

Methane emissions from floodplains in the Amazon basin: ~~towards a process-based model for global applications~~ challenges in developing a process-based model for global applications

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Abstract:

30 Tropical wetlands are estimated to represent about 50% of the natural wetland methane (CH₄) emissions and explain a large fraction of the observed CH₄ variability on time scales ranging from glacial-interglacial cycles to the currently observed year-to-year variability. Despite their importance, however, tropical wetlands are poorly represented in global models aiming to predict global CH₄ emissions. This study documents the first regional-scale, process-based model of CH₄ emissions from tropical floodplains. This publication documents a first step in the development of a process-based model of CH₄ emissions from tropical floodplains for global applications. For this purpose, the

40 LPX-Bern Dynamic Global Vegetation Model (LPX hereafter) was slightly modified to represent floodplain hydrology, vegetation and associated CH₄ emissions. The extent of tropical floodplains was prescribed using output from the spatially-explicit hydrology model PCR-GLOBWB. We introduced new Plant Functional Types (PFTs) that explicitly represent floodplain vegetation. The PFT parameterizations were evaluated against available remote sensing datasets (GLC2000 land cover and MODIS Net Primary Productivity). Simulated CH₄ flux densities were evaluated against field observations and regional flux inventories. Simulated CH₄ emissions at Amazon basin scale were compared to model simulations performed in the WETCHIMP intercomparison project. We found that LPX reproduces the average magnitude of observed net CH₄ flux densities for the Amazon basin. However, the model does not reproduce the variability between sites or between years within a site. simulated CH₄ flux densities are in reasonable agreement with observations at the field scale but with a tendency to overestimate the flux observed at specific sites. In addition, the model did not reproduce between-site variations or between-year variations within a site. Unfortunately, site

50 informations are too limited to attest or disprove some model features. At the Amazon basin scale, our results underline the large uncertainty in the magnitude of wetland CH₄ emissions. Sensitivity analyses gave insights into the main drivers of floodplain CH₄ emission and their associated uncertainties. In particular, uncertainties in floodplain extent (i.e., difference between GLC2000 and PCR-GLOBWB output) modulate the simulated emissions by a factor of about 2. Our best estimates, using PCR-GLOBWB in combination with GLC2000, lead to simulated Amazon-integrated emissions of 44.4 ± 4.8 Tg/yr. Additionally, the LPX emissions are highly sensitive to vegetation distribution. Two simulations with the same mean PFT cover, but different spatial distributions of grasslands within the basin modulated emissions by about 20%. Correcting the LPX simulated NPP using MODIS reduces the Amazon emissions by 11.3%. Finally, due to an intrinsic limitation of LPX to account for seasonality in floodplain extent, the model failed to reproduce the full dynamics in seasonality in CH₄ emissions: but we proposed solutions to this issue. The Inter Annual Variability (IAV) of the emissions

60 increases by 90% if the IAV in floodplain extent is account for, but still remains lower than in most of WETCHIMP models. While our model includes more mechanisms specific to tropical floodplains, we were unable to reduce the uncertainty in the magnitude of wetland CH₄ emissions of the Amazon basin. Our results helped identifying and prioritizing directions towards more accurate estimates of tropical CH₄ emissions, and stress the need for more research to constrain floodplain CH₄ emissions and their temporal variability. even before including other fundamental mechanisms as floating macrophytes or lateral water fluxes.-

1 Introduction

Methane (CH₄) is an important atmospheric component because of its contribution to radiative forcing and its role in atmospheric chemistry. Its chemical interactions result in indirect radiative forcings through its impacts on the oxidizing capacity of the atmosphere, and the production of tropospheric ozone and stratospheric water vapor. Wetlands contribute between a quarter and half of the global CH₄ emissions (Kirschke et al., 2013, *in press*). However, both the global magnitude and the latitudinal distribution of wetland emissions are poorly known (e.g., Denman et al., 2007). Tropical (30°N-30°S) wetlands are estimated to represent about 50% of the natural wetland CH₄ emissions. In addition, tropical wetlands also contribute to the variability in atmospheric CH₄ concentration at different time-scales, ranging from glacial-interglacial cycles (Loulergue et al., 2008; Singarayer et al., 2011, Baumgartner et al., 2012) to the currently observed year-to-year variability (e.g., Bousquet et al., 2006).

To estimate regional to global emissions of CH₄, two *modeling* strategies are commonly applied: the top-down and the bottom-up approach. The top-down approach, also referred to as atmospheric inverse modeling, optimally combines atmospheric observations of CH₄, a model of atmospheric chemistry and transport, and a priori information about sources and sinks (e.g., Bergamaschi et al., 2009; Bousquet et al., 2006, Monteil et al., 2011). The bottom-up approach integrates the available information about wetland CH₄ emissions at the process level into regional (Bohn et al., 2007) to global terrestrial models (e.g., Riley et al., 2011; Ringeval et al., 2011). The two approaches are complementary, in that they address different spatial scales and are constrained by observations relevant to different parts of the CH₄ budget. Top-down estimates provide only limited insight into the underlying biogeochemical processes controlling emissions, particularly over regions where several processes and sources overlap. In contrast, the bottom-up approach incorporates knowledge of small-scale processes but extrapolation of their local emission estimates to larger scales compatible with the atmospheric signals is uncertain. Top-down and bottom-up approaches are usually not independent since the top-down approach often uses bottom-up emission maps as *a priori* estimate (e.g., Spahni et al., 2011). Due to the current limitations of each approach (see e.g., Houweling et al., 2013, *submitted* about top-down models), the uncertainties after combining them are still large. This applies to both the size of global wetland emissions and their year-to-year variability (Kirschke et al., 2013). Despite these uncertainties, the Amazon watershed has been identified as a key player in the mismatch between top-down and bottom-up estimates (Pison et al., 2013). For instance, the magnitude of the Amazon wetland CH₄ emissions increases from 44 to 52 TgCH₄/yr when CH₄ retrievals from a remote sensing instrument (e.g., SCIAMACHY) are implemented as constraints in the inverse modeling system of Bergamaschi et al., (2009).

Both SCIAMACHY CH₄ concentrations and airborne measurements (Beck et al., 2012; Miller et al., 2007) showed elevated concentrations over the Amazon, but attributing these high concentrations to specific sources (e.g., wetlands) is not straightforward. In recent years, potentially important - but still debated - new mechanisms of CH₄ production under oxic conditions (see e.g., Keppler et al., 2006; Nisbet et al., 2009; Vigano et al., 2008) or anoxic conditions (Covey et al., 2012) have been proposed. Nonetheless, recent isotope analysis showed that the majority of airborne measured CH₄ concentration elevations can be attributed to microbial CH₄ production (Beck et al., 2012), reducing the number of potential drivers of the observed elevated concentrations. Despite possible alternative sources of CH₄ in the Amazon basin, wetlands thus likely remain the main source of the Amazon CH₄ emissions.

110 Melack et al. (2004) estimated the Amazon basin integrated wetland emissions at 22 Tg/yr by combining flux measurements and remotely-sensed wetland distributions. However, a large fraction of the spatio-temporal variability in the processes that control the CH₄ emissions are not accounted for in this approach, which introduces large uncertainties. The use of ~~Global Vegetation Dynamic Land Surface~~ Models (~~GVMsDLSMs~~) (e.g. Riley et al., 2011) is a promising approach for reducing the uncertainties further. However, the recent WETCHIMP ~~GVMsDLSM~~ intercomparison experiment (Melton et al., 2013; Wania et al., 2013) shows a large range of estimates for the tropics, indicating that tropical wetlands are poorly represented in these models. This is partly explained by the absence of a dedicated parameterization of tropical wetland ecosystems in the current generation of ~~GVMsDLSMs~~, affecting both the estimated wetland extent and CH₄ flux densities (i.e., the flux per m² of wetland).

120 Parameterizations introduced in ~~GVMsDLSMs~~ to simulate the wetland CH₄ flux densities are primarily representative of boreal peatlands (Wania et al., 2010; Zhuang et al., 2004) and do not account for diversity in wetland types (peatlands, swamps, etc.) (Ringeval et al., 2010). Moreover, hydrological models such as the TOPMODEL approach used in some ~~GVMsDLSMs~~, for example in the MetOffice climate model (Gedney and Cox, 2003) and ORCHIDEE (Ringeval et al., 2012) to estimate wetland extent, do not allow the simulation of floodplains, which is the main habitat associated with wetland in the Amazon watershed (Hess et al., 2003; Miguez-Macho and Fan, 2012). Further development of ~~GVMsDLSMs~~ is urgently needed to account for these omissions, given the importance of tropical wetlands for understanding global CH₄.

130 In the present study, we present the first adaptations of regional-scale a global process-based model of CH₄ emissions to Amazon floodplains specificities. This provides in the first step towards a more realistic model for tropical wetlands, dedicated to tropical wetlands to enhance our understanding and predictive ability of tropical CH₄ emissions. We do this in the framework of the LPX-Bern 1.0 Dynamic Global Vegetation Model (Land surface Processes and eXchanges, Bern version 1.0) (Spahni et al., 2013; Stocker et al., 2013) given its ability to simulate transitions of vegetation types, carbon and water pools between different terrestrial ecosystems. LPX includes a parameterization for the simulation of boreal peatland emissions developed by Wania et al. (2010). Besides, wetland CH₄ emissions were estimated in LPX for remotely sensed wetland extents (Prigent et al., 2007) by using simple parameterizations (Spahni et al., 2011, Wania et al., 2013). ~~The aim of this study is to develop a more process-based approach for CH₄ emissions from tropical floodplains by adapting~~ We adapted the

140 Wania et al., (2010) process-based approach to the case of the Amazon floodplainsbasin. To do so, we used the outputs of a hydrological model (PCR-GLOBWB) to prescribe the floodplain extent in the LPX model. ~~In addition, t~~The LPX model was extended with a representation of the Amazon floodplain vegetation, focusing on the contributions of trees and grasses to vegetation cover and productivity. The model extended with floodplains and floodplain CH₄ is tagged as LPX-Bern version 1.1. Finally, the model was used to estimate CH₄ emissions from the Amazonian floodplains. The different parameterizations were tested using remote sensing data (GLC2000, MODIS), in situ flux measurements and results of the WETCHIMP model intercomparison (Melton et al., 2013; Wania et al., 2013). ~~The model extended with floodplains and floodplain CH₄ is tagged as LPX-Bern version 1.1.~~

150 The model was used to estimate the sensitivity of the CH₄ emissions from the Amazonian floodplains to different (uncertain) processes. The comparisons with observations and sensitivity tests allowed identifying and prioritizing the challenges faced to obtain more accurate estimates of Amazon CH₄ emissions.

Section 2 describes the use of PCR-GLOBWB-simulated floodplains extent, the representation of floodplain vegetation and associated CH₄ emissions and the sensitivity analyses. Main results are presented in Section 3. Model performance, uncertainties and priorities for future developments are discussed ~~Finally, results are discussed and summarized~~ in Section 4.

160 2 Methods

2.1 The base LPX model

The LPX-Bern 1.0 (hereafter called LPX) is a subsequent development of the Lund-Potsdam-Jena (LPJ) dynamic global vegetation model (Sitch et al., 2003) which combines process-based, large-scale representations of land-atmosphere carbon and water exchanges and terrestrial vegetation dynamics in a modular framework. In the following, we briefly present the LPX characteristics relevant to understand the modifications described in Sections 2.2 to 2.5 to improve estimates of CH₄ emissions from the Amazonian basin.

2.1.1 Vegetation representation

170 In LPX, following Wania et al. (2009b), boreal (>45°N) grid-cells are treated either as peatland or as mineral soil depending on the soil carbon content derived from Tarnocai et al. (2009). The computation of water and carbon fluxes differs between these soil types. In particular, peatland soils are vertically separated into the acrotelm and the permanently water-saturated catotelm. This separation is not relevant for floodplains and therefore, floodplains are treated as mineral soils (see section 2.2.1). Plant hydrology is treated following Gerten et al. (2004) with an extension of the number of mineral soil layers to 8 following Wania et al. (2009b).

180 Each mineral grid cell of LPX is split into fractions (called hereafter Land Units, LUs). LUs are reserved for natural vegetation, agriculture (including cropland and pasture), and built-up areas (Strassmann et al., 2008). Peatlands are modeled as a separate LU (Spahni et al., 2013). The natural vegetation LU consists of 10 generic plant functional types (PFTs) that may co-exist. Peatland LUs may contain any of these generic PFTs, complemented by two peatland specific PFTs introduced by Wania et al. (2009a): flood tolerant C3 graminoids ([grassessedges](#)) and Sphagnum mosses. Within a given LU, it is assumed that the different PFTs are well mixed. As a result, the PFTs compete locally for resources while different LUs in the same grid-cell are assumed to occupy different environments, without competition of PFTs among them. To model Amazon floodplains, we introduced a new LU as well as new flood-tolerant PFTs (see sections 2.2 and 2.3).

