Improved agreement of AIRS tropospheric carbon monoxide products with other EOS sensors using optimal estimation retrievals

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Abstract

We present in this paper an alternative retrieval algorithm for the Atmospheric Infrared Sounder (AIRS) tropospheric Carbon Monoxide (CO) products using the Optimal Estimation (OE) technique, which is different from AIRS operational algorithm. The primary objective for this study was to compare AIRS CO, as well as the other retrieval properties such as the Averaging Kernels (AKs), the Degrees of Freedom for Signal (DOFS), and the error covariance matrix, against the Tropospheric Emission Spectrometer (TES) and the Measurement of Pollution in the Troposphere (MOPITT) CO, which were also derived using OE technique. We also demonstrate that AIRS OE CO results are much more realistic than AIRS V5 operational CO, especially in the lower troposphere and in the Southern Hemisphere (SH). These products are validated with in situ profiles obtained by the Differential Absorption Carbon Monoxide Measurements (DACOM), which took place as part of NASA’s Intercontinental Chemical Transport Experiment (INTEX-B) field mission that was conducted over the northern Pacific in Spring 2006. To demonstrate the differences existing in the current operational products we first show a detailed direct comparison between AIRS V5 and TES operational V3 CO for the global datasets from December 2005 to July 2008. We then present global CO comparisons between AIRS OE, TES V3, and MOPITT V4 at selected levels as well as for the total column amounts.

1 Introduction

Satellite measurements of atmospheric chemical constituents have enhanced our understanding of how natural and human activities affect climate and air quality in the earth system. The monitoring of tropospheric Carbon Monoxide (CO), identified as a good pollution tracer and a precursor of ozone, provides significant value in air quality studies. Three sensors on NASA Earth Observing System (EOS) satellites, the Measurement of Pollution in the Troposphere (MOPITT) on Terra, the Atmospheric Infrared
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The AIRS instrument on Aqua was launched in 2002 with its primary goal of determining the vertical profiles of temperature and water vapor in the Earth’s atmosphere to improve weather forecasts, as well as for climate monitoring (Aumann, et al., 2003). CO retrievals are obtained from the 2160–2200 cm$^{-1}$ portion of the spectrum on the edge of the 1–0 vibration-rotation band of CO (McMillan et al., 2003). The AIRS operational retrieval uses an information-based method described by Susskind et al. (2003). Warner et al. (2006) summarized AIRS CO validation results against NASA Intercontinental Chemical Transport Experiment-North America (INTEX-NA) field campaign data (Singh et al., 2006) and concluded that in the mid-troposphere AIRS operational CO products agree with in situ measurements to within 20 parts per billion volume (ppbv) or 10% over the US. They further concluded that the performance of AIRS CO profile retrievals in the Northern Hemisphere (NH) mid-troposphere is comparable with MOPITT V3 CO profiles in the same region. Yurganov et al. (2008) summarized the comparison of CO total column measurements between AIRS, MOPITT and two ground spectrometers, and concluded that AIRS data (version 4) for biomass burning events are in agreement or lower than both MOPITT and ground measurements, but CO bursts can...
be seen by AIRS in most cases.

The TES instrument (Beer, et al., 2006) is an imaging infrared Fourier Transform Spectrometer with both nadir and limb-viewing capability covering the spectral range 650–3050 cm\(^{-1}\) at either 0.08 cm\(^{-1}\) or 0.02 cm\(^{-1}\) spectral sample spacing. TES has higher spatial and spectral resolution compared with AIRS, however the global coverage is significantly less with nadir only pixels at 5×8 km\(^2\) in comparison to AIRS’s retrieval grid of 45×45 km\(^2\) along track by 1600 km across track. TES CO retrievals are based on the Optimal Estimation (OE) technique described by Rodgers (2000) and use MOZART (Model of Ozone and Related Tracers, NCAR/Atmospheric Chemistry Division) model profiles as the prior information. Luo, et al. (2007b) summarized TES CO validation using the Intercontinental Chemical Transport Experiment: MILAGRO and Pacific phases (INTEX-B) (Singh et al., 2008) (http://www.espo.nasa.gov/intex-b) differential-Absorption Carbon Monoxide Measurements (DACOM) in situ profiles. They pointed out that for certain types of CO distributions, such as during the Houston, TX portion of the experiment, the difference between TES and DACOM CO observations is smaller than the variability of both the TES (10–15% the standard deviations (SDV)) and DACOM measurements (15–20% SDV), while in other areas, the differences could be as large as ±35%.

