Trajectory model simulations of ozone (O$_3$) and carbon monoxide (CO) transport in the lower stratosphere (LS)

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Abstract

A domain-filling, forward trajectory model originally developed for simulating stratospheric water vapor is used to simulate ozone (O$_3$) and carbon monoxide (CO) in the lower stratosphere (LS). Trajectories are initialized in the upper troposphere, and the circulation is based on reanalysis wind fields. In addition, chemical production and loss rates along trajectories are included using calculations from the Whole Atmosphere Community Climate Model (WACCM). The trajectory model results show good overall agreement with satellite observations from the Aura Microwave Limb Sounder (MLS) and the Atmospheric Chemistry Experiment Fourier Transform Spectrometer (ACE-FTS) in terms of spatial structure and seasonal variability. The trajectory model results also agree well with the Eulerian WACCM simulations. Analysis of the simulated tracers shows that seasonal variations in tropical upwelling exerts strong influence on O$_3$ and CO in the tropical lower stratosphere, and the coupled seasonal cycles provide a useful test of the transport simulations. Interannual variations in the tracers are also closely coupled to changes in upwelling, and the trajectory model can accurately capture and explain observed changes during 2005-2011. This demonstrates the importance of variability in tropical upwelling in forcing chemical changes in the tropical lower stratosphere.

1. Introduction

The influx of water vapor (H$_2$O) to the stratosphere is largely determined by the large-scale troposphere-to-stratosphere transport in the tropics, during which air is dehydrated across the cold tropical tropopause (e.g., Fueglistaler et al., 2009 and references therein). Observations such as the entry mixing ratios (Dessler, 1998; Dessler et al., 2013), the
coherent relations between water vapor and temperature (Mote et al., 1996), and the extensive cirrus clouds near the tropopause (e.g., Winker and Trepte, 1998; Wang and Dessler, 2012) all support this understanding. Back trajectory models have provided detailed simulations of stratospheric H$_2$O (e.g. Fueglistaler et al., 2005). More recently, a newly designed domain-filling forward trajectory model driven by reanalysis wind and temperature has demonstrated success at simulating the transport of H$_2$O in the stratosphere (Schoeberl and Dessler, 2011; Schoeberl et al., 2012; Schoeberl et al., 2013). In this trajectory model, winds determine the pathways of parcels and temperature determines the H$_2$O content through an idealized saturation calculation. This simple advection-condensation strategy yields reasonable results for H$_2$O in the stratosphere, although the detailed results depend on the wind and temperature fields utilized and assumptions regarding supersaturation (Liu et al., 2010; Schoeberl et al., 2012).

Besides H$_2$O, ozone (O$_3$) and carbon monoxide (CO) are also important trace gases in the stratosphere linked to circulation, transport and climate forcing (for O$_3$). Ozone abundances in the stratosphere cover a wide dynamic range (~10’s of ppbv to a few ppmv), and are influenced by a variety of chemical and dynamical processes, including photochemical production and loss and large- and small-scale transport (for example, deep convective lofting of boundary layer low O$_3$ air to the upper troposphere, e.g. Folkins et al., 2002). CO behaves as a tropospheric source gas, originating from natural and anthropogenic emissions, including combustion processes near the surface and oxidation of methane and other hydrocarbons within the troposphere. The main sink of CO is oxidation by the hydroxyl radical (OH) (Logan et al., 1981). The CO photochemical lifetime of 2-3 months makes it a useful tracer for transport studies in the troposphere and lower stratosphere (e.g., Bowman, 2006; Schoeberl et al., 2006, 2008).

The purpose of this work is to simulate O$_3$ and CO in the lower stratosphere (LS) using the domain-filling, forward trajectory model used previously for stratospheric H$_2$O. Trajectory modeling of O$_3$ and CO can provide useful tests for simplified understanding of transport and chemical processes in the stratosphere, and provide complementary information to the H$_2$O simulations (which are primarily constrained by tropopause temperatures). In addition to testing circulation and transport within the trajectory model, the O$_3$ and CO simulations can provide understanding of mechanisms leading to observed
chemical behavior, including transport history and pathways, seasonal and interannual variations, and relation to the stratospheric age-of-air (Waugh and Hall, 2002). Ozone and CO are complementary tracers representing primarily stratospheric and tropospheric sources; both tracers exhibit strong gradients across the tropopause, and they are often combined using tracer correlations to understand transport and chemical behavior in the upper troposphere and lower stratosphere (UTLS) (e.g., Fischer et al., 2000). Furthermore, O₃ and CO exhibit relatively large out-of-phase seasonal cycles in the tropical lower stratosphere (Randel et al., 2007), and these coupled variations provide a sensitive test of the trajectory model simulations in this region.

