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Biomass burning emission inventory with daily resolution: Application to aircraft observations of Asian outflow

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[1] We develop a daily-resolved global emission inventory for biomass burning using AVHRR satellite observations of fire activity corrected for data gaps and scan angle biases. We implemented this inventory in a global three-dimensional model (GEOSCHEM) to simulate aircraft CO observations during the TRACE-P mission over the NW Pacific in February–April 2001. Seasonal biomass burning in SE Asia was a major contributor to the outflow of Asian pollution observed in TRACE-P and shows large day-to-day fluctuations that vary depending on location. Three simulations were conducted with the same 3-month total (February–April) emissions but different temporal distributions: 2001 daily resolved, 2001 monthly resolved, and climatological monthly resolved. The effect of daily resolved versus monthly resolved 2001 emissions in the simulation of CO is less than 8 ppbv in Asian outflow over the NW Pacific but can exceed 100 ppbv over source regions. The relatively small effect in Asian outflow reflects spatial and temporal averaging of emissions during ageing in the continental boundary layer. Significant improvement in the simulation of TRACE-P observations (as diagnosed by the resolved variance) is found when using 2001 monthly versus climatological monthly emissions, but using 2001 daily emissions does not offer further improvement.

INDEX TERMS: 0322 Atmospheric Composition and Structure: Constituent sources and sinks; 0345 Atmospheric Composition and Structure: Pollution—urban and regional (0305); 0365 Atmospheric Composition and Structure: Troposphere—composition and chemistry; KEYWORDS: biomass burning, TRACE-P, AVHRR, Asia, pollution


1. Introduction

[2] Biomass burning in the tropics is a major source of trace gases and aerosols to the atmosphere [Crutzen and Andreae, 1990; Andreae and Merlet, 2001]. It exhibits large day-to-day variability reflecting weather, agricultural practices, and accidental fires. Previous global three-dimensional (3-D) model studies of tropospheric composition have not accounted for this variability. This may lead to problems when using the models to interpret observed atmospheric concentrations in terms of the magnitude and distribution of biomass burning emissions. We present here an algorithm for resolving the day-to-day variability in biomass burning emissions using space-based fire count data, and we apply it to a 3-D model interpretation of Asian pollution outflow observed during the NASA Transport and Chemical Evolution over the Pacific (TRACE-P) aircraft mission over the NW Pacific in February–April 2001 [Jacob et al., 2003].

[3] Measurements of fire from space are available from a number of sensors. These include the Along Track Scanning Radiometer (ATSR) [Arino and Rosaz, 1999], the Tropical Rainfall Measuring Mission (TRMM) [Giglio et al., 2000], the Advanced Very High Resolution Radiometer (AVHRR) [Stroppiana et al., 2000] and the Moderate Resolution Imaging Spectrometer (MODIS) [Kauffman et al., 1998]. All of these instruments are presently flying on polar-orbiting satellites. Coverage is global but data density can be low, particularly after considering data loss from cloudy scenes. Observations in geostationary orbit would afford higher density, but so far only a few such data sets are available, such as the GOES observations in the Western Hemisphere [Prins et al., 1998]. We use here the AVHRR data, which are available as daily global fire counts for the February–April 2001 period. AVHRR fire observations have been employed previously [Kauffman et al., 1990; Christopher et al., 1998] to make regional estimates of biomass burning emissions.

[4] TRACE-P was conducted at the peak of the biomass burning season in SE Asia and India [Duncan et al., 2003].
Biomass burning outflow from SE Asia is transported to the North Pacific primarily by warm conveyor belts ahead of cold fronts at 25–30°N, and by convective outflow at lower latitudes [Bey et al., 2001b; Liu et al., 2003]. We evaluate our biomass burning inventory and the impact of day-to-day variability for the TRACE-P period with a global 3-D model (GEOS-CHEM) [Bey et al., 2001a] simulation of aircraft observations of carbon monoxide (CO). CO is a product of incomplete combustion with large sources from both biomass and fossil fuel burning. We also use TRACE-P measurements of HCN [Singh et al., 2003] and C2Cl4 [Blake et al., 2003] to separate biomass burning and fossil fuel contributions to the CO observations.