MOPITT was specifically designed to measure the tropospheric carbon monoxide and methane by using the cell correlation technique as described by Drummond (1989). The CO information is obtained from the spectral region at 4.7 \(\mu\)m with a spatial coverage of 22×22 km\(^2\). The MOPITT retrieval algorithm follows the OE method similarly to TES and for version 3 (V3) and earlier versions the retrieval method was discussed by Pan et al., 1998 and Deeter et al., 2003. MOPITT latest version (V4) has been delivered recently and it adapts to use a set of dynamic a priori profiles, in 1×1 degree grids computed from the MOZART model, and a fixed covariance matrix (Deeter et al., 2010). MOPITT V4 also uses a log-normal VMR for the state variable vertical distribution as opposed to the previously used in V3 of VMR values for CO vertical distribution. Deeter et al. (2010) also summarized the retrieval improvements for V4.
due to other factors such as the slow instrument drifting first pointed out by Emmons et al. (2009). MOPITT CO products have been validated against a variety of in situ measurements (Emmons et al., 2004, 2007, 2009; Deeter et al., 2010), and therefore, have been established as a well-recognized reference source.

For nadir-viewing satellite instruments using the thermal spectral regions, the trace gas measurement sensitivities are lacking in certain vertical portions of the atmosphere. For example, near the earth surface, prior information for the trace gases is needed in the retrieval algorithm. To identify the true measurement gained from satellite retrievals, the sensitivities of the retrievals need to be considered (Rodgers and Connor, 2003). Furthermore, to understand the CO retrievals from multiple instruments we need to understand the factors affecting retrievals, such as the Averaging Kernels (AKs), Degree of Freedom for Signal (DOFS), error covariance matrix, etc. Because these quantities from AIRS operational algorithm are obtained using a formulation that is different from that used in the OE method, a proper comparison of the sensors requires that we use the same retrieval method.

We present in this paper an alternative retrieval algorithm for the AIRS tropospheric CO product using an OE technique that is formulated as similarly as possible to that used by MOPITT and TES. We will demonstrate that AIRS OE CO results are much more realistic than AIRS V5 operational CO, especially in the lower troposphere regions and in the Southern Hemisphere (SH). These products are validated with in situ profiles obtained by DACOM measurements (Sachse et al., 1987), which took place as part of INTEX-B. In Sect. 2, we first show a detailed comparison between AIRS V5 and TES operational CO for the global datasets during December 2005 to July 2008 period to demonstrate the large differences existing in the current operational products. In Sect. 3, we document the mathematical formulations used by both AIRS V5 operational algorithm and AIRS OE technique. In Sect. 4, we will present AIRS OE CO retrievals and the comparisons with AIRS operational V5 CO profiles. We also include validations of CO profiles using in situ measurements from DACOM instrument onboard DC-8 during the INTEX-B experiment. We then present the global CO comparisons between
AIRS, MOPITT, and TES at the selected levels and for the total column amounts before summarizing our results.

2 Background: CO differences between AIRS and TES operational products

AIRS and MOPITT CO comparisons of the operational products were carried out by Warner et al. (2007), and they concluded that the two datasets agree to within approximately 10% or 20 ppbv near 500 hPa where the measurement sensitivities are high for both sensors. The differences between the two sensors are much larger in the lower troposphere where the measurement sensitivities are lower. In this section, we will summarize the CO mixing ratio differences at selected levels between AIRS and TES for the period of approximately 3 years.

