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Remote Sensing of Environment xxx (2011) xxx



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	Highlights	
Characterization of terrestrial wa	ter dynamics in the Congo Basin using GRACE	Remote Sensing of Environment xxx (2011) xxx – x.
Hyongki Lee ^{a,b,*} , R. Edward Beighley ^c , Dai Yamazaki ^e , Konstantinos Andreadis	Douglas Alsdorf ^{a,b} , Hahn Chul Jung ^d , C.K. Shum ^{a,b} , Jianbin ^b	Duan ^a , Junyi Guo ^a ,
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► We provide the first-ever measurement from mainstem. ► Differences between	nts of the Congo wetlands water volume change. ► Wetland the Congo wetland and the Amazon floodplain are highlig	l water is dominated by local upland runoff and much le hted.

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Remote Sensing of Environment xxx (2011) xxx-xxx



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Characterization of terrestrial water dynamics in the Congo Basin using GRACE and satellite radar altimetry

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ABSTRACT

The Congo Basin is the world's third largest in size (~3.7 million km²), and second only to the Amazon River 26 in discharge (~40,200 m³ s⁻¹ annual average). However, the hydrological dynamics of seasonally flooded 27 wetlands and floodplains remains poorly quantified. Here, we separate the Congo wetland into four $3^{\circ} \times 3^{\circ}$ 28 regions, and use remote sensing measurements (i.e., GRACE, satellite radar altimeter, GPCP, JERS-1, SRTM, 29 and MODIS) to estimate the amounts of water filling and draining from the Congo wetland, and to determine 30 the source of the water. We find that the amount of water annually filling and draining the Congo wetlands is 31 111 km³, which is about one-third the size of the water volumes found on the mainstem Amazon floodplain. 32 Based on amplitude comparisons among the water volume changes and timing comparisons among their 33 fluxes, we conclude that the local upland runoff is the main source of the Congo wetland water, not the fluvial 34 process of river-floodplain water exchange as in the Amazon. Our hydraulic analysis using altimeter measurements also supports our conclusion by demonstrating that water surface elevations in the wetlands 36 are consistently higher than the adjacent river water levels. Our research highlights differences in the hydrol-37 ogy and hydrodynamics between the Congo wetland and the mainstem Amazon floodplain. 38

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44 **1. Introduction**

The Congo Basin is the world's third largest in size (~ 3.7 million km²), 45 46 and second only to the Amazon River in discharge (~40,200 $\text{m}^3 \text{ s}^{-1}$ annual average). The impact and connections of this hydrologic flux with 47 the region's climate, biogeochemical cycling, and terrestrial water stor-48age, especially in wetlands, is of great importance. For example, the ex-4950tent of the differences in chemistry, seasonality, rate and volume of water input to the floodplain and wetland systems from upland runoff, 51direct rainfall and mainstem flooding are likely to supply substantially 5253 different amounts of nutrients and other solutes (Melack & Engle, 2009). However, the hydrological dynamics of seasonally flooded 54

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wetlands and floodplains remains poorly quantified through ground ob-55servations, satellite observations or modeling. As a consequence, esti-56mates of the magnitude of other processes driven by such dynamics,57such as methane emissions from flooded wetlands that form a significant58contribution to global atmospheric methane, also cannot be well estimat-59ed. Given the vast size and remote location of the Congo Basin, satellite-60borne observations provide the only viable approach to understanding61the spatial and temporal distributions of its water balances.62

Recently, Alsdorf et al. (2010) have estimated the amounts of water 63 filling and draining from the mainstem Amazon floodplain using data 64 from the Gravity Recovery and Climate Experiment (GRACE) and 65 other satellite measurements. They showed that the majority of water 66 on the mainstem Amazon floodplain is derived from the river with a 67 much less amount from local upland runoff. However, there has been 68 no attempt to estimate the Congo wetland water storage and its flux. 69 In this study, we use satellite-borne observations to suggest a baseline 70 measurement of these storages and fluxes by examining 1) the amount 71 of water stored and drained from the Congo wetland, and 2) whether 72 the water comes from rivers or adjacent upland areas. 73

We use total storage change in the form of equivalent water height 74 (EWH) change (Wahr et al., 1998) from the GRACE measurements 75 (Tapley et al., 2004), precipitation (P) estimates from the Global 76