2. Model Description and Data Used

2.1 Trajectory Model

The trajectory model used here follows the details described in Schoeberl and Dessler (2011) and Schoeberl et al. (2012), with trajectories calculated using the Bowman trajectory code (Bowman, 1993; Bowman and Carrie, 2002; Bowman et al., 2013). Because of the overly dispersive behavior of kinematic trajectories (e.g., Schoeberl et al., 2003; Liu et al., 2010; Ploeger et al., 2010; Schoeberl and Dessler, 2011), we perform diabatic trajectories using isentropic coordinates, in which the vertical velocity is the potential temperature tendency converted from the diabatic heating rates via the thermodynamic equation (Andrews et al., 1987).

The methodology for the trajectory simulations of O₃ and CO follows Schoeberl and Dessler (2011) and Schoeberl et al. (2012), wherein the parcels are initialized at the 370-K isentrope, below the tropical tropopause, using climatological O₃ and CO from the MLS (monthly means averaged over 2005-2011) to provide approximate entry level values in the upper troposphere. The 370 K level is chosen as the initialization level because it is above the level of zero net heating rates and parcels there tend to ascend to the stratosphere (see Discussion). To account for chemical changes along the trajectories, we use chemical production and loss rates output from WACCM. Specifically, the O₃ and
CO concentration carried by each parcel is modified from the previous time step using the production and loss frequencies calculated from WACCM.

For each day 1350 parcels are initiated on an equal area grid covering 40° N-S latitude and advected forward in time by reanalysis winds. At the end of each day, any parcels that have descended below the 250-hPa level are removed since in most cases they have re-entered the troposphere. The upper boundary is chosen to be the ~2200-K isentrope (~1 hPa or ~50 km) to cover the entire stratosphere. Parcels are initialized and added to the ensemble consecutively each day and the combined set of parcels is then run forward. This process is repeated over the entire integration period. As more and more parcels are injected into the model, the stratospheric domain is filled up with parcels – this is the concept of domain-filling used in our model. The trajectory model simulations are started in January 1990 and integrated to the end of 2011; after 3-4 years the system reaches steady state with approximately 1 million parcels in the domain. We focus on analyzing the model results during 2005-2011, to overlap the MLS and ACE-FTS observations.

Upwelling across the tropical tropopause in the trajectory model is determined by the reanalysis total diabatic heating rates ($Q_{tot}$, hereinafter $Q$), which include heating due to long-wave and short-wave radiation, moist physics, friction, gravity wave drags, etc. As shown below, our simulations of O3 and CO are sensitive to the imposed upwelling. There is substantial uncertainty in the detailed magnitude and spatial structure of $Q$, as seen in the differences among separate reanalysis results (Schoeberl et al. 2012; Randel and Jensen, 2013; Wright and Fueglistaler, 2013). Figure 1 illustrates the differences in $Q$ in the tropics (15° N-S) based on several reanalysis data sets, highlighting large differences in the UTLS. Given this uncertainty, we tested the sensitivity of our calculations to variations in the heating rates by comparing results based on the NASA Modern Era Retrospective-Analysis for Research and Applications (MERRA, Rienecker et al., 2011) and the ECMWF ERA Interim reanalysis (ERAi, Dee et al., 2011). Below we highlight the sensitivity of the resulting O3 and CO simulations to these imposed circulations.

Changes in chemical concentrations in the trajectory model are calculated using the chemical continuity equation (e.g., Dessler, 2000):
Here, $\chi$ represents either O$_3$ or CO. Current chemical concentrations $[\chi]_{\text{current}}$ in volume mixing ratio (VMR) are determined by concentrations in previous time step $[\chi]_{\text{previous}}$ and the net change, derived from the production minus loss occurring in each time step. The production rate $P_{\chi}$ in VMR per unit time is obtained from WACCM. The loss rate $(LF_{\chi} \cdot [\chi]_{\text{previous}})$ in VMR per unit time is calculated as a product of loss frequency $LF_{\chi}$ (per unit time) times the chemical concentration (VMR), representing a linear chemical loss. The loss frequencies are estimated from WACCM by dividing the model loss rate $L_{\chi}$ by the chemical concentration $[\chi]$, i.e., $LF_{\chi} = L_{\chi}/[\chi]$. In our simulation $P_{\chi}$ and $LF_{\chi}$ are calculated from WACCM as a function of latitude, altitude, and climatological months. The time step for calculating trajectories is 45-min while the time step for calculating chemical changes is 1-day.

