CO₂ semiannual oscillation in the middle troposphere and at the surface

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[1] Using in situ measurements, we find a semiannual oscillation (SAO) in the midtropospheric and surface CO₂. Chemistry transport models (2-D Caltech/JPL model, 3-D GEOS-Chem, and 3-D MOZART-2) are used to investigate possible sources for the SAO signal in the midtropospheric and surface CO₂. From model sensitivity studies, it is revealed that the SAO signal in the midtropospheric CO₂ originates mainly from surface CO₂ with a small contribution from transport fields. It is also found that the source for the SAO signal in surface CO₂ is mostly related to the CO₂ exchange between the biosphere and the atmosphere. By comparing model CO₂ with in situ CO₂ measurements at the surface, we find that models are able to capture both annual and semiannual cycles well at the surface. Model simulations of the annual and semiannual cycles of CO₂ in the tropical middle troposphere agree reasonably well with aircraft measurements.

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1. Introduction

[2] Atmospheric CO_2 has a trend of ~ 2 ppm/yr based on measurements from Mauna Loa covering from 1958 to 1994 [Keeling et al., 1995]. The increasing atmospheric CO_2 has a significant impact on the global climate system [Dickinson and Cicerone, 1986]. Superimposed upon this trend is an annual cycle resulting from the uptake and release of CO_2 by vegetation whose amplitude is greatest in the northern hemisphere (NH). Using CO_2 measurements at Mauna Loa, Buermann et al. [2007] found that variations of CO_2 seasonal cycle amplitudes are closely related to carbon sequestration in the biosphere, and are influenced by precipitation and circulation. In addition to the trend and annual cycle, atmospheric CO_2 also demonstrates intraseasonal and interannual variabilities.

[3] El Niño is the most important tropical interannual variability that can influence the CO₂ concentrations. During El Niño (La Niña) events, the atmospheric CO₂ growth rate

increases (decreases) at tropical surface stations [Keeling et al., 1995; Jones et al., 2001; Nevison et al., 2008]. Using midtropospheric CO₂ data from the Atmospheric Infrared Sounder, Jiang et al. [2010] found that El Niño can influence the midtropospheric CO₂ concentrations. Midtropospheric CO₂ is enhanced in the central Pacific Ocean and diminished in the western Pacific Ocean during El Niño [Jiang et al., 2010]. In the high latitudes, midtropospheric CO₂ concentration can be influenced by the strength of the polar vortex. Polar midtropospheric CO₂ is reduced (enhanced) when the polar vortex is strong (weak) [Jiang et al., 2010]. Recently, Li et al. [2010] demonstrated that midtropospheric CO₂ concentrations are also modulated by the Madden-Julian oscillation.

[4] In this paper, we will focus on investigating the intraseasonal variability of midtropospheric CO₂, especially on the CO₂ semiannual oscillation (SAO) and its possible causes. This work will yield a quantitative study of how SAO influences the midtropospheric CO₂. It also offers an opportunity to investigate the possible cause for the SAO signal in the midtropospheric CO₂.

2. Data and Models

[5] In this paper, we use aircraft CO₂ from *Matsueda et al.* [2002], which are incorporated into the Comprehensive Observation Network for Trace gases by AIrLiner (CONTRAIL). Aircraft CO₂ from *Matsueda et al.* [2002] are measured at 8–13 km biweekly since April 1993 to the present over the western Pacific from Australia to Japan. The latitudinal coverage is approximately from 25°S to 35°N. The longitudinal coverage is from 135°E to 150°E. We also use surface CO₂ flask measurements from NOAA ESRL network [*Tans et al.* 1998; *Earth System Research Laboratory*

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(ESRL), 2007]. Site information for NOAA surface CO₂ is available at http://www.esrl.noaa.gov/gmd/dv/site/site_table.html.