We have collocated each pair of AIRS and TES pixels by time and geographical location for all available TES data from December 2005 to July 2008, and summarized the global statistics of the CO profile differences. We define the CO difference at each vertical level between each AIRS and TES pixel pair as \( \delta(\text{CO}) = \text{CO}_{\text{AIRS}} - \text{CO}_{\text{TES}} \) and the Probability Distribution Function (PDF) for each monthly ensemble as \( P(\delta\text{CO}) \) (Sparing et al., 1995). Figure 1a shows an example of the CO volume mixing ratios (VMRs) (ppbv) difference PDFs at 500 hPa for May 2006 over the NH land (top left), NH ocean (top right), SH land (bottom left), and SH ocean (bottom right). The PDFs are depicted as the thin black histograms and the red solid curves are the Gaussian function fits to the PDFs for each area. The vertical red dotted lines at the centers of the Gaussian functions are the modes of the function and the horizontal red dotted lines are the full-width half-maxima (FWHMs) of the Gaussian fits indicating SDVs of the dataset. We are defining the bias as the mode of the distribution of differences, rather than the average, between the two datasets, represented by each pair of AIRS and TES measurements, which were taken approximately 15 min apart over the same locations. For this example the biases are 5 to 10 ppbv, with the NH ocean being the smallest and the SH land being the largest, and the SDVs are 10–15 ppbv with the smallest over SH
ocean and the largest over NH land.

Figure 1b illustrates the time series of the differences between the two products. The biases and the SDVs of the monthly Gaussian fits to the PDFs of the CO VMR differences at 500hPa for the period of December 2005 to July 2008 for NH land (top left), NH ocean (top right), SH land (bottom left), and SH ocean (bottom right), respectively. Solid curves show the biases of the Gaussian fits and the dotted curves marks the bounds of the SDVs varying with time from December 2005 to July 2008. The CO VMR biases between the two sensors at 500 hPa are below 10 ppbv globally. In the NH, the seasonal variation of the differences is correlated with the magnitudes of the CO concentrations, since the CO differences are larger during the winter-spring phase and smaller during the summer as is the case for the NH CO concentrations. The seasonal variations of the differences are smaller in the SH, and show local peaks associated with strong SH biomass burning events. The SDVs are also larger when the CO concentrations are higher and smaller when the CO concentrations are lower, and range from 10 to 15 ppbv.

Figures 2a and b show an example of the $P(\delta CO)$ with Gaussian fits at 800 hPa for May 2006 and the time series of the biases and SDVs of the monthly Gaussian fits, respectively. The CO biases between the two sensors at 800 hPa are significantly larger than at 500 hPa except for the NH ocean cases, and meanwhile, the SDVs are smaller at 800 hPa than at 500 hPa. In the NH over land, the biases range from 0 to 30 ppbv with TES CO higher than AIRS CO, and over the oceans the biases range from −10 to 10 ppbv. AIRS CO at 800 hPa in the SH are uniformly higher than TES CO, with the biases as large as 30–50 ppbv. The fact that the biases are larger closer to the earth surface is an indication that the sensitivities of the two products are very different in the lower troposphere in comparison to the mid-troposphere.
3 Algorithm descriptions

3.1 AIRS V5 operational algorithm

The current AIRS physical retrieval algorithm seeks to minimize the weighted difference between the clear column radiance observations and the radiance computed using the AIRS forward model (SARTA) (Strow, et al., 2003) by varying the geophysical state (Susskind, et al., 2003). Furthermore, changes to a group of the geophysical states are represented by a geophysical perturbation parameter and a perturbation function with trapezoidal shapes. An eigenvector decomposition technique is employed in the algorithm to solve for the geophysical state, and a damping process is used to stabilize the solution (Susskind, et al., 2003). The selection of the number and levels of the trapezoidal functions, the constraint magnitude for damping, and the choice of the first guess profile all affect the performance of the retrieval. A subset of 36 spectral channels out of 52 channels in the CO region is selected for the operational retrievals using principle component analysis. The parameters used in the retrievals for this study are described by the AIRS Version 5.0 Released Files Description (http://disc.sci.gsfc.nasa.gov/AIRS/documentation). Due to the lack of observational sensitivity in certain portions of the atmosphere for thermal sensors, an appreciable amount of the first guess profile often is retained in AIRS CO operational products (Warner et al., 2007). This first guess profile is analogous to the a priori profile in the optimal estimation algorithm. AIRS first guess for the V5 operational products use the MOPITT V3 a priori mean profile, adding the Air Force Geophysics Laboratory (AFGL) climatology for the levels above 10 hPa. Based on Rodgers and Connor (2003), an estimated profile from a retrieval algorithm is represented as the combination of a true profile and an a priori profile through the knowledge of an averaging kernel. The AIRS operational averaging kernel, shown as the A matrix in Eq. (1), indicates the sensitivity of the measurements to the CO concentrations at the retrieval levels defined by the trapezoids (Susskind, et al., 2003). The details of the derivation and application of the averaging kernels was discussed by Maddy et al. (2008), and also documented at NASA Goddard Earth
Sciences Data and Information Services Center (GES DISC). The convolution of an in situ profile uses the following formula for AIRS V5:

\[ x' = x_0 \left[ 1 + A \left( \frac{x - x_0}{x_0} \right) \right] \]  

