Interannual variations of river water storage from a multiple satellite approach: A case study for the Rio Negro River basin

Frédéric Frappart,1,2 Fabrice Papa,3 James S. Famiglietti,1 Catherine Prigent,4 William B. Rossow,3 and Frédérique Seyler5

Received 29 September 2007; revised 27 June 2008; accepted 25 August 2008; published 5 November 2008.

[1] Spatiotemporal variations of water volume over inundated areas located in a large river basin have been determined using combined observations from a multisatellite inundation data set, the TOPEX/POSEIDON (T/P) altimetry satellite, and in situ hydrographic stations for the water levels over rivers and floodplains. We computed maps of monthly surface water volume change over the period of common availability of T/P and the multisatellite data (1993–2000). The basin of the Negro River, the largest tributary in terms of discharge to the Amazon River, was selected as a test site. A strong seasonal signal is observed with minima in October and maxima in June. A strong interannual component is also present, particularly important during ENSO years. The surface water change was estimated to be 167 ± 39 km³ between October 1995 (low water) and June 1996 (high water). This result is consistent with previous estimates obtained for the 1995–1996 hydrological cycle over the same area using the JERS mosaic data. The surface water volume change is then compared to the total water volume change inferred from the GRACE satellite for an average annual cycle. The difference between the surface storage change and the total storage change derived from GRACE was computed to estimate the contribution of the soil moisture and groundwater to the total storage change. Our study supports the hypothesis that total water storage is almost equally partitioned between surface water and the combination of soil moisture and groundwater for the Negro River basin. The water volume changes are also evaluated using in situ discharge measurements and the GPCP precipitation product (correlation of 0.61). The results show the high potential for the new technique to provide valuable information to improve our understanding of large river basin hydrologic processes.


1. Introduction and Background

[2] Terrestrial waters represent less than 1% of the total amount of water on Earth. However, they have a crucial impact on terrestrial life and human needs, and play a major role in climate variability [Cosandey and Robinson, 2000; Perrier and Tuzet, 2005]. Among the various reservoirs in which fresh water on land is stored (e.g., ice caps, glaciers, snowpack, soil moisture and groundwater), surface waters (rivers, lakes, reservoirs, wetlands and inundated areas) play a crucial role in the global biogeochemical and hydro-
[Hamilton et al., 2002]. Water storage in these wetlands and its outflow represent a significant part of the water balance in the basin [Richey et al., 1989; Alsdorf et al., 2001].

[5] Analysis of the flow and storage of fresh water over land is thus a key issue for understanding the terrestrial branch of the global water cycle, and is now recognized to have major importance for climate research as well as for inventory and management of water resources [Bullock and Acreman, 2003]. Our current knowledge of the interseasonal and interannual variability of the land surface water cycle at the regional to global scales is still rather incomplete [Matthews, 2000].

[4] Quantifying temporal variations in water volume stored in river floodplains has many practical applications, and further, would help improve our understanding of the controlling mechanisms, inundation extent and their influence on the river water budget. For inundated areas permanently or temporarily connected to main channels, determining water volume variations is equivalent to estimating the potential water volume stored and/or released by a valley reach during flood stage. These water volume variations are an important variable for hydrodynamic modeling of river flow and the determination of river transport capacity. For inundated areas that never connect to the main channel, volume variations are essentially a function of base flow variations, inputs from the local basin and rainfall. In all cases, the inundated area is a buffer zone between the river and the upland watershed, and its water volume variation represents the flood pulse of floodplains as expressed by Junk et al. [1989]. It is also a key biogeochemical and ecological characteristic, which unfortunately, cannot be easily measured in the field [Alsdorf and Lettenmaier, 2003].

[5] Until recently, estimates of floodplain volume variations within large river basins have essentially relied on hydrological models [Coe et al., 2002; Winsemius et al., 2006, Goteti et al., 2008]. In situ gauge measurements have helped to quantify the movement of water (discharge, height) in river channels, but provide comparatively little information about the spatial dynamics, height variations and volume storage of surface water, within floodplains and wetlands. In addition, some regions are completely ungauged and the number of ground-based stations has dramatically decreased during the last decade [Alsdorf and Lettenmaier, 2003]. Lacking spatial measurements of surface water volumetric changes, hydrological models are unable to properly represent the effects of surface storage on river discharge [Alsdorf, 2003; Alsdorf et al., 2007].