190 The fundamental entity simulated in LPX is the average individual of a PFT. Each PFT is characterized by its own set of parameters describing growth, carbon uptake, etc. Photosynthesis and water balance are coupled in a two step approach. First, LPX first calculates the non-water-stressed photosynthesis rate, and then optimizes the canopy conductance based on water-limited transpiration (Sitch et al., 2003). Contrary to most of the commonly used [GVMsDglobal vegetation models](#) (e.g., see Krinner et al., (2005) for the ORCHIDEE model), there are no PFT-specific parameters for the optimal maximum rubisco-limited potential photosynthetic capacity (v_{cmax}) and the potential rate of Ribulosis-1,5-bisphosphate (RuBP) regeneration (v_{jmax}). This has an effect on our strategy to model flood tolerance for the newly introduced PFTs.

200 Each woody PFT population is characterized by the population density (n) and a set of variables describing the state of the average individual. E.g., the fractional plant coverage (FPC) of a given woody PFT in a LU depends on the foliage projected cover of the average individual and n . For

herbaceous PFTs, the big leaf approximation ($n=1$) is used. Vegetation dynamics is computed at a yearly time-scale. Mortality is imposed as a reduction in population density at the end of each simulation year. Such reductions can be caused by depressed growth efficiency, heat stress, negative NPP and by exceeded PFT-specific bioclimatic limits. For a realistic simulation of floodplain vegetation cover, tree mortality required specific attention as will be described in Section 2.3.3.

In addition, two rules control the coexistence of grasses vs. trees in a LU of a given grid-cell (Sitch et al., 2003), which plays a key-role in our study: i) self-thinning of woody vegetation (i.e., a reduction in the population) if the LU FPC sum of woody vegetation exceeds an arbitrary limit of 95%, ii) competitive dominance of taller growing woody PFTs by first reducing herbaceous PFT biomass if tissue growth leads to a grid-cell FPC sum greater than unity.

Finally, note that for this study ~~dynamical nitrogen cycling-limitationN~~ (Stocker et al., 2013, Spahni et al., 2013) has been turned off.

2.1.2 Wetland CH₄ emissions

Wetland CH₄ emissions are the product of the wetland extent and the CH₄ flux density. In LPX, natural wetland CH₄ emissions are computed for different classes of wetlands: boreal peatland, inundated wetland and wet mineral soils. Originally, for peatland CH₄ emissions, an additional scaling factor was used to account for the peatland microtopography (see Wania et al., 2010; Zürcher et al., 2013). In the present study, such scaling procedure is not used for tropical floodplains (see Section 2.5).

The areal extents of boreal peatlands and inundated wetlands are derived from different maps: soil survey maps derived from Tarnocai et al. (2009), Yu et al. (2010) and Wania et al. (2012) for peatlands and maps of remote-sensed inundation extent given by Prigent et al. (2007) for inundated wetland. The extent of wet mineral soils is defined as the grid-cell fraction that is not occupied by peatland and inundated wetland (and rice), but for which the soil water content is above a given threshold (Spahni et al., 2011). Originally, in LPX, floodplains were included in the inundated wetland class and thus, the extent was not explicitly defined but derived from Prigent et al. (2007). In our approach, however, the floodplain extent is prescribed using outputs of the PCR-GLOBWB model (see section 2.2.1).

In previous studies (Spahni et al., 2011; Wania et al., 2010), process-based computation of CH₄ flux densities were restricted to the case of boreal peatlands. CH₄ flux densities for inundated wetland and wet mineral soils were estimated using scaling factors, such as the simple CH₄/CO₂ ratio and CO₂ heterotrophic respiration (HR) (Spahni et al., 2011; Wania et al., 2013). A similar approach is used in the LPJ version of Hodson et al. (2011a).

In the case of boreal peatlands, the CH₄ flux density that escapes to the atmosphere results from three processes: production, oxidation and transport. Water saturation and O₂ concentration are key-variables for estimating the balance between production and oxidation and thus the resulting CH₄ concentration in each soil layer. The O₂ transport by diffusion and through plants is explicitly represented in the model. In Wania et al. (2010), the potential carbon pool for methanogenesis is estimated from the heterotrophic respiration (HR). Part of the NPP is attributed to root exudates, which contributes to HR without passing through the litter and soil pools. The potential carbon pool for methanogenesis is distributed over all soil layers, weighted by the root distribution. This carbon is split into CH₄ and CO₂ depending on a (fixed) maximum ratio and the anoxic status of each soil layer. The anoxic status of

each soil layer is computed in LPX as a function of the local soil water content rather than the actual O₂ concentration. The dissolved CH₄ concentration and the gaseous CH₄ fraction are calculated from the amount of CH₄ available in each layer. The rate of CH₄ oxidation is computed as a function of the O₂ concentration. Remaining dissolved CH₄ escape to the atmosphere by diffusion either through the soil (saturated or not) or through plant tissue pores (aerenchyma). Gaseous CH₄ escape to the atmosphere by ebullition. Thus, as in Wania et al. (2010), a total of three transport processes is accounted for: diffusion, plant-mediated transport and ebullition. The ebullition parameterization makes use of the partial pressure of CO₂ for triggering ebullition events, following (Zürcher et al., 2013). Our modifications mainly consist of accounting for the O₂ concentration to compute the anoxic status of soil layers and removal of the catotelm/acrotelm distinction in the tropical zone (see Section 2.5).

2.2 Floodplain hydrology

2.2.1 Floodplain extent

Global wetlands vary widely in hydrologic, soil and vegetation characteristics. Floodplains differ fundamentally from other wetland types, such as bogs, mires and fen which depend on local precipitation and groundwater fluctuations (Mistch and Gosselink, 2000). Floodplains result from temporarily increased river discharge, importing sediment and nutrients from elsewhere. Instead, floodplains tend to have mineral soils and can sustain productive vegetation. The Amazon basin covers 7 million km², with floodplains as the dominant type (Hess et al., 2003; Miguez-Macho and Fan, 2012).

To account for the specific conditions encountered in floodplains, a new Land-Unit (LU) “floodplain” was introduced in the LPX model. The new LU allows us to impose the water conditions of floodplains without affecting the water budget of the rest of the grid-cell, but prevents to account for seasonal variation in the wetland extent. This is because in the current LPX version, the area fraction of each LU is updated only once a year. Consequently, seasonal variations in wetland extent are not explicitly represented. Nevertheless, seasonality in the flooding depth is accounted for within the floodplain LU on a daily time-step (see section 2.2.3).

In the Amazon basin, this new floodplain LU grows or shrinks over time mainly in exchange with the natural vegetation LU, hereafter referred to as the non-floodplain LU. The hydrological conditions (annual-mean extent and seasonally varying water depth) in the LPX floodplain LU are prescribed according to the outputs of PCR-GLOBWB. This hydrological model includes a river routing routine capable of simulating the hydrology of floodplains (Van Beek and Bierkens, 2009; van Beek et al., 2011). A similar approach using PCR-GLOBWB has been used in the PEATLAND-VU model for simulation of CH₄ emissions from northern wetlands (Petrescu et al., 2010). The use of a hydrological model instead of remote-sensing measurements (e.g., Papa et al., 2010) allows us to represent the two components of the wetland CH₄ emissions (flux density and wetland extent), which can be applied to years for which no inundation data are available. The simulation of floodplain extent may introduce potential biases, in particular because evaluation of model performances is difficult for the Amazon basin (cf. Discussion).

PCR-GLOBWB is a global hydrological model and has been developed primarily for estimating the

availability of fresh water (also called “blue” water) (van Beek and Bierkens, 2009; van Beek et al., 2011; Gleeson et al., 2012). PCR-GLOBWB calculates the water storage on a cell by cell basis in two vertically stacked soil layers and an underlying groundwater reservoir on a daily time step. The exchange between the soil column and the atmosphere includes rainfall, snowmelt and evaporation from plants and interception. The soil column produces runoff, comprising direct surface runoff, subsurface storm flow and base flow, that is routed as discharge along the drainage network using the kinematic wave approximation for channels. Where interrupted by lakes or reservoirs, parameterized on the basis of the GLWD1 (Lehner and Döll, 2004), discharge is controlled by storage-outflow relationships, including a prognostic operation scheme that optimizes the release of each reservoir (van Beek et al., 2011). In each grid-cell containing a river channel, floodplains can form if the simulated channel storage exceeds the capacity at bank-full discharge. The resulting floodplain area and inundation depth follow from the distribution of elevations above the river bed, which is parameterized using the cumulative land area in a cell (subdivisions of 0.01, 0.05 and 0.1 through 1.0 by increments of 0.1). The basis of this subdivision is the HYDRO1k (http://gcmd.nasa.gov/records/GCMD_HYDRO1k.html) which provides a global, hydrologically correct elevation data set with a spatial resolution of 1×1km². The height of the floodwaters (called hereafter 'flooding depth') and the extent of the sub-merged area thus follow from intersecting the cumulative floodplain volume with the discharge in excess of channel storage. In terms of variables, for a given day *t* and a given grid-cell, the total volume that is stored in a channel reach of a grid-cell (*wst*), the resulting floodplain extent (*fldf*) and the flooding depth above ground (*fldd*) are given by PCR-GLOBWB. The *fldd* given by PCR-GLOBWB is always positive or null. River channel area is subtracted *a-posteriori* from *fldf*, while the river channel volume is subtracted from *wst*. Contrary to the earlier application of PCR-GLOBWB to estimate CH₄ emissions over northern wetlands (Petrescu et al., 2010), the extent of the floodplain is computed here online, leading to a dynamic interaction between floodplain extent and depth and flood wave propagation as a result of increased resistance compared to bank-full discharge (Winsemius et al., 2013).

In LPX, the annual extent of the floodplain LU per grid cell (*fldf_{mean}*) is defined as the area that corresponds to the annual mean flood volume (*wst*) as calculated by PCR-GLOBWB (cf. Fig A1, top and middle panels for an example). Because only two LUs are considered in the present study, changes in the floodplain LU extent are balanced by corresponding changes in the non-floodplain LU.

2.2.2 Flooding depth seasonality

The Amazon basin is characterized by a period of more frequent rain (approximatively from February through April) and a period of less intense rainfall, with the driest period between July and September. This leads to large water level fluctuations and thus seasonality in both floodplain extent and flooding depth. The river level lags behind the seasonal pattern of precipitation by a few months because of storage of water in different water pools (see e.g., Figure 1 of Bartlett et al., 1990). The Amazon River is mainly monomodal, i.e., characterized by a single pulse of flooding per year, contrary to other basins in South America.

As for peatland (Wania et al., 2009b), we computed a water table **depthposition** (WTP) variable for the floodplain LU. For a given time-step, depending on the value of the PCR-GLOBWB-derived flooding

depth that is prescribed to LPX (fdd_{LPX}), two cases arise :

- if fdd_{LPX} is equal to 0, the LPX computed soil water content is used for the WTP calculation. In this case, WTP is negative, and equal to the difference (expressed in meter) between the maximum and actual soil water content.
- if fdd_{LPX} is positive, the soil water content is set to full saturation over the floodplain LU and fdd_{LPX} is used as WTP.

The PCR-GLOBWB derived flooding depth is consistent with a seasonally varying floodplain extent. In contrast, the simulation of vegetation growth and distribution in LPX is calculated for an annually constant floodplain fraction (f_{mean}). Therefore a method is needed to calculate fdd_{LPX} , which takes into account the neglected seasonality in floodplain extent. Two alternative approaches were implemented and tested in LPX. In the 1st approach (called “redist” hereafter), the flooding depth varies proportionally with the variation of flood volume (wst) relative to its annual mean. In the 2nd approach (named “product”), the flood depth varies proportional to the product of flood depth and flood fraction. This differs from the flood volume, because floodplains have a nonuniform depth. On the other hand, the floodplain LU in LPX has a uniform depth. Approach 2 is a sensitivity test to evaluate the importance of this inconsistency.

In the “redist” approach, for a given day t and grid-cell, $fdd_{LPX} = fdd_{redist}$ where fdd_{redist} was defined as follows:

$$\begin{aligned} \text{If } wst(t) \geq \overline{wst}, fdd_{redist}(t) &= \overline{fdd} + \frac{wst(t) - \overline{wst}}{\overline{f_{mean}}} \\ \text{If } wst(t) < \overline{wst}, fdd_{redist}(t) &= \overline{fdd} \cdot \frac{wst(t)}{\overline{wst}} \end{aligned} \quad (\text{Eq. 1})$$

with wst , fdd , f_{mean} respectively in m³, m and m². Over-lined variables denote annual means.