(1)

where \( A \) stands for the averaging kernel, \( x' \) represents the transformed in situ profile, \( x \) is the true profile, and \( x_0 \) is the first guess profile. Equation (1) is used to convolve the in situ DACOM profiles in the comparisons with AIRS operational retrievals as described in the following section.

### 3.2 Optimal Estimation method

To use the Optimal Estimation (OE) method in AIRS retrievals we follow the formulations given by Rodgers (2000), and also described by Pan et al. (1998). Given a model of the instrument’s signals, the forward equation for the CO profile retrieval problem can be written as

\[ y = f(x, b) + n_\varepsilon \]  

(2)

where \( y \) is the vector of measured radiances, \( x \) is the state vector (variables to be retrieved from the measurements), \( b \) represents all other parameters used by the forward model, \( f(x, b) \) is the forward model function, and \( n_\varepsilon \) is the instrument noise.

The meteorological parameters, such as pressures, temperatures, the profiles of constituents other than CO, and the surface parameters are retrieved in the earlier steps of the AIRS operational retrieval system and used as input to the forward model. The retrieval methods for atmospheric variables other than CO is described in Sect. 3.1. The retrieval process inverts Eq. (2) to find \( x \) for a given \( y \). As in the case of atmospheric remote soundings, the inverse problem is usually ill posed, because the number of independent measurements is less than the number of variables to be retrieved. The OE method introduces the a priori information as an additional constraint. The
solution then can be understood as the combination of the present measurements and the prior knowledge.

For the variables that obey a Gaussian distribution, this inverse problem is equivalent to the maximum likelihood solution, and by using a Newtonian iteration, the solution to Eq. (2) can be written as (Rodgers, 1976; 2000; Pan et al., 1998):

\[ x_{n+1} = x_a + C_a K_n^T (K_n C_a K_n^T + C_e)^{-1} [y_n - K_n (x_a - x_n)] \]  

where \( n \) is the order of iteration and \( C_e \) is the measurement error covariance matrix. \( K_n \) is the Jacobian matrix for iteration \( n \), \( x_a \) is the mean of the a priori distribution, and \( C_a \) is the a priori covariance matrix for \( x_a \). We used the a priori matrix developed by MOPITT version 3 (Deeter et al., 2003) in this study.

Similar to the averaging kernel discussion in the previous section, it is important to apply the averaging kernel information in the sensitivity analyses using the OE method. In pairing with the retrieval formulations, the averaging kernels are computed using the following:

\[ A = C_a K^T (K C_a K^T + C_e)^{-1} K \]  

and,

\[ x' \approx AX + (I - A)x_a, \]  

where \( I \) represents the identity matrix. Equation (5) states that in the absence of other error sources the retrieved state is a weighted mean of the true state and the a priori state, with the weight \( A \) for the true state and \( I - A \) for the a priori. This shows the importance of averaging kernels as diagnostics of the retrieval. The closer the matrix \( A \) is to the identity matrix the better the retrieved state resembles the true state. Equation (5) also shows that unless the matrix \( A \) is an identity matrix, each layer of the retrieved state is an averaged contribution of multiple layers in the true state and the a priori state.

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AIRS CO retrievals using OE

As stated in Sect. 2, the agreement between AIRS and TES CO is the best in the NH mid-troposphere and the largest differences are in the lower troposphere and in the SH. We discuss, in this section, the advantage of using OE for AIRS CO retrievals and the improvements in areas where the two operational products vary the most. We are using AIRS operational codes as the base line for this study and only changing the minimization formulations for the CO retrievals by using the OE method. The treatment for all other state variables retrieved before CO, such as temperature, water vapor, ozone, cloud, and surface properties, etc. are kept the same as in the AIRS V5 operational products. This will facilitate the comparisons between the two retrieval algorithms for CO without introducing additional factors.