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2

H. Lee et al. / Remote Sensing of Environment xxx (2011) xxx-xxx

Precipitation Climatology Project (GPCP; Adler et al., 2003), evapotrans-77 78 piration (ET) estimates from the Hillslope River Routing (HRR) model (Beighley et al., 2009), water elevation changes from Environmental 79 80 Satellite (Envisat) altimeter measurements, and hydrological maps from HydroSHEDS (Lehner et al., 2008). Measurements of inundated 81 area are made from a combination of (1) the Japanese Earth Resources 82 Satellite-1 (JERS-1) Synthetic Aperture Radar (SAR) mosaics developed 83 by the Global Rain Forest Mapping project (GRFM), (2) the Shuttle 84 85 Radar Topography Mission (SRTM) Digital Elevation Model (DEM), and 86 (3) Moderate-resolution Imaging Spectroradiometer (MODIS) mosaics 87 (Jung et al., 2010a). Unfortunately, we have no available contemporaneous in situ discharge or water stage measurements. We combine these 88 satellite-based measurements to: (1) estimate the wetland storage 89 90 changes in four regions along the Congo mainstem and its major tributaries, and (2) determine whether the water comes from rivers or adja-91 cent upland areas. 92

The methods presented here are improved compared to the previous 93 94 study over the Amazon Basin (Alsdorf et al., 2010) because 1) Hydro-SHEDS is used to estimate the upland area that contributes directly to 95 the wetland instead of using a ratio between estimates of upland area 96 compared to the wetland area; 2) more realistic ET estimates are used 97 instead of a single number representing the whole basin; and 3) a hy-98 99 draulic analysis from altimeter measurements is also presented. We 100 also use a longer time span (6 years compared to 2.5 years) of GRACE 101 data.

102 2. Methods

2.1. Study area 103

We select four $3^{\circ} \times 3^{\circ}$ study regions to cover the wetlands of the 104 105Congo River mainstem and its major tributaries (Fig. 1). Study region 1 includes the Ubangi River (~3800 $\text{m}^3 \text{ s}^{-1}$ annual discharge, Laraque 106 et al. (2001)), which is the largest right-bank tributary of the Congo 107 mainstem. Study region 2 includes the Sangha River (~1600 $m^3 s^{-1}$ 108 annual discharge, Laraque et al. (2001)) and represents the majority 109 of the northern tributary wetlands. Study regions 3 and 4 include east-110 ern and southern tributaries, respectively. The box size is chosen based 111 on the limit of the spatial resolution of GRACE which is determined 112from the maximum degree $(n_{\text{max}} = 60)$ of the Stokes coefficients. 113



Fig. 1. Locations of four 3°×3° study regions in the Congo Basin. Background shows topography from the SRTM C-band DEM. Intersections between Envisat altimeter and the Congo River are indicated with "+".

2.2. Wetland storage changes from satellite measurements

Total storage changes for a given area, ΔS , are a summation of the 115 storage changes in wetlands (ΔS_w), rivers (ΔS_r), groundwater (ΔS_σ), 116 and soil moisture (ΔS_{sm}): 117

$$\Delta S = \Delta S_{w} + \Delta S_{r} + \Delta S_{g} + \Delta S_{sm}. \tag{1}$$

Measurements from GRACE provide ΔS in terms of anomalies with 120 respect to a mean total storage value. We processed the Release 4 121 (RL04) Center for Space Research (CSR) GRACE Level 2 (L2) data 122 product (Bettadpur, 2007) from January 2003 to December 2008. To 123 reduce the GRACE longitudinal stripes associated with correlations 124 among even or odd degree Stokes coefficients at resonant orders 125 (Swenson & Wahr, 2006), decorrelation based on Duan et al. (2009) 126 was used. We also applied smoothing using a 3-degree Gaussian filter 127 (Guo et al., 2010). EWHs are computed at $1^{\circ} \times 1^{\circ}$ grid spacings, and 128 spatially averaged over each study region. Finally, total storage anom- 129 alies are obtained by multiplying the EWHs by the box area. More de- 130 tails on the GRACE measurements are provided in Section 3.1. 131

The channel storage anomalies are estimated by multiplying water 132 stage anomalies, obtained from the Envisat altimeter, with open 133 channel areas estimated from the classification of GRFM image data 134 (Table 1, see discussion below). The Envisat Geophysical Data Records 135 (GDRs) contain 35-day repeat, 18-Hz data (twenty-measurements- 136 per-frame), which corresponds to a ground spacing of approximately 137 350 m. The GDRs include range measurements from four different 138 retracking algorithms. In this study, we use the retracked measure- 139 ments from the ICE-1 retracker (Bamber, 1994), which generally 140 Q3 performs well over inland water bodies (Frappart et al., 2006; Lee 141 Q4 et al., 2010). The water stage anomalies over the intersections 142 between the altimeter and the open water bodies are averaged for 143 each tributary, and are then multiplied by the corresponding channel 144 areas. 145