2.2 WACCM Chemical Production and Loss Rates

The chemical production and loss rates from the Whole Atmosphere Community Climate Model (WACCM4) (Garcia et al., 2007) are applied in our model to represent chemical processes. The vertical domain of WACCM4 extends from the surface to the lower thermosphere, with horizontal resolution of 2.5° × 1.9° in longitude and latitude and 88 levels up to ~150 km. In the UTLS the vertical resolution is 1.1–1.4 km. The WACCM4 simulation used here is run with specified dynamics (SD) fields (Lamarque et al., 2012), in which the model is nudged by the MERRA meteorological fields (Rienecker et al., 2011) from January 2004 to December 2011 (hereafter we only use WACCM to represent SD-WACCM4).

Figure 2 shows the annual zonal mean ratio of net tendency (production rate minus loss rate) divided by the loss rate for O$_x$ and CO. These show the mean differences of production to loss, with positive numbers indicating net chemical increase and negative numbers indicating net chemical decrease; values close to zero imply balanced production vs. loss. On the background are the respective photochemical lifetimes evaluated from the loss rates ($\tau_L = [\chi]/L_{\chi}$). Both O$_x$ and CO exhibit relatively long
lifetimes (> 3 months) in the lower-middle stratosphere, and hence transport will have a
dominant influence on their distributions. Above ~30 km the lifetimes are shorter, and
the photochemical behavior will determine chemical structure.

Here $O_x$ is the odd oxygen – the sum of $O_3$ and $O$. We use the production and loss
rates of $O_x$ because (1) the time step of trajectory model is 1 day, which is much longer
than the lifetime of $O_3$ in the middle and upper stratosphere, and (2) the abundances of $O_3$
only change on time scales comparable to or longer than the lifetime of $O_x$ (see e.g.
Dessler, 2000, chapter 3). Therefore, it is reasonable to use production and loss rate of $O_x$
instead of $O_3$ for our purpose of simulation. Because $O_3$ abundance is much greater than
$O$, i.e., $O_x \approx O_3$, throughout the rest of the paper we use $O_3$ and $O_x$ interchangeably.

$O_x$ production in the stratosphere is almost entirely due to the photolysis of $O_2$.
$P_{O_x}$ increases with altitude over most of the stratosphere because the photolysis rate
(proportional to solar radiation) increases faster with altitude than $[O_2]$ decreases, so the
net effect is increasing. Figure 2a shows that from 150 to ~10 hPa $O_x$ production exceeds
loss in the tropics, whereas a net chemical decrease of $O_3$ occurs in mid- to high latitudes
(transport of $O_3$ from the tropics to higher latitudes closes the budget). Between about 10
and 2 hPa the daily production and loss of $O_x$ become comparable and the system is in
diurnal steady state. The system gradually reaches photochemical steady states above ~2
hPa, where the lifetime of $O_x$ is less than a day and the instantaneous production and loss
can be treated equal throughout the day. Here we simply set the $[O_x]$ to be $(P_{O_x}/LF_{O_x})$ in
our trajectory calculations.

In the UTLS region, the main source of CO is from the troposphere (from direct
emissions and photochemical production), and this source is accounted for by initializing
trajectories with observed values of CO. There is a net chemical decrease (Fig. 2b) in the
UTLS, where CO is predominantly removed by oxidation with OH. CO experiences a net
increase in the middle stratosphere (above ~24 km) due to oxidation of methane ($CH_4$),
although this has little influence on the results shown here.

In our simulation $P_x$ and $LF_x$ are calculated from WACCM as a function of
latitude, altitude, and climatological months averaged over 2005-2011, so that a constant
annual cycle is applied throughout the model integration. As discussed in Sect. 5, this
calculation will not accurately handle the situation where chemical losses are linked to meteorological behavior, such as for ozone losses in the polar winter stratosphere.

2.3 MLS Observations of O$_3$ and CO

We will compare the model to observations of O$_3$ and CO from the Aura Microwave Limb Sounder (MLS) (Waters et al., 2006). MLS climatology (constant annual cycle) of O$_3$ and CO averaged from August 2004 to December 2012 is used to set the initial abundances when parcels are initialized at the 370 K level, and the observations at higher levels are used to evaluate the trajectory model results. We use the MLS version 3.3 (v3.3) Level 2 products, described in the data quality and description document (http://mls.jpl.nasa.gov/data/v3-3_data_quality_document.pdf). O$_3$ profiles are available at 12 levels per decade from 261 to 0.02 hPa and CO profiles are available between 215 and 0.0046 hPa at 6 levels per decade. The vertical resolution of O$_3$ in the UTLS is approximately 2.5-3 km while for CO it is ~ 4.5-5 km. The detailed validation for these data sets can be found in Froidevaux et al., (2008), Pumphrey et al., (2007), and Livesey et al., (2008).