For comparison purposes, we list in Table 1 the prior information used for AIRS OE, V5, MOPITT V3, V4, and TES as either a form of first guess (X) or a priori (Y). The a priori used for AIRS OE is the same as the MOPITT V3 global uniform a priori, which include both the mean profile and the covariance matrix. AIRS V5 only uses a first guess (not a priori), which is equal to the mean profile of AIRS OE or MOPITT V3 a priori. TES uses a set of dynamic a priori profiles, averaged over several areas over the globe, with a fixed covariance matrix (Luo et al., 2007). MOPITT V4 recently adapted to use a set of dynamic a priori profiles, in 1×1 degree grids, and a fixed covariance matrix (Deeter et al., 2010). Although all of these retrievals use a first guess it does not contribute to the final results in the OE method, except for the case of AIRS V5. In addition, the AIRS OE CO retrievals are performed in 34 retrieval layers to strike a balance between the computational time, retrieval stability, and the smoothness of output profiles, etc. AIRS V5 CO VMRs in the standard products is output at 9 vertical levels. TES CO VMR products are at 67 levels, and MOPITT V3 CO are output at 7 levels.

Figure 3 shows two CO profiles collected during INTEX-B from DC-8 spirals on 4 March 2006 where the in situ measurements from DACOM are depicted by the green
curves. This geographical region on that particular day was affected by agricultural fires in the southeast US, which resulted in high CO concentrations in the lower troposphere. The spiral profile on the left panel was collected over the Gulf of Mexico and the one on the right was collected over land to the west of Birmingham, AL, near the fires. The dark blue curves in Fig. 3 depict AIRS V5 CO profiles using the operational retrieval algorithm and the cyan curves are the in situ profiles convolved to AIRS V5 CO using Eq. (1). The red curves represent the OE CO profiles and the orange curves are the in situ profiles convolved to the OE retrieval space using Eq. (5).

In general, the differences between the retrievals and the convolved in situ profiles using the AKs given by the retrieval algorithm indicate how well a remote sensor measures the atmospheric property. In this sense, the two algorithms provide similar information where the AIRS instrument is sensitive to CO, namely in the mid-troposphere, where the retrievals agree with the in situ measurements to within 5–10 ppbv. The OE algorithm provides some slight improvements as shown by the differences between the red and orange curves in comparison with the blue and cyan curves. The OE technique leads to significantly greater agreement between the retrievals and the in situ profiles in the lower troposphere, especially below 800 hPa, even though the retrieved minus convolved differences for the two algorithms are similar. The much more realistic results for the AIRS OE CO, compared to the in situ profiles, are largely due to how the a priori information is used in the retrieval algorithm. The OE method constrains the retrieved profiles to the a priori, whereas AIRS operational results converge to the first guess where the measurement information is lacking.

We have produced full-day retrievals for AIRS CO using OE for the period of Jan. to June 2006 and compared these retrievals with the collocated TES CO profiles. Note that TES CO was retrieved using the OE algorithm and a globally varying a priori developed from the MOZART model. A typical NH CO profile comparison is shown in Fig. 4 from the retrievals on 4 March 2006. The black curve on the top panel represents the first guess, and/or the mean a priori profile, which is the same as MOPITT V3 globally uniform a priori. The dark blue curve on the top panel represents the 9-level AIRS
V5 CO from NASA/DAAC, the red curve represents the AIRS OE CO, and the green line shows the collocated TES CO profile. All three retrieved CO profiles agree well (within approximately 10 ppbv) in the mid-troposphere (300–600 hPa) region while the differences are significantly larger between AIRS V5 and TES CO closer to the surface. Below 600 hPa, AIRS operational retrievals converge towards the first guess since the measurement sensitivities are weaker in these layers, resulting underestimation in the NH where the CO emissions are stronger in this season. AIRS OE retrievals are constrained largely by the shape of the a priori and result in much better agreement with TES CO.