[6] In recent years, remote sensing techniques have clearly shown the capability to monitor components of the water balance in large river basins [Famiglietti, 2004]. They are particularly very useful to surface water hydrology investigations [Smith, 1997; Alsdorf et al., 2007] as they provide a unique mean to continuously observe large regions and are the only alternative to the lack of in situ data in remote areas. For example, satellite altimetry has been used for systematic monitoring of water levels of large rivers, lakes and floodplains [e.g., Birkett, 1998; Mercier et al., 2002; Maheu et al., 2003]. Synthetic Aperture Radar (SAR) Interferometry [Alsdorf et al., 2000, 2001] and passive and active microwave observations [Sippel et al., 1998] also give crucial information on land surface water dynamics. The multisatellite technique from Prigent et al. [2007] now offers the first global estimates of monthly inundation extents over most of a decade (1993–2000) and at a 0.25° horizontal resolution. In addition, new space gravity missions, such as the GRACE mission, offer for the first time the possibility of directly measuring the spatio-temporal variations of total terrestrial water storage [Wahr et al., 2004; Tapley et al., 2004b; Chen et al., 2005a, 2005b; Ramillien et al., 2005; Seo et al., 2006; Lettenmaier and Famiglietti, 2006; Syed et al., 2008].

[7] Frappart et al. [2005, 2006a] estimate water storage changes using remote sensing by combining high-resolution imagery—derived inundation extents and altimetry-derived water level measurements in the Negro River and Mekong basins respectively. The technique showed encouraging results, but suffered from a lack of temporal coverage, as the SAR data from JERS are only available for two months in 1995–1996. In this paper, we propose a new approach to estimate river volume changes in a large drainage basin that combines estimates of spatial and temporal surface water extent from a multisatellite technique [Prigent et al., 2007] with water level measurements from satellite altimetry (along with measurements from in situ gauge stations). The method is developed and a case study is presented over the Negro river in the Amazon basin. The results of this study are compared to GPCP (Global Precipitation Climatology Project) precipitation estimates and total water storage derived from GRACE data for validation purposes to test the timing and relative amplitude of the seasonal variations. These results will improve our understanding of large river basin hydrologic processes and modeling.

2. Study Region

[8] The Negro River subbasin (700,000 km², Figure 1) covers 12% of the Amazon basin. It is the largest tributary to the Amazon River and ranks as the fifth largest river globally in terms of mean annual discharge [Meade et al., 1991; Molinier et al., 1995]. The Negro River joins the Solimões River to form the Amazon River downstream from Manaus, and drains an area of around 700,000 km² in Colombia (10%), Venezuela (6%), Guyana (2%) and Brazil (82%). It extends from 73.25° to 59.35° longitude west and from 5.4° north to 3.35° latitude south. Water coloration is typically dark, owing to the high content of dissolved organic matter and a low sediment load [Sternberg, 1975]. It is a low-gradient river, which partly accounts for the considerable extent of the floodplains (along with the large amount of precipitation). Rainfall in the subbasin varies greatly both in space and time. Mean annual precipitation rates vary by more than 50% within the Negro river basin, from less than 2000 mm/a (minimum values less than 1700 mm/a are recorded in the northern part of the Branco River basin), to around 2500 mm/a near Manaus and up to 3000 mm/a in the northwest [Liebmann and Marengo, 2001]. The timing of the rainy season differs widely along a south to north gradient: the beginning of the rainy season occurs in December in the south and in March or April in...
the north, whereas the rainy period ends from May to October [Marengo et al., 2001].

3. Data Sets

3.1. Multisatellite Inundation Data Set

[9] The methodology developed to quantify the extent and seasonality of land surface inundation at the global scale with a suite of satellites is described in detail by Prigent et al. [2001a, 2001b, 2007]. Global monthly mean maps of inundation extent are created with a 0.25° × 0.25° spatial resolution at the equator. The technique is globally applicable without any tuning for individual environments [Prigent et al., 2007]. Case studies for specific regions and environments are presented for India by Papa et al. [2006], for the Ob River basin by Papa et al. [2007] and for the

Figure 1. Overview map of South America with the location of the Negro River basin (red square). Negro River basin extends from 73.25°W to 59.35°W and from 5.4°N to 3.35°S. Map of the Negro River subbasin extracted from JERS-1 radar mosaic. Each thin white line delineates a TOPEX/POSEIDON track. Black crosses in a white circle represent in situ gauge stations, and black dots in a white circle represent altimetric stations over the Negro River subbasin.
large Siberian watersheds by Papa et al. [2008]. This data set has been also recently used for hydrologic and climatic analyses, such as the evaluation of the methane surface emissions models [Bousquet et al., 2006] or the validation of river flooding scheme performances in land surface models [Decharme et al., 2008].

