The African contribution to the global climate-carbon cycle feedback of the 21st century

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Abstract. Future climate change will have impact on global and regional terrestrial carbon balances. The fate of African tropical forests over the 21st century has been investigated through global coupled climate carbon cycle model simulations. Under the SRES-A2 socio-economic CO2 emission scenario of the IPCC, and using the Institut Pierre Simon Laplace coupled ocean-terrestrial carbon cycle and climate model, IPSL-CM4-LOOP, we found that the warming over African ecosystems induces a reduction of net ecosystem productivity, making a 38% contribution to the global climate-carbon cycle positive feedback. Most of this contribution comes from African grasslands, followed by African savannahs, African tropical forest contributing little to the global climate-carbon feedback. However, the vulnerability of the African rainforest ecosystem is quite large. In contrast, the Amazon forest, despite its lower vulnerability, has a much larger overall contribution due to its 6 times larger extent.

1 Introduction

Coupled climate carbon cycle studies highlighted the vulnerability of the continental biosphere to human induced climate change (Cox et al., 2000; Dufresne et al., 2002). In particular tropical forest ecosystems may play a key role in future changes in global carbon balance. These ecosystems represent the largest reservoir of living biomass. The balance between carbon uptake through photosynthesis and release from decomposition of organic matter is sensitive to changes in climate regime, such as increased aridity. Climate variability and rising CO2 have also an impact on tropical forest biomass growth rates and mortality (Clark et al., 2003), and on CO2 emissions related to fire disturbances (Nepstad et al., 2004). There is a consensus among the coupled climate carbon cycle models to simulate a decrease of terrestrial carbon uptake in the tropics under future climate change (Friedlingstein et al., 2006, hereafter F06). This decreased tropical land uptake is a dominant contributor to the positive carbon cycle – climate feedbacks found in coupled models. Its causes have to be found in a combination of reduced photosynthesis due to a warming, generally combined with an increase in soil aridity, as well as an increase in soil oxidation, due to the warming. Some models also simulate a decrease of forest density, and a gradual replacement by savannah type ecosystems.

On shorter time scales, such as the interannual variability, recent studies also highlighted the dominant role of tropical ecosystem in the control of atmospheric CO2 growth rate (e.g. Rodenbeck et al., 2003; Peylin et al., 2005; Baker et al., 2006). The El Niño/Southern Oscillation (ENSO) climate variability induces large excursions in photosynthesis and/or decomposition and fires (Page et al., 2002) leading to anomalous CO2 release from tropical ecosystems during El Niño events, and conversely to anomalous uptake during the cooler and wetter La Niña episodes. However, the effect of drought in decreasing C uptake as simulated by all global models, was recently challenged by field measurements showing locally more C uptake by tropical forests during the dry periods (Saleska et al., 2007; Bonal et al., 2008).
However, so far, within the tropical forest, most of the focus has been given to the Amazon basin (Cox et al., 2004; Betts et al., 2004; Huntingford et al., 2004). For example, the fate of the Amazon forest has been largely investigated, as this ecosystem show a very strong positive feedback, i.e., carbon release, through its dieback in the Hadley Centre future climate-carbon simulations (Cox et al., 2000, 2004). Also the Large-Scale Biosphere – Atmosphere Experiment in Amazonia (LBA) program (http://lba.inpa.gov.br/lba) provided ecosystem and atmospheric data that were used by models in order to gain process understanding of the Amazon forest. Oppositely, very few studies focused on the African tropical forest and savannas biomes, their role in the global present carbon cycle and its vulnerability to future climate change.

2 Methodology

In a companion paper (Ciais et al., 2008), we used the ORCHIDEE global land ecosystem model (Krinner et al., 2005) forced by the observed evolution of the climate of the 20th century to investigate the present and historical African carbon balance. Here, we used the same land surface model, ORCHIDEE, but embedded within the Atmosphere Ocean General circulation model of IPSL in order to simulate the past and future evolution of the climate and carbon cycle. The IPSL-CM4-LOOP model couples the ocean-atmosphere general circulation model IPSL-CM4 (Dufresne et al., 2007) and the land and ocean carbon cycle models, ORCHIDEE (Krinner et al., 2005) and PISCES (Aumont and Bopp, 2006).

