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Correlations between interannual variations of simulated global and regional CO₂ fluxes from terrestrial ecosystems and El Niño Southern Oscillation

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ABSTRACT
Using a terrestrial ecosystem model (Sim-CYCLE), CO₂ exchange between the biosphere and the atmosphere was simulated for the period of 1980–2000. As a result, the biosphere was totally a sink of CO₂, and the phase of the interannual variation of the global annual CO₂ flux from the biosphere was similar to that of the annual increase of the global atmospheric CO₂. The interannual variation of the global annual CO₂ flux showed high correlation with that of the annual mean ENSO index (NINO.3 monitoring index) for a time lag of 1 yr. As for the regional annual carbon fluxes, some showed high correlation with the annual mean ENSO index for a time lag of 1 yr and some without time lag. Among the regional carbon fluxes those showed high correlations with ENSO index without time lag, some showed positive correlations and some showed negative correlations, and they canceled out each other in total. These negative correlations were found in tropical regions. All the regional carbon fluxes those showed high correlations with ENSO index for a time lag of 1 yr showed positive correlations.

1. Introduction
The atmospheric CO₂ keeps increasing year after year, but its annual growth rate has large interannual variation (Conway et al., 1994). The CO₂ emission from combustion of fossil fuel, the largest source of the atmospheric CO₂, is estimated to be quite constant with a small uncertainty (Marland et al., 2009). The estimation of the CO₂ emission from land use change has a large uncertainty (Houghton, 2003), though the interannual variation of the annual flux from this category is thought to be not as large as that of the annual increase of the atmospheric CO₂ because land use change is a human activity. Therefore, the absorption of the atmospheric CO₂ by terrestrial ecosystems and oceans seems to have a large interannual variation.

It is well known that there is a lagged correlation between the increase of the atmospheric CO₂ and ENSO index (Bacastow, 1976; Thompson et al., 1986; Elliott et al., 1991). Rayner et al. (1999) suggested that the correlation might be caused by initial response of the CO₂ fluxes from tropical oceans and later response of the terrestrial CO₂ fluxes to ENSO. However responses of terrestrial ecosystems to ENSO should not be globally simultaneous and uniform. It is very interesting why such a distinct lagged correlation is shown between ENSO and the global total carbon flux.

Interannual variations of terrestrial regional CO₂ fluxes were estimated by synthesis inversion methods using atmospheric transport models (Bousquet et al., 2000; Patra et al., 2005; Peylin et al., 2005; Gurney et al., 2008) and numerical simulations using terrestrial ecosystem models (Kindermann et al., 1996; Jones et al., 2001; Cao et al., 2005). Peylin et al. (2005) improved understanding of the interannual variation of the atmospheric CO₂ by comparing the results of an inversion method and model simulations. In most of the studies above, correlations between ENSO and anomalies of CO₂ fluxes were approached. Especially for the interannual variation of the anomaly of the carbon flux from tropical area, it is reported that its amplitude is dominant (Kindermann et al., 1996; Bousquet et al., 2000) and it tends to be positive in the year of El Niño (Bousquet et al., 2000; Jones et al., 2001; Patra et al., 2005). However the observed global total carbon flux to the atmosphere shows correlation with ENSO for a time lag of 1 yr. It is also an interesting point.

To investigate the contributions of the terrestrial regional carbon fluxes to the atmospheric CO₂, carbon exchange between the atmosphere and the biosphere was simulated using a
physiological process-based terrestrial ecosystem model (Sim-CYCLE). From the result, the interannual variation of the global annual carbon flux was compared with that of the annual increase of the atmospheric CO2. Then, correlations between the global and regional annual carbon fluxes and the annual mean ENSO index were analysed, and contributions of the regional carbon fluxes to the global carbon flux were investigated.

2. Model and data

Sim-CYCLE (Simulation model of Carbon Cycle in Land Ecosystems) (Ito and Oikawa, 2002) calculates production by photosynthesis, distribution, respiration, litter fall and decomposition of organic matter on physiological processes. Sum of the emissions by respiration and decomposition minus the absorption by photosynthesis is treated as the carbon flux from the biosphere to the atmosphere. It is assumed that each grid area of the model is dominated by one of 11 vegetable types. In this version of the model, photosynthesis process does not depend on soil water. The horizontal resolution is 2.5 × 2.5 degree, and the time step is 1 d.

