Future changes in precipitation and impacts on extreme streamflow over Amazonian sub-basins

M Guimberteau, J Ronchail, J C Espinoza, M Lengaigne, B Sultan, J Polcher, G Drapeau, J-L Guyot, A Duchateau, and P Ciais

1 Laboratoire d'Océanographie et du Climat: expérimentations et approches numériques (LOCEAN), UMR7159, Paris, France
2 Institut Pierre Simon Laplace (IPSL), Paris, France
3 Université Paris Diderot, Sorbonne Paris Cité, Paris, France
4 Instituto Geofisico del Perú, Lima, Peru
5 Universidad Agraria La Molina, Lima, Peru
6 Institut de Recherche pour le Développement (IRD), Paris, France
7 Laboratoire de Météorologie Dynamique (LMD), Paris, France
8 Centre National de la Recherche Scientifique (CNRS), Paris, France
9 Pôle de Recherche pour l’Organisation et la Diffusion de l’Information Géographique (PRODIG), Paris, France
10 Institut de Recherche pour le Développement (IRD), Brasilia, Brazil
11 Unité Mixte de Recherche (UMR) Sisyphé, Paris, France
12 Université Pierre et Marie Curie (UPMC), Paris, France
13 Laboratoire des Sciences du Climat et de l’Environnement (LSCE), Gif-sur-Yvette, France
14 Joint Unit of Commissariat à l’énergie atomique et aux énergies alternatives (CEA), Gif-sur-Yvette, France
15 Université de Versailles Saint-Quentin (UVSQ), Versailles, France

E-mail: matthieu.guimberteau@upmc.fr

Received 21 December 2012
Accepted for publication 14 February 2013
Published 7 March 2013
Online at stacks.iop.org/ERL/8/014035

Abstract

Because of climate change, much attention is drawn to the Amazon River basin, whose hydrology has already been strongly affected by extreme events during the past 20 years. Hydrological annual extreme variations (i.e. low/high flows) associated with precipitation (and evapotranspiration) changes are investigated over the Amazon River sub-basins using the land surface model ORCHIDEE and a multimodel approach. Climate change scenarios from up to eight AR4 Global Climate Models based on three emission scenarios were used to build future hydrological projections in the region, for two periods of the 21st century. For the middle of the century under the SRESA1B scenario, no change is found in high flow on the main stem of the Amazon River (Óbidos station), but a systematic discharge decrease is simulated during the recession period, leading to a 10% low-flow decrease. Contrasting discharge variations are pointed out depending on the location in the basin. In the western upper part of the basin, which undergoes an annual persistent increase in precipitation, high flow shows a 7% relative increase for the middle of the 21st century and the signal is enhanced for the end of the century (12%). By contrast, simulated precipitation decreases during the dry seasons over the southern, eastern and northern parts of the basin lead to significant low-flow decrease at several stations, especially in the Xingu River, where it reaches −50%, associated with a 9%
reduction in the runoff coefficient. A 18% high-flow decrease is also found in this river. In the north, the low-flow decrease becomes higher toward the east: a 55% significant decrease in the eastern Branco River is associated with a 13% reduction in the runoff coefficient. The estimation of the streamflow elasticity to precipitation indicates that southern sub-basins (except for the mountainous Beni River), that have low runoff coefficients, will become more responsive to precipitation change (with a 5 to near 35% increase in elasticity) than the western sub-basins, experiencing high runoff coefficient and no change in streamflow elasticity to precipitation. These projections raise important issues for populations living near the rivers whose activity is regulated by the present annual cycle of waters. The question of their adaptability has already arisen.

**Keywords:** Amazon, ORCHIDEE, streamflow extreme, climate change, precipitation

Online supplementary data available from stacks.iop.org/ERL/8/014035/mmedia

1. Introduction


Despite the uncertainties of future climate projections, related to CO2 emission scenarios and to General and Regional Circulation Models (GCMs and RCMs respectively) (Li et al 2006, Vera et al 2006, Boulanger et al 2007, Marengo 2009), some consensus is found regarding climate change in the Amazon River basin. Indeed, a 4 °C warming is expected in tropical South America (Boulanger et al 2006) and a consensus is found about a precipitation decrease in austral winter (Vera et al 2006), a longer dry season and more severe droughts (Li et al 2006, 2008). Moreover, annual precipitation is expected to decrease in the eastern Amazon River basin and, in contrast, to increase and be more intense over the western Amazon (Meehl et al 2007, Alves and Marengo 2010).

