Decadal record of satellite carbon monoxide observations

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Abstract. Atmospheric carbon monoxide (CO) distributions are controlled by anthropogenic emissions, biomass burning, transport and oxidation by reaction with the hydroxyl radical (OH). Quantifying trends in CO is therefore important for understanding changes related to all of these contributions. Here we present a comprehensive record of satellite observations from 2000 through 2011 of total column CO using the available measurements from nadir-viewing thermal infrared instruments: MOPITT, AIRS, TES and IASI. We examine trends for CO in the Northern and Southern Hemispheres along with regional trends for Eastern China, Eastern USA, Europe and India. We find that all the satellite observations are consistent with a modest decreasing trend ∼ −1 % yr⁻¹ in total column CO over the Northern Hemisphere for this time period and a less significant, but still decreasing trend in the Southern Hemisphere. Although decreasing trends in the United States and Europe have been observed from surface CO measurements, we also find a decrease in CO over E. China that, to our knowledge, has not been reported previously. Some of the interannual variability in the observations can be explained by global fire emissions, but the overall decrease needs further study to understand the implications for changes in anthropogenic emissions.

1 Introduction

Carbon monoxide (CO) is measured from space by several instruments due to its important role in atmospheric chemistry and climate. CO emissions have an indirect radiative forcing around 0.2 W m⁻² (Forster et al., 2007, IPCC AR4) since CO is a chemical precursor of greenhouse gases CO₂ and tropospheric O₃. The sources of CO are incomplete combustion processes, including fossil fuel and biofuel burning, wildfires and agricultural biomass burning, as well as secondary chemical production from hydrocarbon oxidation. The primary sink of CO is oxidation by the hydroxyl radical (OH). Because CO has a medium lifetime (weeks to months), it can be transported globally, but does not become evenly mixed in the troposphere. This makes CO an ideal tracer of transport processes from pollution sources which often produce significant enhancements over background values (e.g., Edwards et al., 2004; 2006).

Global direct emissions of CO are dominated by relatively stable anthropogenic emissions (∼500–600 Tg yr⁻¹) and by biomass burning with significant interannual variability (∼300–600 Tg yr⁻¹), (e.g., Granier et al., 2011, van der Werf et al., 2010). Satellite observations of CO can clearly identify areas of intense burning (e.g., Edwards et al., 2006;
In Sect. 3 we show the time series and trends derived from the instruments that make nadir observations of CO. Consequently, no obvious trend (increasing or decreasing) was observed in the 6-yr (2003–2008) SCIAMACHY total column CO over any of the regions. The contribution of stratospheric CO from pressures below 100 hPa to the total column is typically less than 5 % for mid-latitudes and less than 9 % for the tropics. The MOZART climatology has monthly, but not interannually variability, hence it is not a source of any decadal trends observed in the MOPITT data. In this study we use MOPITT V5 data (Worden et al., 2010; Deeter et al., 2011; Deeter et al., 2012). In order to compare with the other TIR instruments (AIRS, TES and IASI), we use the retrievals of CO from the TIR channels only (V5T data). The retrievals from the NIR channel are only possible over land for MOPITT. CO total columns from the MOPITT multispectral retrievals combining TIR and NIR channels are very similar to the TIR-only results, but are not directly comparable to any of the other instruments.
Table 1. NOAA Validation Sites for in-situ CO profiles.