In the “product” case, $fdd_{LPX} = fdd_{product}$ as computed in a given year by:

$$\begin{aligned} \text{If } f_{mean}(t) \leq \overline{f_{mean}}, fdd_{product}(t) &= fdd(t) \cdot f_{mean}(t) \\ \text{If } f_{mean}(t) > \overline{f_{mean}}, fdd_{product}(t) &= fdd(t) \end{aligned} \quad (\text{Eq. 2})$$

The two cases (Eq. 1 and 2) are displayed for a given grid-cell in the Figure A1 (bottom panel). The seasonal cycle of fdd_{redist} and $fdd_{product}$ over the whole Amazon basin are given in Figure A2. In both cases, a large proportion of the horizontal seasonality (i.e., the seasonality in wetland extent) is transferred into the seasonality of the flooding depth. The seasonality in wetland extents is assumed to be a major driver of the variability in wetland CH₄ emissions (Bloom et al., 2010; Ringeval et al., 2010). There are no observations on flooding depth available at large scale. However, we compared the simulated flooding depth to information available on particular sites (Section 2.6.3 and Fig. A4). In addition, the effects of the seasonality in the floodplain extent on the simulated CH₄ emissions is evaluated in the Section 3.3 and discussed in Section 4.

2.3 Floodplain vegetation

As mentioned in Section 2.1.2, heterotrophic respiration is used as a proxy for the potential carbon pool

for methanogenesis, including contributions from litter/soil carbon and root exudates. In equilibrium, and without accounting for fire-related fluxes, heterotrophic respiration is in balance with NPP. To get NPP right requires a realistic representation of the floodplain vegetation carbon balance in the Amazon basin. Note also that the vegetation type and structure are important since they influence the CH₄ transport pathway from the soil to the atmosphere (diffusion/ebullition vs. plant-mediated transport).

To simulate floodplain vegetation, flood-tolerant tropical PFTs were introduced. This allows incorporating flood-related stress on productivity in the model. In addition, we changed the mortality parameterization for trees in the floodplain LU. The aim of these modifications was to improve the representation of i) the fractional coverage of grasses versus trees and ii) the floodplain NPP. The grasses/trees contribution to total vegetation was evaluated against the GLC2000 dataset. Floodplain NPP was evaluated against MODIS-derived NPP in combination with GLC2000 (see Sect. 2.6.1).

2.3.1 Flood-tolerant tropical PFTs

The original model version accounts for three tropical PFTs: Tropical broad-leaved evergreen (TrBE), Tropical broad-leaved raingreen (TrBR) and C4 grasses (C4G). This approach was extended by defining three new PFTs, which are flood-tolerant versions of the existing tropical PFTs, increasing the total number of natural PFTs in LPX from 10 to 13.

2.3.2 Photosynthesis

The flooding of vegetation causes anoxic conditions in the rooting zone, which causes accumulation of ethylene within plants, and lower redox potentials. Eventually, phytotoxins accumulate in the rooting zone, impeding plant growth. To account for this inundation stress, we modified the anoxic stress factor on productivity as introduced by Wania et al. (2009a). Introduced parameterizations are empirical relations to mimic the influence of inundation stress on vegetation productivity and distribution in order to bring these properties into a realistic range. Chosen formulations are similar to Wania et al. (2009a). In a next step, the LPX model could be more deeply modified to be more mechanistic (see Section 4). Here, as in Wania et al. (2009a), Two PFT-specific parameters are used: a threshold water table (WTP_{max}), above which a PFT experiences *inundation*, and a maximum survival duration of inundation (t_{inund}), which counts how many days a PFT can survive under *inundation*.

Thus, for a given day d and a given PFT,

$$\text{If } WTP(d) \geq WTP_{max}, \text{icount}(d) = \min(t_{inund}, \text{icount}(d-1) + 1) \quad (\text{Eq. 3})$$

$$\text{If } WTP(d) < WTP_{max}, \text{icount}(d) = 0$$

The monthly gross primary production is reduced by $\overline{\text{icount}}/t_{inund}$ where $\overline{\text{icount}}$ is the monthly mean value of $\text{icount}(d)$. We assumed that one day with WTP below WTP_{max} is sufficient to reset the stress to 0. Contrary to Wania et al. (2009a), the icount variable is not reset to 0 at the beginning of each month to avoid unwarranted influences of the division of a year into months and has an upper limit (t_{inund}) to limit the range of $\overline{\text{icount}}/t_{inund}$ between 0 and 1. As in Wania et al. (2009a), the monthly respiration is reduced by this scaling factor.

Besides, the growth of flood-tolerant PFTs is reduced under drought-stress, following treatment of flood-tolerant C3 graminoids in Wania et al. (2009b). As soon as WTP drops below -20cm, flood-tolerant PFTs experience stress and GPP is reduced proportional to WTP by a factor ranging from 0 to 1 for WTPs between [-20cm,-40cm]. Below -40cm, the stress is equal to 1 corresponding to a

primary production of 0.

The GPP of flood-tolerant PFTs in the non-floodplain LU is reduced as well. A simplified approach is taken, independent of the hydrological conditions encountered in the non-floodplain LU. The prescribed stress corresponds to the drought stress in the floodplain LU at a WTP of -25 cm. This way, the flood tolerant PFTs are effectively outcompeted by classical PFTs and the impact of flood tolerant PFTs on the non-floodplain classical “natural” LU is limited.

To parameterize the flood-tolerant PFTs, we chose representative values for graminoids and flood-tolerant forest trees. While such representative values will never be able to represent the high variability in flooding susceptibility of this diverse ecosystem (Junk and Piedade, 1993; Piedade et al., 2010), we chose a value for emergent C4 grasses with high productivity, dominating the herbaceous Amazon ecosystem, like *Echinochloa polystachya* and *Paspalum repens*. In our approach, we did not account for floating macrophytes (e.g., *Paspalum fasciculatum*) whose ~~the~~ specificities (Wassmann et al. 1992) would require a more fundamental recoding of LPX (see Discussion). Nevertheless, the sensitivity of CH₄ emissions to key-characteristics of floating macrophytes —(no plant-mediated transport and no exudates) will be tested in the ~~next s~~Section 2.6.3. Amazon floodplain water is turbid, and in turbid water the submerged grasses cannot photosynthesize. Plant species, such as *Echinochloa polystachya*, are characterized by high productivity and long shoots to maintain leaves above the water. The higher production potential of C4 as compared to C3 plants may explain the dominance of C4 plants in tropical floodplains (Piedade et al., 1991). Piedade et al. (1991) report shoots longer than 10m (see Fig 1 of Piedade et al., 1991).

The most commonly encountered species of trees are *Laetia corymbulosa* (evergreen), *Crataeva benthamii* (deciduous) and *Pseudobombax munguba*. Most trees species occur both in flooded and in non-flooded ecosystems, except for some species, such as *Pseudobombax munguba* which is stem-succulent. Trees are characterized by reduced metabolism during the aquatic phase (Wittmann et al., 2006). Forests tolerate extended periods of flooding, up to 270d/yr (Wittmann et al., 2002).

Following these descriptions, we chose a higher WTP_{max} for flood-tolerant grasses than for flood-tolerant trees, and the opposite pattern for t_{inund} (i.e., longer for flood-tolerant trees than for flood-tolerant grasses). This parameterization makes trees more adapted to long periods of inundation than grasses, but with an immediate impact on productivity as soon as the flooded conditions are encountered. Grasses are adapted to high WTP. However, as soon as WTP exceeds WTP_{max}, grass productivity is strongly reduced to account for the impact of submergence. Unfortunately, there are no direct observations available to describe how flood duration and flooding depth affect productivity for PFTs in tropical floodplains. Instead, we evaluated ~~The sensitivity of vegetation dynamics and productivity to these parameters settings—were evaluated~~ (see section 3.2). Note also that given values of (WTP_{max}, t_{inund}) will have a different effect on vegetation dynamics and productivity depending on the chosen flooding depth description (i.e., fldd_{product} or fldd_{redist}). Negative WTP_{max} values (between -100 and -300 cm as in Wania, 2007) were assigned to non-flood tolerant PFTs, ensuring that flood-tolerant PFTs are more adapted to flooded conditions.

2.3.3 Tree mortality

Mortality corresponds to a reduction in population density and is computed at a yearly time-scale (Sitch et al., 2003). Mortality can occur as a result of light competition, a negative annual carbon balance, heat stress or when bioclimatic limits of a PFT are exceeded for an extended period (Sitch et al., 2003). A new mortality term was introduced to represent reduction in population density of flood-tolerant trees in the floodplain LU. This mortality term represents enhanced tree mortality even of adapted trees upon flooding (Bailey-Serres and Voesenek, 2008). A constant additional mortality for flood-tolerant trees at flooding, hereafter called M_{add} , is chosen and is used as tuning parameter (see the next section).

2.3.4 Soil carbon decomposition

As for peatlands (Wania et al., 2009a), the sensitivity of carbon decomposition to soil water content (R_{moist}) was set to a constant value. While the WTP could punctually fall below the soil surface (see section 2.2.3), the floodplain LU soil remains almost always saturated, justifying this approach. Under anoxic conditions, decomposition is slow because anoxia limits phenol oxidase activity and causes phenolic compounds to accumulate. For peatlands, R_{moist} of 0.35 was used. For floodplains, an arbitrary R_{moist} of 0.5 was chosen to account for a faster anaerobic degradation in floodplains than in peatlands due to a more neutral pH. The CH₄ emissions sensitivity to R_{moist} is discussed in Section 4.

2.4 Year-to-year variability in floodplain extent

The following procedure was used for the LPX simulations: first, a spin-up of 1500 years was done, by recycling the input data for the 1901-1931 period. Then, a transient run was performed for the 1931-2009 period. In the Result section, we focused on the 1990-2009 period (see Section 2.6 for more details about the general experimental set-up):

At the end of the spin-up, performed in the absence of interannual variation (IAV) in floodplain LU extent, an equilibrium in vegetation and soil carbon content is reached for both floodplain and non-floodplain LUs. The modifications introduced in LPX to simulate floodplains alter the equilibrium state. For example, floodplain forests are characterized by a large number of small trees while non-floodplain forests consist of a smaller number of larger trees.

In regional simulation 7 (see Sect. 2.6.3),

From the 1979 onwards in the transient run, inter-annual variation in floodplain extent was accounted for as forced by PCR-GLOBWB outputs (see Sect. 2.6.3.2). This leads to the conversion of land from one LU (called hereafter lu_2) to the other LU (lu_1) and of the various PFTs accordingly. All variables describing soil and vegetation were converted from one LU to the other. As mentioned earlier, nNon-floodplain and floodplain LU classes use the same PFTs; however, they behave differently in each LU. For any tree PFT of the expanding LU lu_1 , the properties (biomass, LAI, crown area, hereafter called variables V) of a tree individual were modified according to Equation 4. This equation describes the transfer of any variable V of a given PFT _{a} in lu_1 from year t to year $t+1$:

$$V(PFT_{lu_1,a}, t+1) = \frac{V(PFT_{lu_1,a}, t) * A(lu_1, t) * n(PFT_{lu_1,a}, t) + \sum_{b \in L} \Delta A * V(PFT_{lu_2,b}, t) * n(PFT_{lu_2,b}, t)}{A(lu_1, t+1) * n(PFT_{lu_1,a}, t+1)}$$

(Eq. 4)

with $\Delta A = A(lu_1, t+1) - A(lu_1, t) > 0$, and where, at time t , $A(lu_x, t)$ is the extent of the lu_x LU; $n(PFT_y, t)$ is the number of individual trees for the PFT_y , and L is the list of PFTs in lu_2 “corresponding” to $PFT_{lu_1,a}$. Indeed, each PFT is characterized by a rank defining it among the 13 possible PFTs (2nd subscript corresponding to e.g., TrBE, TrBR, C4G, flood-tolerant TrBR, etc.) and the LU in which it grows (1st subscript corresponding to floodplains or non-floodplains). As a first approach, we chose to make each PFT correspond to the same PFT of the other LU (i.e., $L = \{a\}$ in Equation 4). For instance, if the floodplain LU shrinks ($lu_2 = lu_{flood}$ in Eq. 4), TrBR of the floodplain LU is converted into TrBr of non-floodplain LU. The modifications introduced in LPX to simulate floodplains alter the forest characteristics: e.g. floodplain forests are characterized by a large number of small trees while non-floodplain forests consist of a smaller number of larger trees. Due to these differences in forest characteristics between floodplains and non-floodplains, any conversion from one LU to the other leads to a high mortality in the tree population. e.g., this happens for flood-tolerant PFTs that experience stress on the classical “natural” LU when floodplain extent shrinks.

The different litter and soil carbon pools as well as the number of individual trees after conversion of a given fraction of lu_1 to lu_2 were computed following an equation similar to Equation 4. However, these variables were not computed for the average individual in the model and thus no weighting by the number of individuals is necessary. Instead, a simple mean of the given variables of two corresponding PFTs was taken and weighted by the extent of lu_1 and ΔA . Finally, in first attempt, the LU conversion did not account for redistribution of water in order to prevent any major modifications in the non-floodplain LU.