[6] To investigate the possible causes of the semiannual oscillation of midtropospheric and surface CO₂, we use three different chemistry transport models. These models are the Caltech/JPL 2-D chemistry transport model (CTM) [Shia et al., 2006], 3-D GEOS-Chem [Suntharalingam et al., 2004], and 3-D MOZART-2 [Horowitz et al., 2003]. The 2-D CTM has 18 latitudes, equally spaced from pole to pole. It has 40 vertical layers, equally spaced in log scale of pressure from the surface to the upper boundary at 0.01 hPa. Transport in the model is by the stream function and the horizontal and vertical diffusivities taken from Jiang et al. [2004]. The stream function is derived from the National Center for Climate Prediction (NCEP) Reanalysis 2 data [Jiang et al., 2004]. An important feature of the 2-D CTM is its ability to reproduce the age of air in the stratosphere [Morgan et al., 2004].

[7] GEOS-Chem (v7.3.3) is driven by the Goddard Earth Observing System (GEOS-4) assimilated meteorological data from the NASA Global Modeling Assimilation Office (GMAO). Spatial resolution for GEOS-Chem is 2° (latitude) $\times 2.5^{\circ}$ (longitude). There are 30 levels in the vertical from the surface to about 0.01 hPa (\sim 70 km). Advection is computed every 15 min with a flux form semi-Lagrangian method [Lin and Rood, 1996]. Moist convection is computed using the GEOS convective, entrainment, and detrainment mass fluxes described by Allen et al. [1996a, 1996b]. The physics in the GEOS-4 analysis system is adopted from the National Center for Atmospheric Research (NCAR) Community Climate Model, Version 3 (CCM3) and Whole Atmosphere Community Climate Model (WACCM) with important modifications to make it suitable for data assimilation [Bloom et al., 2005].

[8] MOZART-2 is driven by the meteorological inputs every 6 h from the NCEP Reanalysis 1 [Kalnay et al., 1996]. Advection is computed every 20 min with a flux form semi-Lagrangian method [Lin and Rood, 1996]. The horizontal resolution is 2.8° (latitude) × 2.8° (longitude) with 28 vertical levels extending up to approximately 40 km altitude [Horowitz et al., 2003]. MOZART-2 is built on the framework of the Model of Atmospheric Transport and Chemistry (MATCH). MATCH includes representations of advection, convective transport, boundary layer mixing, and wet and dry deposition.

[9] Surface emissions and vertical transport in CTMs are both crucial for CO₂ simulation in the free troposphere. We will use two different boundary conditions to investigate how boundary conditions affect the midtropospheric CO₂. The GLOBALVIEW-CO₂ mixing ratio data [*Tans et al.* 1998; *ESRL*, 2007] are used as the lower boundary condition for the Caltech/JPL CTM, GEOS-Chem, and MOZART-2. For convenience, we refer to this hereforth as the GLOBALVIEW-CO₂ boundary condition [*Jiang et al.*, 2008]. Since the GLOBALVIEW-CO₂ data are limited in space, especially over the oceans, we used the GLOBALVIEW-CO₂ to rescale the monthly mean CO₂ mixing ratios at the surface for the GLOBALVIEW-CO₂ boundary condition. First, we use seasonal varying CO₂ source and sink flux boundary condition to drive the model. We also

interpolate monthly mean GLOBALVIEW-CO $_2$ measurements to the model resolution. Then, we rescale the zonal mean CO $_2$ mixing ratio in the boundary by the monthly mean GLOBALVIEW-CO $_2$ measurements for each month and for each latitudinal band. The monthly mean GLOBALVIEW-CO $_2$ flask data are close to the colocated GLOBALVIEW-CO $_2$ boundary condition.

[10] We will also use prescribed CO₂ sources and sinks as the boundary condition for GEOS-Chem and MOZART-2. The exchange of CO₂ between the terrestrial biosphere and atmosphere is based on net primary productivity and respiration fluxes from the Carnegie-Ames-Stanford (CASA) ecosystem model [Randerson et al., 1997]. Monthly mean biospheric CO₂ fluxes from 2000 to 2004 are used in the models, which includes interannual variability as that used in Feng et al. [2011]. Air-sea exchange of CO₂ is taken from Takahashi et al. [1997], which is an annual mean ocean CO₂ flux. Estimates of fossil fuel emissions are from G. Marland et al. (Global, Regional, and National Fossil-Fuel CO₂ Emissions, 2007, http://cdiac.ornl.gov/trends/emis/overview 2007. html), which is also an annual mean CO₂ flux. Monthly mean biomass burning emissions of CO₂ are based on Duncan et al. [2003]. Since there is an unbalanced CO₂ budget associated with the prescribed source and sink boundary condition [Suntharalingam et al., 2003, 2004], we regress surface CO₂ mixing ratio in the GEOS-Chem restart file against the GLOBALVIEW-CO2 surface flask measurements. As a result, the unbalanced CO₂ budget is resolved to some degree [Jiang et al., 2008]. Discrepancies between the model CO₂ simulations (driven by the same meteorological fields) with the above mentioned two boundary conditions would help identify potential issues with the surface sources and/or sinks on simulating CO₂ annual and semiannual cycles.