We use $2.5^{\circ} \times 2.5^{\circ}$ GPCP monthly merged precipitation rates P(t) 146 (Adler et al., 2003), and create anomalies by subtracting a linear fit, \overline{P} , 147 to the integrated sum of P(t) for each study region (see Alsdorf et al., 148 2010 for details). The slopes of the linear-fit lines represent six-year 149 mean precipitation values, as summarized in Table 1. The GPCP data is 150 derived partly from infrared and microwave satellite measurements, 151 and it should be noted that, as stated in Beighley et al. (2011), there is 152 a discrepancy between various satellite derived precipitation datasets 153 over the Congo Basin in terms of their magnitudes, especially in 154 equatorial regions, which correspond to study regions 2 and 3 in 155 this study. For ET, we use model-based estimates from HRR. It is the 156 sum of wet canopy evaporation, dry canopy transpiration and evapora-157 tion from saturated soil surfaces based on the potential ET using Pen- 158 man-Monteith indirectly through the temperature-based method of 159 estimating its data sources (see Beighley et al., 2009, 2011 for details). 160 The ET rates over each Pfafstetter Level 4 sub-divisions are averaged 161 for each of the four study regions (\overline{E}) (Table 1). This Pfafstetter discreti- 162 zation frame-work is a natural system, based on topographic subdivision 163 of the land surface and the resulting topology of the hydrographic net- 164 work (Verdin & Verdin, 1999). Each level of discretization results in 9 165 sub-divisions (i.e., 4 tributaries and 5 local contributing areas to the 166

Table 1	t1.1
Hydrologic and geomorphic characteristics of each study region.	
	t1 2

	Region 1	Region 2	Region 3	Region 4	t1.2 t1.3
Upland (km ²)	83,605	42,905	55,297	58,587	t1.4
Wetland (km ²)	28,052	68,596	56,360	52,914	t1.5
Channels (km ²)	1058	3990	502	2766	t1.6
Annual P (m year $^{-1}$)	1.44	1.53	1.87	1.71	t1.7
Annual ET (m year ⁻¹)	0.90	1.01	1.06	0.92	t1.8
Contributing area (km ²)	121,330	151,596	152,789	141,728	t1.9

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114

119

main channel). The ultimate number of sub-areas is 9^{Level Number}, but often less than that due to lack of network resolution at higher levels. The Congo Basin was ultimately delineated to Pfafstetter level 4 using a threshold area of ~8.1 km² which resulted in 5498 model units (i.e., sub-divisions) with a median model unit drainage area of 670 km² and a mean hillslope length of 5.4 km. Anomalies of P – ET are estimated to be:

$$P - ET = \left(P(t) - \overline{P}\right) \times \frac{\overline{P} - E}{\overline{P}}.$$
(2)

174

The P-ET anomalies are used to estimate the runoff from the local uplands (Section 3.2).

178 In this study, we assume that groundwater changes associated with the shallow water table (ΔS_g) are driven by P – ET. These changes are 179assumed to be negligible beneath wetland areas that do not drain, i.e., 180 the water table is assumed to be consistently at the surface in wetlands 181 that contain water from year to year. P - ET varies seasonally and is 182 expected to account for water table variations in the upland areas of 183 each $3^{\circ} \times 3^{\circ}$ box. Similarly, we assume that P – ET is forcing any soil 184 moisture variations ($\Delta S_{sm}).$ Thus, our estimates of P and ET are used, 185 below in Section 3.2, to account for ΔS_g and ΔS_{sm} . 186

The Congo interfluvial wetlands cover a variety of vegetation and hydrogeomorphic environments. Most of the Congo classifications have been developed based on vegetation type and forest density (e.g., Hansen et al., 2008; Laporte et al., 1995), whereas few classification schemes have focused on flooding in the wetlands (Bwangoy et al., 2009). In this study, we use the hydrogeomorphic flood classification of Jung et al. (2010a).