2.5 Extension of the LPX CH₄ module to floodplains

Since the main processes leading to CH₄ emissions (production, oxidation and transport) are common to all wetlands, only a limited number of modifications of the original CH₄ emission routine were needed in order to adapt it for floodplain CH₄ emissions. Except for the introduction of a methanogenesis reduction in the presence of O₂, which is considered a generic process-based improvement, all other modifications were needed to make the CH₄ routine consistent with the implemented LPX representation of floodplain vegetation.

In Wania et al. (2010), the potential carbon pool for methanogenesis is split into CO₂ and CH₄ as function of i) the CH₄/CO₂ ratio under fully total anoxic conditions (hereafter called r_{CH_4/CO_2} and equal to 0.10) and ii) the degree of anoxia of each soil layer. The degree of anoxia of each soil layer is computed by using the soil water content: $1 - f_{air}$, where f_{air} is the fraction of air in each layer. Here we introduce a decrease of methanogenesis in the presence of O₂ in such a way that the degree of anoxia is now approximated by:

$$anoxia = 1 - (f_{air} + (1 - f_{air}) * \frac{[O_2]}{[O_2]_{eq}}) \quad (\text{Eq. 5})$$

where $[O_2]$ is the computed dissolved O₂ concentration of the given soil layer and $[O_2]_{eq}$ is the

dissolved concentration that is in Henry equilibrium with the atmosphere. This modification makes the anoxia computation related to the oxygen concentration (and not to air fraction alone) which is more in agreement with basic knowledge about methanogenesis (Conrad et al., 1989).

As underlined in previous sections, we treated floodplains as mineral soils in LPX. That means that no distinction between catotelm/acrotelm was made in the floodplain LU. Also porosity was based on a soil texture map as for other mineral soils instead of using constant porosity and minimum gas fraction for peatlands as in Wania et al. (2010). No thresholds on porosity were applied to allow ebullition. All flood-tolerant PFTs contribute to methanogenesis substrate. However, only the flood-tolerant C4 plants contribute to plant-mediated transport. Root distribution with depth is kept as in Wania et al. (2010).
 Note also that, as in Wania et al. (2010), the water column above the soil surface is added to the top soil layer, even though this is less appropriate for large flooding depths (see discussion).

Table A1 summarizes the meaning of main variables/parameters introduced in LPX.

2.6 Experimental set-up and datasets used for evaluation

2.6.1 Floodplain extent

As outlined above, we defined the annual extent of the floodplain LU per grid cell ($fldf_{mean}$) as the area that corresponds to the annual mean flood volume (wst) as calculated by PCR-GLOBWB (cf. Fig A1, top and middle panels for an example). PCR-GLOBWB was implemented on a spatial grid of 0.5° at a daily time step. For these simulations, we made use of the ERA-Interim reanalysis (Dee et al., 2011) to force the model with precipitation, temperature and reference potential evapotranspiration over the period 1979-2009 which the fields were gridded at the required spatial and temporal resolution. This reanalysis data set has limited qualities in reproducing the hydrological cycle, mainly because precipitation is not included in the data assimilation scheme. Therefore, a scaled precipitation product was chosen which corrects the ERA-interim precipitations using the GPCP observational data set (Balsamo et al., 2011). This correction, however, is coarse (2.5 degrees) resulting in a relatively poor performance over highly variable terrain. Therefore, as an additional correction, all forcing variables (precipitation, evapotranspiration and temperature) were scaled to bring their long-term means in accordance with the CRU TS 2.1 (New et al., 2002). This correction was applied on a month by month basis for time period of the CRU data set (1971-2001) and regions where stations were available.

The model was spun-up by iteratively updating the long-term components of the groundwater system over a period of 10 years followed by a transient simulation. This was then repeated by a full run over the entire simulation period to initialize the stores of the routing model (channels, reservoirs and lakes). This simulation strategy closely follows that of Van Beek et al. (2011) who used the CRU TS2.1 data set directly to force the model. Although different climatological data were used, the long-term means over gauged areas are similar. Therefore, the validation exercise undertaken by Van Beek et al. (2011) and the limitations that were found are expected to apply to our model as well.

To evaluate the land cover as simulated using the combined LPX - PCR-GLOBWB setup, we made use of the Global Land Cover 2000 (GLC2000 hereafter) land cover map (<http://bioval.jrc.ec.europa.eu/products/glc2000/products.php>). GLC2000 is based on multi-resolution satellite data (combining 4 independent datasets) and provides the dominant vegetation type, among 40 classes, for each 0° 0' 32.1444" latitude x 0° 0' 32.1444" longitude grid-cell. We made use of the regional dataset for South-America (The Land Cover Map for South America in the Year 2000, 2003).

To compare to PCR-GLOBWB, for each 0.5x0.5 degree grid-cell g , the floodplain extent was estimated by counting all GLC2000 grid-cells at GLC2000 resolution within g covered by any of the following vegetation classes: “fresh water flooded forests”, “permanent swamp forests”, “periodically flooded shrublands” or “periodically flooded savannah”. The GLC2000 class “water bodies” was not used, since it represents lakes and water channels, rather than floodplains. Floodplain fractions were compared to the long-term mean (i.e., 1979-2009) $f_{ldf,mean}$.

2.6.2 Floodplain vegetation

Three LPX simulations were performed at 0.5 degree resolution with different sets of values for f_{ldd} , WTP_{max} , t_{inund} and M_{add} (Table 1). For each simulation, the Climate Research Unit (CRU) TS 3.0 climate data set (Mitchell and Jones, 2005) was used with monthly input data for surface air temperature, total precipitation, sunshine hours from fractional cloud cover, and number of wet days. The used soil dataset was the Harmonized World Soil Database (HWSD, version 1.0., FAO/IIASA/ISRIC/ISSCAS/JRC, 2009). The following procedure was used for the LPX simulations: first, a spin-up of 1500 years was done, by recycling the input data for the 1901-1931 period. Then, a transient run was performed for the 1931-2009 period. In the Results section, we focus on the 1990-2009 period. As described in Sect. 2.4, a spin-up of 1500 years was performed before the 1931-2009 transient simulation. In the following, we focus on the 1990-2009 period. For all simulations, except for regional simulation 7 (see Sect. 2.6.3), both spin-up and transient runs were performed without IAV in floodplain LU extent. Thus, the 1979-2009 climatology of PCR-GLOBWB outputs was used to prescribe floodplain extents. Note that the climate data sets used to force LPX and PCR-GLOBWB are not the same. The correction of ERA-Interim to CRU TS 2.1, described in the previous section, accounts largely for differences in the long-term mean. However, at smaller spatial and temporal scales inconsistencies are expected to be more important, in particular because of the low observational coverage over parts of the Amazon Basin.

We compared both the simulated grasses/trees contribution to total vegetation and the simulated-floodplain NPP against observations to evaluate whether our modifications capture the floodplain vs. non-floodplain patterns well.

Besides floodplain extent, GLC2000 was used for evaluating the LPX simulated fractional vegetation cover of floodplain grasses, floodplain trees, non-floodplain grasses and non-floodplain trees (see Section 2.3). For this purpose, the LPX floodplain tree cover was compared to the sum of all the GLC2000 classes listed in Sect. 2.6.1 while “Periodically flooded savannah” was used to evaluate the fraction of flooded grasses in LPX. All other natural tree and grass classes were used to evaluate the non-floodplain grass and tree fractions in LPX at 0.5 degree resolution. For LPX, vegetation cover is calculated by summing the FPCs of the PFTs to be compared.

Floodplain NPP was evaluated against the MODerate Resolution Imaging Spectroradiometer (MODIS) derived NPP. We made use of GLC2000 to identify the location of floodplain ecosystems, extracting the NPP of floodplain ecosystems only. This approach assumes that, at a half-degree resolution, the differences in NPP between the floodplain and non-floodplain LUs are explained by the flooding conditions. MODIS-NPP was obtained from the Numerical Terradynamic Simulation Group (NTSG)

(Zhao and Running, 2010; Zhao et al., 2005) (<http://www.ntsg.umd.edu>). We regridded MODIS-derived NPP to the GLC2000 grid. Then, for each grid-cell at LPX resolution, NPP was estimated for different ecosystem types (floodplain and non-floodplain) as well as for the two vegetation meta-classes (grasses and trees) in these ecosystems. To do so, the GLC2000 maps were used to identify the ecosystem and vegetation type of each MODIS grid-cell at GLC2000 resolution. Figure A3 displays the resulting NPP maps at GLC2000 resolution, for the quadrant along the main Amazon River defined in Hess et al. (2003). This approach has the following limitations:

- The MODIS data coverage in the Amazon region is strongly limited by cloud cover, which requires gap filling (Zhao et al., 2005).
- Inconsistencies between MODIS-derived NPP and GLC2000 could occur due to the different resolutions and projection (sinusoidal and regular cylindrical) of the original data sets.
- In MODIS, no NPP values are provided for grid-cells corresponding to water-bodies, which are set to zero. Note that this mainly concerns the main stem of rivers and lakes, rather than the floodplains themselves. By changing the resolution of the MODIS-derived NPP maps, the missing values influence partially overlapping grid cells. This “contamination” is likely to be larger for floodplain than non-floodplain ecosystems, as they are closer to water-bodies. This introduces some underestimation of floodplain NPP, but given the overall uncertainty of the approach this is still considered acceptable.

The main stem of the Amazon River in Brazil is called the Solimoes River upstream from its confluence with the Negro River at Manaus. Hydrologic regimes, vegetation cover and nutrient status differ between the Solimoes river, the Negro river and the main Amazon stem after Manaus (e.g., Junk and Furch, 1993). The Solimoes River region is also called white Amazon (or *varzeas*) and is characterized by a neutral pH and a relatively high concentration of dissolved solids (clay) leading to high fertility. In contrast, Negro river region is called black and clear Amazon (or *igapos*) and is characterized by a lower pH, low concentration of dissolved solids (sand) and thus, lower fertility (c.f. the Figure 1 of Junk et al. (2011)). This difference in fertility is related to a difference in phosphorus supply (Arago et al., 2009). This leads to differences in vegetation cover (e.g., Belger et al. (2010) and see Section 4) and productivity between *varzeas* and *igapos* along a west to east gradient (e.g., Gloor et al., 2012). For instance, the diameter-increment growth rates of trees are up to two-thirds lower in *igapos* than those found in *varzeas* forests (Schöngart et al., 2010).

Only recently, **GVMsDLMSs** started to include phosphorus dynamics (Goll et al., 2012). This is not yet the case for the LPX model, which may influence the comparison between the MODIS and LPX-derived NPP of Amazon floodplains. The influence of a neglect of phosphorus dynamics and other shortcomings in the LPX simulated floodplain NPP is quantified by the following ratio:

$$a_{NPP}(g) = NPP_{MODIS}^{non-flood}(g) / NPP_{LPX}^{non-flood}(g) \quad (Eq. 6)$$

which has been computed for each grid-cell g using the mean annual non-floodplain NPP over the 2000-2009 period. To evaluate the sensitivity of LPX simulated CH₄ fluxes to uncertainties in NPP (i.e., a_{NPP}), a second set of simulations was performed (see Table 1) in which the LPX computed NPPs for both floodplain and non-floodplain LUs were corrected using a_{NPP} . The scaling factors were applied in LPX at each time-step and for all years from the beginning of the spin-up.

2.6.3 CH₄ emissions

For the estimation of CH₄ emissions, LPX simulations were performed at two spatial scales: the site

660 scale and the Amazon basin scale. LPX simulated CH₄ flux densities and CH₄ emissions were compared to available information from observations at different sites and to results of bottom-up models participating in the WETCHIMP inter-comparison (Melton et al., 2013; Wania et al., 2013).

2.6.3.1 Simulations at individual sites

LPX simulations were performed for 5 grid-cells in the Amazon basin, where measurements of floodplain CH₄ flux densities are available, using the same simulation set-up as discussed in section 2.6.2. Only few studies report measurements from floodplains in this region of the World. All these sites were used for sensitivity analysis and model evaluation. Table 2 gives the main characteristics of each site including its coordinates, the period of measurements, some technical information about the measurements, as well as vegetation cover at the sampled locations. Among the sites, three are located in the Negro river floodplain (sites 1, 2 and 3; Belger et al., 2010) and two other sites (site 4, [Wassman et al., 1992](#), and site 5, [Bartlett et al., 1988](#)) are located in the central Amazon at about 80km distance from Manaus. At all sites, measurements have been performed using flux chambers, funnels and/or through determination of gas concentration in air and water. Thus, they are representative of a very small spatial scale (the typical chamber area is 0.2 m²).