3.2. TOPEX/POSEIDON (T/P) Derived Water Levels

[10] The T/P radar altimeter is the first dual frequency sensor of a joint French and U.S. mission, whose goal is to substantially improve our understanding of global ocean dynamics by making accurate measurements of the ocean surface topography [Fu and Cazenave, 2001]. It operates in Ku and C bands, at frequencies (wavelengths) of 13.6 GHz (2.3 cm) and 5.3 GHz (5.8 cm), respectively. Since its October 1992 launch, T/P has provided along-track nadir measurements of Earth surface elevation (ocean and continental surfaces) between 66° latitude north and 66° latitude south, with a 10-day repeat cycle. The equatorial cross-track separation is 315 km and the along-track spatial resolution is 560 m for the high rates measurements (AVISO, 1996). The water levels are given by the difference between the satellite orbit information and the range or altimetric height [Fu and Cazenave, 2001].

[11] We use the Geophysical Data Records (standard ocean data, AVISO database) which are commonly adopted for monitoring water levels over rivers and floodplains [e.g., Birkett, 1998; Birkett et al., 2002; Maheu et al., 2003]. The altimetry data have been corrected for the typical geophysical and environmental corrections required over land [e.g., Frappart et al., 2006b]. The accuracy of T/P derived water levels in the Negro River basin is discussed in section IV A.

3.3. In Situ Water Level Time Series

[12] The Brazilian Water Agency (Agencia Nacional de Aguas or ANA) is in charge of managing a network of 571 gauging stations in the Brazilian part of the Amazon basin (http://www.ana.gov.br). At each station, daily measurements of water stage are collected, and daily estimates of discharge are produced using rating curves, obtained from periodic (sometimes several times a year) simultaneous measurement of stage and discharge. Among these 571 inventoried gauges, 46 are located in the Negro River subbasin and 25 of them have records over the last 20 years. From these 25 in situ gauge stations, only 8 are leveled and thus can be used for this study (Figure 1). We also differentiated daily river discharge observations from Jatuarana (first in situ station downstream from the confluence of the Solimões and Negro Rivers) and Manacapuru (outlet of the Solimões River) to estimate the monthly river discharge in Manaus.

3.4. GRACE-Derived Land Water Solutions

[13] The GRACE mission, launched in March 2002, is devoted to measuring spatiotemporal changes in Earth’s gravity field that results mainly from water mass redistribution among the surface fluid envelopes [Tapley et al., 2004b]. Several recent studies have shown that GRACE data over the continents provide important new information on the total land water storage (surface waters, soil moisture and groundwater, and where appropriate on snow mass) [Tapley et al., 2004a; Wahr et al., 2004; Chen et al., 2005a, 2005b; Ramillien et al., 2005; Schmidt et al., 2006; Yeh et al., 2006; Rodell et al., 2007].

[14] We use the land water and snow solutions derived from the inversion of 35 GRACE geoids from the third data release by GeoForschungZentrum (GFZ-RL03), as presented by Ramillien et al. [2005, 2006]. These solutions range from February 2003 to February 2006, with a few missing months (June 2003 and January 2004) and a spatial resolution of 400 km. Studies made with previous and less accurate releases of GRACE products estimated the error on the land water storage to be 18 mm for 750 km spatial average GRACE-based land water solutions [Wahr et al., 2004] and Ramillien et al. [2005, 2006] found an error 15 mm as the final a posteriori uncertainty on the land water solutions, with spatial resolution of 660 km.

3.5. Precipitation From GPCP

[15] The Global Precipitation Climatology Project (GPCP), established in 1986 by the World Climate Research Program, provides data that quantify the distribution of precipitation over the globe [Adler et al., 2003]. We use here the Satellite-Gauge Combined Precipitation Data product of GPCP Version 2 data for evaluating our estimates of monthly surface water volume variations in the Negro River basin. The GPCP products we used are monthly means with a spatial resolution of 2.5° in latitude and longitude; and are available from January 1979 to present. Over land surfaces, the uncertainty in the rate estimates from GPCP is generally lower than over the oceans owing to the in situ gauge input (in addition to satellite) from the GPCC (Global Precipitation Climatology Center). Over land, validation experiments have been conducted in a variety of location worldwide and suggest that while there are known problems in regions of persistent convective precipitation, nonprecipitating cirrus or regions of complex terrain, the estimate uncertainties range between 10 and 30% [Adler et al., 2003].