2. Construction of Daily Biomass Burning Emission Inventory

[5] Our approach is to start from a climatological biomass burning emission inventory with 1° × 1° spatial resolution and monthly temporal resolution [Duncan et al., 2003] and apply day-to-day variability to that inventory for February–April 2001 using a daily compilation of AVHRR fire count data. Major gaps and biases need to be removed from the AVHRR data, as described below.

2.1. World Fire Web (WFW) Daily Fire Product

[6] The AVHRR data employed in this work were obtained from the World Fire Web (WFW) maintained by the Global Vegetation Monitoring Unit at the European Union Joint Research Centre in Italy (available at http://www.gvm.sai.jrc.it/fire/wfw/wfw.htm). AVHRR is a satellite-borne scanning radiometer which operates in the visible and infrared and was first launched in 1978 aboard TIROS-N [Kidwell, 1998]. Current generations of this instrument are aboard the National Oceanic and Atmospheric Administration (NOAA)-14 and NOAA-15 platforms. Satellite data from AVHRR are processed by a global network of 22 receiving stations or nodes. Daily fire maps assembled by the WFW from these 22 nodes provide nearly global coverage. The data obtained for this study cover February–April 2001 and were obtained from the NOAA-14 AVHRR instrument. This instrument has a 2400 km across-track swath and provides global coverage every day with a pixel resolution of 1.1 × 1.1 km at nadir and up to 2.4 × 6.9 km off-nadir. It orbits the Earth 14 times a day with an 8-day repeat period and has an ascending orbit time of 1340 local time [Kidwell, 1998]. Agricultural fire activity tends to peak in early afternoon [Langaas, 1992] and thus AVHRR is well suited for detecting biomass burning.

[7] The WFW active fire algorithm for AVHRR is presented by Stroppiana et al. [2000] and is based on an examination of absolute temperatures and temperature gradients between neighbouring pixels. It is a modified version of the Flasse and Ceccato [1996] algorithm. Fire counts are registered for hot pixels which emit stronger infrared radiation than their nonfire neighbors. A cloud detection algorithm, based on reflectance and temperature, is applied to identify pixels obscured by cloud. Sun-glint and hot-ground false detections have been recognised as the principal sources of error [Stroppiana et al., 2000]. The former is caused by reflection of sunlight from nonpermanent water or ice. The latter can be the consequence of heterogeneous land types where hot and cold pixels are adjacent to one another (e.g., paved highways). Validation of the WFW fire product by Giglio et al. [1999] suggests that while the detection of large fires (>1000 m²) is robust there is greater uncertainty in small fire detection.

[8] The WFW provides a 0.5° × 0.5° gridded product of fire data which we regridded to 1° × 1° to match the Duncan et al. [2003] biomass burning inventory. The data include the number of fire counts per grid square (nF), the number of pixels which are obscured by cloud per grid square (nC) and the total number of pixels per grid square (nT) (Figure 1). The WFW product has not been corrected for satellite overpass strip overlap (i.e., double-counting) but we apply this correction through normalization of the data, as described below.

[9] Coverage of the WFW data (Figure 1a) is limited both by the satellite instrument measurement track and by the contributing receiving stations in the network. Receiving stations for eastern and southern Africa and for northern South America did not submit data in February–April 2001. Cloud cover (nC/nT) is less than 40% for much of the

![Figure 1](image-url)
northern tropics (dry season), including SE Asia and India. Comparison with the International Satellite Cloud Climatology (available at http://isccp.giss.nasa.gov/) indicates that extensive clouds are erroneously reported in the AVHRR product over high northern latitudes. This is likely due to reflection from snow and ice.