As external input data of Sim-CYCLE, downward short-wave radiation, total cloud cover, specific humidity, air temperature, pressure, soil moisture and soil temperature of NCEP/NCAR (National Centers for Environmental Prediction, National Oceanic and Atmospheric Administration/The National Center for Atmospheric Research, University Corporation for Atmospheric Research) Re-analysis data set and monthly CO2 data of WMO/WDCGG (World Meteorological Organization/World Data Centre for Greenhouse Gases) were used. These data sets were spatially and temporarily interpolated according to the grids and the time step of Sim-CYCLE.

3. Methods

To simulate the carbon exchange between the biosphere and the atmosphere, Sim-CYCLE was run for the period of 1751–2000. This section explains the details of the methods.

3.1. Change of atmospheric CO2 for 1751–2000

This simulation needs atmospheric CO2 data for the period of 1751–2000. To estimate the global mean atmospheric CO2 concentration, ice core records (Nefel et al., 1985; Friedli et al., 1986; Siegenthaler et al., 1988; Barnola et al., 1995) were used. The global mean atmospheric CO2 is assumed to be constant at 280 ppm until 1750. After 1959, it is assumed to be same as the annual mean data at Mauna Loa. For the period of 1751–1958, it is interpolated using the ice core records (Fig. 1). To obtain the horizontal distribution and its seasonal cycle of the CO2 anomaly, WMO/WDCGG 1990 monthly data were interpolated spatially and temporarily. The anomaly distribution was added to the global annual mean CO2 value of each year, and used as input data of Sim-CYCLE.

3.2. Parameter adjustment and initial condition setting

Using 1951–2000 NCEP/NCAR Re-analysis climate data and 1990 WMO/WDCGG CO2 data repeatedly, Sim-CYCLE was run until it reached a stable state. The model parameters were adjusted so as the plant and the soil carbon mass of all vegetation types in the stable state is within ±10% of the values of Whittaker and Likens (1975) and Post et al. (1982), respectively.

Then, using 1951–2000 NCEP/NCAR data and 1751 CO2 data (obtained by the method of 3.1) repeatedly, Sim-CYCLE was run until it reached a stable state. The state is used as the initial condition of the simulation.

3.3. CO2 increase experiment

From the initial condition obtained by the method of 3.2, Sim-CYCLE was run for the period of 1751–2000. Also in this experiment, 1951–2000 NCEP/NCAR data and 1751 CO2 data (obtained by the method of 3.1) repeatedly, Sim-CYCLE was run until it reached a stable state. The state is used as the initial condition of the simulation.

4. Results

From the results of the simulation, correlations between the global and regional carbon fluxes and ENSO were verified for the period of 1980–2000. The land regions are same as TransCom 3
Fig. 2. Map of land regions and NINO.3 monitoring area (the red striped rectangle). The land regions are same as TransCom 3 experiment (Baker et al., 2006).

(Baker et al., 2006). As an index of ENSO, NINO.3 monitoring index of Japan Meteorological Agency (2009) was used. Figure 2 is a map of land regions and NINO.3 monitoring area.

4.1. Correlation between the global annual carbon flux and ENSO

As a result of the simulation, the total of terrestrial ecosystems was a carbon sink of 1.36 GtC yr$^{-1}$ for the period of 1980–2000. The anomaly of the global annual carbon flux calculated by Sim-CYCLE from which its long-term trend was eliminated is shown in Fig. 3 (the solid line). Figure 3 also shows NOAA/ESRL (National Oceanic & Atmospheric Administration/Earth System Research Laboratory) annual increase data of the global atmospheric CO$_2$ (Tans and Conway, 2009) (the broken line) and the annual mean NINO.3 monitoring index (the dashed line). The interannual variation of the global carbon flux was comparable in phase to the variation of the atmospheric CO$_2$ increase, though its amplitude was about a half.

Next, correlation coefficients between the annual mean NINO.3 index and the global carbon flux or the atmospheric CO$_2$ increase were calculated. The correlation coefficients are calculated for time lags of 0 yr and 1 yr. The results are shown in Table 1. Both the global carbon flux and the atmospheric CO$_2$ increase showed higher correlations with ENSO for a time lag of 1 yr than 0 yr.

Figure 4 shows the interannual variations of the anomalies of the carbon fluxes from three latitudinal zones (90N–20N, 20N–20S and 20S–90S). The variation of the flux anomaly of tropical zone (20N–20S) was dominant in amplitude, and its phase was similar to the variation of the global carbon flux anomaly (Fig. 3).

4.2. Correlations between the regional annual carbon fluxes and ENSO

Figure 5 shows interannual variations of the regional and global annual carbon flux anomalies calculated by Sim-CYCLE (the black solid line). Figure 5 also shows the interannual variations of the anomalies of the regional and global mean NCEP/NCAR re-analysis data. The plotted parameters are air temperature (the red line), total cloud cover (the green line), specific humidity (the blue line) and soil moisture (the dark yellow line).