Following the coupled climate carbon cycle model intercomparison project (C4MIP) protocol described in F06, we performed two climate-carbon coupled simulations (UNC and COU) over the 1860–2100 period. In both experiments, CO2 emissions are prescribed from historical data for 1860–2000 (Marland et al., 2005; Houghton and Hackler, 2002) and from the SRES-A2 scenario for the 21st century (Nakićenović et al., 2000). The other greenhouse gas concentrations are set to pre-industrial values. Land cover was prescribed in all simulations. COU and UNC differ because in UNC, CO2 is treated as a non-radiatively active gas, so that the carbon cycle experiences no CO2-induced climate change. The difference between these two runs defines the climate carbon cycle feedback. We note that the UNC simulation does still simulate a moderate warming. This is due to the stomatal response to increasing atmospheric CO2, which leads to a shift from latent to sensible heat release from the surface, as reported before (e.g. Sellers et al., 1996).

3 Results

The global results of the IPSL-CM4-LOOP model have been described in previous papers (F06, Cadule et al., 2010). Here we will shortly summarize the main performance of the model. The COU simulation leads to a global surface warming of 3.7 K by 2100 and an atmospheric CO2 concentration of 807 ppm. The UNC simulation, with no climate change, leads to a CO2 concentration of 776 ppm. The difference, i.e., 31 ppm, is due to a decrease of both land and ocean carbon uptake under future climate conditions. The cumulated land sinks decreases by 80 PgC while the ocean sinks decrease by 65 PgC between 1901 and 2100.

When compared to the other C4MIP models, IPSL-CM4-LOOP has a lower than the average climate carbon cycle feedback. The CO2 amplification, 31 ppm, translates in a gain of 6%, where the C4MIP models show an average amplification of 85 ppm, that is, an average gain is 15%. This low gain cannot be explained by the climate sensitivity of the climate model as it is on the higher end of the C4MIP models. It is rather a direct result of the low sensitivity of the land carbon cycle to the climate change (called γL in F06). We defined γL as the difference in cumulated NEP in COU and UNC (corrected by the effect of atmospheric CO2 difference between the two simulations), normalized by the surface temperature change between COU and UNC (see F06 and Friedlingstein et al., 2003 for details). In other words, γL describes the sensitivity of the land carbon cycle to warming. In IPSL-CM4-LOOP, γL amounts to −20 PgC K−1, where the C4MIP models average is −80 PgC K−1. ORCHIDEE simulates a climate induced reduction of Net Ecosystem Productivity (NEP) in the tropics but largely balanced by an increased NEP in mid and high latitudes ecosystems. This positive NEP response is partly due to a cold bias in the UNC climate of the model in these regions.

Over Africa, the model simulates a regional warming by the end of the 21st century. This warming amounts to 3 K in the Congo basin, but reaches 6 K in South Africa and in northern Sahel (Fig. 1). The associated change in precipitation is an increase in precipitation around the equator (up to 300 mm yr−1; +20%) and a reduction in precipitation outside of the tropical belt (up to 100 mm yr−1) (Fig. 1). The regions with more warming also experience drying, while regions (Central Africa) which warm moderately experience wetter conditions in the future. These results are consistent with the ones obtained by the IPCC 4th Assessment Report climate models for the SRES-A2 scenario (see Fig. 11.2 of IPCC Working Group 1, 4th Assessment Report, Christensen et al., 2007). This climate change pattern is rather similar to the conditions of a permanent El Niño episode over the African continent (see Ciais et al., 2010).

When looking at the carbon cycle, the Gross Primary Productivity (GPP) of the COU coupled run increases everywhere across Africa, the largest increase being, as could be expected, in the already most productive regions of the African rainforest (Fig. 2). Overall GPP increases from 25 PgC yr−1 at pre-industrial up to 40 PgC yr−1 by 2100. Across the tropical rainforest ecosystem, this increase is a essentially driven by a GPP enhancement by atmospheric CO2.
Climate change has a small negative impact on the GPP of the African continent. Indeed, when looking at the difference between the coupled and the uncoupled runs, which allow to isolate the climate effect of the carbon cycle, we find a reduction of GPP over Africa (−1 PgC yr−1). This climate induced change in GPP varies widely within African ecosystems. The ORCHIDEE model uses a Plant Functional Type (PFT) approach to represent the different types of ecosystems. In order to further analyze the model results we separate the contribution from African forests (defined as the model grid cells where the tree PFTs fraction is above 0.66), African savannahs (tree PFTs fraction between 0.33 and 0.66) and African grasslands (tree PFTs fraction below 0.33). Doing so, we estimate that forests show a 5% increase in GPP, savannahs show no significant changes and grasslands show a 10% decrease in GPP (Fig. 3). Although we do not focus on cropland here, as there representation in the model is quite crude; they do also show a decrease of GPP.