194 **3. Results and discussions**

195 3.1. GRACE measurements over the Congo Basin

The Congo River is the only major river to cross the equator twice. In 196 doing so, the basin lies in both the Northern and Southern Hemisphere 197such that it receives year-round rainfall from the migration of the Inter-198Tropical Convergence Zone (ITCZ). After the north has its wet season in 199 July-September, the ITCZ moves south and the remainder of the basin 200 receives large amounts of rain. Fig. 2 shows the spatial variations in 201the storage changes from the CSR GRACE data after decorrelation and 202 smoothing. It can be seen that the positive anomaly in September 2032006, which is present outside of the Congo Basin, becomes stronger 204 as it moves southward and into the basin. Likewise, the positive anom-205206 aly observed in the southeastern part of the basin in January 2007 becomes stronger and widely spread over the southern boundary of 207 the Congo Basin. This spatial pattern of the storage changes is different 208 from that over the Amazon Basin, where the strongest positive or neg-209 ative annual water storage anomalies are observed to be centered inside 210 the basin (e.g., Alsdorf et al., 2010; Han et al., 2005). 211

We also examine the basin-averaged time series of EWH anomalies 212 obtained using four different GRACE data products (from CSR, Jet Pro- 213 pulsion Laboratory (JPL), GeoForschungsZentrum (GFZ), Institut für 214 Geodäsie und Geoinformation (ITG)) using equivalent decorrelation 215 and smoothing (Fig. 3 (top)). They generally agree in terms of their an- 216 nual increases and decreases in the time series. In addition, all of them 217 show a drying trend until 2006, and then a sharp increase at the end 218 of 2006. However, there are important differences in their amplitudes. 219 There are at least 1 cm EWH differences among the GRACE products; 220 for example, the CSR and ITG solutions differ by at least 1 cm during 221 the last two months of 2005. If we convert this 1 cm EWH difference 222 to streamflow by multiplying it by the basin area $(3.7 \text{ million } \text{km}^2)$ 223 and dividing it by the time duration, we get about 7000 m³ s⁻¹. As a 224 comparison, this approximately corresponds to the mean annual dis- 225 charge of the Ohio River in the United States. Moreover, there are at 226 least 5 cm EWH differences between the CSR and JPL solutions that 227 last about 5 months in the first half of 2008. If we again convert this to 228 discharge, we get approximately 14,000 m³ s⁻¹ which corresponds to 229 more than one-third of the Congo River mean annual discharge. It also 230 corresponds to about three-quarters of the Mississippi River discharge. 231 This is a significant difference: note that the Congo and Mississippi River 232 basins are similar in size. Furthermore, the four different GRACE prod- 233 ucts do not produce the same errors year after year. For example, in 234 the first half of 2006, the JPL solution has generally less EWH values 235 than the CSR solution, but in the second half of 2006 when the trough 236 occurs, the CSR solution values are less than the JPL values. This can be 237 widely observed every year among all of the GRACE products. Overall, 238 the discrepancy among the GRACE products has important implications 239 for Congo hydrology. In addition to different data processing methods 240 and models adopted at different institutes, these disagreements may 241 also be due, in part, to the movement of ITCZ and the consequent leak- 242 age of strong signal from outside of the basin (e.g., the strong positive 243 anomaly in September 2006) or from inside of the basin (e.g., the strong 244 positive anomaly in May 2007). This leakage is due to the truncated 245 spectral degree (e.g., $n_{\text{max}} = 60$) in the GRACE gravity field solutions 246 and to post-processing smoothing. The leakage phenomenon can 247 occur at all scales including the finest spatial resolution possible with 248 GRACE 249

Recently, in the GRACE science community, there has been an ef- 250 fort to use global simulations of water storage variations when 251



Fig. 2. Monthly Equivalent Water Height (EWH) anomalies from the CSR GRACE product after decorrelation and 3-degree radius Gaussian smoothing. The Congo Basin is shown with a red outer boundary. Red rectangles indicate our study regions. (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)

H. Lee et al. / Remote Sensing of Environment xxx (2011) xxx-xxx



4

Fig. 3. (top) Comparison among GRACE EWH anomalies over the entire basin after equivalent decorrelation and smoothing. (bottom) Comparison among EWH anomalies over the study regions from CSR GRACE data.

restoring the signal loss in GRACE, which is caused by smoothing. It 252has been proposed to estimate scale factors, by comparing the unfil-253 tered model time series with the filtered model simulations, to par-254tially correct for the signal attenuation. We have examined the 255original filtered basin-averaged time series with the scaled time se-256257ries (data courtesy, S. Swenson) from CSR GRACE data. The scale, computed using the Community Land Model (CLM), averaged over 258the Congo Basin is 1.2, and thus the scaled time series has a slightly 259260 lower amplitude than the smoothed time series. This scale is a simple temporally constant number, intended to depend on the statistical 261262characteristic of the model-simulated storage variations. This approach assumes that errors in the global hydrologic model are spatially and 263temporally randomly distributed and thus do not introduce a bias in 264the scaling factor. It is also not entirely clear that a model should be 265used to correct a measurement, especially in the case of the Congo 266 267Basin where model errors are less well-known compared to other re-268gions such as the United States. Moreover, the scaled EWH anomalies cannot resolve the issue of discrepancy among the GRACE products. 269270Therefore, in this study, we do not attempt to correct the leakage error or restore the signal loss due to the smoothing. Rather, we treat 271272the differences among the GRACE products as the error of our storage change estimates. 273