Past studies on climate change and its impact on hydrology at global scale do not show any robust agreement about future runoff trends in the Amazon River basin. Some authors simulate a discharge reduction on the main stem of the Amazon River (Arora and Boer 2001) while others find a 5% increase in discharge at the end of the 21th century (Nohara et al 2006). Salati et al (2009), using 15 AR4 (4th Assessment Report) coupled models and the RCM HadRM3P (Hadley Centre Regional Model version 3P), project a 7–30% discharge decrease on the main stem of the Amazon River depending on the scenario (SRESB2 and SRESA2), the climate model and the period. At the regional scale, literature on future runoff projections in the Amazon River basin remains scarce because of hydrological modeling issues (Coe et al 2007, Decharme et al 2008, Beighley et al 2009, Trigg et al 2009, Getirana et al 2010, Paiva et al 2011a, Yamashima et al 2011, Guimberteau et al 2012, Paiva et al 2012). Milly et al (2005), using ensembles from 12 GCMs for the middle of the 21st century under the SRESA1B scenario, obtained a 5–10% runoff decrease in northeastern Amazon, contrasted to a 10–40% increase in western Amazon. Along the eastern edge of the Amazon River basin, Tomasella et al (2009) found a 30% (60%) mean annual (low-flow) decrease of the Rio Tocantins discharge at the end of the 21th century under the SRESA1B scenario. In the upper Amazon, Lavado Casimiro et al (2011) use two hydrological models, the climatic data from 3 GCMs under 2 SRES scenarios, and found both decreasing (4 basins) and increasing (3 basins) discharge trends in the Peruvian Amazon Andes basins. As a consequence of precipitation and runoff change, flooded areas and flood duration are found to increase in about one third of the basin, especially in Western Amazonia (Langerwisch et al 2012).

The above differences in future extreme flow estimations (low and high flows) for the Amazon River are due to the climatic forcing, model uncertainties but also to the lack of simulations at sub-basin scale. Will annual extreme flows be more severe during the 21st century than the present ones? Will there be differences within sub-basins? To try to answer these questions, our study focuses on sub-basin
scale annual extreme values in the context of a changing climate. This was permitted by the availability of the Land Surface Model (LSM) ORCHIDEE (ORGanising Carbon and Hydrology In Dynamic EcosystEms, section 2.1) which is able to accurately simulate the present-time streamflow in many stations over the basin (Guimberteau et al 2012). We build plausible scenarios of hydrological conditions that populations and ecosystems might have to cope with during the 21st century, based on a multi-scenarios approach from the IPCC’s (Intergovernmental Panel on Climate Change) AR4 (Solomon et al 2007). Our methodology is first described (section 2). We detail the method based on anomalies to build the future climate forcings (section 2.2) and the hydrological signatures chosen to analyze the relationship between precipitation and streamflow (section 2.3). The hydrological signature results for present time are briefly given in section 3. The results from future simulations under SRESA1B at the middle of the 21st century are presented in section 4 and organized according to the different sub-basins. Section 5 summarizes the results obtained at the end of the century and under other scenarios.

2. Methodology

2.1. The land surface model ORCHIDEE

Hydrological simulations under climate change are performed using the hydrological module SECHIBA (Schématisation des EChanges Hydriques à l’Interface Biosphère–Atmosphère, Ducoudré et al 1993) of the LSM ORCHIDEE considering a 11-layer hydrology (De Rosnay et al 2002, D’Orgeval et al 2008, Campoy et al 2013) where a 2 m-soil is vertically discretized to calculate unsaturated soil water fluxes. The routing module (Polcher 2003, Guimberteau et al 2012) is activated in the model to simulate the daily transport of runoff and drainage to the ocean. Flooded areas are taken into account in ORCHIDEE using the representation of floodplains and swamps by D’Orgeval (2006) and a map of their spatial distribution over the Amazon River basin built by Guimberteau et al (2012). A detailed description of the LSM ORCHIDEE is given in the latter study.

2.2. Construction of climate change forcings

GCMs are known to poorly simulate precipitation patterns, even more at sub-basin scale (Solomon et al 2007). Moreover, because of the importance of local precipitation for hydrology, downscaling methods have been developed for hydrological studies and lead to improved results at basin scale (Wilby et al 1999). We applied the delta downscaling method approach to produce climate change forcings from GCMs results. This method is based on the addition of the monthly anomalies between two climatologies (simulated climate change and current climate results from a given GCM) to a baseline meteorological forcing. It appears to be a downscaling method well adapted for hydrological studies (Fowler et al 2007). In fact, if we assume that relationships between variables in the present-time climate are likely to be maintained in the future, the delta method incorporates from the present-time forcing more realistic spatio-temporal patterns in precipitation which are crucial for future streamflow simulation. However, this method prevents the study of future interannual variability as future monthly anomalies of climatologies are applied to the present-time forcing whose variability is maintained.