All sites classify as floodplain, despite the fact that some are referred to as “floodplain lake” or “lake” in the literature. For example, Marani and Alvalá (2007) (Supp. Site 2) define a “lake” as a permanently flooded area. Usually, measurements of CH₄ flux densities are given for different vegetation cover (emergent grasses, floating macrophytes, shrubs or forest) as well as for non-vegetated spots (i.e., open-water). All selected 5 sites include information about CH₄ flux densities for at least one vegetated spot. No information about the grass cover (emergent or floating) of the measured plots is available for any of the sites.

In addition, LPX simulations were performed for 3 additional sites (cf. the last three lines of Table 2). These sites are used in Section 4 to evaluate the ability of our modified LPX version to simulate emission from open-water bodies and emissions outside of the Amazon basin.

All simulations were compared to observation for those years for which measurements are available (Table 2). CH₄ fluxes are evaluated on annual or sub-annual time-scales. A sensitivity analysis is performed using six different settings, including an “optimal” simulation (Table 3). Each of the settings was compared to the site observations. The simulations mainly vary in vegetation cover, flooding depth and the application of the NPP scaling factor (a_{NPP}). The first three simulations were performed for the 3 parameter combinations defined in Table 1. Vegetation was either computed by LPX or prescribed through modifications of WTP_{max} , t_{inund} and M_{add} . To prescribe grasses, M_{add} was set to 1. To prescribe trees at their maximum cover (95% of the grid-cell), WTP_{max} and t_{inund} of trees are set equal to the values of flood-tolerant grasses and M_{add} was set to 0. Note, however, that this strategy slightly modifies the inundation stress value. Available information about the observed flood level on site was used to prescribe the flood level in LPX for simulations 2-7 (see Figure A4). The “optimal” simulation uses conditions that are as close as possible to reality. These conditions vary among sites as described in Table 4.

The “optimal” simulations for sites 1 and 2 include a modification in the root profile and soil porosity according to Belger et al. (2010). For sites 4 and 5, the “optimal” simulation aims at evaluating the sensitivity of the simulated CH₄ flux densities to some floating macrophyte properties, namely a suppression of the transport by plants and the fraction of NPP going to exudates set to 0.

2.6.3.2 Simulations for the Amazon basin

The set-up of simulations performed for the Amazon basin is summarized in Table 5. The simulations mainly differ in vegetation characteristics (WTP_{max} , t_{inund} and M_{add}), the parameterization of flooding depth ($fdd_{product}$, fdd_{redist}), the use of the NPP-scaling factor (a_{NPP}) and the way to account for CH₄ production/transport. The aim is to estimate the CH₄ emission sensitivity to i) the parameterization of vegetation stress and mortality for vegetation (whose major effect is on vegetation distribution), and ii) the flooding depth parameterization, iii) the introduced modifications in the methanogenesis computation (Eq. 5), iv) uncertainties in LPX-computed NPP as well as v) macrophyte properties (last column of Table 5).

Each simulation, except simulation 7, uses the same forcing climate dataset and spin-up procedure as described in section 2.6.2. Note that each simulation reaches its own equilibrium state after spin-up using its own set of parameters. This strategy differs from Wania et al. (2010), where sensitivities were evaluated after perturbing a common equilibrium state by alternative parameter settings. Our simulation 7 is used to investigate the impact of inter-annual variation in the wetland extent and thus required a different forcing dataset from 1979. While the 1979-2009 climatology of PCR-GLOBWB outputs was used for simulations 1-6 (see Section 2.6.2), the year-to-year variability in the floodplain extent given by PCR-GLOBWB was explicitly accounted for to force LPX from 1979 onwards in simulation 7. The modifications implemented in LPX to simulate floodplains (inundation stress, mortality) lead to different ecosystem equilibriums for floodplain and non-floodplain. The implementation of inter-annual variation in wetland extent from 1979 in simulation 7 altered these equilibrium. However, the introduced perturbation was strongly softened in 1993, i.e. when the period of comparison between Simulation 7 and WETCHIMP results started.

The simulation results were compared to estimates of CH₄ flux densities representative of large spatial regions. e.g., Bartlett et al. (1990) use a total of 284 flux measurements from 42 sampling sites along ~1500 km of the Amazon River to compute an average CH₄ flux density. Average fluxes are usually given for floating grass mats, flooded forest, as well as open-water. These measurements are generally similar to those described in “Simulations at individual sites” Section. Table 6 provides an overview of the studies that were used, with the corresponding geographical regions and time periods of the measurements. Smith et al. (2000) give a mean CH₄ flux density for a region along the Orinoco river, which is used in Section 4 to discuss the ability of LPX to simulate CH₄ flux densities outside of the Amazon basin. To allow a comparison of LPX results to upscaling estimates, separate emissions from flooded forest and flooded grasses were required from LPX. However, because in LPX, CH₄ flux densities are computed for the entire floodplain LU without distinguishing among PFTs, we classified each grid-cell as 'flooded forest' or 'flooded grasses' ecosystem using four different criteria (details in Appendix 1). In each case, only grid-cells where floodplain cover was greater than 5% of the grid-cell were retained.

Finally, the floodplain CH₄ emissions were compared to the outputs of the models participating in the WETCHIMP inter-comparison (Melton et al., 2013; Wania et al., 2013). WETCHIMP (Wetland and Wetland CH₄ Intercomparison of Models Project) was organized to evaluate our present ability to simulate large-scale wetland characteristics and corresponding CH₄ emissions. In the present study, we use emission estimates for the Amazon basin from 6 WETCHIMP models (see right panel of Table 7). An overview of the computation in the WETCHIMP models of the two components that determine total CH₄ emissions (i.e., the CH₄ flux densities and the wetland extent) were given in Fig. 7 and Fig. 8

of Wania et al. (2013). Briefly, none of the WETCHIMP model distinctly accounted for floodplain extents. Few models used an hydrological model to estimate the extents of locally-saturated wetland (e.g. SDGVM, Hopcroft et al., (2011)), while some other models represented all wetland types by relying on the Papa et al. (2010) dataset (e.g. CLM4Me, Riley et al. (2011)). No models had been specifically designed for tropical wetland ecosystems and had e.g. floodplain PFTs. The comparison between the current LPX model and the WETCHIMP models focuses on the magnitude of emissions in the Amazon basin, the spatial distribution as well as the temporal variability (both the seasonal and year-to-year variability) of the two above cited CH₄ emissions components. Given the uncertainty in the Amazon floodplain extent and the high estimates thereof by PCR-GLOBWB, as compared to GLC2000 (see Sect. 3.1), the emission sensitivity to floodplain extent was evaluated. To do so, we only retained grid-cells where the GLC2000 flooded vegetation fraction was larger than 2%. For these grid-cells, PCR-GLOBWB outputs were used to provide both floodplain extent and flooding depth.

3 Results

760 3.1 Floodplain extents

Around 6.8% of South America is covered by floodplain according to PCR-GLOBWB against only 3.7% in GLC2000 (Fig. 1). Outside of the Amazon basin, GLC2000 and PCR-GLOBWB agree relatively well; e.g., for floodplains in the Pantanal and along the Parana River. However, the largest differences are found within the Amazon basin, in particular in the South of the basin and along the Amazon River itself: 4.1 and 12.2% of the Amazon basin are covered by floodplains for GLC2000 and PCR-GLOBWB, respectively. If only grid-cells where the GLC2000 flooded vegetation fraction is larger than 2% are retained, $f_{\text{ldf_mean}}$ reaches 6.7%, i.e., a value much closer to the GLC2000 estimates. This means that the mismatch between PCR-GLOBWB and GLC2000 at half-degree resolution is mainly explained by the presence of additional (small) floodplains in PCR-GLOBWB, instead of a large disagreement in the extent of floodplains that are accounted for in both estimates.

This difference may be explained in part by the fact that $f_{\text{ldf_mean}}$ is only based on hydrological processes, whereas GLC2000 represents land fractions where the floodplain vegetation is dominant. $f_{\text{ldf_mean}}$ might be interpreted as the potentially vegetated floodplain extent, of which the actual occurring floodplains given by GLC2000 is a subset. In addition, some wetlands are difficult to observe by remote sensing (Miguez-Macho and Fan, 2012). For instance, the western reach of the Solimoes/Amazon mainstem floodplain has a considerably greater proportion of flooded forest than the eastern reach (Melack et al., 2004), while this is not apparent in GLC2000. The difficulty to estimate flooded area under forest canopies suggests that floodplain extent in the west of the basin may be underestimated in GLC2000. The true floodplain extent of the Amazon basin is however poorly known: e.g., Junk et al. (2011) indicate that about 30% of the 7 million km² comply with international criteria for wetland definition while Melack and Hess (2010) give an estimate of 14% of the lowland Amazon basin (<500m).

From a model point of view, the realism from simulated floodplain extent could be doubted by different model limitations. First, the used parameters as the dimensions of the river bed, its resistance and that of the inundated floodplain (Manning's coefficient set here to uniform values respectively of 0.04 and 0.10) are relatively uncertain. Anthropogenic effects are not accounted for. While the model includes the effect of additional resistance on water depths and travel times from floodplain extent, it does this as part of the channel flow. Thus, delay introduced by the flooding and draining of the inundated area is only partially accounted for. Moreover, processes such as the increased evaporation under flooded forest and the losses due to infiltration are neglected although they may become perceptible over the large flooded areas of the Amazon Basin. Nevertheless, the inclusion of the floodplain in the PCR-GLOBWB simulations results in a more realistic hydrograph for the Amazon River (Figure A5)

To test the influence of the differences in the PCR-GLOBWB and GLC2000 estimated floodplain extents, we estimated floodplain CH₄ emissions using PCR-GLOBWB outputs with and without filtering for grid-cells that contain floodplains in GLC2000 (as described in section 2.5.2).

800 3.2 Floodplain vegetation

3.2.1 Trees/grasses contribution to total vegetation cover

Overall, GLC2000 shows a larger grass contribution to the total vegetation cover in floodplain ecosystems (y) than in non-floodplain ecosystems (x) (Fig. 2; y/x ratio $\gg 1$). The exception to this pattern is Llaonos de Mojos, which is characterized by a much smaller difference in vegetation between floodplains and non-floodplains ($y/x=1.8$) compared to the rest of the Amazon ($y/x=14.1$). Llaonos de Mojos is a tropical savanna in the Bolivian Amazon, characterized by a large human impact and specific meteorological conditions with a larger seasonal cycle in precipitation (Walker, 2008). The distinct vegetation cover of Llaonos de Mojos is not represented by LPX.

810 In LPX, trees dominate over grasses in non-floodplain ecosystems. The fractional tree cover is limited to a maximum coverage of 95% in each LU (see section 2.1). At this maximum, grasses can only occupy the remaining 5%; a situation that is simulated in the non-floodplain LU of most grid-cells (see Figure 2: dots on the $x=0.05$ line). The floodplain LU was modeled to decrease tree growth, facilitating the competition of grasses ($y/x \gg 1$). Consequently, the grass contribution to total vegetation is much larger for the floodplain than the non-floodplain LU and ranges typically between 20% and 80%. The use of M_{add} (additional mortality for flood tolerant trees on floodplain LU, see Section 2.3.3) as a tuning parameter allows to bring the mean y/x ratio of LPX in agreement with GLC2000 for the Amazon basin (after excluding Llaonos de Mojos).

820 Different parameter settings, even if reproducing the same mean y/x ratio averaged at Amazon basin scale, however, still lead to very different spatial patterns at smaller scales. These differences are particularly clear when comparing the parameterizations using fdd_{redist} and $fdd_{product}$, respectively. Using fdd_{redist} , the floodplain LU is either entirely covered by trees or by grasses, whereas for $fdd_{product}$, mixed conditions are more common. This is related to the influence of the choice of fdd on productivity, affecting the inundation stress distribution (Figure A6).

3.2.2 NPP

Figure 3 compares the LPX-simulated NPP with the MODIS-derived NPP for non-floodplain (top panel) and floodplain (bottom) ecosystems. It shows the improved agreement between LPX and MODIS on both LUs when the a_{NPP} scaling factor is applied. The improvement for the non-floodplain is trivial since a_{NPP} was calibrated using non-floodplain ecosystems. A significant deviation from the 1:1 line in panel 3b remains: the slope of the regression after forcing it to have an intercept equal to 0 is 1.015 (std error: 0.003, paired t-test < 0.05). This is explained by the fact that a_{NPP} was computed from the last 10 years of the simulation plotted in panel 3a and then applied to the whole simulation (including spin-up) plotted in panel 3b (cf. Sect. 2.3.5). The application of a_{NPP} to the floodplain LU improves the agreement in floodplain NPP between MODIS and LPX: the slope of the MODIS vs. LPX linear regression increases from -0.08 in panel 3c to 0.51 in panel 3d, with a R^2 of 0.016 and 0.42, respectively. However, a mismatch in floodplain NPP between MODIS and LPX remains: slopes of regression after forcing it to have an intercept equal to 0 are 0.782, 0.901, 0.966 and are significantly different to 1 (paired t-test < 0.05). This could not be attributed to LPX shortcomings in accounting for processes relevant to both ecosystems (e.g., phosphorus limitation).

