

# Deforestation in Amazonia impacts riverine carbon dynamics

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21

## 22 Abstract

23 Fluxes of organic and inorganic carbon within the Amazon basin are considerably controlled  
24 by annual flooding, which triggers the export of terrigenous organic material to the river and  
25 ultimately to the Atlantic Ocean. The amount of carbon imported to the river and the further  
26 conversion, transport and export of it depend on temperature, atmospheric CO<sub>2</sub>, terrestrial  
27 productivity and carbon storage, as well as discharge. Both, terrestrial productivity and  
28 discharge, are influenced by climate and land use change. The coupled LPJmL and RivCM  
29 model system (Langerwisch et al., 2015) has been applied to assess the combined impacts of  
30 climate and land use change on the Amazon riverine carbon dynamics. Vegetation dynamics  
31 (in LPJmL) as well as export and conversion of terrigenous carbon to and within the river  
32 (RivCM) are included. The model system has been applied for the years 1901 to 2099 under  
33 two deforestation scenarios and with climate forcing of three SRES emission scenarios, each  
34 for five climate models. We find that high deforestation (BAU scenario) will strongly  
35 decrease (locally by up to 90%) riverine particulate and dissolved organic carbon amount until  
36 the end of the current century. At the same time, increase in discharge leaves net carbon  
37 transport during the first decades of the century roughly unchanged only if a sufficient area is  
38 still forested. After 2050 the amount of transported carbon will decrease drastically. In  
39 contrast to that, increased temperature and atmospheric CO<sub>2</sub> concentration determine the  
40 amount of riverine inorganic carbon stored in the Amazon basin. Higher atmospheric CO<sub>2</sub>  
41 concentrations increase riverine inorganic carbon amount by up to 20% (SRES A2). The  
42 changes in riverine carbon fluxes have direct effects on carbon export, either to the  
43 atmosphere via outgassing, or to the Atlantic Ocean via discharge. The outgassed carbon will  
44 increase slightly in the Amazon basin, but can be regionally reduced by up to 60% due to  
45 deforestation. The discharge of organic carbon to the ocean will be reduced by about 40%  
46 under the most severe deforestation and climate change scenario. These changes would have  
47 local and regional consequences on the carbon balance and habitat characteristics in the  
48 Amazon basin itself but also in the adjacent Atlantic Ocean.

49

## 50 1 Introduction

51 The Amazon basin, defined as the drainage area of the Amazon River, covers approximately  
52 six million square kilometres, and more than 70% of it is still covered with intact rainforest  
53 (Nobre, 2014). The amount of carbon in biomass in Amazonian rainforest is estimated to be  
54  $93 \pm 23 \times 10^{15}$  g C (Malhi et al., 2006). This biomass is stored in a wide range of diverse  
55 habitats, including tropical rainforest and savannahs, as well as numerous aquatic habitats,  
56 like lakes and wetlands (Goulding et al., 2003; Eva et al., 2004; Keller et al., 2009; Junk,  
57 1997). The large diversity in habitats, partly already founded in the geologic formation of  
58 Amazonia, leads to a high diversity of animal and plant species (Hoorn et al., 2010), making  
59 the Amazon rainforest one of Earth's greatest collections of biodiversity.

The Amazon River, which floods annually large parts of the forest, plays an important role in supporting the diversity of Amazonian ecosystems. The flooding is most decisive for the coupling of terrestrial and aquatic processes by transporting organic material from the terrestrial ecosystems to the river (Hedges et al., 2000). The input of terrigenous organic material (Melack and Forsberg, 2001; Waterloo et al., 2006), acts, for instance, as fertilizer and food source (Anderson et al., 2011; Horn et al., 2011), and is a modifier of habitats and interacting local carbon cycles (Hedges et al., 2000; Irmiler, 1982; Johnson et al., 2006; McClain and Elsenbeer, 2001). Across the Amazon basin, the outgassing carbon from the river to the atmosphere and export of it to the ocean are the two most important processes that have to be included, when assessing the effects on riverine carbon dynamics under climate and land use change. Approximately  $470 \times 10^{12}$  g C yr<sup>-1</sup> is exported to the atmosphere as CO<sub>2</sub> (Richey et al., 2002), in comparison with about  $32.7 \times 10^{12}$  g C yr<sup>-1</sup> of total organic carbon (TOC) is exported to the Atlantic Ocean (Moreira-Turcq et al., 2003). It is estimated that the large scale outgassing of carbon from the Amazon River plays an important role in assessing the future carbon balance of the Amazon basin, integrating riverine as well as terrestrial processes.

Deforestation continues to be the largest threat to Amazonia. The transformation of tropical rainforest to cropland and pasture impacts ecosystem stability profoundly due to altered climate regulation and species richness (Foley et al., 2007; Lawrence and Vandecar, 2014; Malhi et al., 2008; Spracklen et al., 2012). Until the year 2012 approximately 20% of the original forest of the Brazilian part of the Amazon basin has been deforested, corresponding to an area of about 750,000 km<sup>2</sup> (Godar et al., 2014; INPE, 2013). This deforestation was mainly driven by the land expansion for soybean and cattle production and the expansion of the road network (Malhi et al., 2008; Soares-Filho et al., 2006). Governmental and conservation efforts have helped to decrease recent deforestation rates (Nepstad et al., 2014) but economic instability might reverse this trend (Aguiar et al., 2016; Fearnside, 2015). Deforestation also alters the soil stability and increases erosion (Yang et al., 2003). Together with climate change effects and forest burning, land cover change is predicted to release carbon at rates of  $0.5\text{--}1.0 \times 10^{15}$  g C yr<sup>-1</sup> from this area (Potter et al., 2009). Furthermore, the effects of deforestation on terrestrial carbon storage and fluxes persist several decades after logging because the forest needs about 25 years to recover approximately 70% of its original biomass, and at least another 50 years for the remaining 30% after abandonment of agriculture (Houghton et al., 2000; Poorter et al., 2016).

Deforestation immediately reduces the terrestrial organic carbon pools, which fuel riverine respiration (Mayorga et al., 2005), while increasing the velocity and amount of runoff, as well as the discharge (Foley et al., 2002; Costa et al., 2003). Additionally, climate change alters precipitation which then affects inundation patterns (Langerwisch et al., 2013), such as temporal shifts in high and low water months and changes of inundated area. The combined effects of deforestation and climate change have the potential to tremendously alter the exported terrigenous carbon fluxes, the amount of carbon emitted to the atmosphere and exported the ocean. The local export of terrestrial organic carbon to the river changes the nutrient supply and therefore alters the habitat for riverine plants and animals, (Hamilton, 2010).

The aim of our study is to elaborate on these combined effects of climate change and deforestation on the riverine carbon fluxes, on the export of organic material into the Atlantic Ocean and on the outgassing of riverine carbon to the atmosphere. By considering the interactions between riverine and terrestrial carbon processes a complete view on future changes in the regional and basin-wide carbon balance can be achieved for the Amazon basin. We mean in this study the effects of replacing tropical forest with soy bean fields and pasture, i.e. deforestation, and the effects of changed carbon fluxes and pool sizes on crop and pasture land, i.e. changed land use.

To address these issues basin-wide data are needed, which not only describe the current situation but also assess future changes, expanding our knowledge obtained from on-site measurements. To partly overcome these limitations we make use of the well-established dynamic global vegetation model LPJmL together with the riverine carbon model RivCM. While LPJmL (Bondeau et al., 2007; Gerten et al., 2004; Rost et al., 2008; Sitch et al., 2003) provides plausible estimates for the carbon and water pools and fluxes within the coupled soil-vegetation system, RivCM (Langerwisch et al., 2015) focuses on the export, conversion and transport of terrestrial fixed carbon in the river and to the atmosphere and ocean. In Langerwisch et al. (2015) the solely effects of climate change have been estimated. The results of the mentioned study show that climate change causes a doubling of riverine organic carbon in the Southern and Western basin while reducing it by 20% in the eastern and northern parts towards the end of this century. In contrast, the amount of riverine inorganic carbon shows a 2- to 3-fold increase in the entire basin, independent of the climate change scenario (SRES). The export of carbon to the atmosphere increases on average by about 30%. The amount of organic carbon exported to the Atlantic Ocean depends on the SRES scenario and is projected to either decrease by about 8.9% (SRES A1B) or increase by about 9.1% (SRES A2). The current study, which is an extension of Langerwisch et al. (2015) goes one step further and investigates the combined effects of climate change and deforestation on the riverine carbon dynamics. The coupled model LPJmL-RivCM was forced by several climate change and deforestation scenarios that cover a wide range of uncertainties. We estimated temporal and spatial changes in three riverine carbon pools as well as changes in carbon emissions to the atmosphere and carbon export the ocean.