Climate change anomalies are derived from up to 8 AR4 GCMs (see table s1 available at stacks.iop.org/ERL/8/014035/mmedia), 2 different 20 yr running mean periods (2046–2065 and 2079–2098) and 3 emission scenarios (SRESB1, A1B and A2). We mainly focus on the results coming from SRESA1B, which projects a very rapid economic growth, a low population growth, a rapid introduction of new and more efficient technology, leading to a 720 ppm CO2 concentration by the year 2100. This choice was driven by the better availability of the 8 atmospheric variables required to run ORCHIDEE (see table s2 available at stacks.iop.org/ERL/8/014035/mmedia) under SRESA1B scenario compared to SRESA2. The global meteorological dataset NCC (NCEP/NCAR Corrected by CRU data, Ngo-Duc et al 2005) is used as the baseline for this study. The 21 yr period representing the current climate in NCC dataset is the period 1980–2000, during which the precipitation dataset was corrected by ORE (Environmental Research Observatory) HYBAM (Geodynamical, hydrological and biogeochemical control of erosion/alteration and material transport in the Amazon River basin) (Cochonneau et al 2006) observations that have improved the present-time streamflow simulation (called ‘ORCH4’) by ORCHIDEE (Guimberteau et al 2012).

In each GCM grid-cell $i$, monthly anomalies ($\text{ANO}(X_{\text{GCM},i,t})$) have been calculated from climatologies of the atmospheric variables outputs derived from each GCM ($X_{\text{GCM},i,t}$) (figure 1 and equation (1)). Linear spatial interpolation has been applied to the GCM outputs whose resolution was coarser than NCC (1.0°). The resulting monthly anomalies of a given variable of a GCM were applied to the corresponding NCC variable ($X_{\text{NCC},i,t}$) at each NCC time-step ($t_{\text{NCC}} = 6$ h) for the corresponding month ($t_{\text{GCM}} = 1$ month) of each year of the 1980–2000 period (figure 1 and equation (2)). Thus, simulations are performed with ORCHIDEE forced by these new climate change forcings starting from the same state of equilibrium of the model (5 yr spin-up over the 1975–1979 period).

$$\text{ANO}(X_{\text{GCM},i,t}) = \left( \frac{X_{\text{SRES}a,i} - X_{\text{20C3M},i}}{X_{\text{20C3M},i}} \right)_{\text{GCM},i,t}$$

(1)

$$X_{\text{NCC},i,t} = X_{\text{NCC},i,t} + (\text{ANO}(X_{\text{GCM},i,t}) \ast X_{\text{NCC},i,t}).$$

(2)

2.3. Hydrological signatures

The variations of three hydrological signatures (runoff coefficient, precipitation elasticity of streamflow and first/last deciles of streamflow) are studied at sub-basin scale using 16 ORE HYBAM gauge stations (table 1) to characterize the
hydrological response to future precipitation change. In the stations where it accounts, the backwater effect (Meade et al 1991) that is not represented in ORCHIDEE, is eliminated in the measurements (Espinoza et al 2009a). For each signature, differences are performed between one simulated year in climate change condition and the baseline ORCH4 climatology. Thus we estimate for each year the signature variation between future simulation and average present-time condition. The results are presented from an ensemble which is an average of 168 differences (8 models ∗ 21 yearly differences) between future and present conditions.

2.3.1. Annual runoff coefficient. The annual average runoff coefficient ($R_{\text{coeff}}$, equation (3)) varying from 0 to 1, is the proportion of the long-term average precipitation ($\bar{P}$ in kg m$^{-2}$ s$^{-1}$) that runs off into streams (long-term average discharge $\bar{Q}$ in kg m$^{-2}$ s$^{-1}$), assuming no net change in storage. A high (low) runoff coefficient points out that a large amount of water leaves the basin as streamflow (as ET).

$$R_{\text{coeff}} = \frac{\bar{Q}}{\bar{P}}.$$ (3)

Three runoff coefficients are computed for each sub-basin:

- $R_{\text{OBS}}$ is computed from ORE HYBAM discharge data (except for the river mouth discharge whose value is from Callede et al (2010), for the 1972–2003 period average) and from the precipitation dataset used to correct NCC forcing (see section 2.2), on average over the 1980–2000 period.
2.3.2. Precipitation elasticity of streamflow. In order to quantify the response of discharge to the precipitation change, we compute the precipitation elasticity ($\varepsilon_p$, equation (4)) which is the relative change in mean annual streamflow divided by the relative change in mean precipitation (Schaaeke 1990, Sankarasubramanian et al. 2001). The streamflow elasticity can vary usually between 1.0 and 4.0, i.e. a 10% change in mean annual precipitation results in a 10–40% change in mean annual runoff. The precipitation elasticity of streamflow is strongly correlated to the runoff coefficient, the mean annual precipitation and the streamflow according to Chiew (2006). Thus, river basins with low (high) runoff coefficient will be more (less) sensitive to precipitation change.

$$\varepsilon_p = \text{median} \left( \frac{Q_y - Q_s}{(P_y - P_s) \cdot \bar{Q}} \right)$$

($P_y$ and $Q_y$ (both in kg m$^{-2}$ s$^{-1}$) are the annual precipitation and discharge respectively, averaged over the year $y$ of the time period.