In the same way as the global carbon flux, correlation coefficients between the annual mean NINO.3 index and the regional annual fluxes were calculated. The results are shown in Table 2. In Table 2, Temperate North America, Tropical America, Temperate South America, Southern Africa and Tropical...
Table 1. Correlation coefficients \(r\) between the annual mean NINO.3 monitoring index and the anomaly of the annual increase of the global atmospheric CO2 or the global annual carbon flux simulated by Sim-CYCLE (shown in Fig. 3).

<table>
<thead>
<tr>
<th></th>
<th>0 yr lag</th>
<th>1 yr lag</th>
</tr>
</thead>
<tbody>
<tr>
<td>Anomaly of the annual increase of the atmospheric CO2</td>
<td>0.12</td>
<td>0.41</td>
</tr>
<tr>
<td>Anomaly of the global annual carbon flux of Sim-CYCLE</td>
<td>0.28</td>
<td>0.57</td>
</tr>
</tbody>
</table>

Note: Significant level is 95% for \(|r| \geq 0.43\) and 90% for \(|r| \geq 0.37\).

Fig. 4. Anomalies of annual carbon fluxes from three latitudinal belts for the period of 1980–2000 simulated by Sim-CYCLE. The three belts are 90N–20N (the broken line), 20N–20S (the solid line) and 20S–90S (the dashed line). Long-term trend was eliminated. Flux from the biosphere to the atmosphere is positive.

Asia showed higher correlations for a time lag of 0 yr than 1 yr. Northern Africa showed high correlation both for time lags of 0 yr and 1 yr. It is noticeable that Tropical America and Tropical Asia showed negative high correlations for a time lag of 0 yr. As for the regions where the correlation coefficients are larger for a time lag of 1 yr than 0 yr, only positive correlation coefficients were calculated.

To understand the difference of the regional responses, correlations between the regional and global annual mean NCEP/NCAR parameters and the regional and global carbon fluxes were analysed. The results are shown in Table 3. In most of the regions, the regional carbon flux showed high correlation with the total cloud cover (Table 3). For the air temperature, several regional fluxes showed low correlations, but the global carbon flux showed higher correlation with the global mean air temperature than with the global mean total cloud cover. The correlation between the regional carbon flux and the specific humidity was high in most of the regions, but the correlation coefficient was a positive value. Like the case of the total cloud cover, the regional carbon flux showed high correlation with the regional mean soil moisture in most of the regions, but the correlation between the global carbon flux and the global mean soil moisture was not high. Correlations between the annual mean NINO.3 index and the regional and global annual mean NCEP/NCAR parameters were also analysed. The results were shown in Table 4.

5. Discussion

The interannual variation of the global annual carbon flux anomaly calculated by Sim-CYCLE was similar to that of the annual increase of the atmospheric CO2 in phase and showed high correlation with the annual mean NINO.3 monitoring index for a time lag of 1 yr (Fig. 3). These two variations are similar even in the first half of 1990s, in which the correlation between the increase of atmospheric CO2 and the NINO.3 index is low. In this period, the interannual variation of the global annual carbon flux of Sim-CYCLE was very similar to that of the global annual mean NCEP/NCAR temperature data of land grids (Fig. 5). The global annual mean temperature declined for the period of 1992–1993 in spite of El Niño. It is suggested that this decline was due to the eruption of Mt. Pinatubo (Hansen et al., 1996). Considering that responses of the fluxes from the biosphere and oceans to ENSO cycle differ (Jones et al., 2001), the results of the simulation may suggest that the global flux from the biosphere play a primary role in the interannual variation of the atmospheric CO2. However, as shown in Fig. 3, the amplitude of the interannual variation of the global annual flux calculated by Sim-CYCLE was much smaller than that of the annual increase of the atmospheric CO2. To explain this difference, sensitivity of the model to climate change must be inspected. In addition, the fluxes from oceans, intensification of the flux from the biosphere by water stress in drought and the flux by fire must be taken into account.