<table>
<thead>
<tr>
<th>NOAA Validation Site</th>
<th>Latitude</th>
<th>Longitude</th>
<th>Observational Period</th>
</tr>
</thead>
<tbody>
<tr>
<td>Briggsdale, Colorado</td>
<td>40.37° N</td>
<td>104.3° W</td>
<td>13/01/00–30/08/11</td>
</tr>
<tr>
<td>Molokai, Hawaii</td>
<td>21.23° N</td>
<td>158.95° W</td>
<td>31/01/00–22/04/08</td>
</tr>
<tr>
<td>Poker Flat, Alaska</td>
<td>65.07° N</td>
<td>147.29° W</td>
<td>07/02/00–26/08/11</td>
</tr>
<tr>
<td>Harvard Forest, Massachusetts</td>
<td>42.54° N</td>
<td>72.17° W</td>
<td>08/02/00–18/11/07</td>
</tr>
<tr>
<td>Rarotonga</td>
<td>21.25° S</td>
<td>159.83° W</td>
<td>17/04/00–25/06/11</td>
</tr>
<tr>
<td>Charleston, South Carolina</td>
<td>32.77° N</td>
<td>79.55° W</td>
<td>22/08/03–28/09/09</td>
</tr>
<tr>
<td>Norfork, Virginia</td>
<td>36.70° N</td>
<td>75.50° W</td>
<td>28/07/04–29/04/07</td>
</tr>
<tr>
<td>Worcester, Massachusetts</td>
<td>42.95° N</td>
<td>70.63° W</td>
<td>05/01/08–18/08/11</td>
</tr>
<tr>
<td>Sinton, Texas</td>
<td>27.73° N</td>
<td>96.86° W</td>
<td>08/01/08–16/08/11</td>
</tr>
<tr>
<td>Cape May, New Jersey</td>
<td>38.83° N</td>
<td>74.32° W</td>
<td>10/01/08–17/08/11</td>
</tr>
<tr>
<td>Beaver Crossing, Nebraska</td>
<td>40.80° N</td>
<td>97.18° W</td>
<td>19/01/08–17/04/11</td>
</tr>
<tr>
<td>West Branch, Iowa</td>
<td>41.72° N</td>
<td>91.35° W</td>
<td>19/01/08–31/08/11</td>
</tr>
<tr>
<td>East Trout Lake, Saskatchewan</td>
<td>54.35° N</td>
<td>104.98° W</td>
<td>26/01/08–19/08/11</td>
</tr>
<tr>
<td>Dahlen, North Dakota</td>
<td>47.50° N</td>
<td>99.24° W</td>
<td>11/03/08–28/06/11</td>
</tr>
<tr>
<td>Trinidad Head, California</td>
<td>41.05° N</td>
<td>124.15° W</td>
<td>22/03/08–08/08/11</td>
</tr>
</tbody>
</table>

* more station information listed at http://www.esrl.noaa.gov/gmd/ccgg/aircraft/sites.html.

Fig. 1. Time series of MOPITT V4 and VST retrieval bias for total column CO determined from validation results using in-situ CO profiles from NOAA sites (listed in text). Retrieval bias statistics (including long-term drift) for individual sites are consistent with the overall statistics, i.e., no clear geographical dependence is observed in the validation results.

MOPITT V3 and V4 CO profiles and columns were validated with respect to NOAA aircraft in situ CO profile measurements (Emmons et al., 2009; Deeter et al., 2010) and we show a similar comparison for the MOPITT VST total column with respect to validation data in Fig. 1, along with the latest comparison for V4. The NOAA sites used for validation are listed in Table 1. Routine in situ profiles are collected at bi-weekly to monthly frequency, depending on the site, with ~1 ppb precision for CO (Novelli et al., 1998). MOPITT retrievals taken within 12 hours and 50 km of the NOAA flights were used for these comparisons. For MOPITT VST, the total column does not have a significant time dependent bias with respect to in-situ data. For MOPITT V4, the total column bias with respect to in situ data increased by

\[
(0.021 \pm 0.002) \times 10^{18} \text{ mol cm}^{-2} \text{ yr}^{-1}, \text{ or } \sim 1 \% \text{ yr}^{-1}. \]

This was mostly due to the use of static instrument parameters in the V4 forward model. The V5 forward model accounts for the instrument time dependence by including monthly mean values for the cell parameters. The increase in total column CO due to instrument drift in V4 is enough to obscure the decreasing trend we see in V5 data. Previous studies of global trends using MOPITT V4 data (Yurganov et al., 2010 and Fortems-Cheiney et al., 2011) show some of the same features as seen here with MOPITT V5, especially after applying a correction to agree with surface validation sites in Yurganov et al. (2010). However, with the explicit correction for instrument cell parameter drift in MOPITT V5, we now see better agreement with the other satellite observations and in-situ measurements.