The potential evaporation elasticity of streamflow is computed in the same way according to equation (4).

2.3.3. First and last deciles of streamflow. The deciles are used to distinguish the monthly low flow (i.e. first decile) and monthly high flow (i.e. last decile) for a given sub-basin. The first (last) decile of streamflow is the value not exceeded by the lowest (highest) 10% of all streamflow values. They are calculated for each sub-basin, each year of future climate forcing and scenario, for present and future time horizons. For each future simulation, the significance of the median change of the first and last deciles is assessed using the Wilcoxon signed-rank test at the 95% level.

3. Brief overview of simulated streamflow related to precipitation in present time

The ability of ORCHIDEE to simulate the streamflow variations over the end of the 20th century was thoroughly tested over the Amazonian sub-basins by Guimberteau et al. (2012). The streamflow simulation has been improved using the new ORE HYBAM precipitation forcing which better takes into account the precipitation regimes. They are very different according to the regions of the Amazon River basin due to its huge size and its extension in both hemispheres. We can distinguish different seasons of precipitation according to the regions (Figueroa and Nobre 1990, Nobre et al. 1991, Ronchail et al. 2002, Espinoza et al. 2009a). From southeastern to southwestern regions, the precipitation season occurs in DJF (December, January and February) and the dry season in JJA (June, July and August). The precipitation seasonality in western regions is similar but less pronounced. The rainiest and driest seasons in northwestern regions are slightly more marked; they occur in MAM (March, April and May) and DJF respectively. The precipitation regime slightly differs in the northermmost and the northeastern regions where the rainiest seasons occur in MJJ (May, June and July) and MAM, respectively.

The hydrological signatures describing the link between precipitation and runoff (runoff coefficient and precipitation elasticity) in present time are briefly analyzed in sections 3.1 and 3.2, then used to assess future streamflow variations in a climate change perspective (section 4). In the latter section and section 5, we focus on the future change of the annual extreme values of streamflow using the hydrological signature of streamflow deciles.

3.1. Annual runoff coefficient

The runoff coefficient observed at the mouth ($R_{\text{OBS}}^{\text{Coeff}} = 0.49$) (table 1) indicates that about half of the precipitation reaching the Amazon River basin runs off to the mouth. It is in good agreement with Callede et al. (2008) who found values between 0.41 and 0.59 for the 1940–2003 period, with Shuttleworth (1988) who evaluated that one half of the incoming precipitation is returned to the atmosphere as ET above the Amazonian forest, and with Marengo’s (2006) compilation of results indicating that no more than 46% of precipitation leaves the basin as runoff. The high $R_{\text{OBS}}^{\text{Coeff}}$ values in the stations at the outlet of the Andean basins show the influence of the high precipitation values (hotspots) occurring in these regions and the topography that induces a fast runoff generation from soil layers toward the exit of the Andes. This is observed at Rurrenabaque (0.68) and Tamshiyacu where $R_{\text{OBS}}^{\text{Coeff}} = 0.81$ is close to the value given by Espinoza (2009) (0.79). On the contrary, southern runoff coefficients varying from 0.28 (Guajara-Mirim) to 0.39 (Itaituba) are very low in regions retaining large amount of water in floodplains. Southern $R_{\text{OBS}}^{\text{Coeff}}$ is in good agreement with Molinier (1992), Molinier et al’s (1995) estimations which give 0.36, 0.29 and 0.39 at Fazenda Vista Alegre, Altamira and Itaituba, respectively. Northern runoff coefficients generally exceed 0.5, with exception of the northernmost and tropical basins (Caracaraí and Sao Francisco basins). They are overestimated compared to Molinier (1992) and Molinier et al (1995) (about 0.50 at Serrinha for the 1973–1989 period).

$R_{\text{ORCH}}^{\text{Coeff}}$ (table 1) sums up the streamflow results obtained in Guimberteau et al. (2012), where the northern and western streamflow simulation was improved by ORE HYBAM precipitation forcing and new floodplains distribution in the model. The main western basins at Tamshiyacu and Sao Paulo de Olivença are subjected to a runoff coefficient underestimation, probably due to the lack of precipitation data in the northwestern Amazon (Azaradarkhsh et al. 2011). Sub-basins from southern regions present a large overestimation in runoff coefficient maybe due to an ET underestimation and an exaggerated extension of rainy spots in the forcing.
3.2. Precipitation elasticity of streamflow

The streamflow elasticity simulated in this study for present-time climate ($\varepsilon^{ORCH4}$) varies between 0.8 (at Óbidos) to 2.15 (Labrea) over the Amazon River basin and is generally overestimated compared to the observations ($\varepsilon^{OBS}$) (table 1). The streamflow elasticity computed over the Amazon River sub-basins is of the same order of magnitude as other estimations in the world: it ranges from 1.0 to 2.5 over the US (Sankarasubramanian et al. 2001), from 2.0 to 3.5 in 219 catchments across Australia (Chiew 2006), from 1.0 to 4.0 in the Quarai River which is a tributary of the Uruguay River (part of the La Plata basin) (Paiva et al. 2011a), from 2.0 to 6.0 for different LSMs in the Colorado River (Vano et al. 2012) and from 1.0 to 3.0 at global scale according to Tang and Lettenmaier (2012).