3. Results

[11] Figure 1 presents a comparison between Matsueda's aircraft CO₂ (red dots) and model CO₂ mixing ratios averaged between 9 km and 13 km (solid lines) from 2000 to 2004. Figures 1a–1f show 25°S, 15°S, 5°S, 5°N, 15°N, and 25°N, respectively. Different color lines are for different model simulations. There are two GEOS-Chem model outputs. One is forced by the GLOBALVIEW-CO₂ boundary condition (green line). The other is forced by the prescribed CO₂ sources and sinks boundary condition (orange line). GEOS-Chem CO₂ forced by the prescribed CO₂ source/sink boundary condition (orange line) have higher CO₂ concentrations in the summer seasons than that forced by GLOBALVIEW-CO₂ boundary condition (green line). It suggests that there might be a missing sink in the prescribed CO₂ source/sink boundary condition in the summer season. The purple line is CO₂ from Caltech 2D model. The blue line is CO₂ from MOZART2 forced by NCEP1 meteorology. The model results match the high-precision aircraft measurements of CO₂ in the middle troposphere remarkably well. Seasonal cycle and trend for CO₂ are simulated well by different models. The amplitude of CO₂ seasonal cycle increases with latitudes, with larger seasonal cycle in the northern hemisphere compared with that in the southern hemisphere. In addition to the annual cycle, there is a

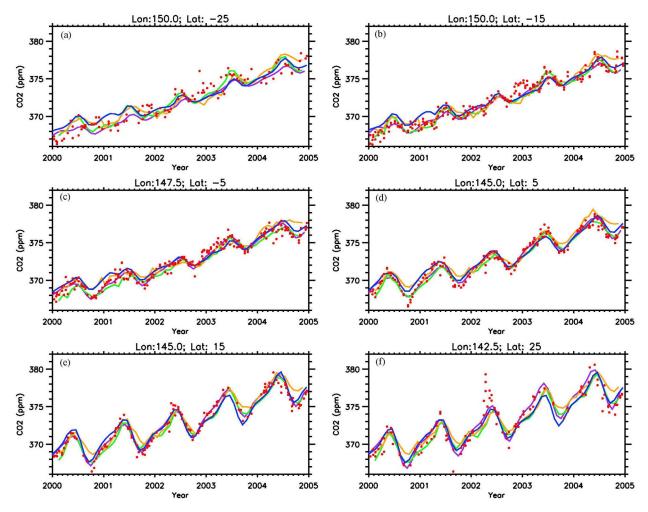


Figure 1. (a–f) Aircraft observations between 9 km and 13 km (red dots) [*Matsueda et al.*, 2002] and model CO₂ mixing ratios (color lines). The CO₂ mixing ratios from the GEOS-Chem model forced by the GLOBALVIEW-CO₂ boundary condition and prescribed CO₂ source/sink boundary condition are shown by the green and orange lines, respectively. The CO₂ mixing ratios from the Caltech/JPL 2-D model forced by NCEP2 and GLOBALVIEW-CO₂ BC are shown by the purple line. The CO₂ mixing ratios from MOZART-2 forced by NCEP1 and GLOBALVIEW-CO₂ BC are shown by the blue line.

6 month signal appearing in the CO_2 from both aircraft and model simulations. To investigate the 6 month signal in more detail, power spectral analysis is applied to the detrended CO_2 from aircraft and model simulations. Linear trends have been removed from CO_2 time series. Power spectra for the detrended CO_2 are shown in Figure 2. In addition to the spectral peak at 12 months, there are 6 month signals appearing in the power spectra of Matsueda's CO_2 and model CO_2 .