## 2 Methods

To assess the impacts of climate change and deforestation on riverine carbon pools and fluxes in the Amazonian watershed we applied the model system of LPJmL and RivCM. RivCM is a grid-based model that assesses the transport and export of carbon at monthly time steps and is driven climate data and terrestrial carbon pools (Langerwisch et al., 2015). Climate inputs are taken from different global climate model simulations driven by three SRES scenarios (A1B, A2 and B1; Nakićenović et al., 2000). Terrestrial carbon inputs are calculated by the process-based dynamic global vegetation and hydrology model LPJmL (Bondeau et al., 2007; Gerten et al., 2004; Rost et al., 2008; Sitch et al., 2003). To estimate soil and vegetation carbon, LPJmL uses the above mentioned climate data and a set of deforestation scenarios from a regional projections by SimAmazonia (Soares-Filho et al., 2006). An overview of the interconnection between the two models and the scenarios is given in

## Figure 1.

### 2.1 Model descriptions

#### 2.1.1 LPJmL – a dynamic global vegetation and hydrology model

The process-based global vegetation and hydrology model LPJmL (Bondeau et al., 2007; Gerten et al., 2004; Rost et al., 2008; Sitch et al., 2003) simulates the dynamics of potential natural vegetation and thus carbon pools for vegetation, litter and soil and corresponding water fluxes, in daily time steps and on a spatial resolution of  $0.5 \times 0.5$  degree (lat/lon). The main processes included are photosynthesis (modelled according to Collatz et al., 1992; Farquhar et al., 1980), auto- and heterotrophic respiration, establishment, mortality, and phenology. For calculating these main processes LPJmL uses climate data (temperature, precipitation, and cloud cover), atmospheric CO<sub>2</sub> concentration, and soil type as input. The simulated water fluxes include evaporation, soil moisture, snowmelt, runoff, discharge, interception, and transpiration, which are directly linked to abiotic and biotic properties. In each grid cell LPJmL calculates the performance of nine plant functional types, which represent an assortment of species classified as being functionally similar. In the Amazon basin primarily three of these types are present, namely tropical evergreen and deciduous trees and C4 grasses. In addition to the potential natural vegetation LPJmL can simulate the dynamics of 16 user-defined crops and pasture on area that is not covered by natural vegetation. In analogy to natural vegetation, LPJmL evaluates carbon storage in vegetation, litter and soil as well as water fluxes for these areas. On areas, which are converted to crops and pasture, the vegetation carbon stored in natural vegetation (carbon in living above- and belowground biomass) is removed from the terrestrial domain and added to the litter pool. Due to deforestation, a large amount of carbon is removed from the living biomass, i.e. after some years, the pool size of potential carbon that can be washed out to the river is decreasing dramatically. On the deforested areas growth and harvest of soy bean and managed grasslands is simulated. We distinguished these two types of land use, because soy bean farming and pasture leave different amounts of litter carbon on site. In LPJmL, during soy harvest a maximum of 30% of the aboveground soy biomass, representing the beans, is removed as harvest every year. The remaining aboveground biomass as well as all belowground biomass is left on site and enters the litter pool. Managed grasslands are harvested regularly as well, but always 50% of the aboveground biomass is removed. The remaining aboveground biomass and the total belowground biomass enter the litter pool. Once a stand is harvested the remaining above- and belowground biomass is added to the litter pool. The soil pool remains unchanged. Only after litter decomposition this carbon enters the soil carbon pool. Therefore, after deforestation the amount of carbon washed out from managed land to the river, and entering the riverine carbon system, is much less in size compared to litter exported to the river from undisturbed forests. Changes of soil characteristics and soil carbon pools due to erosion, which is a common consequence of deforestation (Yang et al., 2003) is not included in the model. In summary, the terrestrial ecosystem is losing carbon due to deforestation followed by harvest. Therefore, the riverine ecosystem is receiving less carbon due to reduced terrestrial carbon input after forest was converted to managed land.

LPJmL has been shown to reproduce current patterns of biomass production (Cramer et al., 2001; Sitch et al., 2003), carbon emission through fire (Thonicke et al., 2010), also including managed land (Bondeau et al., 2007; Fader et al., 2010; Rost et al., 2008), and water dynamics (Biemans et al., 2009; Gerten et al., 2004, 2008; Gordon et al., 2004; Wagner et al., 2003). The simulated patterns in water fluxes, like evapotranspiration, runoff and soil moisture, are comparable to stand-alone global hydrological models (Biemans et al., 2009; Gerten et al., 2004; Wagner et al., 2003).

### 2.1.2 RivCM – a riverine carbon model

RivCM is a process-based model that calculates four major ecological processes related to the carbon budget of the Amazon River (Figure 1B). These processes include (1) mobilization, (2) decomposition and (3) respiration within the river, and (4) outgassing of CO<sub>2</sub> to the atmosphere (Langerwisch et al., 2015). During mobilization parts of terrigenous litter and soil carbon, as it is provided by LPJmL, is imported to the river, depending on inundated area. The further processing of the terrigenous carbon in the river happens during its decomposition, which represents the manual breakup, and its respiration, representing the biochemical breakup. Finally the CO<sub>2</sub> that is produced during respiration can outgas if the saturation concentration is exceeded (Langerwisch et al., 2015). These four processes directly control the most relevant riverine carbon pools, namely particulate organic carbon (POC), dissolved organic carbon (DOC), and inorganic carbon (IC), as well as outgassed atmospheric carbon (representing CO<sub>2</sub>), and exported riverine carbon to the ocean (either as POC, DOC, or IC).

The model is coupled to LPJmL by using the calculated monthly litter and soil carbon and water amounts as inputs. It operates at the spatial resolution of  $0.5 \times 0.5$  degree (lat/lon) and on monthly time steps. The ability of the coupled model LPJmL-RivCM to reproduce current conditions in riverine carbon concentration and export to either the atmosphere or the ocean has been shown and discussed by Langerwisch et al. (2015). A validation of the carbon pools and fluxes with observed data shows that RivCM produces results that are within the range of observed concentrations of both organic and inorganic carbon pools. Model results strongly underestimate the amount of outgassed carbon while the carbon discharged to the ocean is overestimated. There are still large uncertainties in the process understanding of riverine carbon processes that translates to uncertainty in the parameter estimation. Therefore, a respective model like we have applied here can currently only reproduce broad estimations of exported CO<sub>2</sub> (outgassing) and exported organic carbon (discharge). In general the model reaction to climate change alone and in combination with deforestation and land-use change is as expected (e.g. reduction of organic carbon due to deforestation, increase of inorganic carbon due to climate change). Therefore, we think it is reasonable to use our model to estimate changes in process relations and general trends. Further data-model comparison and improved parameterization are still required to allow assessing the simulated absolute numbers model. Despite these shortcomings we make use of the coupled model system of LPJmL and RivCM to assess the combined impacts of climate change and deforestation.



## 2.2 Model simulation

All transient LPJmL runs were preceded by a 1000-year spin-up during which the pre-industrial CO<sub>2</sub> level of 280 ppm and the climate of the years 1901-1930 have been repeated to obtain equilibria for vegetation, carbon, and water pools. All transient runs of the coupled model LPJmL-RivCM have been preceded by a 90-years-spinup during which the climate and CO<sub>2</sub> levels of 1901-1930 have been repeated to obtain equilibria for riverine carbon pools.

LPJmL-RivCM was run on a  $0.5^\circ \times 0.5^\circ$  degree (lat/lon) spatial resolution for the years 1901 to 2099. For the estimation of the impact of projected climate change (CC) and deforestation (Defor), simulations have been conducted driven by five General Circulation Models (GCMs), each calculated for three SRES emission scenarios, and three LUC scenarios.

### 2.2.1 Climate change and deforestation data sets

To assess the effect of future climate change, projections of five GCMs (see also Jupp et al., 2010; Randall et al., 2007), using three SRES scenarios (A1B, A2, B1) (Nakićenović et al., 2000) have been applied (Figure 1A). The GCMs, namely MIUB-ECHO-G, MPI-ECHAM5, MRI-CGCM2.3.2a, NCAR-CCSM3.0, UKMO-HadCM3, cover a wide range in terms of temperature and precipitation and have therefore been chosen to account for uncertainty in climate projections. The emission scenario SRES A1B describes a development of very rapid economic growth with convergence among regions, and a balanced future energy source between fossil and non-fossil. SRES A2 describes a development of a very heterogeneous world with slow economic growth. And SRES B1 describes a development of converging world similar to A1B but with more emphasis on service and information economy.

To estimate the additional effects of deforestation on riverine carbon pools and fluxes three land use scenarios were applied: two scenarios directly relate to different intensity of deforestation, and one represents a reference scenario with complete coverage by natural vegetation (NatVeg scenario, hereafter). The two deforestation scenarios are based on the SimAmazonia projections (Soares-Filho et al., 2006, see also **Figure 2**). The authors estimate the development of deforestation in the Amazon basin until 2050 based on historical trends and projected developments. In the business-as-usual scenario (BAU) they assume that recent deforestation trends continue, the number of paved highways increases, and new protected areas are not established. In contrast, deforestation is more efficiently controlled in the governance scenario (GOV). For this scenario the authors assume that the Brazilian environmental legislation is implemented across the Amazon basin and the size of the area under the *Protected Areas Program* increases. The SimAmazonia scenarios cover the years from 2001 to 2050. **After 2050 the fraction of deforested area is kept constant. From 2051 until the end of the century the only driver of change is the continuing climate change. This approach enables us to estimate the consequences of combined dynamics of deforestation and climate change until 2050 and the effects of intensified climate change after 2050, when deforestation is halted at its maximum.** Deforestation rates preceding the scenarios (before 2001) were derived from extrapolating the data into the past. **LPJmL requires historic land-cover information to correctly capture transient carbon dynamics. The model starts to simulate vegetation dynamics from bare ground and can't be initialized with a land-cover map of a**

particulate year. It was therefore necessary to develop an approach which produced consistent land-cover information for the (undisturbed) past and the deforestation scenarios. For that, the mean annual rate of deforestation was calculated for the reference period of 2001 to 2005 (Eq. (1)) and this rate was applied to calculate the fraction of deforested area  $F_t$  for the years 1901 to 2000 for each cell (Eq. (2)).

$$r = \left( \sum_{t=2001}^{2005} \frac{F_t}{F_{t+1}} \right) \times \frac{1}{2006 - 2001} \quad (1)$$

$$F_t = F_{2001} \times r^{2001-t} \quad (2)$$

To evaluated spatial differences in the basin we defined three sub-regions (see Table 1). Three regions were selected for further detailed analysis and differ in projected changes in inundation patterns and in deforestation intensity. R1 is located in the Western basin with projected increase in inundation length and inundated area (Langerwisch et al., 2013) combined with low land use intensity. R2 is a region covering the Amazon main stem with intermediate changes in inundation (Langerwisch et al., 2013) and intermediate land use intensity. And R3 is a region with projected decrease in duration of inundation and inundated area (Langerwisch et al., 2013) combined with high land use intensity. In the deforestation scenarios we assume that on 15% of the deforested area soy bean is grown and 85% of the area is used as pasture for beef production (Costa et al., 2007).