4. Methods

4.1. Water Level Time Series Derived From T/P Altimetry

[16] Along the satellite tracks shown in Figure 1, water level time series from T/P altimetry data are estimated at 86 altimetry stations. These altimetry stations are identified in Figure 1 by black dots in white circles. The selected altimetry stations selected provide good quality water level time series (according to data editing and error bars associated with each time series; see below) with limited data gaps in the time series. Most stations are located over the floodplains but other stations correspond to intersections of the satellite tracks with the river.

[17] To construct a water level time series, we consider all 10 Hz altimetry data along a portion of satellite track. The intersections between satellite tracks and rivers or floodplains were determined using the flooded areas identified by a classification of the JERS-1 dual mosaics [Frappart et al., 2005]. Once selected, the data are expressed in terms of water height (water level) above the geoid: for this purpose, the GRACE geoid GGM02C, complete to degree and order 150, has been used [Tapley et al., 2005]. For each intersection between the river (or the floodplain) and the satellite ground track, we define a so-called “altimetry station,”
represented by a rectangular window. Outliers are deleted using a 3-σ criterion over the whole time span of analysis. For each 10 day cycle, the water level at a given virtual station is obtained by computing the median of all the high-rate data (10 Hz) included in the rectangular window. This process, repeated for each cycle, allows the construction of a water level time series at the virtual station. The dispersion in L1 norm is given by the estimator known as Median Absolute Deviation:

\[
MAD(x) = \frac{1}{N-1} \sum_{i=1}^{N} |x_i - x_{\text{med}}|,
\]

where \(N\) is the number of observations, \(x_i\) is the \(i^{th}\) observation, and \(x_{\text{med}}\) is the median of the observations.

[18] The accuracy of T/P water level time series over the river and floodplains has been discussed in several previously published papers [i.e., Birkett, 1998; de Oliveira Campos et al., 2001; Birkett et al., 2002; Maheu et al., 2003]. For this study, the accuracy of the altimetry derived water levels over the Negro River was previously estimated by Frappart et al. [2005]. Minimum standard deviations of around 10 cm are measured on downstream rivers with large open water areas during high water when the pulse emitted by the altimeter is not scattered by vegetation, whereas maximum standard deviations of around 50 cm are observed on flooded areas covered with dense vegetation during low water.

### 4.2. Water Level Maps

[19] Monthly maps of water level can be estimated over the Negro River basin. As the temporal resolution of the inundation map is one month, we create monthly averages of the water levels for each altimetry or in situ station. For a given month during the flood season, water levels were linearly interpolated over the flooded zones of the Negro River basin. Each pixel of 25 km \(\times\) 25 km is considered inundated when its percentage of inundated area is greater than 0. Maps of interpolated surface water levels with 25 km resolution have been constructed for each month between January 1993 and December 2000. As mentioned above, water levels are expressed with respect to the geoid. For the purpose of hydrological interpretation of water volumes, referring to the topography would be best as far as floodplains are concerned. However, available topographic databases are not necessarily precise enough for the present study.

[20] For a given month, the water levels within flooded zones of the Negro River basin were mapped using a bilinear interpolation scheme to estimate the water level at each grid point. The upstream portion of the Negro River is oriented north/south. It flows from west to east after the confluence with the Uaupes, a west/east flowing tributary, until Manaus, the outlet of the Negro River. The Branco River, the largest tributary of the Negro River, flows north/south. Two types of cross section are observed: the satellite crosses the river and the satellite track runs along the river. As T/P cross track (or intertrack) is 315 km at the equator, large floodplain areas are not measured by the altimeter. Over the western/eastern flowing parts of the river network, the T/P tracks cross the river nearly perpendicularly, allowing clear separation of the contributions of the river mainstream from those of associated floodplains (see Figure 1). As a consequence, the interpolation in the along-track direction follows the difference of water levels between the mainstream and the floodplain. In the cross-track direction, interpolation over several tens of kilometers will only reflect the mean slope of the river.

[21] Over the northern/southern flowing parts of the river network, the T/P tracks run parallel to the river. In these cases, depending on the choice of the geographical coordinates of the virtual station, the time series can be influenced by the elevation variation within the adjacent floodplain [Birkett et al., 2002]. These authors did not report obvious amplitude or phase differences due to the inclusion of some floodplain areas. The water level variations between the mainstream and the inundated floodplain, as reported by Alsdorf [2003] using interferometric SAR observations on the Amazon floodplain (lower than 11 cm), are generally lower than the altimeter-derived water level dispersion.