[12] To investigate the causes of the semiannual oscillations in the midtropospheric CO₂, we first apply sensitivity studies to the 2-D Caltech/JPL chemistry and transport model. Averaged CO₂ at 9–13 km forced by the GLOBALVIEW-CO₂ boundary condition and NCEP2 meteorology is shown by the solid line in Figure 3. To investigate the annual cycle and semiannual cycle amplitudes in the midtropospheric CO₂, we fit the data by a series of Legendre polynomials and harmonic functions

[Jiang et al., 2008]. We use the sum of the first, second, and third Legendre polynomials to remove the trend from the data. The harmonic functions represent annual and semiannual cycles. Annual cycle, calculated by e cos $(2\pi t) + f\sin(2\pi t)$, is shown in Figure 3b, where e and f are the amplitudes of the annual cycle. The amplitude for the annual cycle of 2-D model CO₂ at 25°N is about 2.3 ppm. The semiannual cycle, calculated by $g\cos(4\pi t) + h\sin(4\pi t)$, is shown in Figure 3c, where g and h are the amplitudes of the semiannual cycle. The amplitude for the semiannual cycle of 2-D model CO₂ at 25°N is about 0.8 ppm.

[13] In a sensitivity test, we remove the annual and semiannual oscillations from surface CO_2 and use linear trends as the boundary condition at the model surface. As such, there is no semiannual cycle and annual cycle source originating from surface in this model run. This results in the reduction of the amplitude of semiannual cycle for midtropospheric CO_2 to 0.09 ppm (black dotted line in Figure 3c), which is

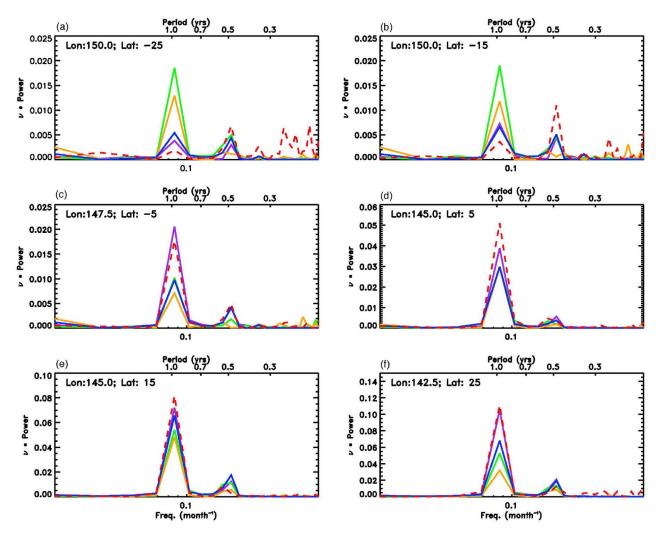


Figure 2. Power spectra for aircraft and model CO_2 time series. Red dashed line is the power spectra for the *Matsueda et al.* [2002] data. Colors for solid lines are the same as in Figure 1.

about 11% of the total amplitude of the semiannual oscillation in the midtropospheric CO₂. It clearly suggests that the dominant cause of the semiannual cycle in the middle troposphere is from the surface sources. Weak semiannual and annual cycles in the midtropospheric CO₂ shown by black dotted lines in Figures 3b and 3c originate from transport fields. The phase of the CO2 seasonal cycle due to the transport (black dotted line in Figure 3b) is shifted relative to that forced by the GLOBALVIEW-CO₂ boundary condition and NCEP2 reanalysis meteorology field (black solid line in Figure 3b). The CO₂ seasonal signal due to the transport (black dotted line in Figure 3b) is related to the strength of the vertical velocity in the 2D CTM. The vertical velocity is stronger in the summer season and weaker in the winter season. More CO₂ can be lifted to the middle troposphere during the summer than in the winter season. Thus the midtropospheric CO₂ (black dotted line in Figure 3b) reaches maximum value in the summer season at 25°N. Although the vertical velocity peaks in summer, the CO₂ seasonal cycle (forced by the GLOBALVIEW-CO₂ boundary condition and NCEP2 reanalysis meteorology field;

black solid line in Figure 3b) reaches maximum in April as a result of summertime drawdown by the biosphere. In another sensitivity test, we force the model with linear CO_2 trend boundary condition and climatological transport fields. Climatological transport fields are the average of the transport fields from 2000 to 2004. There is no semiannual oscillation originating from either the surface or the transport fields. As a result, the semiannual oscillation disappears in midtropospheric CO_2 as shown by the black dashed line in Figure 3c.

