigma q \vec{v}) - \sigma K \nabla q = -\sigma \frac{q}{\tau_{chem}}. \quad (\text{A1})$$

Note that K includes isentropic diffusivity and diabatic diffusivity. The τ_{chem} is lifetime of the trace gas due to the chemical oxidation. Multiplying the equation by q gives

$$q \frac{\partial \sigma q}{\partial t} + q \nabla \cdot (\sigma q \vec{v}) = q \nabla \cdot (\sigma K \nabla q) - \sigma \frac{q^2}{\tau_{chem}}. \quad (\text{A2})$$

Rearranging each term in Equation A2, that is,

$$\begin{aligned} q \frac{\partial \sigma q}{\partial t} &= \frac{1}{2} \frac{\partial \sigma q^2}{\partial t} + \frac{1}{2} q^2 \frac{\partial \sigma}{\partial t} \\ q \nabla \cdot (\sigma q \vec{v}) &= q^2 \nabla \cdot (\sigma \vec{v}) + \frac{1}{2} \sigma \vec{v} \cdot \nabla q^2 = \frac{1}{2} q^2 \nabla \cdot (\sigma \vec{v}) + \frac{1}{2} \nabla \cdot (\sigma q^2 \vec{v}) \\ q \nabla (\sigma K \nabla q) &= \nabla (\sigma q K \nabla q) - (\sigma K \nabla q) \nabla q = \frac{1}{2} \nabla (\sigma K \nabla q^2) - \sigma k_h |\nabla_h q|^2 - \sigma k_\theta \left(\frac{\partial q}{\partial \theta} \right)^2, \end{aligned}$$

with this rearrangement Equation A2 can be transformed to

$$\frac{1}{2} \frac{\partial \sigma q^2}{\partial t} + \frac{1}{2} q^2 \left(\frac{\partial \sigma}{\partial t} + \nabla(\sigma \vec{v}) \right) + \frac{1}{2} \nabla(\sigma q^2 \vec{v}) = \frac{1}{2} \nabla(\sigma K \nabla q^2) - \sigma k_h |\nabla_h q|^2 - \sigma k_\theta \left(\frac{\partial q}{\partial \theta} \right)^2 - \sigma \frac{q^2}{\tau_{chem}}.$$

Integrating over the stratospheric region between the isentropic layers 500 and 1200 K and using the divergence theorem and by denoting the air mass weighted mean by an overbar gives

$$\frac{1}{2} \frac{d \bar{q}^2}{dt} = -\frac{1}{2M} \int_S \sigma q^2 \vec{v} \cdot \vec{n} dS + \frac{1}{2M} \int_S \sigma (K \nabla q^2) \cdot \vec{n} dS - \overline{k_h |\nabla_h q|^2} - \overline{k_\theta \left(\frac{\partial q}{\partial \theta} \right)^2} - \frac{\bar{q}^2}{\tau_{chem}}, \quad (A3)$$

where S denotes the boundary and $\vec{n}$ the outward pointing normal of S . In Equation A3 the diffusion coefficients and chemical lifetime are averages in following ways

$$k_h^* = \frac{\overline{k_h |\nabla_h q|^2}}{|\overline{\nabla_h q}|^2}, \quad k_\theta^* = \frac{\overline{k_\theta \left(\frac{\partial q}{\partial \theta} \right)^2}}{\left(\overline{\frac{\partial q}{\partial \theta}} \right)^2}, \quad \frac{1}{\tau_{chem}^*} = \frac{\overline{q^2 / \tau_{chem}}}{\bar{q}^2}$$

over the volume weighted by air mass and square of the gradient of the tracer mixing ratio. Temporal variation of M compared to that of $\bar{q}^2$ is overlooked in arriving at Equation A3. Recognizing that S is just the lower and the upper bounding isentrope gives the final result (cp. Equation 4).