[10] Cloudy pixels may contain wet vegetation which does not readily burn, however fire activity has been observed in persistently cloudy regions [Robinson, 1991]. In the absence of evidence to the contrary, we assume that the probability of fires in cloudy pixels for a given $1^\circ \times 1^\circ$ square is the same as in clear-sky pixels. The resulting Fraction On Fire (FOF) for the $1^\circ \times 1^\circ$ grid square is calculated as:

$$FOF = \frac{n_F}{n_T-n_c}$$  \hspace{1cm} (1)

This approach corrects not only for clouds but also for double counting when polar orbital tracks overlap. FOF values in fires regions are typically in the $10^{-4} - 10^{-2}$ range. We do not calculate FOF values for grid squares with cloud fractions greater than 90%, because of excessive uncertainty.

### 2.2. Correction of Scan Angle Bias

[11] Our initial analysis of the WFW fire data revealed a bias in the fire detection efficiency for the off-nadir pixels. This bias is illustrated for two scenes in Figure 2. Nadir view for a given scene occurs only every 8 days. As scan angle increases, the coverage, or the total number of pixels observed, decreases (Figure 2, top panel) [Frulla et al., 1995]. This decrease simply reflects the viewing geometry and should not affect the FOF. However, we find that the FOF peaks for nadir viewing and drops to near zero at large scan angles (Figure 2, middle panel), reflecting a bias in the detection efficiency previously identified qualitatively by Giglio et al. [1999]. This is further illustrated in Figure 3 (top panel), which shows the probability of nonzero FOF in a $1^\circ \times 1^\circ$ grid square as a function of the number of pixels viewed ($n_T$) for the ensemble of the East Asian region (10–50\textdegree N, 70–160\textdegree E) in February–April 2001. Nadir viewing yields $n_T = 1.25 \times 10^4$ daily pixel observations per $1^\circ \times 1^\circ$ grid square. Low values of $n_T$ are associated with high scan angles. Values of $n_T > 1.25 \times 10^4$ reflect double counting and are also associated with large scan angles. For $n_T$ values within the near-nadir range [$1 \times 10^4$, $1.25 \times 10^4$], the probability of nonzero FOF is within 50% of that for the nadir view.

[12] The bottom panel of Figure 3 shows the mean values of FOF as a function of $n_T$ for the subset of grid squares with nonzero FOF. A linear relationship is found for $n_T$ in the range [$5 \times 10^3$, $1.25 \times 10^4$], implying that nonzero FOF values observed within that $n_T$ range provide useful infor-
data are available (Figure 1). Figure 2 (bottom panel) shows the corrected FOF product for the two illustrative sites grid squares.

2.3. Construction of Daily Emission Inventory

[14] We apply the daily global distribution of FOFs from section 2.2 to impose a day-to-day variability for February–April 2001 to the climatological biomass burning emission inventory from Duncan et al. [2003]. This inventory, originally described by Lobert et al. [1999], is based on data from the last 15 years for savannah burning, deforestation, forest fires, shifting agriculture, the burning of shrub and grassland, and burning in fields. It includes seasonal variation with monthly resolution prescribed from a 5-year record of ATSR and AVHRR fire observations. The global emission of CO is 437 Tg CO yr\(^{-1}\); approximately 50% of February–April emissions are in Asia (Figure 4).

[15] Duncan et al. [2003] employed the TOMS Aerosol Index (AI) [Herman et al., 1997] to quantify the 1979–2001 interannual variability in biomass burning emissions. Figure 5 shows that burning in SE Asia in February–April 2001 was 87% of the climatological average, although there are large month-to-month departures as described below. The TOMS AI does not provide useful information on interannual variability in biomass burning over India due to excessive anthropogenic haze.

[16] We assume here that the geographical distribution of the 3-month average emissions for February–April 2001 is that given by the climatology and we use our FOF product to distribute the emissions temporally over these three months. For regions with less than 5 days of nonzero FOF data for February–April 2001, we default to a 31-day running average of the climatological emissions. This affects mainly eastern Africa and northern South America (Figure 1). Figure 6 shows the resulting emissions of CO for three regions in SE Asia and India. There is large day-to-day variability as well as significant, regionally variable departures of the 2001 monthly means from the climatology (Figure 4). Emissions in March 2001 were generally less than the climatology, while emissions in February 2001 were slightly higher.