After applying a_{NPP} , we evaluated the effect of flooding conditions on NPP by comparing, for each grid-cell, floodplain and non-floodplains NPP for both MODIS-GLC2000 and LPX (Figure 4). Floodplain conditions have effects on i) vegetation cover (contribution of grass vs. trees, see 3.2.1) and ii) NPP for each vegetation type, both contributing to differences in total NPP (i.e., without vegetation type distinction) between floodplain and non-floodplain ecosystems. In LPX, the differences in vegetation cover drive the difference in total NPP between simulations 1-3 (see the larger difference in Fig. 4d than in either Fig. 4e or 4f). This difference in vegetation cover results from differences in flood tolerance settings (Figure A7). To test the significance of non-flooded vs. flooded NPP in both MODIS and LPX, we forced linear regressions to have an intercept equal to 0 (called hereafter FLR). Results of the significance test are summarized in Table A2.

In MODIS-GLC2000, the flooding conditions have only a very weak effect on the NPP of trees (Fig 4b, Table A2: slope of FLR not significantly different from 1). In contrast, grasses show a decrease in NPP of 11% on average (Fig. 4c; Table A2: paired t-test<0.05). In agreement with MODIS-GLC2000, the floodplain and non-floodplain ecosystems in LPX show almost no difference in the simulated NPP of trees (Figure 4e): while significantly different from 1 (Paired t-test in Table A2<0.05), the computed FLR slopes are very close to 1 (0.898, 0.936 and 0.949 for the LPX simulations with different parameter combinations). Despite the introduction of some stress, LPX tries to optimize ecosystem's productivity given the available resources (light, water, temperature, CO₂, nutrients, etc.). As in MODIS, grasses do show a larger difference between floodplain and non-floodplain NPP. However, the sign is opposite: in LPX, the effect of flooding is to increase the NPP (Figure 4f; Table A2: slopes of FLR significantly lower than 1).

The high productivity of flooded grasses in LPX displayed in Fig. 4f is explained by WTP_{max} and t_{inund} of grass flood-tolerant PFTs. Because of the short t_{inund} , the flood-related stress for grasses tends to be either 0 or 1. In the latter case, no grasses are present at all and by default, those grid-cells are not represented on Figure 4e. For those grid cells where a significant grass cover is simulated, their inundation stress is close to 0 and saturated conditions lead to enhanced NPP. The NPP characteristics found by MODIS-GLC2000 are not confirmed by field studies: a reduced metabolism is observed in flooded conditions for some tree species (e.g., Wittmann et al., 2006), while flooded grasses may be very productive (Engle et al., 2007; Piedade et al., 1991). As the relation from MODIS is not confirmed, it seems premature to adjust the WTP_{max} and t_{inund} parameters in LPX to better reproduce the MODIS-derived NPP of grasses.

3.3 Floodplain CH₄ emissions

3.3.1 On site evaluation of CH₄ flux density

The evaluation of LPX CH₄ emissions with site data is hampered by the scarcity of available measurements as well as by the mismatch between the spatial coverage of such measurements and the spatial resolution of Land Surface Models and the potentially large variability in emissions on small spatial scales not resolved by LPX. Figure 5 displays a comparison between the CH₄ flux densities simulated by LPX and measurements at Cuini, Itu and Araca (sites 1-3) where information is available at a yearly time scale. For sites 4 (Isla Marchentaria) and 5 (Marrecao), information is available at a monthly time-scale (Fig. 6). In the following, the two groups of sites are treated separately. Regarding

LPX, we rely on a set of sensitivity simulations with different parameter settings and vegetation cover (Table 3).

Sites 1-3 concern interfluvial wetlands in the Negro river basin. We determined the total measured flux at each site by summing up the ebullition and diffusive fluxes. A mean diffusive flux for different types of vegetation was computed by weighting grasses and non-grass vegetation (i.e., shrub, forest and palms) equally. For ebullition, the measurements do not distinguish between different vegetation types. Mean annual CH₄ flux densities of 294.8±56.5, 28.4±130.6 and 29.4±12.6 mgCH₄.m².d⁻¹ were observed for Cuini, Itu and Araca, respectively. Annual CH₄ flux densities simulated by LPX (mean over the simulation settings 4 and 5) are 307.0, 308.0 and 310.3 mgCH₄.m².d⁻¹ for each site, respectively. Thus, LPX captures the total flux at Cuini rather well, but overestimates it at Itu. The evaluation of LPX performance at Araca is difficult since no information about ebullition is available.

Total emissions range between 257 and 361 mgCH₄.m².d⁻¹ at Cuini and between 296 and 386 mgCH₄.m².d⁻¹ at Itu for the different LPX sensitivity simulations (Figure 5, Table 3). The difference between simulations 1 and 2 indicates the sensitivity of CH₄ flux densities to the use of regional PCR-GLOBWB hydrology instead of local information. Prescribed WTD leads to larger CH₄ emissions than the use of PCR-GLOBWB (3 sites average of 388.0±22.9 vs. 333.1±14.0). This is partly explained by WTD-induced changes in vegetation (cf. pie-charts in Figure 4). In LPX, the total emissions are on average lower when grasses (mean of simulation 4 over the 3 sites: 266.0±11.8) than when trees dominate vegetation cover (mean of simulation 5 over the 3 sites: 350.9±14.5). Prescribing the reported local vegetation leads to slightly higher emissions than using simulated vegetation (simulation 2 vs. either 4 or 5). The total flux is reduced by about 10% for the first 3 sites when the α_{NPP} scaling factor is applied. Finally, prescribing the local conditions for sites 1 and 2 ('optim' simulation) does not improve the agreement between LPX and observations (e.g., the observed lower ebullition at Itu).

While total emissions vary within a relatively confined range, the contribution of individual transports pathways to the total emissions can be very different for these different model setups and parameter choices. For example, plant-mediated transport is preferred over ebullition in case of an extensive grass cover (simulations 4 vs. 5). Belger et al. (2010) assume that measured diffusive fluxes include plant-mediated transport. Because of the larger difference between vegetated and non-vegetated spots in Itu than in Cuini, they concluded that plant-mediated transport is absent at Cuini. If the LPX-derived plant-mediated transport and diffusion are combined, a good agreement with the measured diffusive fluxes is obtained at Cuini when tree cover is prescribed. However, grass cover leads to a large overestimation of the combined diffusive component. This overestimation seems to be more relevant than the agreement found when prescribing trees, because the site description (Table 2) only mentions shrubs and no trees. At Itu, the overestimated ebullition cannot be evaluated with respect to the other transports pathways, in absence of vegetation specific ebullition measurements.

Figure 6 displays a comparison at monthly time-scale between LPX-simulated CH₄ flux densities and observations for Isla Marchentaria (site 4; Wassmann et al., 1992) and Marrecao (site 5; Bartlett et al., 1988). In the following, we give LPX estimates that represent means over the months for which measurements are available. Overall, the net CH₄ flux densities are captured reasonably well for forests (LPX: 233.7 vs. observations: 192±26.8 mgCH₄.m².d⁻¹) and grasses (LPX: 209.9 vs observations: 230±72.2 mgCH₄.m².d⁻¹) at site 5. For site 4, LPX-simulated CH₄ flux densities are on average

over-estimated for grasses (LPX: 200.5 vs. observations: 35.3 ± 4.8 mgCH₄.m².d⁻¹) and trees (LPX: 229.5 vs observations: 76.7 ± 8.6 mgCH₄.m².d⁻¹). Increased CH₄ flux densities are temporarily observed over forests (e.g., 200 mgCH₄.m².d⁻¹ in August) and for these particular time-periods, observations and LPX are in agreement. Indeed, a large seasonality in CH₄ flux density in forests is observed at site 4 and is characterized by high fluxes during high water level (May-July) and low fluxes during rising water levels (Feb-Apr). According to Wassmann et al. (1992), these low fluxes are caused by an increased oxygen concentrations in the water column due to entrainment of oxygen rich river water into the floodplain. Such lateral and small-scale water fluxes are not accounted for in the model and could explain the underestimated seasonality in LPX-simulated CH₄ flux densities.

At site 5, there is no significant difference in the observed total flux density between tree- and grass-covered areas. This is not the case for site 4, where flux densities in tree-covered areas (averaged over the measurement period) are about two times the flux over grasses. For both sites, LPX reproduces a slightly larger flux over trees than over grasses, but the size of the difference (~+12.6 and +10.2% resp. for site 4 and 5) is almost insignificant.

At site 4, the measured contribution of different transport pathways to total emissions are similar for tree- and grass-covered areas and vary seasonally. A predominantly diffusive contribution is observed during rising water tables (up to 82.8% of total flux in forest during Feb-April) while ebullition dominates during high water levels (up to 96.3% of total flux in forest during May-August). Because of floating macrophytes, plant-mediated transport is assumed to be absent at site 4 (Wassmann et al., 1992). Site 5 shows a larger contribution of diffusion for grasses than for trees. As for site 4, 'grasses' mainly include floating macrophytes (Bartlett et al., 1988). LPX-simulated CH₄ flux densities are characterized by a predominance of plant-mediated transport when grasses are present and of ebullition in the presence of trees at both sites 4 and 5. In both cases, the contribution of diffusion to total flux is very small. Floating macrophytes are not accounted for in the model, which could explain the difference with the observations. In the optimized run, ebullition replaces plant-mediated transport with little impact on the contribution of diffusion. Thus, the contribution of diffusion in the optimal run is still smaller than the observed yearly minimum (i.e., during high water level). There is no seasonality in the contribution of each type of transport to the total flux in LPX.

For site 4, prescribed WTD leads to only a slight reduction in flux densities (from 107.1 in simulation 1 to 101.6 in simulation 2). There are no floodplains in PCR-GLOBWB at site 5 and therefore simulation 1 shows no CH₄ flux density at this site. At both sites, the correction of NPP by α_{NPP} reduces the emissions by about 23%. In simulations 1-3, grid-cell is entirely covered by grasses for both sites 4 and 5; thus, these simulations are similar to simulation 5.

To summarize the comparison with site observations, it seems that LPX roughly reproduces the magnitude of observed net CH₄ flux densities but with a tendency to overestimate emissions at some sites. LPX is less well able to reproduce the variability among sites. Also variability in time at a given site seems to be difficult to capture, which could be explained by the absence of lateral fluxes of oxygen-rich water in the model. LPX simulates larger CH₄ flux densities at flooded forest sites than at grass covered sites but it is difficult to verify this difference with the available observations (e.g., no vegetation distinction in ebullition in Belger et al. (2010)). Floating macrophytes have so far not been represented in the model. Despite the obvious impact of this omission on the representation of

transport, it seems to have a weak effect on total CH₄ flux density. LPX-simulated diffusion seems to be under-estimated in floodplains along the Amazon main stem. LPX shows a large sensitivity of transport pathways to vegetation cover, which is difficult to validate since the contribution of plant-mediated transport to the observed fluxes is often unclear. In absence of floating macrophytes and lateral exchange in LPX, our current parameterization may be better suited for simulating CH₄ flux densities in interfluvial wetlands in the Negro River basin as described in Belger et al. (2010) than along the Amazon main stem.

3.3.2 CH₄ emissions for the Amazon basin

3.3.2.1 Evaluation against measurement inventories

Figure 7 compares mean CH₄ flux densities from LPX to interpolated observations along the Amazon River main stem (~1700km) (Area n°1 in Table 6, top panels of Figure 7) and in a region around Manaus (Area n°2 in Table 6, middle panels of Figure 7). No upscaling estimates are available for the emissions from the Negro river floodplains and these floodplains are therefore not included here. No information derived from the site scale has been used to force LPX; vegetation cover was simulated internally and WTD is from PCR-GLOBWB as described in Section 2. Open water measurements (1st bar of each left hand panel) will be discussed in Section 4.

An interesting feature derived from the observations in floodplains along the Amazon River main stem (Fig 7 - top panel) concerns the larger CH₄ flux densities associated with floating macrophytes than with flooded forests (average over measurement period: 240.7±64.0 for floating macrophytes (grasses) vs. 69.4±18.1 mgCH₄.m².d⁻¹ for flooded forests). This difference is not significant for the region around Manaus, where most of the studies on sites discussed in section 3.3.1 were performed (Figure 7 - middle line). Note, however, that the measurements from geographical regions 1 and 2 do not represent the same time of year (cf. Table 6). In addition, an unknown contribution to this difference may come from differences in flooding depth between these forest and grassland ecosystems. LPX simulates larger CH₄ flux densities for tree than for grass-dominated areas for all studied periods (e.g., along the Amazon River main stem: 225.0±30.0 for grass and 349.3±58.0 for forest). The mismatch between LPX and observations is larger for forests than for grasses. In the region around Manaus, LPX captures the magnitude of CH₄ flux densities well over grass (LPX: 226.5 ±31.7 vs 214.0±64.0) and over-estimates fluxes over forest (359.9±0.8 for LPX vs 150.0±98.0). These results are relatively insensitive to the specifics of the procedure used to attribute a vegetation type to each grid-cell (see Appendix 1).