4. How do the different Amazonian sub-basins respond to climate change (middle of the 21st century under SRESA1B scenario)?

We assess that low flow (high flow) is strongly related to the dry season (rainy season). This simple hypothesis is limited by the complex relation between precipitation and streamflow regimes within the Amazon River basin. Floodplains extension can modify the relationship between precipitation amount and extreme stream flows. However, for a small sub-basin such as the Mamore at Guaraja-Mirim, the lag between precipitation and runoff is not significant at monthly timescale. In contrast, it can be significant (about two months) over larger sub-basins including floodplains areas (Óbidos, Manacapuru, Fazenda Vista Alegre). After a brief overview of the climate change projected over the basin for the 2046–2065 time horizon under the SRESA1B scenario (section 4.1), the regional changes in extreme flows related to the seasonal precipitation changes (see section 3) are presented for three large coherent regions of the basin (see color code in table 1) in section 4.2: southeastern/southern region (SE–S, section 4.2.1), western region (W, section 4.2.2) and northern region including northwestern/northern/northeastern regions (NW–N–NE, section 4.2.3). A separate section is dedicated to Óbidos results (section 4.2.4), representing the hydrological behavior of the Amazon River basin.

4.1. Overview of projected climate change

Annual ensemble mean temperature increases by 2.04 °C (1.49 °C–2.63 °C according to the different simulations) on average over the basin. Higher annual increase occurs in southern sub-basins with the highest ensemble increase over the Mamore River basin (2.15 °C). The temperature increase leads to an enhanced ensemble mean ET by 7.0% on average over the basin. A slight precipitation increase by 1.1% is simulated over the basin. The precipitation change is spatially contrasted. At the annual timescale, few GCMs (generally 3 out of 8) simulate a precipitation increase (i.e. most GCMs project a precipitation decrease) in eastern (Xingu and Tapajos Rivers), northernmost (Branco River) and, to a lesser proportion, southern (Mamore and Beni Rivers) regions (figure 2(a)). In contrast, wetter conditions are found by a majority of GCMs (generally 7 out of 8) in western (Amazonas River at Tamshiyaacu outlet, Purus and Jurua Rivers) and northwestern regions (Japura and Negro Rivers).

4.2. Regional response of streamflow to precipitation change

4.2.1. Southeastern/southern region. The annual precipitation decrease predicted by most models, associated to an ET increase (not shown), leads to a runoff coefficient decrease by more than 5% at the six stations of the region (figure 3). The persistent south-easterly precipitation decrease from March to November predicted by most GCMs (figures 2(c)–(e)) particularly affects streamflow at Altamira where a near 10% decrease in runoff coefficient occurs (figure 3). A south-easterly decrease in dry-season precipitation (figure 2(d)) leads to more pronounced low flows at all the stations of the region (figure 4(b)). The signal is rather robust in the region because at each station, 75% of the ensemble low-flow differences between future and present are negative. The effect is significant over the Madeira (Porto Velho and Fazenda Vista Alegre stations) and the Xingu (Altamira station) rivers where JJA precipitation decreases by 9% and 22% respectively and is associated with a 5% ET increase due to temperature increase (about 2.5 °C) over these basins (not shown). Up to 7 out of 8 simulations give a significant average decrease in the median low flow by about 30% (50%) at Porto Velho (Altamira). No change in high flow is simulated at southerly stations except at Altamira where 6 future simulations out of 8 give a significant decrease by 18% in average (figure 4(a)). Southern sub-basins tend to become more responsive to precipitation. Indeed, precipitation elasticity of streamflow increases by more than 15% at Guajara-Mirim, Altamira and Itaituba (figure 5). Potential evaporation elasticity of streamflow also increases in these stations and in Porto Velho (between 5 and 19%), except at Altamira where it decreases by 66% (not shown) suggesting that the Xingu River basin becomes more (less) responsive to precipitation (evaporation) change.