2.2 AIRS

AIRS is a TIR grating spectrometer on board EOS-Aqua launched in 2002 into a sun-synchronous polar orbit with ~13:15 local time for the ascending node equator crossing (Aumman et al., 2003). AIRS has 13.5 km × 13.5 km horizontal spatial resolution, however, the AIRS algorithm utilizes cloud-clearing processes (Susskind et al., 2003) that increase the global coverage significantly, but reduce the spatial resolution to 45 km × 45 km. AIRS has a 1650 km swath providing near global coverage twice daily. AIRS resolving power is \( \lambda / \Delta \lambda = 1200 \), resulting in a spectral resolution ~1.8 cm\(^{-1}\) for the 4.6 µm CO absorption. AIRS radiance spectra are used to determine cloud and surface properties along with vertical profiles of atmospheric trace gases and temperature. AIRS CO retrievals have been compared to MOPITT and in situ data (Warner et al., 2007), which showed good agreement in AIRS-MOPITT horizontal spatial
variability. This comparison found AIRS CO higher than MOPITT (V3) by 15–20 ppb at 500 hPa over ocean observations, however, Warner et al. (2007) used older versions of both AIRS and MOPITT data than the comparisons presented here.

2.3 TES

TES is a TIR Fourier Transform Spectrometer (FTS) onboard the NASA EOS-Aura platform launched in 2004 into sun-synchronous orbit with ~ 13:40 ascending node equator crossing time. Aura follows Aqua in the so-called “A-Train” orbit, so that TES nadir observations are around 25 min later than AIRS. TES measures radiance spectra at frequencies between 650–2250 cm⁻¹ of the earth’s surface and atmosphere, with relatively fine spectral resolution (0.10 cm⁻¹, apodized) (Beer, 2006). These spectra are used to derive vertical profiles of atmospheric temperature and trace gases, along with effective cloud pressure and optical depth, surface temperature and land emissivity. TES CO profiles and total column amounts have been validated with respect to in situ measurements (Luo et al., 2007a) and compared to MOPITT V3 data (Luo et al., 2007b) and to MOPITT data processed with the TES a priori profiles and covariances (Ho et al., 2009). When a priori profiles and covariances are the same for both TES and MOPITT retrievals, column differences are less than 6.5 % globally, with MOPITT higher than TES. Unlike MOPITT, AIRS and IASI, TES has relatively sparse sampling, only along the orbit track. However, sampling was sufficient for comparisons of monthly averages in the regions chosen for trend analysis.

2.4 IASI

IASI is a TIR-FTS launched in October 2006 on the polar orbiting MetOp-A satellite, with a 9:30 a.m. local time equator crossing for the descending node. IASI has 4 pixels with 12 km diameter (at nadir) and a 2200 km swath which provide global coverage twice daily. IASI spectral resolution is 0.5 cm⁻¹ (apodized) with continuous spectral coverage from 645 to 2760 cm⁻¹ that allows column and profile retrievals of several different trace gases (Clerbaux et al., 2009). Up to now, 24 different gases have been monitored, or detected in special events such as pollution, fire or volcanic plumes (Clarisse et al., 2011). For carbon monoxide, profiles are retrieved using an optimal estimation approach, implemented in the Fast Optimal Retrievals on Layers for IASI (FORLI) software (Hurtmans et al., 2012). The spectra are processed in near real time and the CO profiles are distributed with error covariance, averaging kernels and quality flags information (see http://ether.ipsl.jussieu.fr). The IASI CO product was validated against ground based observations (Kerzenmacher et al., 2012), aircraft data (Pommier et al., 2010, Klonecki et al., 2012) and satellite measurements (George et al., 2009).

3 Trend estimation

3.1 Data selection

Figure 2 shows a global map of monthly averaged MOPITT total column CO for March, 2004 indicating the regions considered for trend estimates in this paper.

These regions were chosen to provide trends over highly populated regions in contrast to the hemispherical trends. Table 2 gives a summary of the data versions and filtering applied for each instrument and Table 3 gives the average number of observations per month (N/mo) and CO total column standard deviation for the latitude/longitude ranges used in this study. The values for N/mo and standard deviation vary significantly across the instruments due to the type of data used (gridded L3 averages vs. single L2 observations),
To account for this error, we also used the data from the MOPITT V5 Users Guide, March 2011, available at
http://www.acd.ucar.edu/mopitt/publications.shtml. MOPITT retrievals are performed for cloud free scenes using a
cloud mask determined by data from Terra-MODIS (Ackerman et al., 1998) and MOPITT (Warner et al., 2001).