Data Availability Statement

The MIPAS data are operational V7 data processed by ESA, which is publicly available at <https://earth.esa.int/web/guest/data-access/view-data-product/-/article/mipas-atmospheric-pressure-temperature-data-constituents-profiles-1547> with registration. The MLS data are available at https://acdsc.gesdisc.eosdis.nasa.gov/data/Aura_MLS_Level2/ website. ERA-Interim described in Dee et al. (2011, doi: 10.1002/qj.828) and MERRA-2 in Gelaro et al. (2017, doi: 10.1175/JCLI-D-16-0758.1) are publicly available. The model TM5-4DVAR is v1.4 and available online (<https://sourceforge.net/projects/tm5/>).

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References

- Abalos, M., Randel, W. J., & Birner, T. (2016). Phase-speed spectra of eddy tracer fluxes linked to isentropic stirring and mixing in the upper troposphere and lower stratosphere. *Journal of the Atmospheric Sciences*, 73(12), 4711–4730. <https://doi.org/10.1175/JAS-D-16-0167.1>
- Abernathy, R., Marshall, J., Mazloff, M., & Shuckburgh, E. (2010). Enhancement of mesoscale eddy stirring at steering levels in the Southern Ocean. *Journal of Physical Oceanography*, 40(1), 170–184. <https://doi.org/10.1175/2009JPO4201.1>
- Allen, D. R., & Nakamura, N. (2001). A seasonal climatology of effective diffusivity in the stratosphere. *Journal of Geophysical Research*, 106(D8), 7917–7935. <https://doi.org/10.1029/2000JD900717>
- Andrews, D. G., Holton, J. R., & Leovy, C. B. (1987). *Middle atmospheric dynamics*, *International Geophysics Series* (Vol. 40, p. 489). New York: Academic Press.
- Bergamaschi, P., Corazza, M., Karstens, U., Athanassiadou, M., Thompson, R. L., Pison, I., et al. (2015). Top-down estimates of European CH₄ and N₂O emissions based on four different inverse models, *Atmospheric Chemistry and Physics*, 15, 715–736. <https://doi.org/10.5194/acp-15-715-2015>
- Bergamaschi, P., Franzenberg, C., Meirink, J. F., Krol, M., Villani, M. G., Houweling, S., et al. (2009). Inverse modeling of global and regional CH₄ emissions using SCIAMACHY satellite retrievals. *Journal of Geophysical Research*, 114, D22301. <https://doi.org/10.1029/2009JD012287>
- Bergamaschi, P., Houweling, S., Segers, A., Krol, M., Frankenberg, C., Scheepmaker, R. A., et al. (2013). Atmospheric CH₄ in the first decade of the 21st century: Inverse modeling analysis using SCIAMACHY satellite retrievals and NOAA surface measurements. *Journal of Geophysical Research: Atmospheres*, 118, 7350–7369. <https://doi.org/10.1002/jgrd.50480>
- Bovensmann, H., Burrows, J. P., Buchwitz, M., Frerick, J., Noël, S., Rozanov, V. V., et al. (1999). SCIAMACHY—Mission objectives and measurement modes. *Journal of the Atmospheric Sciences*, 56(2), 127–150. [https://doi.org/10.1175/1520-0469\(1999\)056<0127:SMOAMM>3.0.CO;2](https://doi.org/10.1175/1520-0469(1999)056<0127:SMOAMM>3.0.CO;2)
- Chen, P. (1994). The permeability of the Antarctic vortex edge. *Journal of Geophysical Research*, 99(D10), 20563. <https://doi.org/10.1029/94JD01754>
- Chen, P., Holton, J. R., O'Neill, A., & Swinbank, R. (1994a). Isentropic mass exchange between the tropics and extratropics in the stratosphere. *Journal of the Atmospheric Sciences*, 51(20), 3006–3018. [https://doi.org/10.1175/1520-0469\(1994\)051<3006:IMEBT%3E2.0.CO;2](https://doi.org/10.1175/1520-0469(1994)051<3006:IMEBT%3E2.0.CO;2)
- Chen, P., Holton, J. R., O'Neill, A., & Swinbank, R. (1994b). Quasi-horizontal transport and mixing in the Antarctic stratosphere. *Journal of Geophysical Research*, 99(D8), 16851. <https://doi.org/10.1029/94JD00784>
- Crevoisier, C., Nobileau, D., Armante, R., Crépeau, L., Machida, T., Sawa, Y., et al. (2013). The 2007–2011 evolution of tropical methane in the mid-troposphere as seen from space by MetOp-A/IASI. *Atmospheric Chemistry and Physics*, 13(8), 4279–4289. <https://doi.org/10.5194/acp-13-4279-2013>
- Daley, R., & Bourassa, Y. (1978). Rhomboidal versus triangular spherical harmonic truncation: Some verification statistics. *Atmosphere-Ocean*, 16(2), 187–196. <https://doi.org/10.1080/07055900.1978.9649026>