We now compare the EWH anomalies over our four study regions 274 (Fig. 3, bottom) to examine whether the $3^{\circ} \times 3^{\circ}$ box size is appropriate 275and if the storage changes among them can be distinguished. From 276Fig. 3 (bottom), the EWH anomalies in region 4 are clearly different 277from the other three regions in terms of their timing and amplitudes. 278For example, in 2004 region 4 has a trough in August whereas it oc-279curred in February over the other regions. Region 4 also has the smallest 280281 peak in December compared to the other regions. Although the timing among regions 1–3 appears to agree, there are differences in anomaly 282 amplitudes. For example, there are about 13 cm EWH differences be-283 tween regions 1 and 2 during July–September 2006. Converting this dif-284 ference to river discharge, yields about $3000 \text{ m}^3 \text{ s}^{-1}$. As another 285 example, about 9 cm EWH differences, lasting about three months at 286 the end of 2005, correspond to about 2000 m³s⁻¹ of streamflow and 287 can be observed between regions 2 and 3. In general, the amplitudes 288 and occasionally the timing of the major peaks and troughs are different 289 among the study regions. This distinction supports our choice of the box 290 size and the resultant wetland volume.

3.2. Wetland water volume change and its flux

292

We observe from Fig. 4 that the total storage anomalies from GRACE, 293 P – ET anomalies, and river storage anomalies within a given study re-294 gion are well timed with each other. However, in terms of amplitudes, 295 the river channel storage anomalies are significantly less than the 296 GRACE anomalies, which suggests that storage changes in rivers ac-297 count for little of the total storage anomalies (note that river anomalies 298 are multiplied by 5, 10, or 20). The P – ET anomaly amplitudes are significantly greater than those of the rivers and typically less than tho 300 total storage anomaly amplitudes, which suggest that P – ET accounts 301 for an important fraction of the GRACE measured total volume change. 302 Thus we concluded that hydrological processes associated with P – ET 303 (e.g., runoff) are significant contributors to the total storage change ob-304 served in each $3^{\circ} \times 3^{\circ}$ study region and that in-channel fluvial processes are not significant contributors. 301

The wetland storage anomalies have two contributors, which are 307 (1) direct precipitation on the wetlands as well as runoff supplied 308 to the wetlands from the surrounding uplands, and (2) water ex- 309 changed between the wetlands and the adjacent river channels. It 310 should be noted that the groundwater contribution to the wetland 311 water levels is considered in the upland P-ET runoff because the 312 groundwater is controlled by the infiltration of rainfall. The volumes 313 of runoff from the local uplands and direct rainfall on the wetlands 314 can be estimated by multiplying P-ET with the contributing area 315 or with the inundated area, respectively. The land areas contributing 316 to the wetlands are computed using the following procedures. 317 First, flow directions from HydroSHEDS are obtained to determine 318 flow accumulation and the associated drainage network. Next, we as- 319 sume that major rivers have width greater than 100 m. This threshold 320 is chosen based on the resolution of the GRFM mosaic, which is used 321 to extract channel areas and to compute the river storage changes in 322 Section 2.2. Based on the relationship between the channel width and 323 the upstream drainage area $(w(m) = 0.438 \text{ A}_{\text{II}}(\text{km}^2)^{0.592}$, Beighley 324 et al., 2011), this 100 m threshold approximately corresponds to riv- 325 ers with drainage areas larger than 10,000 km². We remove these 326 major rivers and their contributing areas from the flow direction 327 grid. Thus, we distinguish the contribution of large river drainage 328 areas from the wetland drainage areas. Then, we extract the wetland 329 pixels for each study region using the classification map (Jung et al., 330 2010a). Finally, we delineate the area that drains to each wetland 331 pixel for each study region (Fig. 5 and Table 1). Essentially, the frac- 332 tion of contributing area that is outside of each $3^{\circ} \times 3^{\circ}$ study region 333 is connected with streams having a drainage area smaller than 334 10,000 km² and that drain directly to a wetland pixel. To further ex- 335 amine whether the 100 m channel width is a reasonable number to 336 distinguish between the contributing areas that flow to the wetlands 337 and the contributing areas that flow to the major rivers, we tracked 338 discharges for all of the rivers which have the contributing areas larg- 339 er than 10,000 km² and that flow into our study regions (red lines in 340Fig. 5). As summarized in Table 2, we used the HRR model to estimate 341 these discharges during the period 2003-2008 (Beighley et al., 2011). 342 The mean annual discharge for all of the major rivers from a unit con- 343 tributing area is estimated at $0.016 \text{ m}^3 \text{ s}^{-1} \text{ km}^{-2}$. For a drainage 344of 10,000 km², this corresponds to 160 m³ s⁻¹ of discharge. So, the 345