For the Northern and Southern Hemispheres (NH and SH) monthly averages, we use MOPITT level 3 (L3) monthly
averages with latitudes 0 to 60°, all longitudes, daytime only. MOPITT L3 data are weighted by an observation qual-
ity index (OQI) before averaging. The OQI is defined in the MOPITT V5 Users Guide, March 2011, available at
http://www.acd.ucar.edu/mopitt/publications.shtml. Weighting by the OQI gives more weight to observations with less
geophysical noise (Deeter et al., 2011). In order to reject retrievals that have a significant influence from the apriori
profile while maintaining relatively uniform sampling across latitudes, we selected data with degrees of freedom for sig-
nal (DFS) > 0.75. For the regional averages, we use MOPITT L2 data, also weighted by the OQI and we selected
daytime, land-only scenes with DFS > 1.0. MOPITT L2 data were used so that we could apply a correction for ge-
olocation errors in MOPITT data processing (see the Ge-
edu/mopitt/publications.shtml.) To account for this error, we
added 0.35° to the longitude values before selecting and av-
ering over the latitude and longitude ranges used for this study.

### Table 2. Data selection information by instrument. Single observation uncertainty is the average retrieval error for individual total column

<table>
<thead>
<tr>
<th>Instrument type</th>
<th>MOPITT</th>
<th>AIRS</th>
<th>TES</th>
<th>IASI</th>
</tr>
</thead>
<tbody>
<tr>
<td>Data version</td>
<td>V5T</td>
<td>V5</td>
<td>V4</td>
<td>v20100815</td>
</tr>
<tr>
<td>Cloud removal method</td>
<td>Cloud-free</td>
<td>Cloud-cleared</td>
<td>Eff. OD*</td>
<td>Cloud-free</td>
</tr>
<tr>
<td>Filters</td>
<td>DFS* &gt; 0.75</td>
<td>QA* = 0 or 1</td>
<td>DFS &gt; 0.9</td>
<td>bias*</td>
</tr>
<tr>
<td>Single obs. uncertainty</td>
<td>5–6 %</td>
<td>10 %</td>
<td>6–7 %</td>
<td>5–7 %</td>
</tr>
</tbody>
</table>

* Terminology: Eff. OD = effective cloud optical depth; DFS = degrees of freedom for signal; QA = Quality Assurance (see text for value definition); QF = quality flag; RMS = root-mean-square. For IASI data, bias and RMS are in W/(cm^2 sr cm^{-1}) with ranges:

\[ -0.15 \times 10^{-9} \leq \text{bias} \leq 0.25 \times 10^{-9} \text{ and RMS} \leq 2.7 \times 10^{-9}. \]

3.1.2 AIRS data

For this study we use AIRS V5 L3 monthly mean products from the Giovanni application provided by GES-DISC (God-
dard Earth Sciences Data and Information Center). Giovanni allows selection of AIRS total column CO for user-selected
geographic areas and time ranges. Monthly average L3 data are gridded by 1° × 1° (latitude × longitude) and filtered by
Quality Assurance (QA) as good (QA = 0 or 1), where values
of 0, 1 and 2 indicate “best”, “good” and “do not use”, re-
spectively. AIRS V5 L3 monthly data were selected for day-
time (ascending orbits) and averaged for the latitude and lon-
titude ranges used in this study without further weighting or
filtering.

3.1.3 TES data

The TES L2 algorithm performs retrievals on both clear and cloudy scenes and cloudy scenes were removed before com-
puting the monthly averages used in this study. TES V4 L2 data were selected for daytime retrievals with quality flag = 1
(good) and DFS > 0.9. Criteria for cloud-free data are an
effective cloud optical depth (OD) < 0.4 and land-only data
were used in the regional averages. Note that starting in 2010,
routine sampling of the SH was limited to northward of 30° S
in order to conserve instrument lifetime. Therefore, TES data
are not included for the SH after 2009.

3.1.4 IASI data

The IASI column values used here are daytime data that
were processed with version 20100815 of the FORLI-CO
retrieval algorithm. All data corresponding to a cloud frac-
tion flag between 0 and 25 % were further analyzed (Cler-
baux et al., 2009). Standard filtering as recommended in
Hurtmans et al. (2012) was applied to residual radiances (in
Table 3. Averages for number of observations per month and CO column standard deviation by instrument. N/mo is the average number of observations per month and \( \sigma \) is the CO total column standard deviation (in \( 10^{18} \) molecules cm\(^{-2}\)). For AIRS L3 data, we show a representative month (4/2006), with N/mo as the number of 1\( ^\circ \) × 1\( ^\circ \) bins in each latitude/longitude range.