[14] To further reveal how the surface annual cycle and semiannual cycle relate to the semiannual oscillation in the midtropospheric CO_2 , we decompose the net exchange between the biosphere and atmosphere (NEP) into the annual cycle and the semiannual cycle using the multiple regression method. Then, we use the combined CO_2 linear trend and annual cycle from the NEP as the boundary condition for the model. Result from this sensitivity study is shown by the green dotted line in Figure 3. We also use combined CO_2 linear trend and semiannual cycle from NEP as the boundary condition for the model, and result is shown

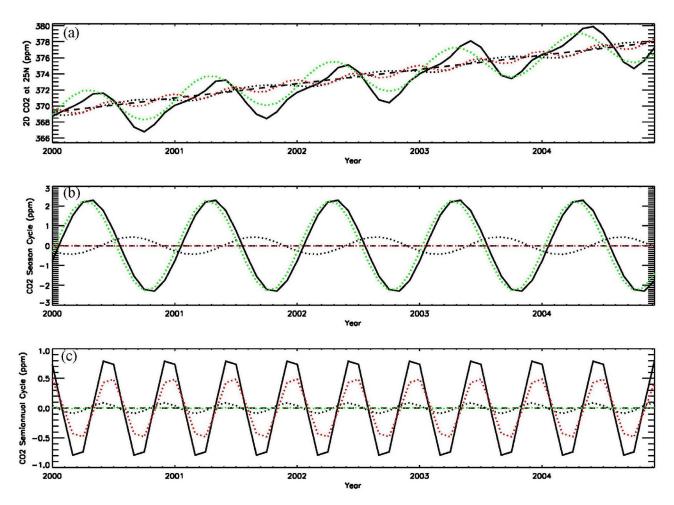


Figure 3. (a) Caltech/JPL 2-D model CO₂ at 25°N. (b) Seasonal cycle of model CO₂ at 25°N. (c) Semiannual cycle of model CO₂ at 25°N. Model CO₂ forced by the GLOBALVIEW-CO₂ boundary condition and NCEP2 reanalysis meteorology field are shown by the black solid line. Model CO₂ forced by the linear CO₂ trend boundary condition and NCEP2 reanalysis meteorology are shown by the black dotted line. Model CO₂ forced by the linear CO₂ trend boundary condition and climatology transport field are shown by the black dashed line. Model CO₂ forced by the combined linear CO₂ trend and annual cycle boundary condition and climatology transport field are shown by the green dotted line. Model CO₂ forced by the combined linear CO₂ trend and semiannual cycle boundary condition and climatology transport field are shown by the red dotted line. Units are ppm.

by the red dotted line in Figure 3. As shown in Figure 3, the semiannual oscillation signal in the midtropospheric CO_2 is mainly from the surface semiannual cycle instead of the surface annual cycle. From Figure 3c, we also find that the NEP is the major contribution to the SAO signal in the midtropospheric CO_2 and counts about 60% of the total SAO signal. In addition to the contribution from transport and the biosphere, CO_2 fluxes from biomass burning and exchange with the ocean can also contribute to the SAO signal in the middle troposphere. When we have better ocean CO_2 surface emissions in the future, we can decompose the contributions to the SAO signal into different components in a future paper.