### 4.3. Surface Water Volume

[22] The variation of water volume corresponds to the difference of surface water levels integrated over the inundated surface. These variations \(\delta V(t_i, t_{i+1})\) between two consecutive months numbered \(i\) and \(i - 1\), over the floodplain \(S\), are the sum of the products of the difference of surface water levels \(\delta h_j(i, i - 1)\), with \(j = 1, 2, \ldots\) inside \(S\), by the elementary surfaces \(R^2 \theta \sin(\theta) \delta \theta \lambda\) and the percentage of inundation \(P_j\):

\[
\delta V(i, i - 1) = R^2 \delta \lambda \delta \theta \sum_{j=1}^{J} P_j \delta h_j(\theta, \lambda, i, i - 1) \sin(\theta),
\]

where \(\delta \lambda\) and \(\delta \theta\) are the sampling grid steps along longitude \(\lambda\) and latitude \(\theta\) (0.22°), respectively, and \(R\) the mean radius of the Earth (6378 km). The surface and total water volume variations are expressed in km³/month.

[23] The error of the method was estimated using:

\[
d\delta V = \sum_{i=1}^{n} (dS_i \delta h_i + S_i d\delta h_i),
\]

where: \(d\delta V\) is the error on the water volume variation \((\delta V)\), \(S_i\) is the \(i^{th}\) elementary surface, \(\delta h_i\) is the \(i^{th}\) elementary water level variation between two consecutive months, \(dS_i\) is the error on the \(i^{th}\) elementary surface, and \(d\delta h_i\) is the error on the \(i^{th}\) elementary water level variation between two consecutive months.

[24] The error sources include misclassifications, T/P altimetry measurements and the linear interpolation method. The maximum error on the volume variation can be estimated as

\[
\Delta(\delta V_{\text{max}}) \leq \Delta S \max \delta h_{\max} + S_{\max} \Delta(\delta h_{\max}),
\]

where: \(\Delta(\delta V_{\text{max}})\) is the maximum error on the water volume variation \((\delta V)\), \(S_{\max}\) is the maximum flooded surface, \(\delta h_{\max}\) is the maximum water level variation between two consecutive months, \(\Delta S_{\max}\) is the maximum error for the flooded surface, and \(\Delta(\delta h_{\max})\) is the maximum
error for the water level variation between two consecutive months.

5. Results

5.1. Spatial Distribution of Inundated Areas

[25] The inundated area fractions for the Negro River basin were extracted from the Prigent et al. [2007] data set. Note that lacking additional external information, the technique captures, but does not discriminate among, inundated wetlands, rivers, small lakes, irrigated agriculture. Prigent et al. [2007] showed that this multisatellite inundation data set exhibits very realistic distributions, with major inundated wetlands well delineated for all latitudes and environments and are considered consistent with existing independent static inventories [Prigent et al., 2001a, 2007]. Moreover, their seasonal and interannual variability has been evaluated over different environments and over different large river basins with respect to altimeter river levels, rain rates and snow estimates, and were found consistent with other related hydrological variables such as in situ river discharge. The only possible independent evaluation of the accuracy at regional scale was performed by comparisons with high-resolution images from Synthetic Aperture Radar and indicated a likely underestimation of the extent of the small wetlands (<80 km², i.e., ~10% of a ~800 km² pixel of our equal-area grid [see Prigent et al., 2007, Figures 6 and 7]).

[26] Figure 2 shows the average fractional inundation in the Negro River basin for 1994. The areas with a very high percentage of inundation (greater than 50%) are located on the lower part of the Negro River and the confluence between Negro and Branco rivers. The pixels corresponding to open waters, temporarily flooded pastures and low vegetation submerged by water during the flood in the Caracarai subbasin (upper part of the Branco River) also exhibit a large percentage of inundation (between 20 and 70%). Inundation associated with the Negro River basin is well identified, even in complex regions characterized by extensive flooding below dense vegetation canopies, with potentially high fractional inundation extent, low variability in the annual maximum, and quasi-permanent flooding (Figure 2).

[27] The seasonal and interannual variations of total inundated area in the Negro River basin are presented on Figure 3. The maximum is observed from May to July whereas the minimum is observed from November to February. Figure 3 shows important interannual variability with maximum inundated area varying from 57,000 km² (June 1996) to 42,000 km² (June 1997) and minimum inundated area varying from 28,000 km² (January 1997) to 15,700 km² (December 1997), while 1998 is characterized by an important peak of inundated area (53,500 km²) which occurs after a year of very low inundation extent. The small inundation in 1997 and large inundation in 1998 corresponds to the 1997/1998 El Niño—Southern Oscillation (ENSO) event and its opposite (1998/1999 La Niña) which are associated with decreases and increases of water levels and discharge in the Negro River basin [Guyot et al., 1998] and water storage in the whole Amazon basin [Stuck et al., 2006].