<table>
<thead>
<tr>
<th></th>
<th>MOPITT</th>
<th>AIRS</th>
<th>TES</th>
<th>IASI</th>
</tr>
</thead>
<tbody>
<tr>
<td>NH</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>0–60(^\circ) N</td>
<td>(L2) 1.091 ( \times 10^6 )</td>
<td>21569</td>
<td>8220</td>
<td>2.021 ( \times 10^6 )</td>
</tr>
<tr>
<td>0–60(^\circ) S</td>
<td>(L3) 19.558</td>
<td>0.27</td>
<td>0.37</td>
<td>0.44</td>
</tr>
<tr>
<td>SH</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>0–60(^\circ) S</td>
<td>(L2) 877 203</td>
<td>21 599</td>
<td>6161</td>
<td>1.924 ( \times 10^6 )</td>
</tr>
<tr>
<td>E. China(^1)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>N/mo</td>
<td>3222</td>
<td>130</td>
<td>86</td>
<td>10 658</td>
</tr>
<tr>
<td>( \sigma )</td>
<td>0.28</td>
<td>0.16</td>
<td>0.34</td>
<td>0.40</td>
</tr>
<tr>
<td>E. USA(^2)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>N/mo</td>
<td>3276</td>
<td>100</td>
<td>76</td>
<td>88 27</td>
</tr>
<tr>
<td>( \sigma )</td>
<td>0.49</td>
<td>0.29</td>
<td>0.69</td>
<td>0.97</td>
</tr>
<tr>
<td>Europe(^3)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>N/mo</td>
<td>1823</td>
<td>149</td>
<td>43</td>
<td>6184</td>
</tr>
<tr>
<td>( \sigma )</td>
<td>0.16</td>
<td>0.14</td>
<td>0.30</td>
<td>0.30</td>
</tr>
<tr>
<td>India(^4)</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>N/mo</td>
<td>12 003</td>
<td>250</td>
<td>63</td>
<td>20 062</td>
</tr>
<tr>
<td>( \sigma )</td>
<td>0.27</td>
<td>0.40</td>
<td>0.33</td>
<td>0.54</td>
</tr>
</tbody>
</table>

\(^1\) E. China [30\(^\circ\)–40\(^\circ\) N, 110\(^\circ\) –123\(^\circ\) E]
\(^2\) E. USA [35\(^\circ\)–40\(^\circ\) N, –95\(^\circ\) to –75\(^\circ\) E]
\(^3\) Europe [45\(^\circ\)–55\(^\circ\) N, 0\(^\circ\)–15\(^\circ\) E]
\(^4\) India [20\(^\circ\)–30\(^\circ\) N, 70\(^\circ\)–95\(^\circ\) E].

3.2 Monthly mean time series and instrument biases

The seasonal behavior of background levels of CO in the atmosphere is dominated by the changes in incoming solar radiation, thus OH radical concentrations. Reactions with OH determine the CO lifetime, which varies from a few weeks in summer to several months in winter at mid-latitudes (e.g., Holloway et al., 2000). Build-up of CO over the winter produces a CO maximum in late winter to early spring, after which chemical destruction results in a minimum for late summer to early fall. For the SH, the maximum due to CO lifetime corresponds to September/October, which is also around the peak in biomass burning for South America and Africa (south of the equator). This results in sharper peaks for the SH compared to NH, where sources of CO are more spread out over the year.

Figure 3 shows the monthly mean time series for the NH (0–60\(^\circ\) N) for all the instrument data records available for March 2000 to May 2012. Figure 4 shows the SH (0–60\(^\circ\) S) monthly mean time series. All instruments show reasonable...
consistency in column amounts and seasonal patterns. The AIRS data show a clear high bias with respect to the other instruments in the SH and a smaller seasonal amplitude in the NH. These biases with MOPITT, TES and IASI have been documented previously by George et al. (2009) and Warner et al. (2010), and are due to differences in instrument spectral resolution and retrieval methods. In particular, the high bias in SH AIRS V5 data is most likely due to the use of a single global first guess profile (Warner et al., 2010). The use of a single global a priori profile was the source of the high bias in SH MOPITT V3 data, which was removed in MOPITT V4 and V5 by the use of the MOZART climatology for a priori profiles (Deeter et al., 2010). However, AIRS seasonal patterns are consistent with the other instruments and here we are comparing the relative time dependence, which should be less influenced by overall measurement bias.