For the comparisons shown here, we applied the MLS O$_3$ and CO averaging kernels to both our trajectory simulations and WACCM model outputs when comparing with the MLS observations. We have followed the detailed instructions for applying averaging kernels as described in http://mls.jpl.nasa.gov/data/v3-3_data_quality_document.pdf.

2.4 Other Verifying Datasets

Besides using chemical production and loss from the WACCM, we also compare O$_3$ and CO modeled by WACCM to the trajectory model simulations, which serves as a sanity check of applying the imposed WACCM chemistry, and also as a simple comparison of Lagrangian vs. Eulerian model results. We also compare the trajectory modeling to the CO measurements from the Atmospheric Chemistry Experiment Fourier Transform Spectrometer (ACE-FTS) (Bernath et al., 2005), which shows some systematic
differences from MLS retrievals in the stratosphere (Clerbaux et al., 2008; Park et al., 2013). A detailed description of the ACE CO observations can be found in Park et al., (2013).

3. Trajectory Modeling Results

3.1 O₃ Results

Figure 3 shows the zonal mean cross section of O₃ during December-February (DJF) from the trajectory model driven by ERAi reanalysis (denoted as “TRAJERAi”), compared to both the MLS observations and the WACCM results. Because O₃ above 10 hPa is in photochemical steady-state (Sect. 2) we focus on O₃ below 10 hPa. Overall the trajectory simulations agree with results from both MLS and WACCM. In the lower stratosphere transport plays an important role in re-distributing chemical species, resulting in contours of O₃ approximately following the isentropes. The enhanced O₃ production due to photolysis at 30 km (~10 hPa) shifts from south during DJF towards north during JJA (not shown), following the seasonal variations of photolysis rates.

Vertical profiles of O₃ averaged over the deep tropics (18° N-S) from 2005 to 2011 are shown in Fig. 4a (note that the x-axis is in log scale). The trajectory model driven by MERRA (denoted as “TRAJ_MER”) shows reasonable agreement with MLS data (and WACCM) in the lower stratosphere, while the results based on ERAi (denoted as “TRAJERAi”) show relatively smaller O₃ values. Above 24 km where photochemical processes dominate, different trajectory runs yield similar results and they both agree with MLS and WACCM data. Note that the MERRA and ERAi simulations use identical O₃ initial values at 370 K, so that the differences in Fig. 4a are primarily a result of differences in upward circulation (Fig. 1). The mean differences in ozone in the lower stratosphere can be explained as a result of the different heating rates (vertical circulations) imposed. The ERAi heating rates are higher than MERRA up to 20 km (Fig. 1). Due to positive vertical gradient in O₃ the stronger circulation moves air with lower O₃ upward, creating a lower relative concentration compared to MERRA.

Monthly time series of O₃ at 100 hPa and 68 hPa averaged over the deep tropics (18° N-S) are shown in Fig. 4b. There is a strong annual cycle in ozone at these levels
related to the seasonal variations in tropical upwelling (Randel et al., 2007; Abalos et al., 2012, 2013a), and this behavior is reproduced by the trajectory model, showing reasonable agreement in amplitude with the MLS observations and WACCM results (in spite of the differences in mean values). There are somewhat larger differences in annual cycle amplitude at 68 hPa, with the ERAi trajectory results showing better agreement with MLS.

The simulated and observed latitudinal structure of zonal mean $O_3$ in the lower stratosphere (68 hPa) throughout the seasonal cycle is shown in Fig. 5. The overall variations are reasonably well simulated by the trajectory model, although low biases are found compared to MLS over middle-to-high latitudes in both hemispheres (and WACCM is systematically higher over the globe). The development of the Antarctic ozone hole is evident in the very low ozone values polewards of 60°$^\circ$ S in October, which is simulated in the trajectory model based on the strong chemical ozone losses in this region derived from WACCM.

Figure 6 compares the horizontal structure of boreal summer (JJA) $O_3$ at 83 hPa from MLS data and the trajectory results driven by MERRA. The trajectory results show a reasonable simulation of the spatial patterns compared to MLS (and WACCM; not shown), with a clear minimum inside the Asian monsoon anticyclone linked to upward transport of ozone-poor air from lower levels (Park et al., 2009). There is also relatively low $O_3$ centered near 15°$^\circ$ S linked to the slow ascending air from the troposphere in this region.