For floodplains along the Amazon River main stem, the observed CH₄ flux densities for the two vegetation types show a different seasonality. The amplitude of the seasonal cycle is larger for floating macrophytes than for forests. In particular, a large increase in CH₄ flux density is observed for floating macrophytes during July-August (~+95% as compared to April-May). The LPX simulated seasonal amplitudes for inundated forest and grassland are much smaller than observed.

For the same reasons as discussed in Section 3.2.1, the LPX simulated contribution of different transport pathways is difficult to evaluate.

3.3.2.2 Comparison to WETCHIMP models

Figures 8, 9 and 10 display CH₄ flux densities, wetland extent and CH₄ emissions in the Amazon basin

for the current LPX version and the WETCHIMP models, respectively. The magnitude of the Amazon-intergrated wetland CH₄ emissions are summarized in Table 7.

The agreement among the various WETCHIMP models is poor for the Amazon basin in terms of both CH₄ flux densities (Figure 8) and CH₄ emissions (Figure 10), as found previously at global scales (Melton et al., 2013). With the exception of SDGVM (Hopcroft et al., 2011), the agreement among the WETCHIMP models is better for wetland extent (Figure 9): all models account for wetlands along the Amazon river main stem, Negro river and in the Pantanal. This pattern could be explained by the common use of the remote sensing derived inundation dataset of Papa et al. (2010) as prescribed or prognostic wetland map in most of the WETCHIMP models. SDGVM is characterized by an extreme imbalance between the contribution of each component (CH₄ flux densities and wetland extent) as compared to other WETCHIMP models (Melton et al., 2013; Ringeval et al., 2013). LPX wetland extents (Figure 9) are characterized by high floodplain fractions in many grid-cells (see section 3.1). The use of GLC2000 to retain only grid-cells where the floodplain extent exceeds 2% of the grid-cell yield a spatial pattern that is in better agreement with the other WETCHIMP models.

Two distinct patterns of CH₄ flux densities in the Amazon basin are found among the WETCHIMP models. LPJ-Bern (Wania et al., 2010; Spahni et al., 2011; Zürcher et al., 2011), DLEM (Tian et al. 2010, 2011; Xu et al., 2010) and LPJ-WSL (Hodson et al., 2011) show larger CH₄ flux densities along the Amazon River than elsewhere in the basin, while SDGVM (Hopcroft et al., 2011), ORCHIDEE (Ringeval et al., 2011) and CLM4Me (Riley et al., 2011) simulate more homogeneous CH₄ flux densities throughout the basin. The pattern simulated by LPX is, however, characterized by i) low (<180 mgCH₄.m⁻².d⁻¹) CH₄ flux densities in grid-cells closest to the rivers, ii) medium CH₄ flux densities (180-420) in grid-cells further away and iii) highest (>420) flux densities at the boundary of the basin. Nevertheless, the maximum values obtained by LPX are lower than those of ORCHIDEE and CLM4Me. In CLM4Me, this is probably due to an overestimation of the NPP simulated in the Amazon region (Bonan et al., 2012). The LPX simulated spatial distribution of CH₄ flux densities could be explained by two factors: larger grass cover in grid-cells around the river (and thus lower CH₄ flux densities (cf. above)) and lower WTD in grid-cells at the boundary of the basin leading to less oxidation.

Concerning the Amazon-integrated wetland CH₄ emissions (Figure 10 and Table 7), two patterns can be distinguished among the model outputs: models with annual emissions lower than 10 Tg/yr (LPJ-Bern and DLEM) and models that simulate very high Amazon emissions (>50 Tg/yr; ORCHIDEE, SDGVM, CLM4Me). WSL is intermediate between these two with emissions around 20 Tg/yr. An interesting feature is also the high spread variation in the Amazon contribution to the total global emissions (between 6 and 30%, cf. Table 7). Amazon LPX-emissions are higher than those of any other model (~89.1±9.3 Tg/yr) and remain in the upper range after the GLC2000 mask is applied (~44.4±4.8Tg/yr). The reduction in LPX CH₄ emissions obtained for the Amazon basin with the application of the GLC2000 mask (factor of 2) is close to the reduction in area fraction (fldf_{mean}) with this same mask (factor of 1.8 smaller; see Sect. 3.1), suggesting that the "masked" area emits per m² just a little bit less than the "non-masked" area. The realism of Amazon LPX-emissions is discussed in Section 4.

Our simulation set-up (simulation scenarios 1-7) allows estimating the sensitivity of floodplain CH₄ emissions to specific processes (last column of Table 5). First, the LPX emissions are very sensitive to vegetation distribution. Simulations 1 and 2 are characterized by the same basin-integrated contribution

of grasses to vegetation cover (Figure 2 and Section 3.2.1). However, the change in vegetation distribution within the basin, induced by the difference in flood stress parameterization, leads to an increase in emissions by about 20.2% from simulation 1 to 2. The stress parameterization indirectly also affects the mortality rate of plants that die during transitions from floodplain to non-floodplain or vice-versa. Dead organic matter ends up being used as a substrate for methanogenesis. A change ~~in vegetation distribution combined with a change~~ in the flooding depth and its indirect effect on vegetation distribution (from simulation 1 to 3) leads to a slight decrease of 3.6%, whereas the application of a NPP correction (a_{NPP} scaling factor) reduces the Amazon emissions by 11.3%. Figures 3a shows that the a_{NPP} ratio is larger (resp. lower) than unity in the east (resp. west) of the basin. Therefore, the reduced emissions in simulation 3 are partly explained by floodplains located in the west of the basin. Accounting for variations in O₂ concentrations in the computation of methanogenesis (Equation 5) leads to almost no modification in Amazon emissions (simulation 1 to 5: +0.1%). The implementation of Equation 5 could have an effect at site scale, when the prescribed flooding depth is equal to 0 during part of the seasonal cycle but the effect is much smaller when PCR-GLOBWB flooding depths are used. Indeed, most emitting areas are characterized by a large flooding depth in PCR-GLOBWB, which strongly limits O₂ diffusion. Finally, a shut down of plant-mediated transport and absence of exudates lead to very small decrease in Amazon emissions (simulation 1 to 6: -0.7%). In simulation 6, the effect of reduction in exudates on methanogenesis substrate is negligible and the ebullition increase almost entirely counterbalances the decrease in plant-mediated transport. The introduction of IAV in floodplain extent in the last 30 years of simulation leads to an emission increase of 6.6% (from simulation 1 to 7). This is explained by an increase in methanogenesis substrate due to higher tree mortality. This increase results partly from the set-up of simulation 7 and the perturbation following the artificial introduction of IAV in floodplain extent in 1979.

Figure 11 (a-b) displays the mean seasonal cycle of wetland extent over 1993-2004 (panel a) and associated CH₄ emissions (panel b). Seasonal cycles were normalized by dividing each curve by its annual maximum. Again, a large variation is found among the WETCHIMP models concerning phasing (month of maximum/minimum) and amplitude of the seasonal cycle of both wetland extent (a) and CH₄ emission (b). For almost all WETCHIMP models, the relative amplitude is larger for CH₄ emissions than for wetland extent. Both wetland extent and emissions show a maximum value between March and April and a minimum extent between September and November. The PCR-GLOBWB-derived floodplain extent shows a similar pattern, but is characterized by a seasonal amplitude in the low end of the range of the WETCHIMP models (max – min is equal to 27% of the maximum annual value). The amplitude of the seasonality in LPX-emissions is very low (20% of the maximum value) and shows an opposite pattern (black curve) with slightly larger emissions during Sept-Oct and lower emissions during March-May. As explained in section 2.2.1, a yearly constant floodplain fraction is used as input of LPX ($f_{ldf,mean}$) but a part of the seasonality in the wetland extent is transferred into seasonality of flood depth (Eq. 1 and 2). To evaluate the sensitivity to yearly constant floodplain fraction, we applied *a posteriori* the seasonality in floodplain extent by multiplying for each grid-cell the simulated CH₄ emission by the seasonal cycle of PCR-GLOBWB-simulated floodplain extent (orange curve in panel 11b). It is worth mentioning that, in this test, the flooding depth seasonality given by Eq. 1&2 was still used-. The *a posteriori* multiplication of emissions cannot allow to account for the effect of seasonality in floodplain extent on whole carbon cycle (vegetation distribution, heterotrophic respiration, etc.) but only for its direct effect on CH₄ emissions (through change in emitting areas). We found that this procedure reconciles the phase of the LPX-seasonality with that of the WETCHIMP models, whereas

the simulated seasonal amplitude remains smaller in LPX than in other models (Fig. 11b). This underlines the difficulty to make a corresponding link between horizontal and vertical seasonality in hydrology. While a flooding depth increase tends to decrease emissions due to increased oxidation, increased floodplain extent leads to increased emissions.

1100 Figure 11c-d shows the IAV in wetland extent and CH₄ emissions of the WETCHIMP models and LPX. At the Amazon basin scale, some common features are found in ORCHIDEE and SDGVM on the one hand and DLEM and WSL on the other hand. LPJ-Bern does not account for IAV in the wetland extent (blue curve in Fig 11 c; cf. Experiment 2 in Wania et al. (2013)). The IAV of the PCR-GLOBWB-derived floodplain extent is smaller than simulated by the WETCHIMP models. The introduction of the GLC2000 filter in LPX does not influence the IAV variability in CH₄ emissions in LPX. The black curve in Figure 11d represents the IAV in emissions averaged over simulations 1-6 and thus does not account for IAV in the floodplain extents. Therefore, this IAV only reflects the IAV of the CH₄ flux densities, which turns out to be insensitive to the differences between simulations 1-6. This black curve can be compared to the blue curve, which represents a simulation of LPJ-Bern without IAV in wetland extent. The absolute CH₄ emission variability in the LPJ-Bern model (blue) is larger than in the LPX model (black) by about +80%. The difference between the two models results either from the modifications in floodplain vegetation introduced in this study or from a difference in the spatial distribution of wetlands. The use of PCR-GLOBWB calculated IAV in floodplain extent increases the IAV of the emissions by about +93% (from black curve to orange one in panel d) even though the IAV of floodplain extent is low as compared to WETCHIMP models (panel c). Despite this +93% increase, the IAV of the LPX-emissions remains lower than most of the WETCHIMP models. The IAV in floodplain extent affects CH₄ emissions directly through a change in emitting areas and indirectly through modifications in vegetation cover, the two effects cannot be separated here.

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4 Discussion and conclusions

We adapted the LPX model, which was initially developed for boreal CH₄ emissions (Wania et al., 2010), to include Amazon floodplain specificities. Our modifications aimed to represent tropical floodplain hydrology, vegetation and associated CH₄ emissions (see below). These modifications lead to substantial improvements in terms of modeling approach but do not reduce the uncertainty in Amazon wetland emissions. However, the advanced treatment of floodplains made uncertainties and challenges about hydrology, vegetation and associated CH₄ emissions explicit. These challenges relate to i) the need for observational and experimental work necessary to mechanistically represent these additional processes, ii) uncertainties in formulation and parameterization of processes that were included and in processes that were neglected in our first step procedure. In addition, we discussed below the challenges towards the modeling of tropical CH₄ emissions outside of the Amazon basin.

4.1 Observations needed to reduce the uncertainty in Amazon CH₄ emissions

Future required model improvements

Our modifications aimed to represent tropical floodplain hydrology, vegetation and associated CH₄ emissions. These modifications lead to improvements in terms of modeling approach but do not reduce the uncertainty in Amazon wetland emissions. CH₄ emissions floodplain simulate tropical the LPX model to LPX simulated CH₄ flux densities are in reasonable agreement with the average magnitude of observed net CH₄ flux densities at site scale. However, variability in CH₄ emissions between sites and in time at the same site were not well modelled. The reasons for these differences are only partly understood as field observations are too scarce to constrain the different transport pathways or the difference of emissions between grass-covered and forest-covered plots. This prevents one to draw quantitative conclusions. Riley et al. (2011) pointed out that it is possible to simulate the fluxes well for a given location, but with incorrect contributing processes, such as production, oxidation, or transport. Thus, not only the net soil/atmosphere flux but also site-level information about the different components of the flux are required. Also, as recommended by Melton et al. (2013), there is a need for observation datasets at appropriate spatial and temporal scales for the coarse resolution of LSMs. E.g. the spatial cover of flux chamber measurements (0.2m²) is not compatible with the measurement of ecosystem variables such as NPP or vegetation cover which are required to evaluate the model variables. Eddy-covariance measurements in the Amazon basin could provide information that is more compatible with spatial resolution of LSMs but up to now, they are restricted to uplands (Querino et al., 2011).