In this study, we have not made any corrections to account for differences in horizontal spatial sampling or vertical sensitivity. Except for the expected bias with AIRS, the remaining spread in CO column values across the instruments is largely due to spatial sampling differences. In the case of TES, the number of observations is 2 orders of magnitude lower than the other instruments, and varied for different years. Therefore, we would not expect perfect agreement for an average of TES observations that are not co-located with the other instruments. In the SH, CO distributions are more driven by biomass burning and observations of any one transient fire event could vary significantly for the different instrument sampling regimes. This is apparent in the IASI data for October, 2010 when IASI sampled intense burning in southern Africa about twice as frequently than MOPITT. Total column results are less sensitive than profile retrievals to differences in vertical sensitivity that are due to instrument parameters, such as spectral resolution, and the choice of constraints in the retrieval algorithms. However there are differences in the instrument column averaging kernels, as shown in George et al. (2009), that would give slightly different results if applied to the same atmospheric distribution.

3.3 Trend calculation

3.3.1 12-month running averages

Since we are interested in decadal scale trends, we first remove the seasonal variations in CO using a 12-month running average. Various options are available for truncating the end-points, but here we simply do not use the first and last 6 months for any of the running average plots or trend calculations since these could introduce a dependence on the smoothing method or at worst, spurious results. To produce a running average that properly removes the seasonal cycles, we found that it was necessary to first perform a cubic spline interpolation over gaps in the monthly averages due to halted instrument operations (shown in Table 2).

3.3.2 Linear trend analysis

To compute linear trends we apply a linear least squares fit to the 12-month running averages, weighted by the standard deviation of the measurements used in the monthly average. For the datasets with months where there were gaps in the data, we use the data record mean standard deviation for each region scaled by a factor of 100 so that these months are de-weighted in the fit. We report 1σ errors from the linear fits. Figure 5 shows the 12-month running averages and linear fits for 11-yr trends with MOPITT CO total column data for all the spatial regions considered. Table 4 lists slope values, in % yr⁻¹, for the time periods covered by each of the instruments (where we have cut out at least 6 months at the beginning and end, to account for edge truncation in the 12-month running average and to avoid fractional years, if possible). This is a very simplified approach to finding trends, and we will explore the variation in trends computed by different methods (such as different data filtering, different smoothing and empirical orthogonal functions (EOF) analysis) in future analyses. However, for this study, we wanted to show the comparison of all the data records using very basic methods.

For comparisons of relative interannual changes in all the instruments, we normalize each of the 12-month running average time series by the average CO column for a common year, 8/2008 to 7/2009. This removes some of the effects of bias between the instruments and allows a simple assessment of fractional changes over any time period. Figures 6 and 7 show MOPITT, AIRS, TES and IASI 12-month running average time series normalized by 8/2008–7/2009 average CO columns for the NH and SH, respectively, along with linear trends over the MOPITT and AIRS data records. Figure 8

Fig. 5. MOPITT 11-yr regional trends in total CO column change per year. Regional time series with 12-month running average and corresponding linear trends are indicated by the different colors, with slope and 1σ error given in molecules/cm² yr⁻¹.
shows the normalized time series and linear trends for China, E. USA, Europe and India. For the TES data records shown in Fig. 8, there are a limited number of observations over these regions, listed in Table 3, which can result in a sampling bias with respect to the other instruments, especially over regions with large variability in CO, such as China. Over China, this is further exacerbated by a lower than average number of samples per month for 2005–2007 (∼60 observations per month) while 2008–2009 had higher than average sampling (∼130 observations per month). For the other regional and hemispherical averages, TES sampling was more uniformly distributed with time.

Table 4 lists the slopes for NH and SH, in % yr$^{-1}$, with 1σ error, for the time range of each data record. None of the time ranges show trends that are significant to >2σ for any of the instruments, so the hemispherical trends only provide the indication of an overall decrease. For the regional trends, we find >2σ significance in China (−1.6 to −1 % yr$^{-1}$), E. USA (−1.4 to −1 % yr$^{-1}$) and Europe (−1.4 to −1 % yr$^{-1}$), for the MOPITT 11-yr and AIRS 8-yr records respectively. The TES and IASI data records do not show significant trends over any of the regions, but show overall consistency with the MOPITT and AIRS time dependence and give confidence to the use of the longer MOPITT and AIRS data records for trend determination.