As for the regional carbon fluxes, some regions showed high correlations with NINO.3 index without time lag (Table 2). By analysing the regional monthly flux and the monthly mean ENSO index, Jones et al. (2001) and Patra et al. (2005) showed that the early reaction starts within a few months. However, this study focuses on the contribution of the regional fluxes to the interannual variation of the increase of the global atmospheric CO2, so the regional annual fluxes were analysed. In Table 2, it seems that the fluxes from the regions that are in tropical area or close to NINO.3 monitoring area tend to show high correlations with ENSO without time lag, though there are exceptions (e.g. Australia).
Fig. 5. Anomalies of regional and global annual carbon fluxes calculated by Sim-CYCLE (the black solid line). Long-term trend was eliminated. Flux from the biosphere to the atmosphere is positive. This figure also shows anomalies of annual mean air temperature (the red line), total cloud cover (the green line), specific humidity (the blue line) and soil moisture (the dark yellow line) of NCEP/NCAR Re-analysis data averaged for each region. The values of soil moisture are averages of 0–10 cm and 10–200 cm layers. The black dashed line is NINO.3 monitoring index.

Among the regional fluxes those have high correlations with NINO.3 index without time lag, fluxes from Tropical America and Tropical Asia showed negative high correlations. The correlation coefficient between the annual mean NINO.3 index and the global annual flux for a time lag of 0 yr is 0.28 (Table 1), however, without these two regions, the correlation coefficient rises to 0.61. This result suggests that the regional fluxes of negative correlations and other regional fluxes of positive correlations are cancelling out each other. As for the global flux without Tropical America and the global flux without Tropical Asia, correlation coefficients with NINO.3 index for a time lag of 0 yr are 0.48 and 0.39, respectively. Though the absolute value of the correlation coefficient of the flux from Tropical America is less than that of the flux from Tropical Asia (Table 2), it seems that the cancelling effect of the flux from Tropical America is greater than that of the flux from Tropical Asia.
Figure 4 shows that the amplitude of the flux anomaly from the tropical zone was dominant. This agrees with Kindermann et al. (1996) and Bousquet et al. (2000). However, its phase is similar to the phase of the global total carbon flux and showed a high correlation for a time lag of 1 yr. If the interannual variation of the flux from the tropical zone has both dominant amplitude and positive correlation without time lag, the global total carbon flux will have a high correlation with ENSO without time lag. The observation of the atmospheric CO2 increase that shows a high correlation with ENSO for a time lag of 1 yr suggests the existence of a cancelling effect of the regional fluxes from tropical area. Although the oceans play a role of cancelling (Feely et al., 1987; Rayner et al., 1999; Jones et al., 2001), Sim-CYCLE suggested the existence of another cancelling effect.

Table 3 shows that there are high correlations between the regional annual carbon fluxes and the total cloud cover. Especially in Tropical America and Tropical Asia, there are high negative correlations between NINO.3 index and the total cloud cover without time lag (Table 4), those are consistent with the high correlations between the total cloud cover and the regional fluxes. These negative correlations between ENSO and the...
Table 2. Correlation coefficients (r) between the annual mean NINO.3 monitoring index and the regional annual carbon fluxes calculated by Sim-CYCLE (shown in Fig. 5)

<table>
<thead>
<tr>
<th>Region</th>
<th>0 yr lag</th>
<th>1 yr lag</th>
</tr>
</thead>
<tbody>
<tr>
<td>Boreal North America</td>
<td>0.29</td>
<td>-0.10</td>
</tr>
<tr>
<td>Temperate North America</td>
<td>0.61</td>
<td>0.12</td>
</tr>
<tr>
<td>Tropical America</td>
<td>-0.58</td>
<td>0.21</td>
</tr>
<tr>
<td>Temperate South America</td>
<td>0.66</td>
<td>-0.08</td>
</tr>
<tr>
<td>Northern Africa</td>
<td>0.24</td>
<td>0.45</td>
</tr>
<tr>
<td>Southern Africa</td>
<td>0.46</td>
<td>0.11</td>
</tr>
<tr>
<td>Boreal Asia</td>
<td>0.13</td>
<td>0.46</td>
</tr>
<tr>
<td>Temperate Asia</td>
<td>-0.07</td>
<td>0.66</td>
</tr>
<tr>
<td>Tropical Asia</td>
<td>-0.71</td>
<td>0.51</td>
</tr>
<tr>
<td>Australia</td>
<td>-0.06</td>
<td>0.61</td>
</tr>
<tr>
<td>Europe</td>
<td>-0.03</td>
<td>0.39</td>
</tr>
</tbody>
</table>

Note: Significant level is 95% for |r| ≥ 0.43 and 90% for |r| ≥ 0.37.