3. Application to CO Simulation in Asian Outflow

[17] We used our daily resolved biomass burning emission inventory in a global 3-D model to simulate observations of CO concentrations during TRACE-P, and we compared with results from simulations using climatological and 2001 monthly resolved emissions. The simulations were done using the GEOS-CHEM model [Bey et al., 2001a; Martin et al., 2002] driven by assimilated meteorological data from the Goddard Earth Observing System (GEOS)-3 of the NASA Data Assimilation Office (DAO). The original GEOS-3 meteorological data have \(1^\circ \times 1^\circ\) horizontal resolution. We use here GEOS-CHEM v. 4.20 (available at http://www-as.harvard.edu/chemistry/trop/geos) with a 2° latitude by 2.5° longitude resolution and 48 vertical sigma levels. Applications of GEOS-CHEM to different aspects of the TRACE-P observations are presented by C.L. Heald et al. (Asian outflow and transpacific transport of carbon monoxide and ozone pollution: An integrated satellite, aircraft, and model perspective, submitted to Journal of Geophysical Research, 2003), Li et al.
Liu et al. [2003], and Palmer et al. [2003]. Kiley et al. [2003] present an evaluation of the GEOS-CHEM simulation of CO transport in TRACE-P and compare it to that of other 3-D models used in TRACE-P.

[18] We conducted a CO-only simulation with specified monthly mean OH fields as described by Bey et al. [2001b]. The CO source is from B.N. Duncan et al. (manuscript in preparation, 2003) and is an improved version of that described by Bey et al. [2001a]. Sources of CO from fossil fuel, biofuel, and biomass burning include both direct emissions and chemical production from the oxidation of methane and nonmethane volatile organic compounds (NMVOCs); the latter is accounted for by appropriate scaling of the CO emissions. The CO originating from biomass burning, biofuels, and fossil fuels is transported as separate tracers, as in the work of Bey et al. [2001b]. A 1-year simulation is used to ensure proper initialization.

[19] We describe results from three simulations. The first uses climatological biomass burning emissions with a monthly temporal resolution and will be referred to as “climatological monthly” in the remainder of this paper. The simulations which employ the 2001 AVHRR-defined emissions with monthly and daily resolution will be referred to as “2001 monthly” and “2001 daily,” respectively. Our total biomass burning emissions for East Asia (the domain of Figure 4) in spring 2001 are 20.8 Tg CO in February, 25.4 Tg CO in March, and 23.1 Tg CO in April.

[20] The mean simulated CO concentrations over the TRACE-P period and the mean absolute differences between the 2001 monthly and 2001 daily model simulations are shown in Figure 7 for three different altitudes. Mean absolute differences were calculated by taking absolute differences in model output sampled every 3 hours and averaging them over the TRACE-P deployment period (25 February–10 April). Biomass burning outflow to the Pacific is mainly restricted to the free troposphere at 15–30°N [Liu et al., 2003]. Mean absolute differences are typically 4–8 ppbv in the free troposphere outflow (10–20% of biomass burning CO concentrations) and up to 100 ppbv in the continental source region. For the ensemble of conditions observed by the TRACE-P aircraft, the model sampled along the flight tracks shows less than 10 ppbv difference between the 2001 daily and monthly simulations in 91% of cases and at most a 30 ppbv difference.

Figure 4. Climatological [Duncan et al., 2003] and 2001 AVHRR-derived monthly total CO emissions from biomass burning in East Asia in February–April. See color version of this figure at back of this issue.
Figure 5. Interannual variability of fire activity in SE Asia (9–29N, 96–119E) for the 3-month period February–April as measured by the TOMS Aerosol Index (1979–2001) [Duncan et al., 2003]. The figure shows the relative deviation from the climatological mean (standard deviation is shown as dashed lines). The data gap from 1994 to 1996 is due to the interruption of TOMS observations between the decommissioning of the Nimbus-7 TOMS and the launch of the Earth-Probe TOMS instrument. The TOMS Aerosol Index for 2001 is a factor of 0.87 lower than climatology, within one standard deviation of 0.21.