- Dee, D. P., Uppala, S. M., Simmons, A. J., Berrisford, P., Poli, P., Kobayashi, S., et al. (2011). The ERA-Interim reanalysis: Configuration and performance of the data assimilation system. *Quarterly Journal of the Royal Meteorological Society*, 137(656), 553–597. <https://doi.org/10.1002/qj.828>
- Edmon, H. J., Hoskins, B. J., & McIntyre, M. E. (1980). Eliassen-Palm cross sections for the troposphere. *Journal of the Atmospheric Sciences*, 37(12), 2600–2616. [https://doi.org/10.1175/1520-0469\(1980\)037%3C2600:EPCSFT%3E2.0.CO;2](https://doi.org/10.1175/1520-0469(1980)037%3C2600:EPCSFT%3E2.0.CO;2)
- Fischer, H., Birk, M., Blom, C., Carli, B., Carlotti, M., von Clarmann, T., et al. (2008). MIPAS: An instrument for atmospheric and climate research. *Atmospheric Chemistry and Physics*, 8(8), 2151–2188. <https://doi.org/10.5194/acp-8-2151-2008>
- Gelaro, R., McCarty, W., Suárez, M. J., Todling, R., Molod, A., Takacs, L., et al. (2017). The modern-era retrospective analysis for research and applications, version 2 (MERRA-2). *Journal of Climate*, 30(14), 5419–5454. <https://doi.org/10.1175/jcli-d-16-0758.1>
- Haynes, P. (2004). *Transport and mixing in the atmosphere, proceedings of 21st ICTAM congress*. Berlin, Germany: Kluwer Academic Publishers, August 15–21.
- Haynes, P., & Anglade, L. (1997). The vertical-scale cascade in atmospheric tracers due to large-scale differential advection. *Journal of the Atmospheric Sciences*, 54(9), 1121–1136. [https://doi.org/10.1175/1520-0469\(1997\)054%3C1121:TVSCIA%3E2.0.CO;2](https://doi.org/10.1175/1520-0469(1997)054%3C1121:TVSCIA%3E2.0.CO;2)
- Haynes, P., & Shuckburgh, E. (2000). Effective diffusivity as a diagnostic of atmospheric transport: 1. Stratosphere. *Journal of Geophysical Research*, 105(D18), 22,777–22,794. <https://doi.org/10.1029/2000JD900093>
- Holton, J. R. (2004). *An introduction to dynamic meteorology, fourth edition, International Geophysics Series* (Vol. 88). New York: Elsevier Academic Press.
- Hoppe, C. M., Hoffmann, L., Konopka, P., Grooß, J.-U., Ploeger, F., Günther, G., et al. (2014). The implementation of the CLaMS Lagrangian transport core into the chemistry climate model EMAC 2.40.1: Application on age of air and transport of long-lived trace species. *Geoscientific Model Development*, 7(6), 2639–2651. <https://doi.org/10.5194/gmd-7-2639-2014>
- Joeckel, P., Tost, H., Pozzer, A., Brühl, C., Buchholz, J., Ganzeveld, L., et al. (2006). The atmospheric chemistry general circulation model ECHAM5/MESy1: Consistent simulation of ozone from the surface to the mesosphere. *Atmospheric Chemistry and Physics*, 6(12), 5067–5104. <https://doi.org/10.5194/acp-6-5067-2006>
- Kalicinsky, C., Grooß, J.-U., Günther, G., Ungermann, J., Blank, J., Höfer, S., et al. (2013). Observations of filamentary structures near the vortex edge in the Arctic winter lower stratosphere. *Atmospheric Chemistry and Physics*, 13(21), 10,859–10,871. <https://doi.org/10.5194/acp-13-10859-2013>
- Khosrawi, F., Grooß, J.-U., Müller, R., Konopka, P., Kouker, W., Ruhnke, R., et al. (2005). Intercomparison between Lagrangian and Eulerian simulations of the development of mid-latitude streamers as observed by CRISTA. *Atmospheric Chemistry and Physics*, 5(1), 85–95. <https://doi.org/10.5194/acp-5-85-2005>
- Krol, M., Houweling, S., Bregman, B., van den Broek, M., Segers, A., van Velthoven, P., et al. (2005). The two-way nested global chemistry-transport zoom model TM5: Algorithm and applications. *Atmospheric Chemistry and Physics*, 5(2), 417–432. <https://doi.org/10.5194/acp-5-417-2005>
- Kuze, A., Suto, H., Shiomi, K., Kawakami, S., Tanaka, M., Ueda, Y., et al. (2016). Update on GOSAT TANSOFTS performance, operations, and data products after more than 6 years in space. *Atmospheric Measurement Techniques*, 9(6), 2445–2461. <https://doi.org/10.5194/amt-9-2445-2016>