H. Lee et al. / Remote Sensing of Environment xxx (2011) xxx-xxx



Fig. 4. Time series of satellite-based measurements of Congo hydrology for each study region. Red lines represent EWH anomalies from CSR GRACE data, and black lines are P – ET anomalies. Blue lines show river storage anomalies, and they are multiplied by 5, 10, or 20 for visual clarity. The river storage anomalies in this plot are generated by weighted averages of river stage anomalies with ratios between the channel area and the box area as the weights. (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)

100 m river width threshold (or 10,000 km² of contributing area) leads to rivers having a discharge greater than 160 m³ s⁻¹ and which do not directly flow into the wetlands.

Fig. 6 shows a comparison of water volume anomalies for the wetlands, rivers, and local upland runoff. In each of the four plots, the amplitudes of the river storages are negligible compared to the GRACE and P – ET anomalies. We suggest that this amplitude is not sufficient



Fig. 5. Colored boundaries represent the contributing land area draining to the wetlands in each study region with the corresponding major rivers (i.e., widths greater than 100 m or contributing areas larger than 10,000 km²; shown as red lines) and their drainage areas excluded from the study regions; see Table 2 for the hydrologic characteristics of the 19 excluded drainages. (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)

to supply any significant water volumes that would sufficiently account 353 for the storage changes measured by GRACE or estimated by P - ET. We 354 further explore this concept, i.e., the negligible amount of fluvial ex-355 change between the wetlands and main river channels, in Section 3.3, 356 below. Instead of river supply, the other potentially significant supply 357 of water to the wetlands is runoff from the surrounding uplands and 358 rainfall directly on the wetlands. The P - ET runoff volume anomalies 359 agree well with the GRACE wetland volume anomalies in terms of tim-360 ing and amplitude in region 1 and reasonably well in region 4. In region 361 2, there is a large discrepancy in their amplitudes in 2003 and 2004. 362 However, both the GRACE and P - ET runoff volume anom-364 alies and the GRACE wetland volume anomalies both reveal a dry 365

Table 2

Summary of the hydrologic characteristics for the major rivers which have contributing areas greater than 10,000 km² based on simulation results from the HRR model for the period 2003–2008.

					- 40.0
ID	Study region ID	Contributing area (km ²)	Annual discharge (m ³ s ⁻¹)	Annual discharge per unit area (m ³ s ⁻¹ km ⁻²)	t2.2 t2.3
1	1	31,270	429	0.014	t2.4
2	1	479,839	5527	0.012	t2.5
3	1	45,642	675	0.015	t2.6
4	1	1,348,434	18,686	0.014	t2.7
5	3	16,073	209	0.013	t2.8
6	3	21,639	322	0.015	t2.9
8	3	39,284	419	0.011	t2.10
9	3	18,809	343	0.018	t2.1
10	4	47,199	720	0.015	t2.1
11	4	453,653	6587	0.015	t2.1
12	4	65,691	1279	0.019	t2.1
13	4	25,752	551	0.021	t2.1
14	4	35,334	753	0.021	t2.1
15	4	136,132	2187	0.016	t2.1'
16	2	10,547	283	0.027	t2.1
17	2	14,145	221	0.016	t2.1
18	2	13,713	190	0.014	t2.2
19	2	158,137	2825	0.018	t2.2
				Mean: 0.016	t2.2

t2.1

H. Lee et al. / Remote Sensing of Environment xxx (2011) xxx-xxx



Fig. 6. Water volume anomalies of river (blue) and runoff (red). The shading illustrates the range wetland water volume anomalies estimated using CSR, JPL, GFZ, and ITG GRACE solutions. The black solid line indicates the mean of the four estimates. (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)

season in 2005 and a rather wet season in late 2007. Essentially, both 366 367 data show somewhat wetter years in 2003 and 2004, dryer years in 2005 and 2006, and returning to wetter years in 2007 and 2008. In re-368 369 gion 3, we observe that the timing of volume increases and decreases 370 do not generally agree between the wetland and the runoff anomalies, although they both show an excessive volume of water in 2007. In sum-371 mary, region 1 annually fills and drains about 20 km³ to 25 km³ of water 372 each year whereas regions 2, 3 and 4 fill and drain about 10 km³ to 37320 km³. 374