![Fig. 6. 12-month running averages for N. Hemisphere total column CO measurements normalized by the 08/2008–07/2009 average CO column for each instrument.](https://www.atmos-chem-phys.net/13/837/2013/)
4 Discussion

4.1 Interannual variability

Figures 6–8 demonstrate good agreement in interannual variability for all instruments except in the cases of known sampling issues discussed in Sec. 3. As further verification of the consistency in the MOPITT and AIRS datasets for detecting patterns in interannual variability, Figure 9 shows a zonal average time record of normalized CO total column. 12-month running averages of MOPITT and AIRS CO total columns in 10° latitude bands are normalized to the time average for each band. The variations in CO for both instruments are very similar in time and latitude, and both observe a deep minimum in 2009–2010.

In Fig. 10, we show the MOPITT and AIRS 12-month running averages, normalized to the 2008 average column, for a 10°×10° spatial average around the NOAA Mauna Loa (MLO) site at 19.5° N and 155.6° W (Novelli and Masarie, 2010). Due to the high altitude of the site, Mauna Loa measurements are more representative of the free troposphere than other surface measurements and often show the influence of air transported from Asia. Although the in-situ measurements of CO vmr are not directly comparable to total column CO, and have significantly more interannual variability, MOPITT and AIRS observe many of the same features seen in the MLO in-situ data time series.

All of the 12 month running average time series of total column CO show clear features due to changes in fire emissions such as the broad peak around 2002–2003 for NH observations due to a significant number of boreal fires (e.g., Yurganov et al., 2005) and an overall decline in global fire emissions from 2007 to 2009 as shown in the GFED3 inventory (van der Werf et al., 2010). There may also be some correlation of lower CO column to lower emissions due to the global financial crisis starting in late 2008. The observed minimum in CO in early 2009 overlaps with the recorded dip in manufacturing (Alcorta and Nixson, 2011) along with related transportation and shipping decreases (de Ruyter de Wildt et al., 2012). Declines due to the economic crisis are clearly observed for NO2 in Europe (Castellanos and Boersma, 2012) and the US (Russell et al., 2012), however, given the longer lifetime of CO, this would require further study (inverse modeling) for attribution.

4.2 Trend evaluation

Only the longer satellite data records (8–11 yr) have trends that are significant above 1σ. This underscores the importance of maintaining consistent long-term CO observations. Using CO measurements from 1996 to 2007, both Zellweger et al. (2009) and Angelbratt et al. (2011) find decreasing trends for CO in Europe. The most comparable measurements to satellite total column CO measurements are those of Angelbratt et al. (2011) that use partial columns of CO (0–15 km) derived from ground-based solar FTIR (Fourier Transform InfraRed) stations at four locations in Europe. They find trends with >2σ significance of −0.45 ± 0.15 ppb yr−1 to −1.00 ± 0.25 ppb yr−1 for 1996–2006. Although we cannot compare to this time range, our MOPITT and AIRS trends in Europe are consistent with the direction and magnitude of these previously computed trends, and show a somewhat steeper decline after 2007.

Using surface in-situ measurements, Novelli et al. (2003) reported a decreasing trend in the NH for 1991 to 2001 of 0.92 ± 0.15 ppb yr−1 (or −0.8 ± 0.2% yr−1 for average NH values of ~120 ppb). Our results are consistent with a continuation of this decreasing trend for the NH. The US Environmental Protection Agency (EPA) reports a national decrease in surface CO of 52% from 2001 to 2010 (EPA, 2011). This corresponds to −5.2 ± 0.2% yr−1, which is larger than the MOPITT trend for E. USA, (−1.4 ± 0.2) % yr−1, but surface CO, measured at mostly urban sites, would be expected to show larger changes compared to total column CO.

Li and Liu (2011) report a decreasing trend in surface CO in Beijing for 2000 to 2009 from in-situ measurements collected from 8 urban sites using commercial gas filter correlation analyzers. Using the values shown in Fig. 6 of Li and Liu, (2011), this trend appears to be similar to the US EPA surface CO trends, ~6% yr−1. Li and Liu (2011) also examined MOPITT V4 CO data over Beijing, which did not show a decreasing trend. As described in Sect. 2.1 and shown in Fig. 1, the uncorrected instrument drift in MOPITT V4 data resulted in an increasing CO column trend that is large enough to obscure the decreasing trends we observe in this study.