regional fluxes do not agree with Patra et al. (2005). The version of Sim-CYCLE of this study may be too sensitive to the solar radiation. In Temperate North America and Temperate South America, there are high positive correlations between NINO.3 index and the total cloud cover without time lag (Table 4), and high positive correlations between the total cloud cover and the regional fluxes (Table 3). In other regions, no correlation between ENSO and the total cloud cover or other parameters of NCEP/NCAR Re-analysis data that are clearly consistent with the correlation between NINO.3 and the regional carbon flux was found. It is interesting that these consistent correlations were found in the regions where the correlation coefficients between ENSO and the regional fluxes are higher for a time lag of 0 yr than 1 yr. The positive correlation coefficients between the specific humidity and the regional carbon fluxes (Table 3) are thought to be due to high correlations between the specific humidity and the total cloud cover and the air temperature, and superiority of the positive effect of the total cloud cover and the air temperature to the negative effect of the specific humidity.

6. Conclusions
As a result of the simulation using the terrestrial ecosystem model, the global annual carbon flux from the biosphere showed an interannual variation, which is similar to that of the annual increase of the global atmospheric CO2 and a high correlation with ENSO for a time lag of 1 yr. This result suggests the domination of the flux from the biosphere in the interannual variation of the atmospheric CO2 increase. As for the regional carbon fluxes, the fluxes from tropical regions that showed negative correlations with ENSO for a time lag of 0 yr cancelled out the fluxes that showed positive correlations for a time lag of 0 yr. In the regions where the regional carbon fluxes showed negative correlations with ENSO, there were also high correlations between ENSO and the total cloud cover and between the total cloud cover and the regional carbon fluxes. This result suggests the important contribution of solar radiation to the 1-yr-lagged correlation between ENSO and the global carbon flux from the biosphere.

7. Acknowledgments
The original Sim-CYCLE model was presented by Akihiko Ito (NIES) and independently modified for this study. NOAA/ESRL atmospheric CO2 increase data, NCEP/NCAR Re-analysis climate data, Japan Meteorological Agency NINO.3 monitoring index data, and TransCom 3 land region map were downloaded.
Table 4. Correlation coefficients ($r$) between the annual mean NINO.3 monitoring index and the regional or global annual mean NCEP/NCAR Re-analysis data (shown in Fig. 5)

<table>
<thead>
<tr>
<th>Region</th>
<th>Air temperature</th>
<th>Total cloud cover</th>
<th>Specific humidity</th>
<th>Soil moisture</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>0 yr</td>
<td>1 yr</td>
<td>0 yr</td>
<td>1 yr</td>
</tr>
<tr>
<td>Boreal North America</td>
<td>−0.15</td>
<td>0.29</td>
<td>−0.07</td>
<td>−0.23</td>
</tr>
<tr>
<td>Temperate North America</td>
<td>0.01</td>
<td>0.23</td>
<td>0.58</td>
<td>0.03</td>
</tr>
<tr>
<td>Tropical America</td>
<td>0.64</td>
<td>0.53</td>
<td>−0.49</td>
<td>−0.05</td>
</tr>
<tr>
<td>Temperate South America</td>
<td>0.52</td>
<td>0.14</td>
<td>0.69</td>
<td>0.10</td>
</tr>
<tr>
<td>Northern Africa</td>
<td>0.08</td>
<td>0.18</td>
<td>0.26</td>
<td>0.05</td>
</tr>
<tr>
<td>Southern Africa</td>
<td>0.28</td>
<td>0.47</td>
<td>0.03</td>
<td>−0.14</td>
</tr>
<tr>
<td>Boreal Asia</td>
<td>−0.10</td>
<td>−0.14</td>
<td>0.02</td>
<td>0.03</td>
</tr>
<tr>
<td>Temperate Asia</td>
<td>−0.05</td>
<td>0.09</td>
<td>0.51</td>
<td>0.38</td>
</tr>
<tr>
<td>Tropical Asia</td>
<td>−0.09</td>
<td>0.22</td>
<td>−0.84</td>
<td>0.06</td>
</tr>
<tr>
<td>Australia</td>
<td>−0.03</td>
<td>0.31</td>
<td>0.13</td>
<td>0.37</td>
</tr>
<tr>
<td>Europe</td>
<td>−0.24</td>
<td>−0.16</td>
<td>0.18</td>
<td>0.20</td>
</tr>
<tr>
<td>Global</td>
<td>0.00</td>
<td>0.23</td>
<td>0.40</td>
<td>0.27</td>
</tr>
</tbody>
</table>

Note: Significant level is 95% for $|r| \geq 0.43$ and 90% for $|r| \geq 0.37$. from their homepages. As for CO$_2$ observation data, use was made of WMO/WDCGG CD-ROM. The figures were produced by GFD-DENNOU Library.

References