[15] Annual and semiannual oscillations in surface CO₂ are also examined. Similar spectral analysis is applied to the GLOBALVIEW-CO₂ and GEOS-Chem model CO₂ at the

surface. In addition to the annual cycle, semiannual oscillation signals are also present in the surface CO₂ from GLOBALVIEW-CO₂ and model CO₂. To compare the annual cycle and semiannual cycle amplitudes in surface CO₂ from observations and model results, we calculate the annual cycle amplitude $(\sqrt{e^2+f^2})$ and semiannual cycle amplitude $(\sqrt{g^2 + h^2})$ for surface CO_2 from GLOBAL-VIEW network and GEOS-Chem model output. Results are shown in Figure 4. The amplitudes for annual and semiannual cycles from GEOS-Chem CO2 are very close to those from GLOBALVIEW-CO2 at the surface. The amplitudes of annual and semiannual cycles are larger in the northern hemisphere compared with those in the southern hemisphere. The maximum amplitude for the annual cycle of surface CO_2 is about ~ 10 ppm. The maximum amplitude is about ~ 3.5 ppm for the semiannual cycle of surface CO_2 .

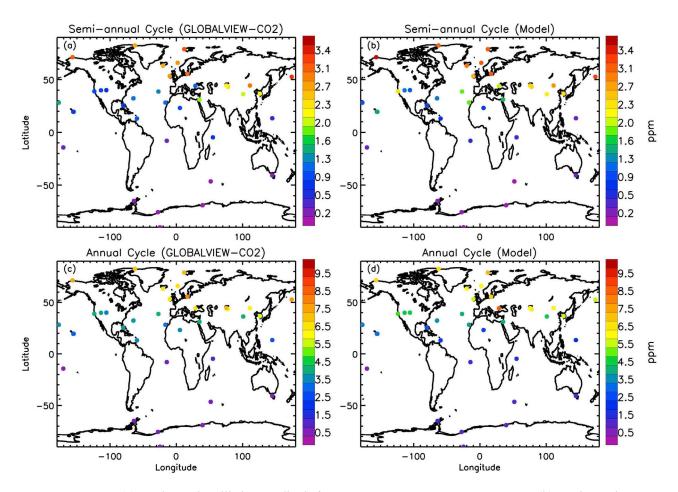


Figure 4. (a) Semiannual oscillation amplitude from GLOBALVIEW-CO₂ measurement. (b) Semiannual oscillation amplitude from GEOS-Chem model CO₂. (c) Annual cycle amplitude from GLOBALVIEW-CO₂ measurement. (d) Annual cycle amplitude from GEOS-Chem model CO₂. Units are ppm.

Scatterplots of the observed and model simulated amplitudes for semiannual cycle and annual cycle of surface CO₂ are shown in Figure 5. As shown in Figure 5, the GEOS-Chem model seems to overestimate the semiannual oscillation

amplitudes at some stations. This might relate to the relatively coarse spatial resolution in the model. The latitudinal distributions of the amplitudes for the semiannual and annual cycles are shown in Figure 6. The amplitude increases with

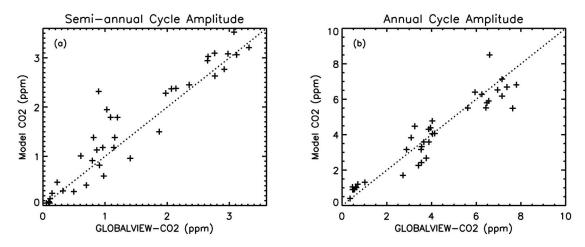


Figure 5. (a) Scatterplot of the semiannual cycle amplitude for GLOBALVIEW-CO₂ and GEOS-Chem model CO₂. (b) Scatterplot of the annual cycle amplitude for GLOBALVIEW-CO₂ and GEOS-Chem model CO₂.

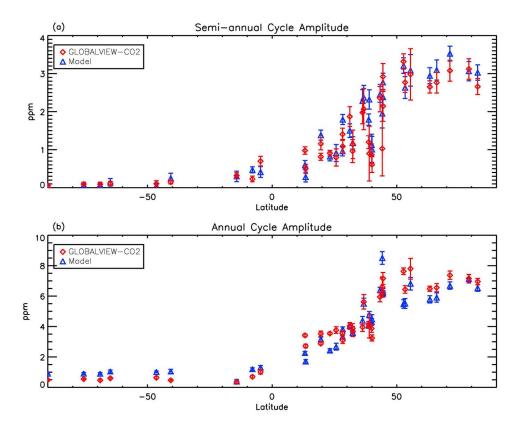


Figure 6. (a) Latitudinal distribution of the semiannual cycle amplitude. (b) Latitudinal distribution of the annual cycle amplitude. Diamonds are the results from GLOBALVIEW-CO₂. Triangles are the results from the GEOS-Chem model. Error bars are the uncertainties of the results at each latitude band.

latitude, which is consistent from both surface GLOBAL-VIEW-CO₂ data and models. Semiannual and annual cycle amplitudes are larger in the northern hemisphere than in the southern hemisphere. This is because the semiannual and annual cycles in surface CO₂ sources (e.g., the net exchange between biosphere and atmosphere) are larger in the northern hemisphere than those in the southern hemisphere.