[28] Long-term and higher spatial resolution surveys of wetland extent over large regions are very scarce. SAR imagery can provide estimates with much better spatial resolution than the multisatellite inundation product but suffers from a lack of temporal coverage. One study of flooding in the Amazon basin for both low-water (September–October 1995) and high-water (May–June 1996) conditions is that of Frappart et al. [2005] based on 100-m resolution L-band SAR observations from the Japanese Earth Resources Satellite-1 (JERS-1). Table 1 compares our estimates with the Frappart et al. [2005] results during low (October 1995) and high water (June 1996) for the Negro River. The total SAR-derived flooded area is 36,000 km² (159,000 km²) for this 700,000 km² region as compared to our results of 26,000 km² (57,000 km²) for low-water (high-water) stage. The difference in total flooded area is larger during high-water stage. With its much better spatial resolution, the SAR can more accurately estimate small areas that are flooded in generally dry conditions or small dry areas in generally

Table 1. Estimates of the Flooded Areas in the Negro River Basin for October 1995 and June 1996*

<table>
<thead>
<tr>
<th>Flooded Area (km²)</th>
<th>Multisatellite</th>
<th>JERS-1</th>
<th>JERS-1 Negro Downstream</th>
</tr>
</thead>
<tbody>
<tr>
<td>Low water (Oct 95)</td>
<td>26,000</td>
<td>36,000</td>
<td>25,000</td>
</tr>
<tr>
<td>High water (June 96)</td>
<td>57,000</td>
<td>159,000</td>
<td>95,000</td>
</tr>
</tbody>
</table>

*From the multisatellite product, a classification of JERS-1 images, and the same classification without the upper Negro (Cucui and Sao Felipe subbasins) and Uaupes (Serrinha subbasin).

Figure 2. Extent of inundation as estimated from the multisatellite data set for 1994 (percent of inundated area per pixel).

Figure 3. Monthly inundated area (km²) in the Negro River basin between January 1993 and December 2000.
flooded conditions, whereas our lower resolution observations may miss some fractional coverage. Nevertheless, the low sensitivity of L-Band to smooth surfaces is responsible for a loss of accuracy for the discrimination between open water surfaces, bare soils and low vegetated areas [Martinez and Le Toan, 2007]. Also, sensitivity analysis showed that the classification accuracy is highly dependent on the number of images used [Martinez and Le Toan, 2007]. The classification results used here for delineating the floodplains in the Negro River basin were obtained with only two acquisitions of SAR images (JERS-1 double mosaic), i.e., the worst case in terms of accuracy.

[29] The SAR-derived product presents a greater sensitivity to small amounts of water under dense canopy cover than the multisatellite product. The inundation in the upper Negro (Cucui and Sao Felipe subbasins) and Uaupes (Serrinha subbasin) is not detected by the multisatellite product. We removed from the inundated area previously derived the estimates for these three subbasins. Without the flooded area of these three subbasins, the SAR-derived flooded areas are now 25,000 km² and 95,000 km² for low- and high-water stage respectively. The two estimates are almost identical for low-water stage but the SAR-derived remains larger at high-water stage.

[30] Additionally, as shown by Prigent et al. [2007], during the low-water stage, more pixels with low fractional inundation are detected by the SAR relative to our product, but also there is a tendency for our analysis to overestimate the higher fractional inundation as well. However, both of these biases are smaller during high-water stage. This result may be explained by more frequent areas with small fractional inundation during low-water stage, as expected, but also by the more frequent occurrence of small dry patches in areas with large fractional inundation. However, a very strict detection threshold produces a systematic overestimate, whereas a less strict threshold allows for both overestimates and underestimates, producing a small average error. Likewise, our lower resolution product can yield a less biased estimate using a finite detection threshold as we did. We note that the SAR product still has some artifacts that have not been removed (radar speckle, lagged gain changes) and that the boundary between flooded and dry signals is still somewhat ambiguous [Hess et al., 2003], so there is uncertainty in the detection threshold for SAR as well.