The trajectory model is also able to capture the spatial behavior of polar ozone. Figure 7 shows a comparison of high latitude $O_3$ in the Northern Hemisphere (NH) during winter (DJF) and in the Southern Hemisphere (SH) during spring (September) between MLS and trajectory modeling driven by both MERRA and ERAi winds. During NH winter, $O_3$ rich air (> 2.2 ppmv) occurs within the polar vortex (denoted with the 24 PVU isopleth in Fig. 7a-c), and the trajectory model captures the observed isolation from middle latitudes, although differences in magnitude exist. During SH springtime, the Antarctic ozone hole (denoted with the 195 K isotherm in Fig. 7d-f) is reasonably well reproduced in the trajectory model based on imposed WACCM chemistry. The trajectory model also captures the spatial structure of the zonal wave ozone maximum near 50°$^\circ$ S
(the so-called “ozone croissant”), linked to the descending branch of the BD circulation. The weaker extra-vortex high in ozone in the trajectory model may be related to the weaker overall circulation in MERRA compared to observations (Schoeberl et al., 2012, 2013). Noted that due to the climatological loss frequency adopted from WACCM, this model cannot account for interactions of circulation and chemistry, which affects $O_3$ mostly in winter polar regions (see Discussion).

3.2 CO Results

Figure 8 shows that zonal mean cross sections of CO from ACE-FTS, WACCM, and the trajectory model agree well in the lower stratosphere. CO has a maximum in the tropical upper troposphere, and decreases with altitude due to OH oxidation to a minimum near 22 km. CO increases above this altitude due to production from methane (CH$_4$) oxidation. The ACE-FTS observations show high CO mixing ratios in the polar middle and upper stratosphere regions, resulting from the downward transport of CO from the mesosphere (from photodissociation of CO$_2$); this behavior is also seen (to a weaker degree) in WACCM, but is not simulated in the trajectory model, which does not include mesospheric processes.

Figure 9a shows the CO vertical profiles and time series averaged in the deep tropics (15° N-S). Due to the differences of MLS and ACE-FTS retrievals in the stratosphere, here we also included ACE CO for comparison. The vertical profiles in Fig. 9a show broad-scale agreements, although there are differences among the trajectory models (with ERAi driven results larger than those driven by MERRA) and also between the ACE-FTS and MLS observations (all of the models agree better with the ACE observations above 22 km). Time series of CO at 100 hPa (Fig. 9b) show a semi-annual cycle linked to initialized variations in the upper troposphere (Liu et al., 2007), with approximate agreement among the models and observations (with slightly larger values in the MLS data). The variability changes to an annual cycle at 68 hPa, as a response to variations in tropical upwelling. At 68 hPa the annual cycle is captured is a reasonable manner in the models, with the ERAi results showing better agreement with MLS and ACE-FTS data.
A further diagnostic to evaluate the model simulations is made by plotting monthly tropical (15° N-S) averages of O₃ vs. CO in the lower stratosphere (68 hPa), as shown in Fig. 10. This includes the observations from MLS, together with trajectory model simulations driven by both MERRA and ERAi, which shows the sensitivity to different Q (see also Fig. 1). There is an overall anti-correlation between O₃ and CO in Fig. 10, mainly representing the out-of-phase annual cycles seen in Figs. 4b and 9b. The comparisons in Fig. 10 show that stronger upwelling in the ERAi simulation produces slightly lower values of O₃ and higher values in CO (> 30 ppbv), in better agreement with MLS data. Moreover, the slope of the CO-O₃ variations in the ERAi simulation approximately matches the MLS result.

The DJF and JJA seasonal distributions of CO at 68 hPa from the ERAi trajectory model are compared to MLS data in Fig. 11. In both seasons the trajectory model shows spatial patterns consistent with MLS data. During DJF the patterns show a center of high CO over central America and enhancements over South East Asia, extending to the tropical western Pacific (largely attributable to fossil fuel emissions, Jiang et al., 2007). The trajectory model also captures the well-known CO maximum linked to the Asian monsoon anticyclone during JJA, which is substantially stronger at the 100-hPa level (e.g., Randel and Park, 2006; Park et al., 2009; Randel et al., 2010).

Overall, the large-scale seasonal behavior of CO simulated by the trajectory model is in agreement with both observations (MLS and ACE-FTS) and Eulerian chemical model (WACCM), although the results are sensitive to the tropical upwelling speed (see Discussion).