[16] To investigate possible causes for the semiannual oscillation of CO₂ at the surface, we examine the signals from different CO₂ surface sources, which include biomass burning, fossil fuel emission, ocean, and biosphere. Fossil fuel emission contributes to the positive trend in CO₂. CO₂ semiannual cycle and annual cycle are mainly from exchange between the atmosphere and the biosphere. Biomass burning also contributes to the semiannual cycle of CO₂. In the current model, the CO₂ flux from ocean is an annual mean data. When we have better CO2 fluxes from ocean in the future, we can explore the SAO signal from the ocean in an independent paper. Gross primary production, respiration, and net ecosystem production at 30°N and 110°E, shown in Figure 7, are an example to illustrate the semiannual oscillation in CO₂ source from the biosphere. Gross primary production (Figure 7a) is related to carbon uptake by plants during photosynthesis. The values are negative since CO2 is taken up by vegetation from the atmosphere. Ecosystem respiration (Figure 7b) is related to the autotrophic and heterotrophic respirations from biosphere. In the winter season, photosynthesis is largely

reduced. The peak for gross primary production (Figure 7a) is relatively flat in winter. However, there are still CO₂ emitted to the atmosphere by respirations from the biosphere in winter, which has a relatively sharp peak compared with the photosynthesis term. The sum of the two terms, gross primary production and ecosystem respiration, leads to the double peaks in each year in the net ecosystem production, as shown in Figure 7c. Thus, phase differences in the gross primary production (photosynthesis) and ecosystem respiration lead to the semiannual oscillation in CO₂ at the surface. Surface semiannual oscillation can propagate to the middle troposphere, where it produces the semiannual oscillation in the midtropospheric CO₂.

4. Conclusions

[17] In addition to the annual cycle, the semiannual oscillation of midtropospheric and surface CO₂ is discussed in this paper by combining the in situ measurements with chemistry transport models. Chemistry and transport models, driven by different transport schemes, are used to simulate the middle tropospheric CO₂. We also apply different boundary conditions to force the 3-D CTMs. The seasonal cycle and semiannual oscillation of surface CO₂ are well simulated by chemistry transport model with the prescribed CO₂ sources and sinks boundary condition. The semiannual oscillation is also found in the midtropospheric CO₂. From the sensitivity study, we found that the semiannual

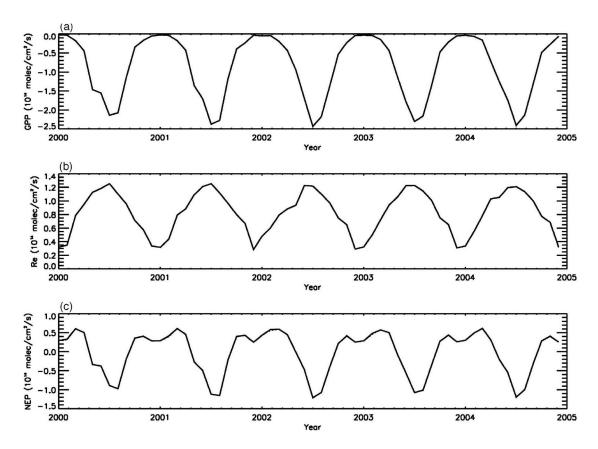


Figure 7. (a) Gross primary production, (b) ecosystem respiration, and (c) net ecosystem production from the Carnegie-Ames-Stanford (CASA) ecosystem model at 30°N and 110°E.

oscillation in the midtropospheric CO₂ originates mainly from sources at the surface. A possible reason for the semi-annual oscillation of surface CO₂ is the CO₂ surface source due to net ecosystem production.

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