## 2.3 Analysis of simulation results

The separate effect of deforestation ( $E_{Defor}$ ) is estimated by calculating the differences between future carbon amounts (2070-2099) produced in the deforestation scenarios (GOV or BAU) and future carbon amounts produced in the potential natural vegetation scenario (NatVeg), where no deforestation is assumed. The combined effect of climate change and deforestation ( $E_{CCDefor}$ ) is estimated by calculating the differences between future carbon amounts produced in the deforestation scenarios and reference carbon amounts (1971-2000) produced in the NatVeg scenario. We analysed all four riverine carbon pools (riverine particulate organic carbon (POC), dissolved organic carbon (DOC), riverine inorganic carbon (IC) and outgassed carbon). The relative changes in POC and DOC show similar patterns (see Figure S1), therefore exemplary POC is shown and discussed in detail.

### 2.3.1 Evaluation of potential future changes

Spatial effects of the two deforestation scenarios (GOV and BAU) on the different riverine carbon pools and fluxes have been estimated by calculating the common logarithm ( $\log_{10}$ ) of the ratio of mean future (2070-2099) carbon amounts of the deforestation scenarios and mean future carbon amounts of the NatVeg scenario ( $E_{Defor}$ , Eq. (3)) for each simulation run.



$$E_{Defor} = \log_{10} \frac{\sum_{t=2070}^{2099} C_{Defor_t}}{\sum_{t=2070}^{2099} C_{NatVeg_t}} \quad (3)$$

298 To estimate changes caused by the combination of climate change and deforestation  $E_{CCDefor}$   
 299 compares future carbon pools in the deforestation scenarios to carbon pools during the  
 300 reference period (1971-2000) in the NatVeg scenario (Eq. (4)).

$$E_{CCDefor} = \log_{10} \frac{\sum_{t1=2070}^{2099} C_{Defor_{t1}}}{\sum_{t2=1971}^{2000} C_{NatVeg_{t2}}} \quad (4)$$

301 Each simulation run combines deforestation and emission scenarios and aggregates the  
 302 outputs for all five climate model inputs used. To identify areas where the differences  
 303 between values in the reference period and future values are significant (p-value <0.05), the  
 304 Wilcoxon Rank Sum Test for not-normally distributed datasets (Bauer, 1972) has been  
 305 applied for each cell.

306 Additionally to the spatial assessment, time series were deduced based on mean values over  
 307 the entire basin and each of the three exemplary regions R1, R2 and R3. These means of the  
 308 carbon pools were calculated for every year during the simulation period. Changes have been  
 309 expressed as the five-year-running-mean of the quotient of annual future carbon amounts in  
 310 the deforestation and in the NatVeg scenarios. These analyses have been conducted both for  
 311 the whole Amazon basin and for three selected sub-regions.

### 312 2.3.2 Estimating the dominant driver for changes

313 We estimated which factor is causing the observed changes the most. To estimate the  
 314 contribution of either climate change ( $D_{CC}$ , Eq. (5)) or deforestation ( $D_{Defor}$ , Eq. (6)),  
 315 reference carbon amounts of the NatVeg scenario have been compared to future amounts of  
 316 the NatVeg scenario ( $D_{CC}$ ), and future carbon amounts of the NatVeg scenario have been  
 317 compared to future amounts of the deforestation scenarios ( $D_{Defor}$ ).

$$D_{CC} = \left| \log_{10} \frac{\sum_{t1=2070}^{2099} C_{NatVeg_{t1}}}{\sum_{t2=1971}^{2000} C_{NatVeg_{t2}}} \right| \quad (5)$$

$$D_{Defor} = |E_{Defor}| \quad (6)$$

318 We define a cell as dominated by climate change effects, if  $D_{CC} > D_{Defor}$  and dominated by  
 319 deforestation effects if  $D_{CC} < D_{Defor}$ . The impact values  $D_{CC}$  and  $D_{Defor}$  ( $median_{POC} = 0.9695$ ,  
 320  $median_{IC} = 1.0106$ , and  $median_{outgassedC} = 0.9982$ ) have been rounded to the second decimal  
 321 place. If both values are equal, the two effects balance each other.

322

## 3 Results

### 3.1 Changes caused by deforestation

Deforestation decreases riverine particulate and dissolved organic carbon (POC and DOC). When continuing high deforestation rates as projected under the BAU deforestation scenario, the decrease in POC is more intense than under GOV deforestation rates (Figure 3A and Figure 3B; for DOC see Figures. S1A and S1B). In some highly deforested sites in the South-East of the basin the amount of POC is only 10% of the amount under no deforestation (indicated by  $E_{Defor}$ ). This pattern is robust between the model realizations with a high agreement of the results amongst the five climate models. In the deforestation scenarios the changes in future POC are drastic, even though the difference between the three emission scenarios A1B, A2, and B1 are very small. However, in some regions within the Amazon basin POC increases (up to 3fold), especially in mountain regions (e.g. Andes and Guiana Shield). Although POC and DOC respond similar in relative terms (see Figure S1), the absolute amounts are approximately twice as high for DOC compared to POC (Table 2). The mean basin-wide loss in POC ranges between  $0.13 \times 10^{12} \text{ g yr}^{-1}$  (A2) and  $0.24 \times 10^{12} \text{ g yr}^{-1}$  (A1B) in the GOV scenario, and between  $0.37 \times 10^{12} \text{ g yr}^{-1}$  (A2) and  $0.48 \times 10^{12} \text{ g yr}^{-1}$  (A1B) in the BAU scenario. The SRES A2 scenario causes the largest changes in POC, further increasing the loss caused by deforestation.

Changes in outgassed riverine carbon caused by deforestation (Figure 3C and Figure 3D) show a similar pattern as the changes in POC, with an even clearer effect of deforestation on a larger area. In both scenarios deforestation decreases outgassed carbon to up to one tenth compared to the amount produced under the NatVeg scenario. The agreement between the five climate models is even larger than in POC. In contrast to the overall pattern, some areas in the Andes and the Guiana Shield show an increase in outgassed carbon of up to a factor of 30, but these areas are an exception. Like in POC the differences between the SRES scenarios are only minor. For the absolute values see Table 2.

For riverine inorganic carbon (IC) deforestation caused significant changes ( $E_{Defor}$ , p-value  $< 0.05$ ) only in small areas (Figure 3E and Figure 3F). In these regions, in the very South of the basin and in single spots in the North, i.e. in the headwaters of the watershed, IC increases by a factor of up to 1.2. Besides these areas of increase, a slight decrease of about 5% is simulated for the region along the main stem of the Amazon River, downstream of Manaus and along the Rio Madeira and the Rio Tapajós. In contrast to POC, the spatial pattern of change in IC does not obviously follow the deforestation patterns. Therefore, the differences between the two deforestation scenarios GOV and BAU scenarios are minor. Whereas POC, DOC, and outgassed carbon show a clear decrease due to deforestation, IC shows a nearly neutral response with maximal mean basin-wide gains (for absolute values see Table 2).

### 3.2 Changes caused by a combination of deforestation and climate change

Climate change and deforestation together will lead to large overall changes in the amount of riverine and exported carbon. Riverine POC and DOC amounts will decrease by about 19.8% and 22.2%, respectively, and exported organic carbon will decrease by about 38.1% (Figure

5). In contrast riverine IC will increase by about 100%, combined with a slight increase of outgassed carbon by about 2.7% (Figure 5). In detail, the basin-wide changes in the amount of POC (Figure 5A-B and Figure S2) caused by deforestation and climate change range between a 2.5-fold increase and a decrease to one tenth. The increase is mainly caused by climate change (indicated as blue area in the inset in Figure 5), whereas the decrease is mainly caused by deforestation (red area in inset). The differences mainly induced by deforestation are larger in the BAU compared to the GOV scenario. In contrast, the differences caused by climate change show no large differences between the two deforestation scenarios. The differences between the emission scenarios are minor (see also Table 2). In some areas the dominance of forcing shifts from climate change dominance ( $D_{CC}$ ) for the GOV scenario (green cell border) to deforestation dominance ( $D_{Defor}$ ) for the BAU scenario (red cell border) due to the higher land use intensity as a result of deforestation (see also Table 3). While in the GOV scenario 20% of all cells are dominated by deforestation impacts, this value increases for the BAU scenario to 30%. During the first decades (2000-2030) basin-wide POC is partly larger in the deforestation scenarios than in the NatVeg scenario by up to 2% in 2000 and about 1% in 2020 (Figure 6A). All climate models show reduced POC amounts in the deforestation scenarios compared to the NatVeg scenario after 2040. The POC amount in the GOV deforestation scenario decreases gradually until the decrease levels off in the late 2060s, i.e. ten years after the constant deforestation area is kept constant. In the BAU scenario, POC decreases strongly in the 2040 to 2060s leading to a loss of about 25% compared to 10% in the GOV scenario. In addition to Figure 6, which shows the temporal development under deforestation only, we provide Figure S2, which shows the developments taking the combination of deforestation and climate change into account.

The three sub-regions R1 to R3 show different patterns (Figure 6A). While in region R1 the difference in the POC amounts between the GOV and the BAU scenario is only small, reflecting the low deforestation in this region, the differences between the two deforestation scenarios are more explicit in regions R2 and especially in R3 (with the largest area deforested), where in addition model uncertainty is low. Starting in the 2050s, the variation between different emission scenarios and climate models increases. Alike the results of the impact of deforestation alone POC and DOC show a similar pattern (see also Table 2).

The changes in outgassed carbon (Figure 5C-D and Figure 6B) are in the same range as changes in POC. Climate change increases outgassed carbon by about 20%, especially in the North-Western basin (Figure 5C-D). The deforestation induces a decrease on outgassed carbon to one tenth in areas with high fraction of deforested area, i.e. in the Eastern and South-Eastern basin. Again, the differences in effects are much larger between the two deforestation scenarios (GOV vs. BAU) than between the different emission scenarios (see also Table 2). After 2050 the rate of deforestation determines the differences in the amount of outgassed carbon (Figure 6B) as well. The outgassed carbon directly depends on the available POC, therefore the time series of both, POC and IC widely match. Under the GOV scenario the basin-wide loss of outgassed carbon is about 16% towards the end of the century. The results of the BAU scenario show an average loss of outgassed carbon of 28%.