4. Interannual Variability of Tracers in the Tropical Lower Stratosphere (LS)

The coherent seasonal variations in O₃ and CO in the tropical LS demonstrate that transport processes have a large impact on the chemical concentrations in this region. The Eulerian-mean calculations of Abalos et al. (2012, 2013a) show that tropical upwelling is the main driver of the annual cycles in O₃ and CO above the tropical tropopause. Our Lagrangian trajectory model results (Figs. 4b and 9b) also show that the annual cycles of O₃ and CO above the tropopause (especially around 70 hPa) are strongly influenced by the tropical upwelling (Brewer-Dobson) circulation.
We further explore interannual variations in the chemical tracers and links to changes in the upwelling circulation. Figure 12 shows the interannual anomalies (by removing the annual cycle) in O$_3$ and CO concentrations in the tropical lower stratosphere from MLS observations and from trajectory calculations, and in addition anomalies in diabatic heating rates (upwelling) from reanalysis at 68 hPa. While there are significant differences in time-mean diabatic heating rates between MERRA and ERAi (Fig. 1), interannual changes in Q (Fig. 12c) show good agreement. Figure 12 shows that interannual anomalies in O$_3$ and CO are strongly anti-correlated (due to oppositely signed vertical gradient) and closely linked to interannual changes in diabatic heating. Furthermore, Fig. 12 shows that trajectory calculations driven by both MERRA and ERAi are able to simulate the observed interannual anomalies in O$_3$ and CO, in spite of significant differences for the background seasonal cycle (Fig. 4b and 9b).

Taking results from the MERRA run as an example, the close relationship between anomalies in diabatic heating and O$_3$ is quantified in Fig. 13a, which shows strong anti-correlation with explained variance $r^2=0.73$ and slope $\Delta O_3/\Delta Q = -1.05 \pm 0.14$ (ppmv)/(K/day). Similar strong correlation is found for CO and Q anomalies (Fig. 13b), with explained variance $r^2=0.85$ and slope of $31.11 \pm 2.86$ (ppbv)/(K day$^{-1}$). The sign of the slopes in Figs. 13a-b are opposite because of the different background vertical gradients for O$_3$ and CO. These strong relationships between diabatic heating (Q) and tracer anomalies highlight the dominant role of tropical upwelling in controlling species with strong vertical gradients near the tropical tropopause.

Figure 12 also highlights strong anti-correlations between O$_3$ and CO anomalies, which is further demonstrated in Fig. 13c. Abalos et al. (2012) have shown that for the idealized case where upwelling dominates tracer transports, the ratio of tendencies for two tracers ($\chi_1, \chi_2$) is closely related to the ratio of the respective background gradients:

$$\frac{\frac{\partial \chi_1}{\partial t}}{\frac{\partial \chi_2}{\partial t}} = \frac{\frac{\partial \chi_1}{\partial z}}{\frac{\partial \chi_2}{\partial z}} \sim \text{constant} \quad (2)$$

Integrating this equation in time for monthly O$_3$ and CO anomalies gives the relationship:

$$\Delta \frac{O_3}{\Delta CO} \approx \frac{\frac{\partial O_3}{\partial z}}{\frac{\partial CO}{\partial z}} \quad (3)$$

i.e. the ratio of monthly anomalies approximately follows the ratio of background vertical gradients for the idealized situation where vertical transport is dominant. For the case of CO and O$_3$ in the tropics near 68 hPa, the MERRA trajectory results yields a background
gradient ratio of ~ -18.34 (ppbv km\(^{-1}\))/(ppmv km\(^{-1}\)) for \((d\text{CO}/dz) / (d\text{O}_3/dz)\). A linear fit of the observed CO/O\(_3\) anomalies (Fig. 13c) gives a ratio of -22.45±3.40 (ppbv / ppmv), which is close to the idealized result (slightly outside of the two-sigma uncertainty). This approximate agreement with highly idealized theory provides further evidence for the control of tropical lower stratosphere O\(_3\) and CO by variations in upwelling.

5. Summary and Discussion

The results presented here demonstrate that the domain-filling, forward trajectory model is useful for studying the transport of trace gases in the LS. The O\(_3\) and CO simulations are complementary to modeling H\(_2\)O (mainly controlled by tropopause temperature) in that O\(_3\) and CO rely on both initial conditions and chemical production and loss rates, and are sensitive to transport. Initial conditions based on observations provide entry values of chemical species into the lower stratosphere; after that the chemical production and loss control the net changes of concentrations along the trajectories. Because the MERRA negative heating rates at 150-130 hPa (Fig. 1) prevent air ascending to the stratosphere, we chose a relatively high initialization level in the upper troposphere (370 K) in this study. However, we could have initiated parcels at lower altitude, such as 355 K when using ERAi circulation, to extend the results to the upper troposphere (UT).