[21] Point-by-point comparison of model results to the ensemble of aircraft observations west of the dateline and south of 30°N is shown as a scatterplot in Figure 9. Model simulations were sampled along the aircraft flight track, and observations along the flight track were averaged over model gridboxes. The climatological monthly simulation shows an $R^2$ of 0.46 compared to the improved $R^2$ (0.52) achieved with the 2001 monthly simulation. The 2001 daily simulation does not capture any more variability in the observations ($R^2 = 0.51$) than the 2001 monthly simulation. The lack of improvement in $R^2$ between the 2001 monthly and daily simulations reflects the small differences between the two simulations (Figure 8) and also model transport errors in lifting biomass burning effluents to the free troposphere [Kiley et al., 2003]. Inadequate resolution of individual convective events may limit the advantage to be gained from fine temporal resolution of emissions. The regressions in Figure 8 indicate a low bias in the model which may reflect a combination of errors associated with the anthropogenic and biomass burning sources, and with the magnitude of the OH sink. Further discussion of the constraints from the model simulation on sources during TRACE-P is given in the work of Palmer et al. [2003]. Palmer et al. [2003] use a lower OH concentration (GEOSCHEM v. 4.33) than we do here, resulting in a higher background CO concentration.

[22] Air masses sampled during TRACE-P represented a complicated mixture of source types [Liu et al., 2003]. Our model simulations suggest that biomass burning contributed approximately 30% of the CO observed south of 30°N during TRACE-P. Further assessment of the effect of daily resolution of biomass burning emissions in our model can be made by isolating the biomass burning component of CO in the observations. For this purpose we use the urban tracer perchloroethene (PCE) [Blake et al., 2003] and the biomass burning tracer hydrogen cyanide (HCN) [Singh et al., 2003; Li et al., 2003]. Both PCE and HCN have atmospheric lifetimes of the order of months [Blake et al., 1996; Li et al., 2000]. We find that a multivariable linear fit of CO to PCE and HCN accounts for 77% of the observed variability of CO in the 5-min average data west of the dateline and above 2 km altitude (where most of the biomass burning influence was observed) (Figure 10). Single-variable fits to PCE and HCN can account for 58% and 38% of the variability, respectively. From the linear relationship $[CO] = a[PCE] + b[HCN] + c$ with coefficients $a$, $b$, and $c$ as given in Figure 10, we estimate the biomass burning contribution to the 5-min CO average concentrations as the corresponding value of $b[HCN]$. Figure 11 shows that the improvement in the quality of the simulation (as measured by $R^2$) when using

Figure 6. Total daily CO emissions from biomass burning in February–April 2001 for 3 regions: NE India, Burma, and Cambodia. Daily emissions are shown as the thin solid line, monthly average emissions as a thick solid line and climatological monthly emissions [Duncan et al., 2003] as the dotted line.
2001 rather than climatological monthly emissions increases as the biomass burning contribution to CO in the observations increases. Again, the 2001 daily simulation does not improve significantly over the 2001 monthly simulation.

4. Conclusions

[24] We developed an approach for incorporating daily variability into a biomass burning emission inventory by using global satellite observations of fire activity. We used for this purpose the daily AVHRR WFW satellite data. Our initial analysis revealed a scan angle bias in these data, which we corrected but at the loss of over half of the data. The corrected fire product was employed to produce a daily biomass burning inventory for the spring of 2001. Most of the burning during that season is in East Asia, and we find large daily variability in biomass burning emissions with different patterns within the region. The daily resolved biomass burning inventory was incorporated in a global 3-D chemical tracer model (GEOS-CHEM) and used to simulate observed CO concentrations in Asian outflow over the NW Pacific from the TRACE-P aircraft mission (February–April 2001). We find that accounting for the daily variability of biomass burning in the model modifies CO concentrations by up to 8 ppbv in Asian outflow over the NW Pacific and by 100 ppbv or more in continental source regions, relative to a simulation using monthly mean emissions.