Changes in inorganic carbon (IC) are mainly driven by climate change (under all emission scenarios) and less by the magnitude of deforestation (Figure 5E-F and Figure 6C, Tables 2

and 3). In about half of the Amazon basin the IC amount significantly changes due to climate change (insignificant changes in the other 50%), but in no cell due to deforestation. The magnitude of change varies between emission scenarios: the increase in IC is up to 4-fold in the A2 scenario and up to 2.5-fold in the B1 scenario (see Table 2). For both deforestation scenarios the gain of IC is dominant until 2050, while the basin-wide trend becomes unclear afterwards. However, sub-regions like R1 and R3 show a slight increase during the whole century (Figure 6F,J,M).

## 4 Discussion

Deforestation is, besides climate change, the largest threat to Amazonia. It leads directly to a decrease in terrestrial biomass and an increase in CO<sub>2</sub> emissions (Potter et al., 2009) and has indirect effects on aquatic biomass, diversity of species and their habitats and the climate (Asner and Alencar, 2010; Bernardes et al., 2004; Costa et al., 2003). Our results show that deforestation is also likely to change the amount of riverine organic carbon as well as exported carbon.

We identified a basin-wide reduction in riverine particulate and dissolved organic carbon pools by about 10% to 25% by the end of this century (Figure 3 and Figure 6). This reduction is particularly pronounced in areas of high deforestation intensity along the *Arc of Deforestation*, at the Rio Madeira and the last 500 km stretch of the Rio Amazon, where large deforestation rates reduce terrestrial carbon storage. In the first decades of the 21<sup>st</sup> century the differences in carbon amounts between the two deforestation scenarios are only small (Figure 6). During these decades the deforestation-induced increase in discharge is able to partly offset the decreasing amount of terrigenous organic matter which is the source of riverine organic matter. In the model, the increase of discharge after deforestation is caused by a less intense use of the available (soil) water by the crops, as compared to natural vegetation, which leaves more water for discharge (as also reported by Costa et al., 2003). After the 2050s, the differences in the organic carbon pools caused by deforestation become more obvious (Figure 6), with larger carbon decrease under the more severe BAU scenario. The same patterns occur in the two regions with the pronounced deforestation (R1 and R2). Here the reduction of terrestrial carbon directly reduces the amount of riverine carbon. The variation in future riverine carbon fluxes within each deforestation scenario can be attributed to the differences climate projections and emission scenarios, especially after 2060 when deforested area remains constant and the lagged deforestation effects vanish. In regions with low deforestation intensity (i.e. R1) the effects of land use change are much smaller and the climate change effects dominate the change in riverine organic carbon and outgassed carbon. Under the GOV scenario litter is constantly provided by the natural vegetation and small scale deforestation, and therefore fills up the litter and soil carbon pools, which are responsible for the POC and the outgassed carbon. There is a much clearer drop in the BAU scenario, where a larger fraction of the cell is subject to deforestation; partly 100% of the cell area is deforested in this scenario. In areas where the drop already starts before 2050 (e.g. Figure 6K and L, showing the results for R3) the deforestation in parts of the area already reached 100% before

2050 (also compare with timelines in Figure 2B). In these cells there is a drastically reduced influx of carbon to the litter pool (only from crops) and therefore we already see the drop earlier than in other areas (e.g. R1).

The reduction in the riverine organic carbon pools will have consequences for the floodplain and the river itself. Floodplains as well as riverine biotopes depend on the annually recurring input of organic material, either as food supply or fertilizer (Junk and Wantzen, 2004). The productivity of the floodplain forests is mainly driven by the input of nutrients which are basically sediments and organic material (Worbes, 1997). While the sediment input (also adding nutrients) might increase due to increased discharge, the input of organic material from upstream areas will decrease, leading to a reduced terrestrial and riverine productivity. This reduced productivity will certainly impact many animal species that rely on the food supplied by trees, such as fruits or leaves. The reduced supply of fertilizer and food will therefore likely affect plant and animal species compositions on local and regional scales (Junk and Wantzen, 2004; Worbes, 1997).

Additionally, deforestation will have secondary effects, including a reduction in evasion of CO<sub>2</sub> from the water (outgassed carbon). Lower terrestrial productivity after deforestation decreases the organic carbon material in the river and thus also the respiration to CO<sub>2</sub>. This is opposed by the higher respiration rate as a result of increased temperatures due to climate change. These indirect effects of deforestation on riverine carbon dynamics have to be included in future carbon balance estimates of the sink/source behaviour of the Amazon basin, since it directly couples the change in land use to the atmospheric, marine and therefore global carbon fluxes.

In contrast to the amount of riverine organic carbon and outgassed carbon the amount of riverine inorganic carbon does not show a significant effect of deforestation. The inorganic carbon in the water is only marginally affected by deforestation because the amount of IC that remains in the water depends on the saturation of the water with of IC, which is calculated depending on the water temperature and the atmospheric CO<sub>2</sub> concentration. Climate change-induced higher water temperature causes a reduction in solubility of CO<sub>2</sub>, and higher atmospheric CO<sub>2</sub> concentrations lead to an increase in dissolved CO<sub>2</sub>. The combination of both effects leads to a slight increase in dissolved inorganic carbon in the beginning and a neutral signal towards the end of the century independently of the deforestation. Any changes in the amount of IC can either be attributed to climate change (increasing temperatures and atmospheric CO<sub>2</sub> concentration) or – to a much smaller extent – to changes in the water amount in the cell. The latter can be an effect of deforestation as it is known that deforestation alters the discharge (Costa et al., 2003).

The deforestation of tropical forests will not only affect processes within the rainforest, but also processes in the adjacent Atlantic Ocean. Currently, the annual export of about 6,300 km<sup>3</sup> of freshwater is accompanied by 40×10<sup>12</sup> g of organic carbon to the Atlantic Ocean (Gaillardet et al., 1997; Moreira-Turcq et al., 2003). The present study shows that deforestation leads to a reduction in the exported organic carbon to the ocean by approximately 40%. In the NatVeg scenario the proportion of exported organic carbon to the ocean makes up about 0.8-0.9% of the net primary production (NPP), whereas in the heavily

deforested BAU scenario this proportion is reduced to about 0.5-0.6%. The reduction in the ratio of exported carbon to NPP by deforestation indicates a less pronounced future sink, since the organic carbon is directly extracted from the forest and additionally indirectly from the ocean. The Amazon basin is considered a carbon sink (Lewis et al., 2011). In central Amazonia net primary production sums up to about  $1 \times 10^9 \text{ g C km}^{-2} \text{ yr}^{-1}$  (Malhi et al., 2009). Earlier results showed that climate change alone will increase the amount of outgassed carbon from the Amazon basin by about 40%, while the export to the Atlantic Ocean remains nearly unchanged (Langerwisch et al., 2016). Our results show that additional deforestation will offset the trend in outgassed carbon (only +3%), but will have larger effects on the export to the ocean (-38%). Therefore, future assessments of climate-change and deforestation-induced changes on the carbon balance of the Amazon basin have to include the amount of carbon exported to the ocean and outgassed from the river basin to the atmosphere.

The import of organic material to the ocean positively impacts the respiration and production of the Atlantic Ocean off the coast of South America (Körtzinger, 2003; Cooley and Yager, 2006; Cooley et al., 2007; Subramaniam et al., 2008). A reduction of the import might therefore reduce the productivity in the coast-near ocean since these coastal zones depend on the imported organic matter (Cooley and Yager, 2006; Körtzinger, 2003; Subramaniam et al., 2008) and might have further impacts along the trophic cascade including herbivorous and piscivorous fish. Besides the reduced organic carbon, higher amounts of nutrients may be imported to the ocean, because the nutrients are only marginally taken up within the river and by the former intact adjacent forests. The imports of both, less organic carbon and more nutrients, might induce changes in oceanic heterotrophy and primary production.

#### 4.1 Shortcomings of the approach

The strong decrease of organic carbon may be overestimated because of our model assumptions, which include a complete removal of the natural vegetation carbon during deforestation (see e.g. Figure 6). In reality, the complete conversion of the floodplain forests to cropland or pasture is not very likely. In the more severe deforestation scenario (BAU) about 6% of the area is deforested (Soares-Filho et al., 2006). In our scenarios this also includes areas which are temporarily flooded. Since temporarily inundated areas cannot be easily converted to agricultural area or settlements, this might lead to an overestimation of deforested area. But, for example in Manaus, floodplains within a radius of about 500 km around the city have been extensively logged for construction purposes between 1960 and 1980 (Goulding et al., 2003).

In our study deforestation is simulated by partial or complete removal of vegetation carbon. This also reduces the litter and soil carbon through respiration over time, since these carbon pools are reduced in size since less dead organic material generated by the crops and managed land remains on site, other is harvested. Therefore, our estimates represent more drastic changes in riverine carbon dynamics. The sharp decrease in POC and outgassed carbon after 2050, as it is one result of our study, is caused by the implementation of carbon removal in the model. During inundation the cells are partly or completely covered with water, which leads



to the export of organic material. After the gradual decrease of forest cover (and therewith input of organic material) before 2050, there is a depletion of the remaining organic material in the following years. By a more gradual implementation of inundation in the model this harsh decrease would be softened.

In this study the mobilization of terrigenous organic material is exclusively controlled by inundation. A model that also considers the impact of precipitation, vegetation cover and slope on erosion would likely lead to an increase in erosion and thus to the import of organic matter to the river (McClain and Elsenbeer, 2001) in the first years after deforestation. However, this additional influx of carbon would only be temporal, since the soil and litter carbon pools would be eroded after some years (McClain and Elsenbeer, 2001). Thus, we assume that for the investigation of the long-term dynamics of carbon pools and fluxes, such erosion effects are only of minor importance.