Net carbon exchange (NEP) is reduced when climate change is taken into account (Fig. 3). The total African NEP cumulated over 1860–2100 amounts to 72 PgC in the COU run while it reaches 129 PgC in UNC (Fig. 4). The NEP reduction is much larger than the one of GPP because of a concomitant increase of total respiration under climate change. Autotrophic and especially heterotrophic respiration shows an important increase due to the regional warming (Fig. 3).

In term of reservoirs, total living biomass increases over the 1860–2100 period in Africa, largely thanks to the CO₂ fertilization of productivity. Total biomass increases from 60 to 115 PgC in the uncoupled simulation. Climate change tends to reduce that increase but only marginally (−3 PgC). The soil carbon response to climate change is much larger. The large increase in heterotrophic respiration leads to a soil carbon loss of 54 PgC (in simulation COU relative to simulation UNC).

A cumulated release of 57 PgC to the atmosphere would translate into an increase in concentration of about 12 ppmv, assuming an airborne fraction of 0.45. The African carbon response to climate change represents then about 38% of the global feedback of the IPSL-CM4-LOOP model (31 ppm). When looking at dominant natural biomes (defined here at tropical forest, savannahs and grasslands) the model simulate a carbon loss from all biomes (Fig. 4). The contribution from the African tropical forest alone is only 4 PgC, which translates into about 1 ppm, i.e. as very small contribution to the global feedback. African savannahs contribute to 18 PgC (4 ppm) and natural grasslands have the largest share, 34 PgC (7 ppm) (Table 1).

We note that the relative climate-carbon feedbacks are not simply additive as some regions, such as the high latitudes of the Northern Hemisphere have a negative contribution to the overall positive feedback.

Also the relative contributions of the African biomes should be weighted by their spatial coverage. Grasslands are by far the most dominant natural biome in Africa, followed by savannahs, and by forest. In order to better compare the
Fig. 3. Climate induced change in carbon fluxes and stocks (2090s) from IPSL-CM4-LOOP calculated as the difference between the coupled (COU) and the uncoupled (UNC) simulations over Africa. Top left panel is for Gross Primary Productivity, top right panel is for Net Primary productivity, middle left panel is for Heterotrophic Respiration, and middle right is for Net Ecosystem Productivity (all in gC m\(^{-2}\) yr\(^{-1}\)). Bottom left panel is for carbon stored in biomass, and bottom right is for carbon stored in litter and soil (both in gC m\(^{-2}\)). Note that last two quantities have to be multiplied by 10.

For the respective climate vulnerability of these three biomes, we define here an area specific land carbon vulnerability to climate (in gC m\(^{-2}\) K\(^{-1}\))\(\bar{\gamma}_L\), which is the standard \(\gamma_L\) (in PgC/K) divided by the area of the considered domain.

![Figure 4](link-to-image)

Fig. 4. Time evolution of cumulated change in NEP from IPSL-CM4-LOOP coupled (in black) and uncoupled (in red) simulations over (a) the Africa continent, (b) the African rainforest, (c) the African savannah, and (d) the African grasslands. Units are PgC.

Doing so, we estimate that African forest and savannahs have very similar \(\bar{\gamma}_L\), \(-950\) gC m\(^{-2}\) K\(^{-1}\) and \(-980\) gC m\(^{-2}\) K\(^{-1}\) respectively, while African grasslands are much more resilient, with a \(\bar{\gamma}_L\) of \(-290\) gC m\(^{-2}\) K\(^{-1}\) only. The large contribution of African grasslands to the overall positive feedback comes from the larger spatial coverage of that biome, despite its lower area specific vulnerability.