The differences observed in regions 2 and 3 may be due to problems with satellite rainfall products in the equatorial region. For example, Beighley et al. (2011) used three satellite derived precipitation datasets (TRMM, CMORPH, PERSIANN) to drive the HRR model throughout the Congo Basin. The results, which were compared to historical discharges, Envisat altimetry measurements and GRACE water storages, show that satellite precipitation products provide unreasonably high rainfall for 381 specific time periods (e.g., all three in Oct–Nov; only CMORPH and PER-382 SIANN in Mar–Apr) in the equatorial regions. These findings are also 383 consistent with previous studies that found large discrepancies between 384 gage and satellite precipitation over equatorial regions of Africa (e.g., 385 McCollum et al., 2000; Nicholson et al., 2003). Although additional re-386 search is needed to resolve this issue, one possible cause may be related 387 to the significant level of lightning activity in the region (Williams and Satori, 2004). 389

The rates of wetland filling and draining (Fig. 7) are computed by 390 taking the temporal derivative of the storage anomalies in Fig. 6 391 (Alsdorf et al., 2010). Regions 1 through 4 have about $\pm 1000 \text{ m}^3 \text{ s}^{-1}$ 392 to 2000 m³ s⁻¹ of wetland discharge during flooding and draining. 393 Summing the maximum and minimum wetland flux rates for all four 394 regions yields $\pm 6400 \text{ m}^3 \text{ s}^{-1}$ during flooding and emptying, or $\pm 16\%$ 395



Fig. 7. Wetland flow rates (blue, left Y-axis) and P – ET anomalies (green, right Y-axis). (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)

of the mainstem annually averaged discharge, i.e., $40,000 \text{ m}^3 \text{ s}^{-1}$ at the historic Kinshasa gage (Fig. 1). Summing the maximum volumes for all four regions yields 111 km³ of water stored and subsequently emptied each year from the Congo wetland. This corresponds to only about 8% of the total volume of water annually discharged from the Congo.

The timing of wetland filling (when the flux becomes positive from 401 negative) and draining (when the flux becomes negative from positive) 402 can be compared with the timing of increasing and decreasing of P - ET 403 404 to examine a temporal connection between them. Note that in Fig. 7 we are comparing wetland flow rates derived from GRACE (blue line) to 405406 changes in P – ET (green line). In regions 1, 2, and 4, P – ET is always increasing when the wetland flux rates change from negative to positive, 407 i.e., from draining to filling. P-ET therefore always comes before the 408 wetland filling, and thus from a temporal perspective, the wetland 409 infilling starts with the P-ET runoff from the surrounding uplands. 410 On the other hand, when the wetland flux rates switch from positive 411 to negative, P - ET is always on the decreasing limb of the annual rain-412 fall. This again is expected where the wetland receives the majority of 413 its water from upland runoff. 414

415 3.3. Hydraulic analysis using altimeter measurements in the Congo and 416 Amazon basins

Our interpretation that the Congo wetlands receive the majority of 417 their water from upland runoff as opposed to exchange with adjacent 418 major tributaries as suggested by comparisons of GRACE anomalies 419 with P-ET anomalies, is also supported by Envisat altimeter obser-420 421 vations. The water elevation changes over the wetland regions, which have low topographic relief and higher radar backscatter, are 422 generated and compared with the water elevation changes over adja-423 cent river channels (for example, red circles in Fig. 8(a)). The vertical 424 425datum of both river and wetland water elevations is referenced to the 426 Earth Gravitational Model 2008 geoid (EGM08; Pavlis et al., 2008). Top panels of Fig. 9(a) and (b) show the surface height profiles 427 along the altimeter tracks obtained from several altimeter samplings 428 over the red circle regions in Fig. 8(a). We observe fluctuations in the 429water elevations of the Congo mainstem and its adjacent wetlands in 430 431each altimeter overpass. We then generate water elevation change time series by combining successive overpasses. It should be noted 432 that the wetland regions closest to the river channels along the tracks 433