Figure 4. Examples of time series of water level (with reference to GGM02C geoid) for (top) river and (bottom) floodplain. The thin black line on top panel represents the water level variations at Manaus in situ gauge station.
5.2. Water Level Time Series

The 86 T/P altimetry stations where water level time series can be constructed are unevenly distributed across the basin (Figure 1). Thirty-two are located on rivers and 54 on wetlands. An important goal of this data set is to give valuable information on water levels for unmonitored regions of the Negro River basin such as the Uaupes River flowing from the Colombian part of the watershed, on the western side of the Negro basin (four stations on the river and one on the floodplain), the upper Negro, near the Venezuelan border and the divide with Orinoco basin (15 mainly on a large unmonitored inundated area), or the right bank of Negro River (9 stations). Examples of water level time series derived from radar altimetry are presented in Figure 4.

This relatively large number of stations (94 encompassing 86 altimetric and 8 in situ), distributed over river channels and wetlands, is necessary to accurately estimate water volume variations. Frappart et al. [2005] reported that the precision is better for river channels than for inundated areas and floodplains. Minimum standard deviations of around 10 cm are measured on downstream rivers with large open water areas when the pulse emitted by the altimeter is not scattered by vegetation, whereas maximum standard deviations of around 60 cm is observed on flooded areas covered with dense vegetation.

5.3. Water Volume Variations

The monthly flood maps indicate that the flood period in the Negro River generally ranges from May to August, whereas low-water period ranges from September to February. As an example for the 1995–1996 hydrological cycle, Figure 5 shows maps of interpolated water levels for October 1995 and June 1996, which corresponds to the minimum and maximum respectively.

Figure 6 presents the differences between two consecutive months of the mean surface water volume (i.e., the monthly change in floodplain water storage) averaged over the study area for 1993–2000. Positive variations are obtained between November and June. Positive peaks are larger for the years 1996 and 1998. We also calculated the water storage changes during the 1995–1996 hydrological cycle.
cycle of the Negro River, i.e., the difference between the minimum and maximum storage anomaly. It corresponds to a volume variation of 167 km$^3$ for the whole floodplain.

Using (5), we have estimated the maximum error for the volume change in the lower Negro River basin with the following values:

- $S_{\text{max}} = 60,000$ km$^2$ in year 2000 (see Figure 3);
- $\delta h_{\text{max}} = 5$ m, maximum water level change between two consecutive months during the study period;
- $\Delta S_{\text{max}} = 10\%$ \cite{Prigent2007} of 60,000 km$^2$;
- $\Delta(\delta h_{\text{max}}) = 0.5$ m, maximum dispersion of the altimeter measurements.

For the whole study zone, we obtained a maximum error of 39 km$^3$ over the period October 1995 to June 1996 for a total positive variation of 167 km$^3$, i.e., an error of 23%.

Comparisons have been made with the results obtained combining T/P altimetry and JERS-1 imagery \cite{Frappart2005}. The results of this comparison are presented in Table 2. The total SAR-derived flooded volume difference between low- and high-water stages is 320 km$^3$, that is to say almost two times larger than our current results. If the subbasins of Cucui, Sao Felipe and Serrinha are not included in the estimate for the reasons previously mentioned, the maximum storage is 220 km$^3$. For the same study area, our result is 30% lower than the maximum storage variations estimated by Frappart et al. \cite{2005}. The difference of maximum of surface water storage variations between the two methods derives primarily from the difference between the flooded area estimates.

We then compared the mean annual cycle of monthly surface storage variations for 1993–2000 period with the mean annual cycle of monthly changes in total water storage from GRACE for 2003–2005 (Figure 7). We notice that the average monthly GRACE-derived total water volume changes present almost in-phase fluctuations with the average monthly changes in surface water volume computed in the present study. The maxima (minima) of both of the time series occurred in May (October). The maximum surface water volume change represents between one third and one half of the total and the minimum roughly one half.

A comparison between water volume stored in the floodplains of the Negro River basin and the water volume that flows to Manaus (outlet of the basin) was performed. The time series of monthly changes in surface water volume and monthly integrated discharge are presented in Figure 8. The two time series present the same range of variations (between $-80$ km$^3$/month and $50$ km$^3$/month). Nevertheless, no obvious relationship can be found between these two parameters. Frappart et al. \cite{2005} already observed this lack of clear relationship between potential storage capacity within the floodplain and integrated discharge during the flood season.