4 Discussion

Doing a similar calculation of the carbon loss from the Amazon forest shows that in the IPSL-CM4-LOOP model, the Amazon forest gives a much larger contribution than the African rain forest, 15 PgC reduction of 1860–2100 cumulated NEP in COU relative to UNC (Fig. 5). Again, this larger contribution comes from the area of the domain considered. The area of the Amazon basin is four times larger than the African rainforest basin. Also, in the Amazon, the carbon loss is largely driven by the reduction of GPP due to the increased aridity. A climatic pattern sometimes referred as a “perpetual El Niño”. In Africa, during such “perpetual El Niño” the precipitation pattern is drastically different with an enhancement of precipitation over tropical rain forest and a reduction over already more arid regions (north equatorial and southern African savannas). In the African rainforest, as explained above, the carbon loss only comes then from an enhanced respiration due to the warming trend.

We also estimate the area specific carbon vulnerability to climate, \(\bar{\gamma}_L\) for the Amazon forest. It amounts to \(-640\) gC m\(^{-2}\) K\(^{-1}\), a value actually slightly lower than the African forest and savannahs.
Table 1. Change in Net Ecosystem Productivity and associated area specific carbon vulnerability over Africa, African forest, savannah and grasslands and Amazon tropical forest.

<table>
<thead>
<tr>
<th></th>
<th>Cumulated NEP Coupled run (PgC)</th>
<th>Cumulated NEP Uncoupled run (PgC)</th>
<th>Climate induced NEP reduction (PgC)</th>
<th>$\gamma_l$ (gC m$^{-2}$ K$^{-1}$)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Africa</td>
<td>72</td>
<td>129</td>
<td>57</td>
<td>−600</td>
</tr>
<tr>
<td>African rainforest</td>
<td>20</td>
<td>24</td>
<td>4</td>
<td>−950</td>
</tr>
<tr>
<td>African savannah</td>
<td>25</td>
<td>43</td>
<td>18</td>
<td>−980</td>
</tr>
<tr>
<td>African grassland</td>
<td>28</td>
<td>62</td>
<td>34</td>
<td>−290</td>
</tr>
<tr>
<td>Amazon rainforest</td>
<td>41</td>
<td>56</td>
<td>15</td>
<td>−640</td>
</tr>
</tbody>
</table>

Fig. 5. As Fig. 4 for (a) the African rainforest, and (b) the Amazon basin. Units are PgC.

Fig. 6. Climate change (2090s relative to 1860s) simulated by the HadCM3-LC climate model. As in Fig. 1, left panel is for surface temperature change (°C), right panel is for precipitation changes (mm yr$^{-1}$).

Stronger than IPSL precipitation reduction over the Amazon rainforest, associated with a twice as larger warming (Fig. 6). For African rainforest basin both models simulate similar changes in temperature and precipitation. Land response would then be dramatically different for the Amazon basin, while being comparable for the African rainforest.

Also, the modeling framework used here has some important limitations in term of process representation. Nitrogen cycle and its impact on terrestrial carbon cycle were not accounted for here. A recent study with a nitrogen enabled version of ORCHIDEE (Zaehle et al., 2010) shows that nitrogen limitation reduces CO$_2$ fertilization but increases photosynthesis under global warming, although these effects are mainly active in the mid and high latitudes, and would not dramatically affect our results over Africa. Natural fires were accounted here, but with a relatively simple global fire model. More elaborate fire models are being developed now and the response of fires to climate variability and trends (e.g. VanderWerf et al., 2004; Golding and Betts, 2008) imply that fires are certainly important players in the future carbon budget of the tropics.
5 Conclusions

Tropical ecosystems will be put at risk over the coming century, due to climate change, but also due to the direct anthropogenic pressure, through deforestation (not accounted for in this study). In the coupled carbon climate simulations that we analyzed, the African continent would gain carbon by the end of the 21st century (relative to preindustrial), thanks to the fertilization effect from the atmospheric CO$_2$ increase. However, the associated climate change tends to reduce this African storage by 25%, from 129 PgC to 72 PgC. The African continent is hence an active player of the climate-carbon cycle positive feedback, with a contribution of about 38% of the global feedback of the IPSL-CM4-LOOP model.

Although usually pointed as the major carbon source, South America tropical forests have a lower contribution than the African savannahs or African grasslands. However, when looking at the tropical rain forest ecosystem alone, although their specific vulnerability is slightly larger for African forests, the overall contribution from Africa is much lower than the one from South America because of its much smaller area. Altogether, more than 50% of the feedback comes from Africa and South America in the IPSL-CM4-LOOP simulations. Finally, we note that there is still a large uncertainty in the climate response of African vs. Amazon rainforests due to the uncertainty of climate projections, especially the hydrological cycle at the regional scale.

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