The bottom panels in Fig. 4 shows the AKs for AIRS V5 (dashed) and OE (solid). The OE AKs are selected from 34th in layers and therefore have smaller values than the V5 AKs, which represent thicker layers (Deeter et al., 2007). The correct quantitative comparison is to use the AKs normalized by a layer thickness, however, we are only aiming to demonstrate the vertical position where the measurement information is taken. As shown in the AKs, the main measurement information is obtained in the mid-troposphere between 400 and 600 hPa for both AIRS operational algorithm and the OE.

Similarly a typical SH CO profile is shown in Fig. 5 for 4 March 2006. The V5 CO overestimates in the SH where the CO concentrations are significantly lower than the first guess/a priori. The AIRS OE CO retrievals agrees with TES CO very well in the entire tropospheric column down to 850 hPa for both NH and SH profiles. For the AKs in the SH profile (see bottom panel in Fig. 5), there is a slight secondary curve in the OE near 850 hPa indicating a small amount of sensitivity in this vertical layer.

Using OE for AIRS improves the accuracy of the lower troposphere CO retrievals and the retrievals in clean atmospheric regions such as the SH. This is best demonstrated by the comparisons of the global CO maps in the lower troposphere as shown in Fig. 6. The top panel shows AIRS V5 CO at 840 hPa, the middle panel shows AIRS CO at the same level but retrieved using the OE technique, and the bottom panel shows TES CO at the similar altitudes. The CO in the NH in the lower troposphere is significantly
enhanced, which is justified by the fact that the emissions and the total concentrations in the NH are high in the spring season. This is also consistent with TES CO (bottom panel) that show similar values. In the SH, the OE CO values are significantly lower than the v5 CO especially over the oceans and at high latitudes. These low values of CO from AIRS OE are closer to reality in that they are consistent with our prior knowledge for these regions. In addition, the closer agreement with TES suggests that the changes in the retrieval have been correct. To summarize, AIRS CO retrievals using the OE technique provide more realistic values, particularly in the lower troposphere and in the SH.

5 Comparison of Global Tropospheric CO of AIRS OE with TES and MOPITT

We have computed the PDF analyses of the differences between AIRS OE CO and TES CO in a same manner as described in Sect. 2. Figure 7 shows the PDF and the Gaussian fit properties of the differences between AIRS OE and TES CO convolved to AIRS AKs at 500 hPa and near 850 hPa, respectively, globally for the period of 1 January to 31 June 2006. We adjusted TES CO using AIRS AKs and the a priori profile to reduce the effect of the a priori in the AIRS retrieval. The global bias at 500 hPa represented by the mode of the Gaussian fit is 5.28 ppbv with a SDV of 6.50 ppbv represented by the FWHMs. Similarly, the bias is 5.87 ppbv and the SDV is 16.42 ppbv at 850 hPa. Larger differences at the right wing of the PDFs were primarily due to the NH when CO concentrations are high (not shown), indicating that AIRS measurements are higher than TES when TES CO was convolved using AIRS AKs. Although the differences between the two sensors using a similar algorithm at 500 hPa are smaller than the differences between the two operational products as illustrated in Sect. 2 (see Fig. 1), the changes are not significant at this vertical level. This is because the AIRS sensor is most sensitive in this vertical range for CO and both V5 and OE retrievals perform well. The improvements between AIRS and TES CO by using different retrieval methods are most significant in the lower troposphere, i.e. near 850 hPa, where the
The comparison of CO total column between AIRS and MOPITT further demonstrates the improved agreement of AIRS CO products using the OE technique to other CO sensors as shown in Fig. 8. The top left panel shows the monthly mean AIRS V5 CO total columns ($\times 10^{17}$ molecules/cm$^2$) for March 2006, the top right shows the same quantities except for AIRS OE CO, and the left bottom panel shows the relative difference between the two: $(OE-V5)/OE\cdot100\%$. The total column CO increases using OE over most of the NH for approximately 10% or $\sim 10–15$ ppbv and decreases in the SH for as large as 30%. The AIRS OE CO total column amounts agree significantly better with MOPITT V4 CO total column values as shown in the right bottom panel. This can also be described by the global relative RMS differences that decrease from (V5-MOPITT) being 20.4% to (OE-MOPITT) at 12.7%. MOPITT CO total column values are higher than AIRS using either V5 or OE method, especially at high latitudes in the NH in this example.