## 5 Conclusion

Deforestation decreases terrestrial biomass and contributes to a further increase in CO<sub>2</sub> emissions, which reduces the terrestrial carbon sequestration potential (Houghton et al., 2000; Potter et al., 2009). Moreover, our results show that deforestation will lead to a significant decrease of exported terrigenous organic carbon, leading to a reduction of the amount of riverine organic carbon. The climate change effects additionally increase in the amount of riverine inorganic carbon. Deforestation further decreases the amount of riverine organic carbon leading to a combined decrease by about 20% compared to 10% under climate change alone (Langerwisch et al. 2015). While climate change alone leaves the export to the Atlantic Ocean with +1% nearly unchanged (Langerwisch et al. 2015), considering deforestation will now decrease the export of organic carbon to the ocean by about 40%. In contrast climate change will strongly increase the outgassed carbon by about 40% (Langerwisch et al. 2015), but including deforestation will reduce this increase to only +3%.

These changes in the hydrological regimes and the fluvial carbon pools might add to the pressures that are being imposed on the Amazon ecosystems (Asner et al., 2006; Asner and Alencar, 2010), with strong consequences for ecosystem stability (Brown and Lugo, 1990; Foley et al., 2002; von Randow et al., 2004). For instance, fish play a key role in seed dispersal along the Amazon. If floodplains turn into less productive grounds for juvenile fish, these changes might have considerable effects on local vegetation recruitment dynamics and regional plant biodiversity (Horn et al., 2011). We therefore strongly advocate the combined terrestrial and fluvial perspective of our approach, and its ability to address both climate and land use change.

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## 821 7 Tables

822 **Table 1: Location and characteristics of the three sub-regions.**

	<i>North-West corner</i>	<i>South-East corner</i>	<i>area [10<sup>3</sup>km<sup>2</sup>]</i>	<i>changes in inundation length*</i>	<i>changes inundated area*</i>	<i>land use intensity</i>
R1	0.5°S / 78.5°W	7.0°S / 72°W	523.03	1 month longer	larger	low
R2	1.0°S / 70.0°W	5.0°S / 52°W	891.32	±½ month shift	heterogeneous	medium
R3	4.5°S / 58.0°W	11.0°S / 52°W	523.03	½ month shorter	smaller	high

823 Regions are depicted in **Figure 2.** Changes in inundation compared to the average of 1961-  
824 1990, as estimated and discussed in Langerwisch et al. (2013)

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827 **Table 2: Basin-wide (B) and region wise (R1-R3) amount of carbon in POC and DOC,**  
828 **outgassed carbon and IC [ $10^{12}$  g month<sup>-1</sup>] averaged over 30 years and five climate**  
829 **models.**

	NatVeg <sub>ref</sub>	NatVeg <sub>fut</sub>	GOV <sub>futA/B</sub>	BAU <sub>futA/B</sub>	GOV <sub>futA2</sub>	BAU <sub>futA2</sub>	GOV <sub>futB1</sub>	BAU <sub>futB1</sub>
<b>POC</b>								
B	1.64±0.06	1.76±0.51	1.52±0.43	1.28±0.35	1.63±0.41	1.39±0.34	1.55±0.31	1.30±0.24
R1	0.16±0.01	0.22±0.05	0.20±0.05	0.20±0.05	0.21±0.05	0.21±0.05	0.18±0.02	0.18±0.02
R2	0.42±0.01	0.43±0.15	0.37±0.12	0.30±0.09	0.40±0.13	0.33±0.10	0.38±0.09	0.31±0.07
R3	0.15±0.01	0.14±0.05	0.11±0.04	0.07±0.03	0.12±0.04	0.08±0.02	0.12±0.03	0.08±0.02
<b>DOC</b>								
B	3.41±0.13	3.58±1.05	3.07±0.87	2.59±0.71	3.29±0.84	2.77±0.69	3.15±0.63	2.64±0.48
R1	0.34±0.02	0.46±0.11	0.43±0.10	0.42±0.10	0.45±0.10	0.44±0.10	0.39±0.05	0.38±0.05
R2	0.93±0.03	0.91±0.32	0.77±0.26	0.64±0.20	0.84±0.27	0.69±0.21	0.81±0.20	0.66±0.15
R3	0.34±0.02	0.30±0.11	0.24±0.09	0.16±0.06	0.26±0.08	0.17±0.05	0.27±0.07	0.17±0.04
<b>outgassed carbon</b>								
B	11.82±0.41	16.63±4.14	14.30±3.44	12.05±2.76	15.75±3.43	13.24±2.80	13.37±2.20	11.15±1.68
R1	1.15±0.06	2.05±0.38	1.93±0.35	1.91±0.35	2.10±0.35	2.08±0.35	1.61±0.13	1.60±0.14
R2	2.52±0.08	3.36±0.99	2.81±0.78	2.37±0.6	3.09±0.85	2.59±0.66	2.66±0.56	2.22±0.43
R3	0.99±0.04	1.12±0.42	0.91±0.34	0.55±0.20	1.03±0.32	0.62±0.18	0.94±0.26	0.56±0.14
<b>IC</b>								
B	0.227±0.003	0.457±0.119	0.457±0.120	0.456±0.121	0.523±0.137	0.522±0.138	0.365±0.063	0.364±0.064
R1	0.005±0.001	0.016±0.003	0.013±0.003	0.013±0.003	0.015±0.004	0.015±0.004	0.009±0.001	0.009±0.001
R2	0.153±0.002	0.308±0.081	0.308±0.082	0.307±0.083	0.351±0.094	0.350±0.096	0.245±0.044	0.244±0.044
R3	0.006±0.000	0.011±0.003	0.011±0.003	0.011±0.003	0.013±0.003	0.013±0.003	0.009±0.001	0.009±0.001

830 ‘ref’ refers to mean amounts during reference period 1971-2000. ‘fut’ refers to mean amounts  
831 during future period 2070-2099. Values given are the mean ± standard deviation of the five  
832 climate models.

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**Table 3: Proportion [%] of area dominated by climate or land use change impacts.**

	<i>significantly changed fraction</i>			<i>climate change dominated<sup>1</sup></i>			<i>land use change dominated<sup>1</sup></i>			<i>balanced<sup>1</sup></i>		
	<i>A1B</i>	<i>A2</i>	<i>B1</i>	<i>A1B</i>	<i>A2</i>	<i>B1</i>	<i>A1B</i>	<i>A2</i>	<i>B1</i>	<i>A1B</i>	<i>A2</i>	<i>B1</i>
<b>POC</b>												
GOV	50.85	50.91	50.86	58.8	58.7	54.9	40.9	40.7	44.6	0.3	0.6	0.5
BAU	50.80	50.85	50.85	42.3	43.7	40.1	57.5	56.2	59.8	0.2	0.1	0.1
<b>IC</b>												
GOV	50.80	50.80	50.80	100.0	100.0	100.0	0.0	0.0	0.0	0.0	0.0	0.0
BAU	50.80	50.80	50.80	100.0	100.0	100.0	0.0	0.0	0.0	0.0	0.0	0.0
<b>outgassed carbon</b>												
GOV	97.6	97.60	97.61	70.5	77.7	68.4	29.3	22.3	31.1	0.2	0.0	0.4
BAU	97.55	97.65	97.60	52.4	56.9	50.2	47.6	43.0	49.7	0.1	0.1	0.1

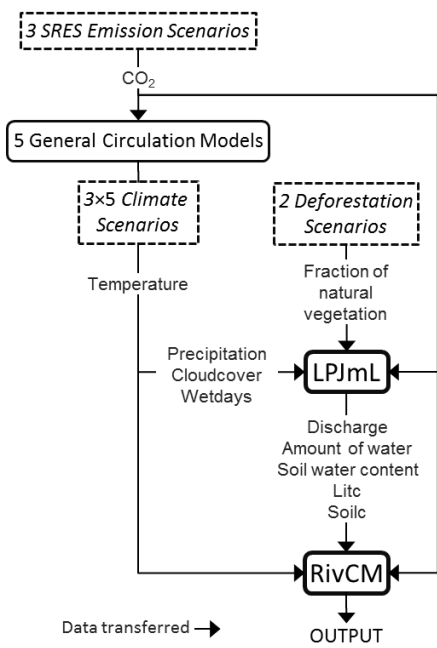
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If both impacts compensate each other the cell is balanced. <sup>1</sup>The proportions refer to the

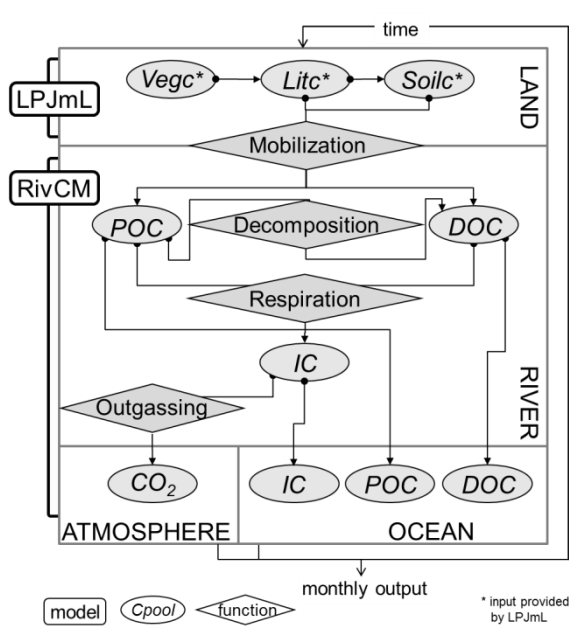
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significantly changed overall fraction (first columns).