In order to understand these trends and the contributions from anthropogenic emissions, fire emissions and possible changes in atmospheric chemistry, a future study for attribution will be needed. This will require data assimilation.
Fig. 8. 12-month running averages for total column CO measurements over E. China, E. USA, Europe and India, normalized by the 08/2008–07/2009 average CO column for each instrument. See text for description of TES bias due to limited sampling, especially for China during 2005–2007.

and inversion with a model that controls for emissions and OH. Previous studies for trends in CO emissions have been shown from inversions using MOPITT V4 (Fortems-Chiney et al., 2011; Hooghiemstra et al., 2012), but this type of study should be updated with MOPITT V5, as well as the AIRS, TES and IASI measurements.

Although this study does not determine how much of the CO trends reported here are due to changes in anthropogenic emissions, it is interesting to note the reported trends from Granier et al. (2011) from comparing several global and regional emission inventories. They show a very slight decrease in global anthropogenic CO emissions (around 1%) from 1990 to 2010, with larger decreases for Europe and USA (~3% yr⁻¹). However, they show increases for India (~1.5% yr⁻¹) and China (~3% yr⁻¹) for 2000 to 2010. For India, our results show an overall slight decrease, but seem to indicate an increase until about 2007. For China, Granier et al. (2011) include the updated emissions in 2006 from Zhang et al. (2009), but likely did not consider more recent updates to activity data in China such as the inventory used to model CO emissions before and during the 2008 Beijing Olympics (Worden et al., 2012). For Beijing, there were reductions in CO emissions by 2008 (even without considering the reductions from restrictions during the Olympics) due to such measures as stricter vehicle emission standards, phasing out of residential coal stove use and changes in the industry sector. These changes in the technology mix will have a significant impact on combustion efficiency and therefore CO emissions. A trend toward improved combustion efficiency from 2005 to 2008 was also found using CO₂-CO correlations at a rural site near Beijing by Wang et al. (2010).

We also note that trends in CO will not necessarily correlate with trends in nitrogen dioxide (NO₂). A significant increase in NO₂ has been observed using satellite data over China in the last decade (e.g., Richter et al., 2005; Zhang et al., 2007). Due to the shorter lifetime of NO₂ (hours to days) compared to CO, observations of high NO₂ concentrations are more co-located with NOx emission sources, and trends in NO₂ abundance are more easily attributed to trends in emissions. For China, NOx emissions are mostly from the power and transportation sectors (Zhang et al., 2009) and have grown significantly since 2000 with the increase in demand for electricity (Zhang et al., 2007). CO emissions, however, have a relatively small contribution (~2%) from the power sector (Zhang et al., 2009). Vehicle emission standards and improved combustion efficiency for newer cars will lower CO emissions, but it is not clear whether NOx emissions for the China transportation sector are reduced by the same factor as CO for these changes.
Fig. 9. Time series of normalized CO total column for zonal, 12-month running averages. Total CO columns are normalized to the average column in each 10° latitudinal band. (a) shows MOPITT data from Sep. 2000 to Nov. 2011. (b) shows AIRS data from April 2003 to November 2011. (First and last 6 months of 12-month running averages are not considered, as described in the text.)

5 Conclusions

We have shown that all nadir-viewing TIR satellite measurements of CO total column are consistent with decreasing trends in CO over the last decade in both Northern and Southern Hemispheres and for specific highly populated regions in the Northern Hemisphere. Hemispherical trends are only significant at the 1σ level but regional trends for China, E. USA and Europe are significant to > 2σ. In particular the decreasing trend in CO over China of 1.6 ± 0.5 % yr⁻¹ over 2002–2012 has not been reported before, to our knowledge.

Future studies are needed to test the hypotheses of what is driving the observed hemispherical and regional trends, i.e., changes in emissions (both anthropogenic and biomass burning) or atmospheric chemistry. Although this study does not provide attribution for the decadal changes in CO, we note that changes in fire emissions may explain some of the observed interannual variability, especially in the Southern Hemisphere. We also discussed trends in anthropogenic CO emissions for China that may need to be updated once new inversion results that use recent satellite data versions are available.

These results clearly demonstrate the need for satellite data records that span more than a decade for quantifying trends. Since the US instruments (MOPITT, AIRS and TES) are maturing, it is important to understand the differences in those data records as compared to IASI, which has planned future missions of the same instrument configuration. IASI-B on MetOp-B launched on 17 September 2012 and a third IASI launch is scheduled for 2016. Combined with the existing data records, the IASI instruments will potentially provide 20+ years of continuous and consistent data for CO and other trace gas species.

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