To understand the measurement capability of a sensor for a particular remotely sensed geophysical parameter, one needs to examine the information content defined by the DOFS in the retrievals (Rodgers, 2000). The DOFS, computed by the trace of the AKs, indicates the amount of independent measurement information. Figure 9 shows an example of the DOFS for March 2006 for AIRS V5 CO (blue), AIRS OE CO (red), TES CO (green) and MOPITT CO (blue), respectively, where the solid lines are the averaged DOFS in 10-degree latitude bands and the dashed lines represent the maximum and minimum DOFS values in each band. For both MOPITT and TES, the DOFS ranges between 1.2 to 1.5 in the tropical and subtropical regions, and decreases to below 1.0 at mid- and high-latitudes. The average DOFS for AIRS OE CO ranges between 0.8 and 1.0 globally. TES CO DOFS are larger than AIRS in low latitudes partly due to the higher spectral resolution, and partly due to algorithm differences, since TES CO retrievals use a variable set of a priori while AIRS OE CO retrievals use a uniform a priori. The DOFS for MOPITT V3 and TES CO agree very well between 65S and 45N and the MOPITT DOFS decreases towards the poles. The latitudinal
dependence of the DOFS for MOPITT and TES are largely due to the change in the surface temperature contrasts, while the flatness in the latitudinal dependence of AIRS maybe partly due to the cloud clearing uncertainties. AIRS OE DOFS are larger than V5 over most of the globe, except the NH mid-latitude region where the two algorithms are similar. Note that the formulation to compute AIRS V5 DOFS is different from that for the OE, and therefore, the changes of AIRS DOFS between the two algorithms should be understood in a relative sense. Nevertheless, the improvements of DOFS using OE are mainly in the SH and high latitudes in the NH.

In addition to the uncertainties and errors from different CO products due to the instrument characteristics and the retrieval algorithms differences, other biases between different sensors exist due to the measurement properties such as the time of overpass and spatial resolutions. MOPITT is on late-morning equator-crossing time and AIRS and TES are on early-afternoon equator-crossing time. In general, the CO VMRs in the mid- to lower-troposphere are relatively larger when there are higher surface thermal contrasts (Deeter et al., 2007) e.g., for AIRS, however, due to the complexity of the retrieval algorithms between MOPITT and AIRS OE, this difference is difficult to quantify without detailed sensitivity studies, which should be a subject of future studies. The effect of spatial resolution differences is best demonstrated by the statistical differences (e.g., biases and SDVs) between AIRS and TES CO. We have studied the PDFs for AIRS and TES CO independently (not shown) and found that the biases are similar for both instruments; however, the SDVs are larger for TES. This is most likely due to the fact that the TES footprints are much smaller, and therefore, the measurements present larger variability. The detailed study in this aspect is also needed, but since this is outside of the scope of this paper, it should be a subject of a future study.

6 Summary

We have developed an offline retrieval algorithm for AIRS CO using the OE technique, similar to MOPITT and TES. This has allowed us to compare the output quantities
of the retrieval with the same formulations as MOPITT and TES CO, i.e., CO VMRs, AKs, error covariance, etc., so that we can directly evaluate the performances of the instruments. We have also found that the new algorithm provides more realistic CO values compared with the in situ measurements and TES and MOPITT CO, and is consistent with prior knowledge of CO distributions.

Tropospheric CO retrievals from AIRS and TES using the OE method agree to within 5–10 ppbv or 5% on average globally and throughout the free troposphere. Since this is not the case for the operational products, we conclude that the differences between operational AIRS and TES CO are largely due to the retrieval algorithms rather than the sensor characteristics. The agreements between AIRS OE CO total columns and MOPITT V4 CO total columns have also been improved compared to AIRS operational V5 CO products. We have also showed the instrument DOFS for AIRS, TES, and MOPITT using the same formulation.

Tropospheric CO products from different space sensors (AIRS, MOPITT, TES, MLS, and IASI, etc.) that provide CO VMRs, AKs, and the error covariance can be used jointly to provide greater temporal and spatial coverage and with higher information content due to more spectral bands or higher spectral resolution. This can be achieved either through data assimilation using coupled chemical transport models and general circulation models and/or data fusion techniques, in which accurate bias estimates between data sources are essential. This study contributes to this direction by evaluating biases between sensors and by providing algorithms that minimize differences between retrieval products. Our future studies will include the cross evaluation of more sensors and possibly for the retrieval of more atmospheric trace species.