4.2.2. Western region. Most GCMs project an annual precipitation increase over all the western part of the Amazon River basin (figure 2(a)). However, annual runoff coefficients are little affected at Tamshiyaacu and Sao Paulo de Olivença (less than 2% decrease), whereas they decrease by more than 5% at Gaviao and Labrea (figure 3), where the lower proportion in swamps and floodplains does not compensate for the change in precipitation contribution to runoff. The wetter projected conditions persist seasonally during the wet season (DDJ), in particular in the northwesternmost part of the basin and along the eastern slopes of the Andes, where all the models simulate a precipitation increase (figure 2(b)). High-flow variation is not significant in the four westerly stations. At Tamshiyaacu (upper Solimões), all the simulations show a median high-flow increase ranging from 5 to 25% (not shown) but only 3 simulations are significant (figure 4(a)). No significant low-flow decrease is simulated over the western
Figure 2. Number of GCMs out of eight that project a precipitation increase over the basin, for the 2046–2065 time horizon under SRESA1B scenario: (a) annual (b) DJF (c) MAM (d) JJA and (e) SON. Contour indicates the precipitation anomaly (mm d$^{-1}$) between ensemble and ORCH4 (solid line when the anomaly is positive and dashed line when negative). The localization of the ORE HYBAM gauge stations are indicated on each map with a color point (see table 1 for their coordinates and the color code).

4.2.3. Northern region. Most GCMs simulate an annual precipitation decrease in the north of the river mouth region and in the northernmost part of the basin (figure 2(a)). The tri-monthly decreases are projected with good confidence in the northeastern region except in MAM (figures 2(b)–(e)). Northerly streamflow decrease is going to be more severe from west (Acanauí) to east (Caracarai and São Francisco).
Runoff coefficient highly decreases at Sao Francisco and Caracarai (10.5 and 13% decrease respectively) while low change (less than 2%) occurs at the most westerly stations (Acauari and Serrinha) (figure 3). The same kind of severity is found regarding northerly low flows. They significantly decrease in at least six simulations out of eight at each station (figure 4(b)). Ensemble median low flows significantly decrease by 20% in the Japura (Acauari station) and the Negro (Serrinha station) rivers, and it is much higher (55%) in the Branco River (Caracarai station) (figure 4(b)). The signal is rather robust for Serrinha and Caracarai, as 75% of the ensemble low-flow differences between future and present are negative. The discharge decreases at Caracarai during the high flow and recession periods. Overall, the seasonal variations of the ensemble are less contrasted than in the ORCH4 simulation (figure 6(a)). No significant change in ensemble high flow is simulated at Acauari and Serrinha stations (figure 4(a)). Generally, the northern sub-basins at Acauari and Serrinha are not responsive to future precipitation change (figure 5) but rather to evaporation; streamflow elasticity to potential evaporation increases by 22% in the Japura River basin (Acauari station) (not shown). In contrast, the streamflow elasticity to precipitation (potential evaporation) increases (decreases) by 12% (22%) at Caracarai, suggesting that the Branco River basin becomes more (less) responsive to precipitation (evaporation) change.

4.2.4. Main stem of the Amazon River. The precipitation change is low on average over the basin, even though it is strongly spatially contrasted. At Óbidos station, the last gauged station before the mouth of the Amazon River, the runoff coefficient slightly decreases (about 4%) (figure 3) and no change in precipitation elasticity of streamflow is found (figure 5). The streamflow seasonal cycle only changes during the recession period when all the future simulations project more pronounced low flow than ORCH4 (figure 6(b)). Ensemble simulations provide a 10% median low-flow decrease (figure 4b) but only five future streamflow simulations out of eight give a significant decrease. No change in high flows is simulated at Óbidos station (figure 4(a)).

5. Hydrological extreme variations for the end of the 21st century and under other scenarios

The annual increase in mean ensemble temperature over the Amazon River basin does not depend much of the SRES scenario at the middle of the 21st century. But differences between scenarios are pointed out at the end of the century (+2.0, +3.0 and +3.8 °C for SRESB1, SRESA1B and SRESA2 respectively). The spatial contrast between an eastern decrease and a western increase in ensemble precipitation occurring during the middle of the century under SRESA1B is maintained for the end of the century. However, rainfall increase is accentuated in the west while the precipitation decline is confined in the north-east. In contrast, the eastern rainfall decrease is enhanced under SRESB1 at the end of the century. Regarding SRESA2, the eastern decrease is much weaker than under SRESA1B at the middle of the century and increased rainfall dominates at the end of the century, being much stronger in the west and in the north.

Considering the A1B scenario, the mean ensemble discharge simulations for the end of the century present a low-flow attenuation in western rivers (Madeira, Purus and Jurua) and on the main stem, when compared to the middle of the century. The most striking are observed at Gaviao and Labrea, on the Purus and Jurua Rivers, where the deficit falls to −5%, when compared to the present, instead of −30% at the middle of the century. In contrast, northwestern high-flow increase (Amazonas and Negro rivers) is enhanced (+12% at Tamshiyacu instead of +7% at the middle of the century) in agreement with the western rainfall increase at the end of the century.