- Leibensperger, E. M., & Plumb, R. A. (2014). Effective diffusivity in baroclinic flow. *Journal of the Atmospheric Sciences*, 71(3), 972–984. <https://doi.org/10.1175/JAS-D-13-0217.1>
- Locatelli, R., Bousquet, P., Saunio, M., Chevallier, F., & Cressot, C. (2015). Sensitivity of the recent methane budget to LMDz sub-grid-scale physical parameterizations. *Atmospheric Chemistry and Physics*, 15, 9765–9780. <https://doi.org/10.5194/acp-15-9765-2015>
- Marshall, J., Shuckburgh, E., Jones, H., & Hill, C. (2006). Estimates and implications of surface eddy diffusivity in the Southern Ocean derived from tracer transport. *Journal of Physical Oceanography*, 36(9), 1806–1821. <https://doi.org/10.1175/JPO2949.1>
- McIntyre, M. E., & Palmer, T. N. (1983). Breaking planetary waves in the stratosphere. *Nature*, 305(5935), 593–600. <https://doi.org/10.1038/305593a0>
- Meirink, J. F., Bergamaschi, P., & Krol, M. C. (2008). Four-dimensional variational data assimilation for inverse modelling of atmospheric methane emissions: Method and comparison with synthesis inversion. *Atmospheric Chemistry and Physics*, 8(21), 6341–6353. <https://doi.org/10.5194/acp-8-6341-2008>
- Nakamura, N. (1996). Two-dimensional mixing, edge formation, and permeability diagnosed in an area coordinate. *Journal of the Atmospheric Sciences*, 53(11), 1524–1537. [https://doi.org/10.1175/1520-0469\(1996\)053%3C1524:TDMEFA%3E2.0.CO;2](https://doi.org/10.1175/1520-0469(1996)053%3C1524:TDMEFA%3E2.0.CO;2)
- Nakamura, N., & Ma, J. (1997). Modified Lagrangian-mean diagnostics of the stratospheric polar vortices: 2. Nitrous oxide and seasonal barrier migration in the cryogenic limb array etalon spectrometer and SKYHI general circulation model. *Journal of Geophysical Research*, 102(D22), 25,721–25,735.
- Neu, J. L. (2003). Variability of the subtropical “edges” in the stratosphere. *Journal of Geophysical Research*, 108(D15), 4482. <https://doi.org/10.1029/2002JD002706>
- Palmer, T. N. (1982). Properties of the Eliassen-Palm flux for planetary scale motions. *Journal of the Atmospheric Sciences*, 39, 992–997.
- Payne, V. H., Clough, S. A., Shephard, M. W., Nassar, R., & Logan, J. A. (2009). Information-centered representation of retrievals with limited degrees of freedom for signal: Application to methane from the Tropospheric Emission Spectrometer. *Journal of Geophysical Research*, 114, D10307. <https://doi.org/10.1029/2008JD010155>
- Pison, I., Ringeval, B., Bousquet, P., Prigent, C., & Papa, F. (2013). Stable atmospheric methane in the 2000s: Key-role of emissions from natural wetlands. *Atmospheric Chemistry and Physics*, 13(23), 11,609–11,623. <https://doi.org/10.5194/acp-13-11609-2013>
- Pommrich, R., Müller, R., Grooß, J.-U., Konopka, P., Ploeger, F., Vogel, B., et al. (2014). Tropical troposphere to stratosphere transport of carbon monoxide and long-lived trace species in the Chemical Lagrangian Model of the Stratosphere (CLaMS). *Geoscientific Model Development*, 7, 2895–2916. <https://doi.org/10.5194/gmd-7-2895-2014>
- Randel, W. J., Gille, J. C., Roche, A. E., Kumer, J. B., Mergenthaler, J. L., Waters, J. W., et al. (1993). Stratospheric transport from the tropics to middle latitudes by planetary-wave mixing. *Nature*, 365(6446), 533–535. <https://doi.org/10.1038/365533a0>
- Sheese, P., Walker, K., Boone, C., Bernath, P., Froidevaux, L., Funke, B., et al. (2017). ACE-FTS ozone, water vapour, nitrous oxide, nitric acid, and carbon monoxide profile comparisons with MIPAS and MLS. *Journal of Quantitative Spectroscopy and Radiative Transfer*, 186, 63–80. <https://doi.org/10.1016/j.jqsrt.2016.06.026>
- Shuckburgh, E., & Haynes, P. (2003). Diagnosing transport and mixing using a tracer-based coordinate system. *Physics of Fluids*, 15(11), 3342–3357. <https://doi.org/10.1063/1.1610471>