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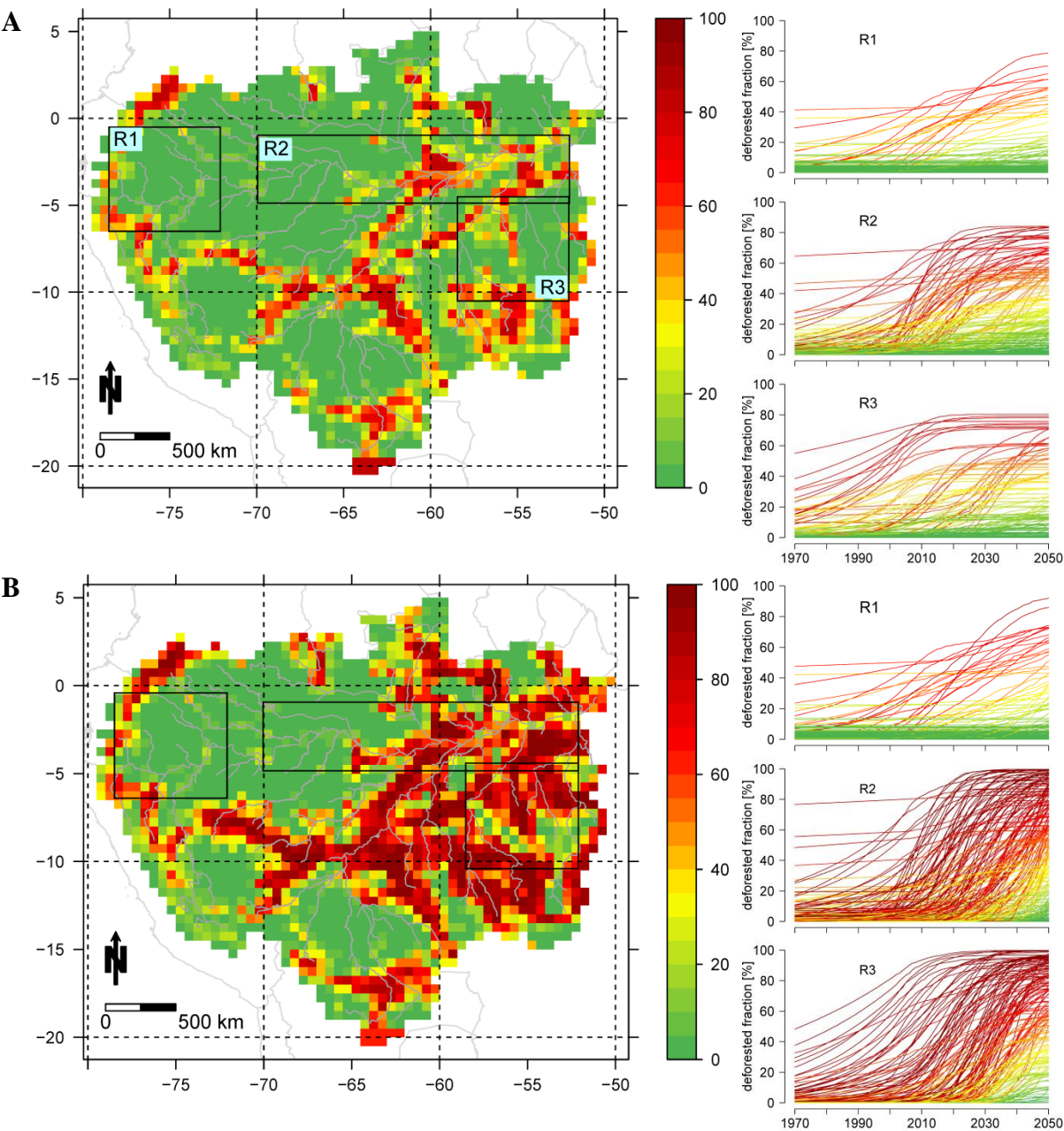


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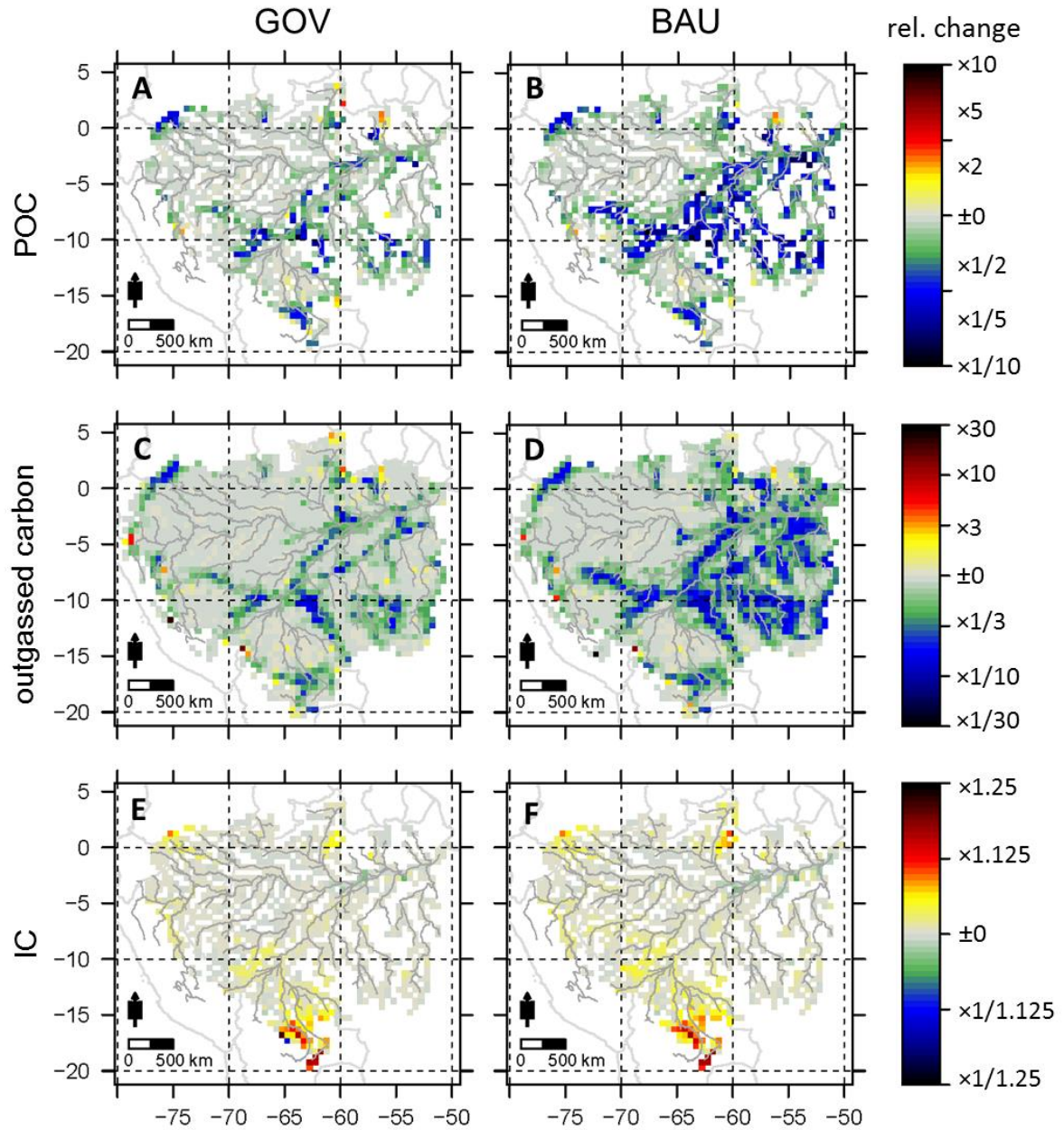


**Figure 1: Overview of the general transfer of data between scenarios and models (A) and the detailed calculation of carbon fluxes within and between LPJmL and RivCM.**