No change in low flow under SRESB1 is found when compared to SRESA1B results for both time horizons. Eastern high flow decreases are slightly attenuated under SRESB1 when compared to SRESA1B at the middle and the end of the

Figure 4. Relative change (%) of the (a) last deciles (i.e. high flow) and (b) first deciles (i.e. low flow) of streamflow between ensemble and ORCH4, for each station, for the 2046–2065 time horizon under SRESA1B scenario. For each year of future-time simulation, differences were performed between the result of the future-time simulation and the climatology of the present-time simulation. The boxes correspond to the interquartile range (IQR, the distance between the 25th and the 75th percentiles), the bold horizontal line in each box is the median and the whiskers extend from the minimum value to the maximum value unless the distance from the minimum (maximum) value to the first (third) quartile is more than 1.5 times the IQR. Circles indicate the outliers that are $1.5 \times \text{IQR}$ below (above) the 25th percentile (75th percentile). The colors of the boxes indicate the localization of the station in the Amazon River basin (see table 1 for the coordinates of the stations and the color code). The numbers of future simulations out of 8 that give a significant change of the median (Wilcoxon signed-rank test at the 95% level) are indicated below each box.

In conclusion, river discharge changes and differences between scenarios are less pronounced at the middle of the century and more accentuated at its end, especially between SRESA2 and the others; indeed, all year round, a large runoff increase is observed in a large part of the western Amazon River basin.

6. Conclusion

Future climate change should substantially modify the Amazon River basin hydrology and impact the water resources availability. In this study, we use the LSM ORCHIDEE to provide discharge projections, according to future climate forcings built from several GCM projections under the emission scenarios B1, A1B and A2. Various hydrological responses are found in the different sub-basins of the Amazon under SRESA1B scenario. Some of them are not significant since climate projections from the different GCMs diverge, as already assessed in previous studies (Bates et al 2008, Kay et al 2009, Blöschl and Montanari 2010, Nobrega et al 2011, Paiva et al 2011a), but in some sub-basins and for specific periods of the year, the projections are more reliable. In particular, low flows are projected to decrease severely in most stations, especially in the southern Madeira and Xingu Rivers and in the northern Branco River, and to a lesser extent in the northern Negro and Japura Rivers. By the middle of the 21st century, the decrease is projected to reach $-50\%$ ($-55\%$) in the Xingu (Branco) Rivers. A low-flow decrease is projected in Obidos, on the main stem of the Amazon River, which has already been experiencing an increasing number of droughts over the last twenty years. In contrast, the western and upper part of the Amazon are projected to undergo an annual precipitation increase so that high-flow discharge is expected to increase by 7% in the middle of the 21st century and even more by the end of the century. Therefore,
more flooding events are expected in this region which is already affected by increasing extreme discharge. In the main stem of the Amazon River, however, there is no consistent signal of more flooding, neither of varying mean discharge, although previous studies projected a positive change (Nohara et al 2006) or a negative change (Milly et al 2005) of the Amazon runoff. By computing elasticity coefficients, we also show that the southern basins with low runoff coefficient are generally more responsive to precipitation change than the western basins with a high runoff coefficient. This result is in agreement with the findings by Chiew (2006) and Paiva et al (2011a).

Some uncertainties relative to the hydrological parameterizations of the LSM can be addressed. The present-time ET underestimation (Guimberteau et al 2012) and the infiltration parametrization are sources of uncertainties in future land water budget. Moreover, the routing module of ORCHIDEE does not represent the river dams and the backwater effect likely to affect the extreme stream flows in some tributaries of the Amazon (Meade et al 1991, Tomasella et al 2011). An advanced large-scale river routine module which represents backwater effect would represent more realistically the simulated water level in the rivers (Paiva et al 2011b, Yamazaki et al 2011, 2012). The modeling of vegetation growth with its interactive environment such as an interannual LAI variation and ET response to elevated CO₂, would lead to better realism using the STOMATE (Saclay-Toulouse-Orsay Model for the Analysis of Terrestrial Ecosystems, Viovy 1996) and LPJ (Lund-Postdam-Jena, Sitch et al 2003) modules. Other errors can result from the downscaling method. First, we assume that climate change varies only over large areas (i.e. as large as a GCM cell area) what leads to underestimate landscape heterogeneity effect due to the coarse resolution of the GCM. Thus, further simulations with better resolution than one degree would test if there is a dependence of simulated Amazonian hydrological processes on the spatial resolution of the forcing, as suggested by Verant et al (2004). Secondly, precipitation interannual variations in response to climate change could not be taken into account due to the simple downscaling method adopted in this study. Some bias-correction methods which conserve the changes of mean and standard deviation of the model simulation data (Watanabe et al 2012) could be adopted to preserve interannual variations. Uncertainties remain also in the emission scenario or the magnitude of rise in mean global temperature and the choice of the GCM. Indeed, a comparison between more AR4 GCMs and a new generation of GCMs (AR5) corroborates high consensus on drying during the dry season but highlights a lower confidence in wettening over the western part of the basin (Joetzjer et al 2013). Finally, considering land surface changes and deforestation in our LSM would probably affect the simulated mean and extreme flows over the Amazon River basin, since several studies reveal that land-use change accounts for at least 50% of the reconstructed global runoff trend over the last century (Piao et al 2007, Sterling et al 2012). The important deforestation in the Amazon River basin may in turn strongly alter the climate and eventually amplify the climate change (Malhi et al 2008). This topic, as well as improvement in the model parametrization, will be addressed in future work.