shows some improvement in the simulation of TRACE-P aircraft observations relative to the simulation with monthly resolved climatology ($R^2$ increases from 0.46 to 0.52). The improvement is particularly pronounced for the TRACE-P data subset most heavily influenced by biomass burning. There is no further improvement in the simulation of observations when the daily resolved 2001 inventory is employed. We attribute this result in part to the spatial and temporal averaging of biomass burning influences in the free tropospheric outflow over the NW Pacific sampled in TRACE-P and in part to model transport noise that limits

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**Figure 8.** Effect of daily and interannual variability in biomass burning emissions on the GEOS-CHEM model simulation of CO concentrations sampled along the TRACE-P flight tracks south of 30°N. The figure shows frequency distributions of the difference between the 2001 monthly and 2001 daily simulations (solid line; mean difference 1.5 ppbv) and the difference between the 2001 monthly and climatological monthly simulations (dotted line; mean difference −6.8 ppbv).

**Figure 9.** Point-by-point comparison of GEOS-CHEM model results to the ensemble of TRACE-P aircraft observations south of 30°N and west of the dateline. The lines of best fit for the climatological monthly (black) 2001 monthly (red) and 2001 daily (blue) simulations are shown. The one-to-one line is shown in dotted black. Statistics were computed with the reduced-major-axis method [Hirsch and Gilroy, 1984]. See color version of this figure at back of this issue.

**Figure 10.** Linear fit of CO to PCE and HCN concentrations for the TRACE-P data subset above 2 km altitude, west of the dateline and excluding stratospheric air. The fit is $[\text{CO}] = a[\text{PCE}] + b[\text{HCN}] + c$, where $a = 9.96 \times 10^{-3}$, $b = 270$ and $c = 24.5$ ppbv.

**Figure 11.** Correlation between simulated (GEOS-CHEM) and observed (DC-8) CO concentrations as a function of the biomass burning CO concentration identified from the observations by linear fit to PCE and HCN concentrations (see text). Results from the simulation with climatological monthly emissions (solid) are compared with those using 2001 monthly emissions (dashed) and 2001 daily emissions (dash-dot). The numbers across the top of the figure indicate the number of points in each averaging interval.
the advantage to be gained from fine temporal resolution of emissions.

[26] Our results indicate that daily resolution of biomass burning emissions is not critical for modeling outflow from tropical regions on a global scale. Such resolution is certainly more important when simulating large isolated wildfire events [Wotawa and Trainer, 2000] or when interpreting satellite observations of biomass burning effluents [Thomas et al., 1998, Spichtinger et al., 2001]. Bottom-up inventories based on satellite observations of burned areas and vegetative cover [Michel et al., 2002] may offer a more accurate picture of biomass burning emissions in the future.

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References


Figure 1. Mean statistics of the AVHRR World Fire Web product for February–April 2001, mapped on a $1^\circ \times 1^\circ$ grid. (a) Percentage of days with available data; (b) cloud cover; (c) daily fire counts. Blank areas have either no fires or no data.
Figure 4. Climatological [Duncan et al., 2003] and 2001 AVHRR-derived monthly total CO emissions from biomass burning in East Asia in February–April.
Figure 7. Simulated CO concentrations for the TRACE-P deployment period (25 February–10 April 2001) and effect of daily resolved biomass burning emissions. Left panels: mean CO concentrations at different altitudes in the 2001 daily simulation. The color bar is saturated at 150 ppbv to emphasize the structure of the Asian outflow over the western Pacific sampled by TRACE-P; surface values over Asia can exceed 600 ppbv. Middle panels: mean biomass burning CO concentrations. The color bar is saturated at 60 ppbv. Right panels: average absolute differences between CO concentrations in the 2001 monthly and 2001 daily simulations. Color bar is saturated at 8 ppbv; absolute differences over Asia can exceed 120 ppbv.
Figure 9. Point-by-point comparison of GEOS-CHEM model results to the ensemble of TRACE-P aircraft observations south of 30°N and west of the dateline. The lines of best fit for the climatological monthly (black) 2001 monthly (red) and 2001 daily (blue) simulations are shown. The one-to-one line is shown in dotted black. Statistics were computed with the reduced-major-axis method [Hirsch and Gilroy, 1984].