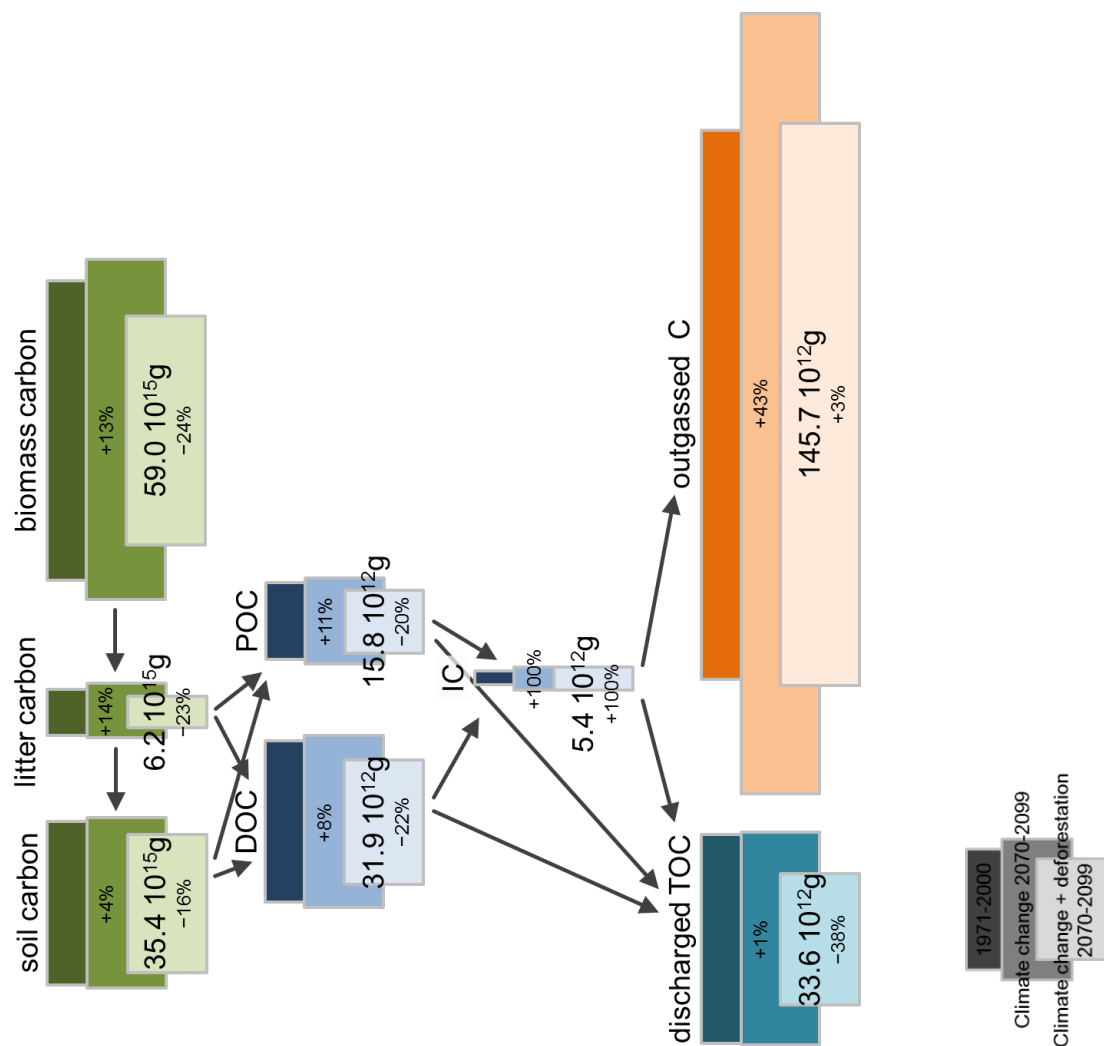




844 **Figure 2: Fraction of deforested area per cell [%] in 2050.** Data are based on Soares-Filho  
845 et al. (2006). **Panel A** refers to the BAU deforestation scenario, whereas **panel B** refers to the  
846 **GOV scenario**. The three sub-regions discussed in the main text are highlighted in the map.  
847 The timelines (right panels) show the development until 2050 for each sub-region  
848 (deforestation kept constant after 2050).

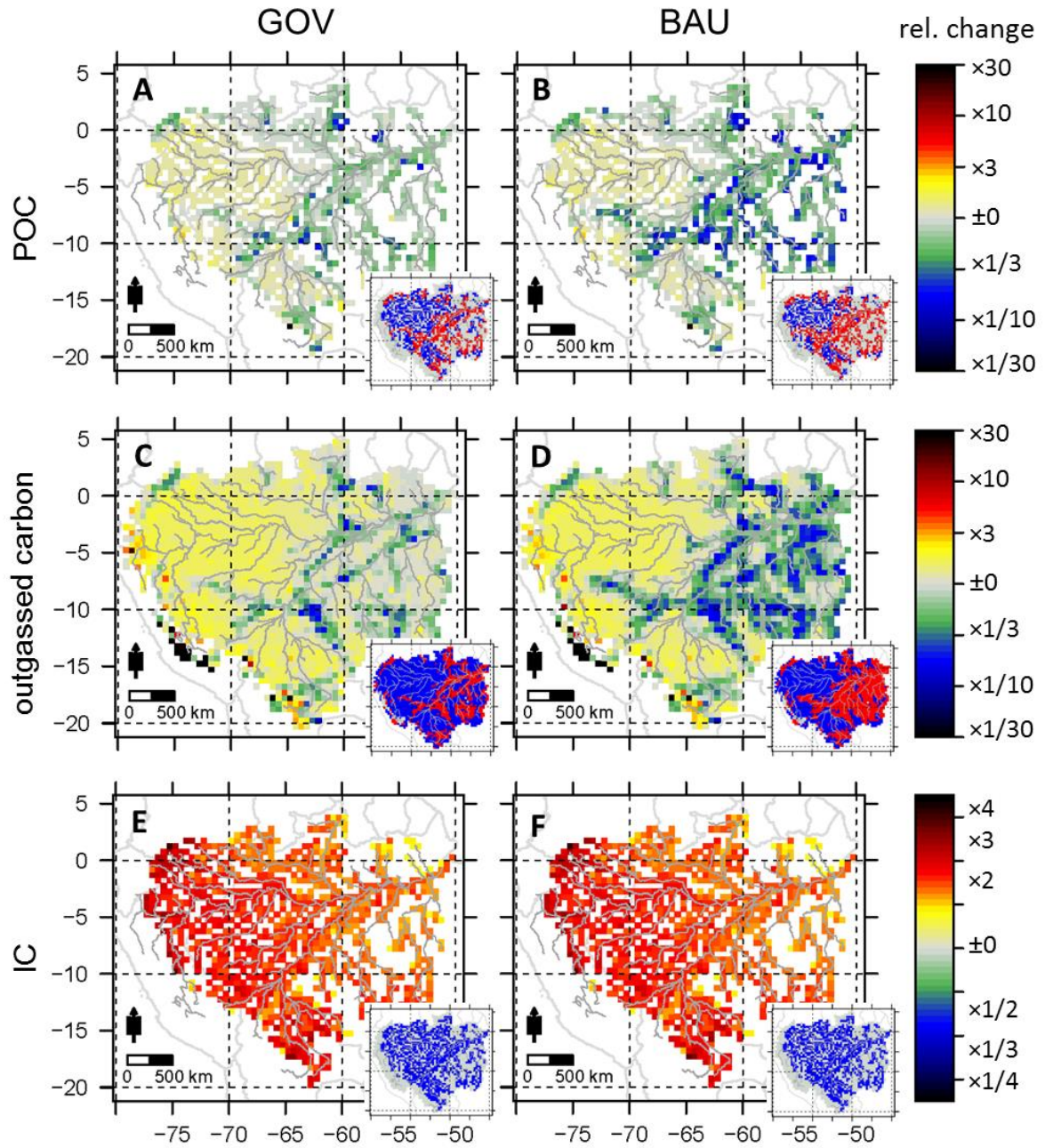


**Figure 3: Change in carbon caused by deforestation.** Climate model mean ( $E_{Defor}$ ) of the change of particulate organic carbon POC (A, B), outgassed carbon (C, D) and inorganic carbon IC (E, F). Results of the SRES emission scenario A1B are averaged over five climate models. Areas in yellow and red indicate a gain and areas in green and blue indicate a loss in carbon caused by deforestation (GOV and BAU). White areas within the Amazon basin represent cells where changes are not significant ( $p$ -value  $> 0.05$ ).



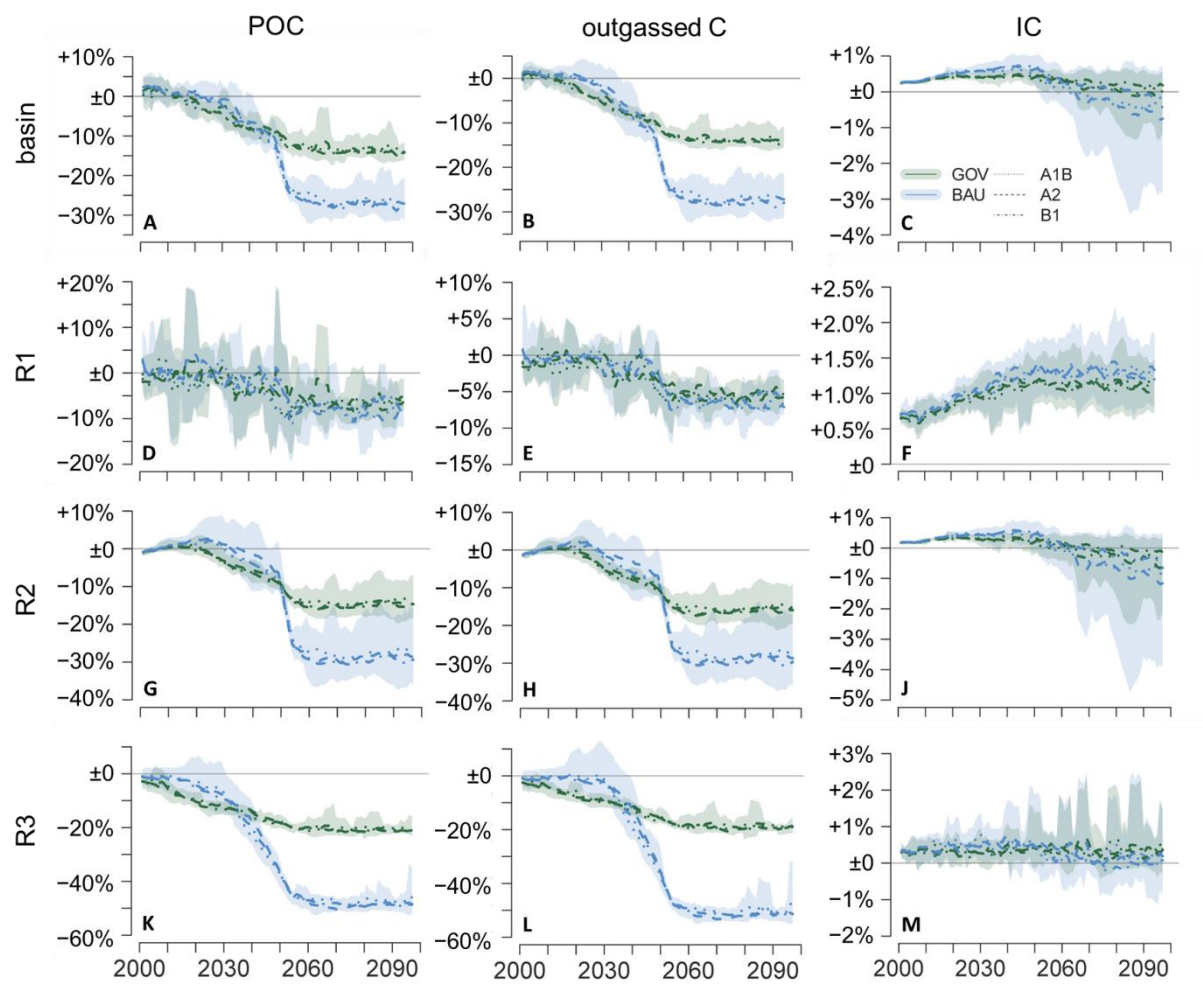
**Figure 4:** Averaged annual amounts and change in the basin carbon budget due to climate change and deforestation. Dark boxes indicate the amount of carbon during the reference period (1971-2000), intermediate boxes during the future period (2070-2099) under climate change only (Langerwisch et al., 2015), light boxes during the future period under the forcing of climate change and deforestation (BAU) together (average over all SRES scenarios and GCMs). Amount is given for future period with relative change compared to reference. Arrows indicate the direction of carbon transfer.