Trajectory modeled O\(_3\) and CO in the tropical lower stratosphere largely depend on the strength of upwelling (and to a lesser degree on the amount of mixing with extratropics, Abalos et al., 2013a). Stronger upwelling is linked to faster transport, which results in less time for chemical production (for O\(_3\)) or loss (for CO), leading to overall lower values of O\(_3\) (Fig. 4a) and higher values of CO (Fig. 9a). The comparisons of MERRA and ERAi simulations that have different tropical upwelling rates (Q), e.g. Fig. 1, clearly demonstrate this sensitivity. The detailed differences in diabatic heating rates among reanalyses have been discussed by Wright and Fueglistaler (2013), who highlight differences in the corresponding long-wave radiative heating rates in the lower stratosphere, which are influenced by both temperature and ozone. We also conducted a sensitivity study of increasing MERRA diabatic heating rates (Q) by constant factors, and
the best overall fit to the observations is 1.5 times the MERRA Q values, consistent with
the ERAi-based results.

Although better agreement with observations of CO in the tropical lower
stratosphere are found using ERAi data, there is reason to suspect that the ERAi diabatic
heating in this region may be too high. For example, Ploeger et al. (2012) performed a
radiative calculation showing that ERAi heating rates are ~40% too high, consistent with
Schoeberl et al. (2012), who show that trajectory modeled water vapor tape recorder
signal based on ERAi heating rates is ~30% too fast compared with MLS observations. In
spite of the detailed differences, the trajectory modeled O₃ shows reasonable simulation
of the large-scale seasonal structure compared to both MLS and WACCM, including both
the tropics and the polar regions. The trajectory modeled CO in the tropical stratosphere
is more sensitive to the MERRA vs. ERAi differences, likely because of the shorter
photochemical lifetime of CO in the lower stratosphere compared to O₃.

The annual cycles in O₃ and CO in the tropical LS are reproduced in the trajectory
model simulations, and the magnitude of variations provides a useful test of the imposed
circulation. The variability of O₃ and CO shows significant correlations with fluctuations
in diabatic heating, for both seasonal and interannual time scales. These close
relationships support the concept that tropical upwelling plays a key role in regulating
variability for chemical species with strong vertical gradients in the lower stratosphere
(and explains the observed compact relationships among interannual anomalies in
diabatic heating, O₃ and CO seen in Figs. 12-13). For the idealized situation where
upwelling dominates tracer transports, the tracer ratios can be expressed as ratios of the
background vertical gradients, and the observed O₃ and CO changes are in approximate
agreement with this expectation (Fig. 13c).

The discussion above linking seasonal or interannual changes in O₃ and CO with
chemical changes along slower or faster upward trajectories is a Lagrangian perspective
on transport (appropriate for our Lagrangian trajectory model). In contrast, the discussion
linking O₃ and CO variations at particular pressure levels to varying circulations acting
on background vertical gradients (Sect. 4) is an Eulerian perspective. These two
perspectives are complementary and do not contradict each other; Abalos et al. (2013b)
have recently shown the equivalence of Lagrangian and Eulerian transport calculations in the tropical lower stratosphere, highlighting that each can provide useful information.

One limitation of this model is that it cannot account for interactions of circulation and chemistry, due to the climatological chemical production rates and loss frequencies adopted from WACCM (Section 2.2). This is most likely important in winter polar regions, where O\textsubscript{3} loss chemistry due to chlorine activation on polar stratospheric clouds is highly temperature dependent. Although we cannot simulate the detailed behavior in any particular (cold) year with climatological chemical ozone loss rates, our model does a reasonable job at simulating the time-average behavior of polar regions.

Our simulations with O\textsubscript{3} and CO have demonstrated the viability of the domain-filling forward trajectory model for simulating species with relatively simple chemistry, and extension to other species would be straightforward. There are several potential applications for such a trajectory model, including describing parcel histories that characterize different transport pathways, and evaluating the importance of tropical-extratropical exchanges. For example, trajectories can allow tracing the sources of CO-rich air in the summertime Asian monsoon region, and quantifying the fate of the parcels after breakup of the anticyclone. The model can also allow detailed comparisons of transport based on different and new reanalysis data sets, or idealized studies of the chemical responses to UTLS circulation in a changing climate.