In addition, we identified important difficulties in constraining some key variables (floodplain extent, vegetation cover, NPP) at regional scale by observations. This is primordial given the results of our sensitivity tests that showed the large CH₄ emissions sensitivity to such variables. This limitation arises first from uncertainties and inconsistencies in global remote sensing datasets (Miguez-Macho and Fan,

1160 2012). Given the intrinsic difference between GLC2000 and MODIS, we experienced difficulties when trying to combine both products to estimate floodplain NPP for trees and grasses. In addition, the comparison of remotely sensed NPP with the few available field observations showed a disagreement (Wittmann et al., 2006; Engle et al., 2007).

Given the large variability among the WETCHIMP models, an interesting question is what the size of the seasonal cycle and the year-to-year variability should be. This may be assessed through the use of a chemical-transport model in combination with aircraft measurements. Such datasets are now available (Beck et al., 2012) and offer a good opportunity to constraint further the CH₄ emissions at Amazon basin scale.

4.2 Current and future required model developments

1170 While LPX cannot simulate the water cycle related to floodplains, our approach allowed to explicitly accounting for floodplain extent through coupling LPX to PCR-GLOBWB outputs. In addition, new PFTs were introduced to improve the representation of floodplain vegetation. Parameterizations ~~have been were~~ introduced to modify both the vegetation cover and productivity of tropical floodplain as compared to non-floodplain. ~~These modifications have been evaluated against GLC2000 and MODIS.~~ From a modeling point of view, our approach is more advanced than most of the WETCHIMP models, because the carbon balance of soils is treated independently for wetlands and uplands. In most of the WETCHIMP models (Melton et al., 2013; Wania et al., 2013) the wetland extent can occupy a fraction of the grid-cell, but there is no subgrid treatment of the carbon cycle fluxes. Thus, inundation has no effect on vegetation, carbon pools and heterotrophic respiration. Instead the mean value of the

1180 heterotrophic respiration over the entire grid-cell is used to compute the CH₄ flux density (see Melton et al., 2013; Ringeval et al., 2013).

~~W~~The advanced and more explicit treatment of floodplains also introduces additional uncertainties about hydrology, vegetation and associated CH₄ emissions into the model. However, we estimated the CH₄ emissions' sensitivity to different processes and could identify which processes are critical for a successful bottom-up estimate of CH₄ emissions from tropical wetlands. Based on this, the following recommendations can be made for further model improvement. Importantly, in the LPX model environment, it is difficult to account for seasonality in floodplain extent. Due to yearly constant LU extent in the LPX model, our approach was to transfer a part of the horizontal seasonality into vertical (i.e., water depth) seasonality. However, this reverses the influences of water availability on CH₄ emissions, in the sense that increases in floodplain extent are expected to increase the CH₄ emissions, whereas the LPX simulated increase in water depth reduces the emissions. A pragmatic solution to this problem would be to account for more than one floodplain LU in LPX. Indeed, introducing a large number of floodplains LUs which can be successively flooded during the year would allow to

1190 ~~mimicking~~ the seasonality in floodplain extent. As a first attempt, introducing one floodplain LU which is flooded during all the year and one floodplain LU with a dry season would already allow to improve the representation of seasonality. This solution would allow to account for the effect of seasonality in floodplain hydrology on both vegetation dynamic and CH₄ emissions.

Besides, at Amazon basin scale, the year-to-year variability in CH₄ emissions from LPX is lower than

1200 most of the WETCHIMP models. While the real magnitude of IAV is unknown as indicated in previous section, LPX-simulated IAV may be enhanced by amplifying the effect of IAV in floodplain extent on the IAV of CH₄ flux densities. This could be done, for example, by modifying the soil carbon

decomposition in floodplains (through the R_{moist} parameter). Under steady state in floodplain extent, heterotrophic respiration \sim NPP and thus the CH₄ emissions is only weakly affected by the value of R_{moist} . However, R_{moist} could be more important when year-to-year variability in floodplain extent is accounted for. e.g., an increase in floodplain extent tends to inject non-floodplain soil carbon into the floodplain carbon pool. The decomposition of such obtained soil carbon to CH₄ is sensitive to R_{moist} . More work is required to estimate the sensitivity of CH₄ emissions to R_{moist} .

1210 A second issue relates to the representation of floodplain vegetation in LPX. We experienced difficulties to simulate floodplain ecosystems with both i) a vegetation cover consistent with GLC2000 and ii) a NPP consistent with MODIS. In LPX, a reduction in tree productivity is required to allow grasses to compete with trees. Following the introduced parameterization, a bimodal behavior is obtained (grid-cell either entirely covered by forest or by grassland) except in the case where the amplitude of the seasonal cycle in flooding depth is increased (Equation 2). A modification of the artificial 5% threshold for maximum tree cover in the floodplain LU as a function of flooded conditions in the previous year may improve the model. Trait-based approaches incorporating more plant strategy components (Van Bodegom et al., 2012) might also help. Also, a more realistic and mechanistic representation of flooding stress, e.g. by separating the effects of partial submergence and soil oxygen demand on plant physiology (van Bodegom et al., 2008), may be considered. Such representations do,

1220 however, require experimental data on adaptation to these stresses representative for tropical PFTs and require accounting for~~Other limitations of our approach include the absence of i) macrophytes and ii)~~ seasonality in vegetation submergence. Through the big leaf approach for grasses commonly used in GVMs, LPX cannot provide any information about the length of grass shoots in the course of the year. ~~and t~~This prevents to varying WTP_{max} in time and an alternative expression needs to be found. Also, a ~~B~~better representation of plant adaptation to flooded conditions is ~~also~~ required. For instance, to account for the fact that some tree species can adapt to both flooded and non-flooded conditions, a corresponding between flood-tolerant and non-flood tolerant trees of the two LUs could be tested (e.g., for flood-tolerant TrBR, setting in Equation 4, $L = \{\text{flood-tolerant TrBR, TrBR}\}$). ~~Finally, t~~The introduction of (nitrogen and) phosphorus limitations (Goll et al., 2012) could help to simulate the fertility gradient in Amazon basin and to improve the representation of the difference in NPP and (through supply of carbon) CH₄ emissions between varzeas and igapos. Also indirectly, nitrogen and phosphorus dynamics may impact CH₄ emissions. The relative availability of nutrient determine litter decomposition and carbon mineralization and therewith methane emissions (Pancotto et al. 2010). Moreover, denitrification and phosphorus-affected soil chemistry will impact soil redox potential and finally nitrogen limitations are known to affect methane oxidation (Bodelier et al. 2000). Given the ubiquitous impacts of nutrient cycling on processes leading to CH₄ emissions, incorporating of these cycles is far from straightforward.

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1240 Finally, the difference in model performance found between the Negro River basin and the Amazon main stem suggests that the implementation of macrophytes and lateral water flow could be important. Our sensitivity tests showed that the direct effect of macrophytes properties on CH₄ emissions (plant-mediated transport, exudates) is likely small. However, differences in productivity and specificities in phenology could have an indirect effect on the CH₄ emissions and this remains to be tested. Lateral fluxes of water containing high O₂ concentration may also play a role. The representation of lateral flux between the grid-cells is problematic in LSMs and we recommend first to implement a parameterization that represents higher O₂ concentrations for floodplains close to the main

channel and characterized by many pulses of flooding per year.

~~LPX-simulated CH₄ flux densities are in reasonable agreement with observations at field scale despite slight overestimation for some sites. However, variations between sites and in time at the same site were not well modelled. On the other hand, field observations are too scarce to constrain the different transport pathways or the difference of emissions between grass-covered and forest-covered plots. Also, at Amazon basin scale, the year-to-year variability in CH₄ emissions from LPX is lower than most of the WETCHIMP models. LPX-simulated IAV could be increased by amplifying the effect of IAV in floodplain extent on the IAV of CH₄ flux densities. This could be done, for example, by modifying the soil carbon decomposition in floodplains (through the R_{moist} parameter). Under steady state in floodplain extent, heterotrophic respiration \sim NPP and thus the CH₄ emissions is only weakly affected by the value of R_{moist} . However, R_{moist} could be more important when year-to-year variability in floodplain extent is accounted for. e.g., an increase in floodplain extent tends to inject non-floodplain soil carbon into the floodplain carbon pool. The decomposition of such obtained soil carbon to CH₄ is sensitive to R_{moist} . More work is required to estimate the sensitivity of CH₄ emissions to R_{moist} .~~

4.3 Extrapolation to other tropical wetland ecosystems

Finally, we investigated if our LPX version could be used to simulate CH₄ emissions of open-water bodies in the Amazon basin. Fig. A8 shows a comparison between the LPX simulation and measurements made on plots not covered by vegetation at Sites 4 and 5. Figure A8 also displays comparison on Supp. Site 1 (Lago Colado; [Crill et al., 1988](#); [Engle and Melack, 2000](#)). For sites 4 and 5, measurements show that average fluxes from plots covered by water (Fig. A8) are significantly lower than those from floating mats and flooded forests (Fig. 6). This difference is bigger for Site 5 than for Site 4. It is difficult to know if the measured plots correspond to vegetated plots which were only punctually covered by water after a rise in the flood height or if they were permanently inundated. Overall, the order of magnitude of measured fluxes in Lago Colado is similar than the one of Sites 4 and 5 and measured CH₄ flux densities are relatively constant for plots covered by water in all sites. LPX simulated CH₄ flux densities are generally much higher and only coincide with measurements in conditions where the water had been mixed: due to the passage of a cold front in September 1987 and the resulting modification of the stratified conditions, a large increase in CH₄ flux densities had occurred (Crill et al., 1988). We thus conclude that LPX is not required to simulate open-water emissions and we propose to estimate CH₄ emissions related to permanent open-water bodies in Amazon basin by using a constant CH₄ flux density.

We also investigated whether our LPX version can be used to simulate CH₄ emissions of floodplains outside of the Amazon basin. Supp. Sites 2 (Pantanal, [Marani and Alvala, 2007](#)) and 3 (Panama, [Keller et al., 1990](#)) as well as Supp. Geographical area n°1 (Orinoco floodplain, [Smith et al., 2000](#)) allow to compare LPX CH₄ flux densities with measurements made outside of the Amazon basin (Fig. 5, 6 and 7). While measurements and LPX are consistent at site scale, simulations overestimate CH₄ flux densities representative to large area in the Orinoco basin. Intensive comparison to measurements outside of the South America is required to assess if ~~our modifications~~the model could be used for all tropical floodplains.

Conclusions

~~In this study, we presented the first regional-scale process-based model dedicated to tropical floodplains. We did this in the framework of the LPX-Bern GVMD. While our model includes more-~~

information about tropical floodplains than before, we were unable to reduce the uncertainty in bottom-up estimates of the magnitude of wetland CH₄ emissions of the Amazon basin, compared to WETCHIMP models. In particular, the uncertainties related to floodplain extent have a large effect on the simulated emissions. Our study shows, however, that the contribution of Amazon floodplains to the Amazon CH₄ budget is potentially very large and thus, no new sources mechanisms (as found by Vigano et al. (2008) and Covey et al. (2012)) are necessarily required. Our results stress the importance of the Amazon basin and urge for more research help narrow the uncertainties in CH₄ emissions. The representation of the specific mechanisms involved as seasonally varying wetland extents has to be improved.

We identified important difficulties to constrain some key variables (floodplain extent, vegetation cover, NPP) by observations. These limitations arise from a lack of field observations and from the uncertainty in global remote sensing datasets. In particular, measurements of CH₄ flux densities provide poor constraints for the simulation and this prevents one to draw quantitative conclusions. Eddy Covariance measurements in the Amazon basin could provide information that is more compatible with spatial resolution of GVMsD.

Given the large variability among the WETCHIMP models, an interesting question is what the size of the seasonal cycle and the year-to-year variability should be. This may be assessed through the use of a chemical transport model in combination with aircraft measurements. Further constraints on the current magnitude of floodplain CH₄ emissions and their sensitivity to climate are still required to better understand the role of such emissions in variations of CH₄ atmospheric concentration in the past (Singarayer et al., 2011) and their potential effect in the future (Stocker et al., 2013).

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Appendix 1

The four criteria applied to select grid-cells in computation of the mean LPX CH₄ flux density over large geographical areas (Figure 7) are :

i : a grid-cell is selected as a 'flooded grass' (or 'flooded forest') ecosystem if the FPC of the corresponding vegetation is larger than 25 % of the floodplain LU area.

1330 ii : a grid-cell is selected as 'flooded grass' (respectively 'flooded forest') ecosystem if the flooded grass NPP is larger than the flooded forest NPP (resp. lower). This allows to account for some seasonality in the phenology (while the FPC is constant at yearly time-scale)

iii: Same as i but keep only grid-cells where flooding depth > 0.1m

iv: Same as i but introduce a floodplain extent area weighting (allow to account for increased probability of sample for large floodplain extent if we assume the measurements are randomly made)

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