Figure 9 compares the time series of the monthly changes in surface water volume and monthly variations of precipitation in the Negro River Basin for the period 1993

\begin{figure*}[h]
\centering
\includegraphics[width=\textwidth]{figure7.png}
\caption{Monthly variation of surface water volume change from T/P radar altimetry and multisatellite derived inundation data set averaged over 1993–2000 period (blue dashed line) and monthly variation of total land water volume change from GRACE averaged over 2003–2005 period (red dashed line). The difference, which represents the sum of soil moisture and groundwater, is represented by a green dashed line.}
\end{figure*}
Both exhibit a similar behavior with a correlation of 0.61 for a time lag of one month (precipitation precedes surface storage variations), although precipitation variations are lower than surface water volume between January and March for years 1994, 1995 and 1998.

The surface water changes averaged over a river basin can be removed from the total water changes detected by the GRACE gravimetry mission to estimate monthly changes in water stored in the rest of the Negro River basin (i.e., soil moisture and groundwater). Owing to the lack of a common period between the estimated surface water volume variations and GRACE observations, we compared the annual cycle of monthly average surface storage variations for 1993–2000 with the annual cycle of changes in total storage from GRACE for 2003–2005. Figure 7 shows difference between these two, or the monthly changes in basin water storage outside of the river channel/floodplain system. The climatology hence defined showed that the total water storage of the Negro River basin is almost equally partitioned between surface water and the combination of soil moisture and groundwater.

6. Conclusion

In this study, we estimated surface water storage variations in the Negro River for the 1993–2000 period. The combined use of altimetric water level observations (from T/P) and inundation patterns derived from multisatellite information to determine water volume variations provides valuable information on the inundation dynamics of river floodplains. Seasonal and interannual variabilities are consistent with precipitation and rivers discharges, especially during ENSO years.

Knowledge of surface water volumes has several potential applications, as flood monitoring and forecasting, sediment and nutrient transport assessment or floodplain geomorphology. We also demonstrate the complementarity among several types of remote sensing data: a multisatellite inundation data set, water levels derived from radar altimetry and GRACE measurements of the total water storage. For the first time, the total water storage from GRACE has been separated into several components. The recent study from Papa et al. [2008] on the variations of surface water extent and total water storage suggests that the technique presented here will be investigated and certainly applied with success to other large river basins.

These results will have implications for better monitoring the water cycle, and in particular, improvements can be expected when additional data from current and future satellite missions are used. The combination of data from the present radar altimeters (T/P, ERS-1&2, Jason-1, ENVISAT RA-2) will allow better sampling of water level variations over rivers and floodplains in both time and space. The use of the future soil moisture products derived

Figure 8. Monthly variation of surface water volume change from T/P radar altimetry and multisatellite derived inundation data set (solid line) and monthly variation of volume change from integrated river discharge in Manaus (dashed line).

Figure 9. Monthly variation of surface water volume change from T/P radar altimetry and multisatellite derived inundation data set (solid line) and monthly precipitation rate over the Negro River basin from GPCP (dashed line).
References


Kerr et al. (2001) and future soil moisture missions will allow for further decomposition of the change in total water storage into its surface water, soil moisture and remaining storage reservoirs (e.g., groundwater).

Acknowledgments. The authors would like to acknowledge the CTOH (Centre de Topographie des Océans et de l’Hydrosphe) at LEGOS for the provision of the TOPEX/POSEIDON GDR data set and the HYDRAm project for the in situ gauge measurements. The first and third authors were supported by NASA GRACE Science Team grant NN04G999G and NASA REASOn grant JPL-1259524. We also want to thank an anonymous reviewer for helping us to improve the quality of the manuscript.


J. S. Famiglietti, Department of Earth System Science, University of California, Irvine, 3317 Croul Hall, Irvine, CA 92697-3100, USA. (jfamigli@uci.edu)

F. Frappart, LMTG, UMR5563, 14 Avenue Edouard Belin, F-31400 Toulouse, France. (frederic.frappart@cesbio.cnrs.fr)

F. Papa and W. B. Rossow, NOAA Cooperative Remote Sensing Science and Technology Center, City College of New York, Convent Avenue at 140th Street, New York, NY 10031, USA. (fpapa@ccny.cuny.edu; wbrossow@ccny.cuny.edu)

C. Prigent, Laboratoire d’Etudes du Rayonnement et de la Matiere en Astrophysique, Observatoire de Paris, CNRS, 61 Avenue de l’Observatoire, F-75014 Paris, France. (Catherine.Prigent@obspm.fr)

F. Seyler, IRD, SHIS, QL 16, Conj. 4, Casa 8, Lago Sul, 71640-245 Brasilia DF, Brazil. (Frederique.Seyler@ird.fr)