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Fig. 1. Comparison of total diabatic heating rates averaged over the deep tropics (18° N-S) in 2000-2010 from four different reanalysis data sets: MER (MERRA), ERAi (ECMWF ERA interim), CFSR (NCEP CFSR) and JRA25.
Fig. 2. The ratio of chemical net tendency (production rate minus loss rate) to loss rate from WACCM for (a) O\textsubscript{x} and (b) CO. Negative numbers are dashed to highlight the net chemical decrease and positive numbers indicate net chemical increase, while zero lines indicate comparable amount of production and loss. For reference, the respective O\textsubscript{x} and CO lifetimes are contoured in color.
**Fig. 3.** Trajectory modeled O$_3$ driven by ERAi reanalysis wind (c, TRAJ_ERAi) in boreal winter (DJF), compared to both MLS observations (a) and WACCM output (b). Both WACCM and trajectory model results are weighted by the MLS averaging kernels. Dashed lines are potential temperature contours, demonstrating that in the lower LS where transport dominates, chemical distributions roughly follow the isentropes.
Fig. 4. Tropical (a) vertical profile and (b) time series (100 hPa, bottom panel; 68 hPa, upper panel) of MLS, WACCM, and trajectory modeled \( \text{O}_3 \) driven by MERRA wind and ERAi wind, averaged over the deep tropics (18°N-S) from 2005 to 2011. Both WACCM and trajectory model results are weighted by the MLS averaging kernels. In (a) the x-axis is in log scale to highlight the differences; the purple square indicates the initial \( \text{O}_3 \) values carried by parcels when they were first injected into the domain at 370 K; the vertical bars in red indicate the MLS vertical resolutions at each of the MLS retrieval pressure levels.
Fig. 5. Zonal mean of O$_3$ at 68-hPa during January (JAN), April (APR), July (JUL), and October (OCT) averaged in 2005-2011 from MLS, WACCM, and trajectory results driven by MERRA wind and ERAi wind. Both WACCM and trajectory model results are weighted by the MLS averaging kernels. In October (Antarctic spring time), the South Pole undergoes exceptional depletion of O$_3$. 

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Fig. 6. Northern Hemisphere Summertime (JJA) tropical O₃ distributions at 83 hPa averaged from 2005 to 2011 between MLS and the MERRA driven trajectory simulations (weighted by the MLS averaging kernels). Horizontal wind vectors from the MERRA reanalysis are overlaid to emphasize the Asian summer monsoon anticyclone.
Fig. 7. Polar O$_3$ distributions shown in MLS (left column), trajectory results driven by MERRA (middle column), and trajectory results driven by ERAi (right column) during Northern Hemisphere winter (DJF, a-c) and Southern Hemisphere spring (September, SEP, d-f) at 68 hPa. Trajectory results are weighted by the MLS averaging kernels. The 24-PVU potential vortices (a-e) and the 195-K temperature (d-f) are overlaid in black dashed lines for both seasons, respectively.
Fig. 8. Zonal mean cross sections of CO from (a) ACE-FTS, (b) WACCM, and (c) trajectory model driven by MERRA reanalysis.
Fig. 9. Tropical (a) vertical profile and (b) time series (100 hPa, bottom panel; 68 hPa, upper panel) of MLS, WACCM, ACE, and trajectory modeled CO driven by MERRA wind and ERAi wind, averaged over 15° N-S from 2005 to 2011. In (b) the WACCM and trajectory model results are weighted by the MLS averaging kernels. In (a) the purple square indicates the initial CO values carried by parcels when they were first injected into the domain at 370 K, and vertical bars in red indicate the MLS vertical resolutions at each of the MLS retrieval pressure levels.
Fig. 10. Monthly variations of $O_3$ vs. CO in the tropical lower stratosphere (15° N-S, 68 hPa) from MLS and trajectory modeling driven by MERRA wind and ERAi wind. Trajectory results are weighted by the MLS averaging kernels.
Fig. 11. Comparison of CO at 68 hPa (∼20 km) during DJF (top row, a and b) and JJA (bottom row, c and d) between MLS (left) and trajectory modeling driven by ERAi wind (right, weighted by the MLS averaging kernels). Horizontal wind vectors from ERAi are overlaid as reference.
Fig. 12. Interannual anomalies of (a) O$_3$ and (b) CO from MLS (red) and trajectory simulations driven by MERRA and ERAi in the tropical (15$^\circ$ N-S) lower stratosphere (68 hPa). (c) shows interannual anomalies of total diabatic heating rates from MERRA and ERAi, which serves in our model as the vertical velocity. The trajectory results are weighted by the MLS averaging kernels.
Figure 13. Scatter plots of the anomalies of (a) O$_3$ vs. Q and (b) CO vs. Q, and (c) CO vs. O$_3$ at 68 hPa. The dots are monthly variations of tracers from trajectory modeling driven by MERRA winds and diabatic heating rates; the black lines show the linear fit. The red line in (c) is the theoretically estimated tracer ratio estimated from the respective background gradients, using simplified relation in Eq. (3).