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Singh, H. B., Brune, W. H., Crawford, J. H., Flocke, F., and Jacob, D. J.: Chemistry and transport


**Table 1.** A priori information used in the retrievals of related satellite products.

<table>
<thead>
<tr>
<th>Method</th>
<th>First Guess</th>
<th>a priori</th>
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<tr>
<td>AIRS OE</td>
<td>Y_\text{MOPITT_V3}: global uniform</td>
<td></td>
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<tr>
<td>AIRS V5</td>
<td>X = \text{MEAN}(Y_\text{MOPITT_V3}) N/A</td>
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<tr>
<td>MOPITT V3</td>
<td>Y_\text{MOPITT_V3}: global uniform</td>
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<tr>
<td>MOPITT V4</td>
<td>Y_\text{MOPITT_V4}: dynamic with MOZART profiles</td>
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<td>TES</td>
<td>Y_\text{TES}: area averaged MOZART profiles</td>
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Fig. 1a. An example of the CO VMRs (ppbv) difference PDFs at 500 hPa for May 2006 over the NH land (top left), NH ocean (top right), SH land (bottom left), and SH ocean (bottom right). The PDFs are depicted as the thin black histograms and the red solid curves are the Gaussian function fits to the PDFs for each area. The vertical red dotted lines at the centers of the Gaussian functions are the modes of the function and the horizontal red dotted lines are the FWHMs of the Gaussian fits indicating the SDVs of the dataset.
Fig. 1b. The time series of the biases and the SDVs of the monthly Gaussian fits to the PDFs of the CO VMR differences at 500 hPa for the period of December 2005 to July 2008 for NH land (top left), NH ocean (top right), SH land (bottom left), and SH ocean (bottom right), respectively. Solid curves show the biases of the Gaussian fits and the dotted curves marks the bounds of the SDVs.
Fig. 2a. Same as in Fig. 1a except for at 800 hPa.
Fig. 2b. Same as in Fig. 1b except for at 800 hPa.
Fig. 3. INTEX-B DC-8 spiral profiles on 4 March 2006, over Gulf of Mexico (left) and over land to the west of Birmingham, AL near the fires (right), collocated with AIRS V5 CO (blue and cyan lines) and AIRS OE CO (red and orange lines). Cyan curves show the convolved in situ CO profile to the V5 retrievals and the orange shows that to the OE retrievals.
Fig. 4. A typical NH CO profile comparison from the retrievals on 4 March 2006, where the black curve on the top panel represents the first guess for V5 and the mean OE a priori profile. The dark blue curve on the top panel represents the 9-level AIRS V5 CO from NASA/DAAC, the red curve represents the AIRS OE CO, and the green line shows the collocated TES CO profile. The bottom panel shows the AKs from the V5 retrievals (dashed) and from the OE (solid).
Fig. 5. Same as in Fig. 4 except for the SH.
Fig. 6. The comparisons of the global CO maps in the lower troposphere where the top panel shows AIRS V5 CO at 840 hPa on 5 March 2006, the middle panel shows AIRS CO at the same level but retrieved using the OE technique, and the bottom panel shows TES CO at the similar altitudes.
Fig. 7. The PDF and the Gaussian fits of the differences between collocated AIRS OE CO and TES CO convoluted to AIRS AKs at 500 hPa and 800 hPa, respectively, globally for the period of 1 January to 31 June 2006.
Fig. 8. The comparisons of the monthly mean CO total columns ($\times 10^{17}$ molecules/cm$^2$) between AIRS OE (top right) and MOPITT V4 (bottom right) for March 2006. The top left panel shows AIRS V5 CO total columns and the left bottom panel shows the relative difference between V5 and OE: (OE-V5)/OE·100%.
Fig. 9. An example of the DOFS for March 2006 for AIRS V5 CO (blue), AIRS OE CO (red), TES CO (green) and MOPITT CO (blue), respectively, where the solid lines are the averaged DOFS in 10-degree latitude bands and the dashed lines represent the maximum and minimum DOFS values in each band.