**Figure 5:** Change in carbon caused by deforestation and climate change. Climate model mean ( $E_{CCDeffor}$ ) of the change of particulate organic carbon POC (A, B), outgassed carbon (C, D) and inorganic carbon IC (E, F). The inset maps show blue areas where changes are predominantly caused by climate change ( $D_{CC}$ ) and red areas where changes are predominantly caused by deforestation ( $D_{Deffor}$ ). For further details see Figure 3. White areas within the Amazon basin represent cells where changes are not significant ( $p$ -value  $> 0.05$ ).

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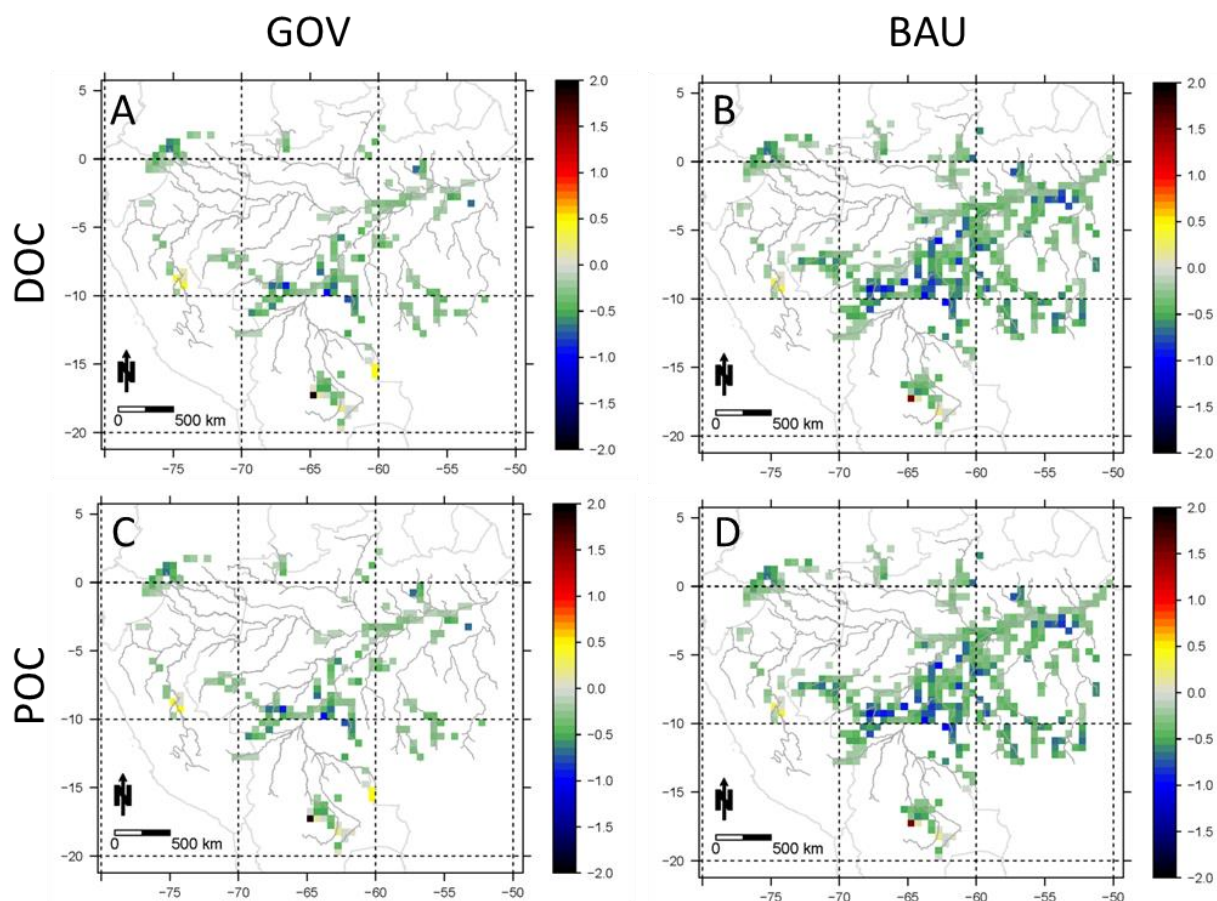
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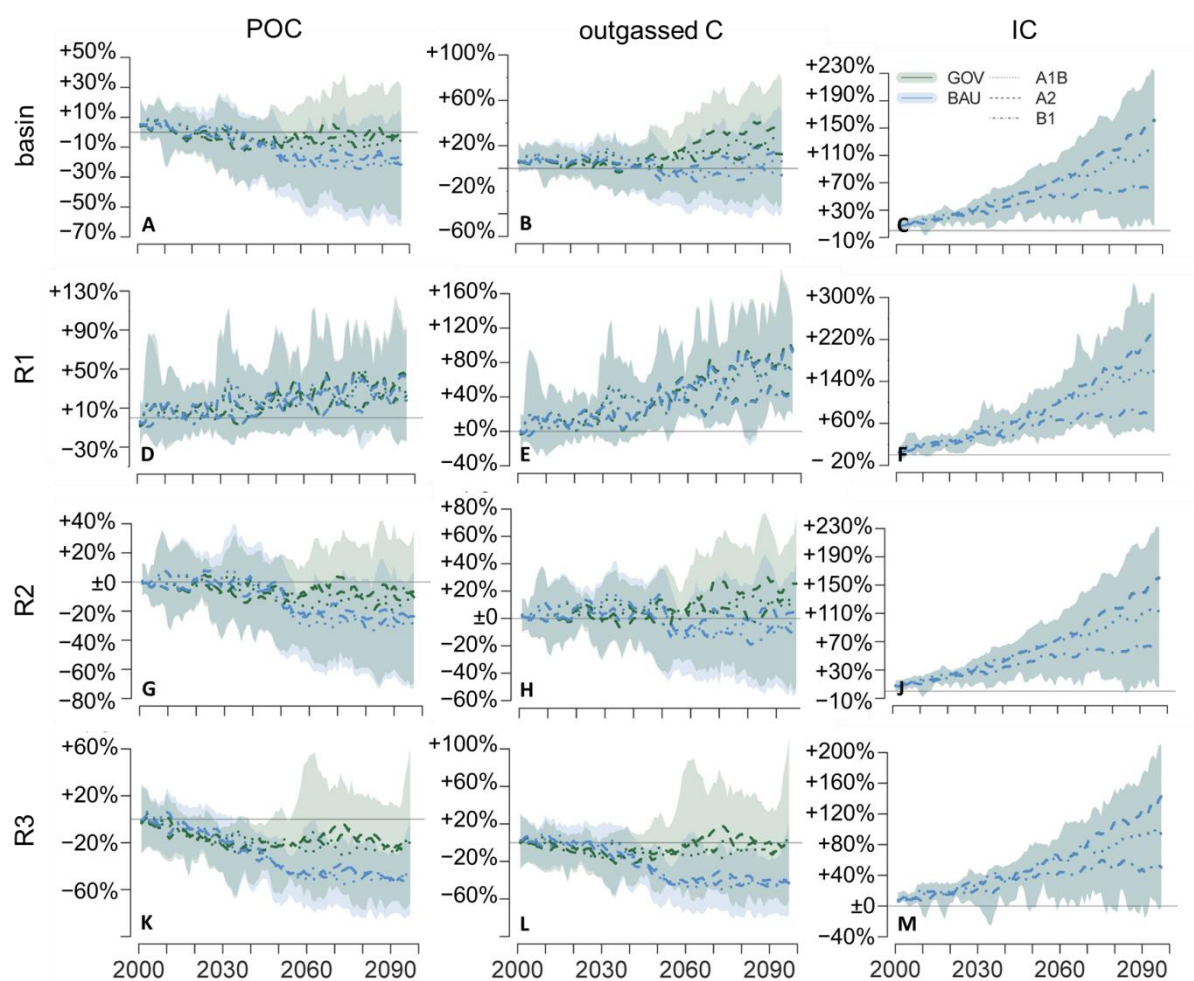
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**Figure 6:** Temporal change in riverine organic carbon due to land use change only. Change of annual sum of carbon in the deforestation scenario (GOV or BAU) compared to the NatVeg scenario for the whole basin (A-C) and the three sub-regions (R1-R3; D-M) as 5-year-mean for GOV (green) and BAU (blue), representing  $E_{\text{Defor}}$ . The shaded areas indicate the full range of values of all five climate models. Bold lines represent the 5-year-mean of the five climate models.



**Figure S1: Similar change in dissolved (A, B) and particulate organic carbon (C, D) due to deforestation.** SRES scenario is A1B, climate model is MPI-ECHAM5. Positive values (yellow and red) indicate a gain and negative values (green and blue) indicate a loss in carbon caused by deforestation (GOV and BAU). Only cells with significant changes ( $p < 0.05$ , Wilcoxon Rank Sum Test) are shown.





**Figure S2: Temporal change in riverine organic carbon due to the combination of climate and land use change.** Change of annual sum of carbon in the deforestation scenario (GOV or BAU) compared to the NatVeg scenario (average over 1971-2000) for the whole basin (A-C) and the three sub-regions (R1-R3; D-M) as 5-year-mean for GOV (green) and BAU (blue), representing  $E_{CCDefor}$ . The shaded areas indicate the full range of values of all five climate models. Bold lines represent the 5-year-mean of the five climate models.