Acknowledgments

This work was financially supported by the GIS (Groupement d’Intérêt Scientifique) REGYNA (REGiOnALisation des précipitations et impacts hydrolgiques et agronomiques) and the EU-FP7 AMAZALERT (Raising the alert about critical feedbacks between climate and long-term land-use change in the Amazon) projects. Simulations with ORCHIDEE were performed using computational facilities of the Institut du Développement et des Ressources en Informatique Scientifique (IDRIS, CNRS, France).
References

Alves L and Marengo J 2010 Assessment of regional seasonal predictability using the PRECIS regional climate modeling system over South America Theor. Appl. Climatol. 100 337–50
Azarderakhsh M, Rossow W, Papa F, Norouzi H and Khabibulaev R 2011 Diagnosing water variations within the Amazon basin using satellite data J. Geophys. Res. 116 D24107
Blöschl G and Montanari A 2010 Climate change impacts—throwing the dice? Hydrol. Process. 24 374–81
Chen J, Wilson C and Tapley B 2010 The 2009 exceptional Amazon flood and interannual terrestrial water storage change observed by GRACE Water Resources Res. 46 W12526
Cochonneau G et al 2006 The environmental observation and research project, ORE HYBAM, and the rivers of the Amazon basin Climate Variability and Change—Hydrological Impacts (IAHS Publication vol 308) (Wallingford: IAHS Press) pp 44–50
Cook B, Zeng N and Yoon J 2012 Will Amazonia dry out?: magnitude and causes of change from IPCC climate model projections Earth Interact. 16 1–27
Decharme B, Douville H, Prigent C, Papa F and Aires F 2008 A new river flooding scheme for global climate applications: off-line evaluation over South America J. Geophys. Res. 113 D11110
D’Orgeval T 2006 Impact of the changing climatic regime on the cycle of the eau en Afrique de l’Ouest: modélisation et incertitudes PhD Thesis Université Paris VI
Ducoudré N, Laval K and Perrier A 1993 SECHIBA, a new set of parameterizations of the hydrologic exchanges at the land atmosphere interface within the LMD atmospheric global circulation model J. Clim. 6 248–73
Figueroa S and Nobre C 1990 Precipitation distribution over central and western tropical South America Climanalise 5 (6) 36–45
Fowler H, Blenkinsop S and Tebaldi C 2007 Linking climate change modelling to impacts studies: recent advances in downscaling techniques for hydrological modelling Int. J. Climatol. 27 1547–78
Guimberteau M et al 2012 Discharge simulation in the sub-basins of the Amazon using ORCHIDEE forced by new datasets Hydrol. Earth Syst. Sci. 16 911–35
Langerwisch F, Rost S, Gerten D, Poulter B, Ramming A and Cramer W 2012 Potential effects of climate change on
Climate Change (New York: Cambridge University Press) pp 235–337
Uvo C and Graham N 1998 Seasonal runoff forecast for northern South America: a statistical model Water Resources Res. 34 3515–24
Vano J, Das T and Lettenmaier D 2012 Hydrologic sensitivities of colorado river Runoff to changes in precipitation and temperature J. Hydrometeorol. 13 932–49
Vera C, Silvestri G, Liebmann B and González P 2006 Climate change scenarios for seasonal precipitation in South America from IPCC-AR4 models Geophys. Res. Lett. 33 L13707
Verant S, Laval K, Polcher J and De Castro M 2004 Sensitivity of the continental hydrological cycle to the spatial resolution over the Iberian Peninsula J. Hydrometeorol. 5 267–85
Watanabe S, Kanae S, Seto S, Yeh P, Hirabayashi Y and Oki T 2012 Intercomparison of bias-correction methods for monthly temperature and precipitation simulated by multiple climate models J. Geophys. Res.—Atmos. 117 D23114
Yamazaki D, Kanae S, Kim H and Oki T 2011 A physically based description of floodplain inundation dynamics in a global river routing model Water Resources Res. 47 W04501
Yamazaki D, Lee H, Alsdorf D, Dutra E, Kim H, Kanae S and Oki T 2012 Analysis of the water level dynamics simulated by a global river model: a case study in the Amazon River Water Resources Res. 48 W09508
Yoon J and Zeng N 2010 An Atlantic influence on Amazon rainfall Clim. Dyn. 34 249–64