- Spivakovsky, C. M., Logan, J. A., Montzka, S. A., Balkanski, Y. J., Foreman-Fowler, M., Jones, D. B. A., et al. (2000). Three-dimensional climatological distribution of tropospheric OH: Update and evaluation. *Journal of Geophysical Research*, 105(D7), 8931–8980. <https://doi.org/10.1029/1999JD901006>
- Steinhorst, H.-M., Konopka, P., Günther, G., & Müller, R. (2005). How permeable is the edge of the Arctic vortex: Model studies of winter 1999–2000. *Journal of Geophysical Research*, 110, D06105. <https://doi.org/10.1029/2004JD005268>
- Stenke, A., Dameris, M., Grewe, V., & Garny, H. (2009). Implications of Lagrangian transport for simulations with a coupled chemistry-climate model. *Atmospheric Chemistry and Physics*, 9(15), 5489–5504. <https://doi.org/10.5194/acp-9-5489-2009>
- Trepte, C. R., & Hitchman, M. H. (1992). Tropical stratospheric circulation deduced from satellite aerosol data. *Nature*, 355(6361), 626–628. <https://doi.org/10.1038/355626a0>
- Wetzel, G., Höpfner, M., & Oelhaf, H. (2016). Long-term validation of MIPAS ESA operational products using MIPAS-B measurements, ESA Technical Note.
- Worden, J. R., Turner, A. J., Bloom, A., Kulawik, S. S., Liu, J., Lee, M., et al. (2015). Quantifying lower tropospheric methane concentrations using GOSAT near-IR and TES thermal IR measurements. *Atmospheric Measurement Techniques*, 8(8), 3433–3445. <https://doi.org/10.5194/amt-8-3433-2015>
- Xiong, X., Barnett, C., Maddy, E. S., Gambacorta, A., King, T. S., & Wofsy, S. C. (2013). Mid-upper tropospheric methane retrieval from IASI and its validation. *Atmospheric Measurement Techniques*, 6(9), 2255–2265. <https://doi.org/10.5194/amt-6-2255-2013>