are selected and compared with the river water fluctuations. As 434 shown in the bottom panels of Fig. 9(a) and (b), the range in wetland 435 water levels is small compared to the river. Moreover, the wetland 436 water topographic elevations are overall between 0.5 and 2.5 m 437 above the river, i.e., the wetland water levels are always greater 438 than the river. Clearly, the river cannot flow "uphill" into the wet- 439 lands. Therefore, the wetlands do not receive water from the adjacent 440 major tributaries or mainstem Congo River and instead can only sup- 441 ply water to the rivers. While these observations are necessarily local 442 to the red-circled areas in Fig. 8(a), they support the previous inter- 443 pretation which used GRACE observations to suggest that the local 444 upland runoff is the main source of the Congo wetland water. We 445 have investigated several more altimeter overpass locations (black- 446 circled in Fig. 8(a)), where the altimeter footprint allows delineation 447 of wetlands and rivers, and nearly all locations demonstrate that wet- 448 land water elevations are consistently higher in elevation than the 449 adjacent river. It may be argued that the river channel at the location 450 of the altimeter transect is located further downstream than the wet- 451 land. In this case, the wetland water levels can always be higher than 452 the river, and it does not necessarily indicate that the river waters 453 cannot flow into the wetlands. However, this is true only if there 454 are abundant floodplain channels that connect the wetlands to the 455 adjacent river channel. Jung et al. (2010b) highlighted the fundamen- 456 tal differences in the water level changes between the Amazon and 457 Congo wetlands, using Interferometric Synthetic Aperture Radar 458 (InSAR) measurements, due to differences in the connectivity of the 459 floodplain-river systems. The result suggests that connectivity of the 460 Congo River to the interfluvial wetland area is limited, compared 461 with the Amazon. 462

To further demonstrate that altimeter measurements are a hydraulic 463 indicator of the direction that water can or cannot flow, we also exam-464 ine Envisat altimeter measurements over the Amazon Basin. These 465 serve as a comparison to the Congo. We generate the time series of 466 water elevation changes over the Amazon mainstem (Solimoes River) 467 and its adjacent floodplain (Fig. 9(c)). We note that the water levels at 468 mid-rising stage in the floodplain are lower than the river, but the low 469 water levels are almost identical. This implies that the water is flowing 470 down the hydraulic slope from the river to the floodplain during mid-471 rising stage. Moreover, in nearly every year, the river clearly rises before 472 the floodplain. These timings, in combination with the elevations noted 473



Fig. 8. Color-coded lines represent Envisat 18-Hz ICE-1 retracked surface heights, referenced to EGM96 geoid, over (a) the Congo Basin from cycle 12 (December 2002), and (b) the Amazon Basin from cycle 18 (August 2003). Background is SRTM DEM. (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)

H. Lee et al. / Remote Sensing of Environment xxx (2011) xxx-xxx



Fig. 9. Water level fluctuations over the Congo mainstem and its adjacent wetlands along (a) Envisat passes 472 and (b) 887 over the red circle regions in Fig. 8. Water level variations over the Amazon River and its floodplain are shown in (c). They are referenced to the EGM08 geoid. (For interpretation of the references to color in this figure legend, the reader is referred to the web version of this article.)

474 above, imply that the Amazon river is flowing into the floodplain. We 475 also observe that the river has a "shoulder" in the mid-rising times of every year where the water level stops rising, or at least slows its rising 476 before rapidly rising again. The floodplain shows a similar shoulder. This 477 further suggests that the floodplain is responding to the river. These ob-478 479servations indicate that the floodplain of the Amazon mainstem derives its water more from the river, than from the uplands: a conclusion al-480 481

ready supported by GRACE observations in Alsdorf et al. (2010).

4. Conclusions 482

The amount of water annually filling and draining the Congo wetland 483 is large, but only about one-third of the mainstem Amazon floodplain 484 volumes (111 km³ compared to 285 km³; Alsdorf et al., 2010). Based 485on the amplitude comparison among the water volume changes and 486 the timing comparison among their fluxes, we conclude that the local 487 upland runoff is the main source of the Congo wetland water, not the 488 fluvial process of river-wetland water exchange. Delineating whether 489 the water comes from local uplands or from distal places via fluvial 490 transport presumably makes a difference in the sediment supplies and 491 492 in the carbon and nutrient exchanges. For example, given the hydraulic gradient analysis of Section 3.3, it is unlikely that the Congo wetlands are 493 filled with sediments derived from erosion processes in the Congo head-494waters around the western flanks of the East Africa Rift system. Our 495analysis using altimeter measurements, although they could be local ob-496 497servations considering the vast size of the basins, supports our conclu-498 sion, highlighting the difference between the Congo wetland and the Amazon floodplain hydraulics. Our finding is in alignment with Jung et 499al. (2010b) which concluded that flow patterns in the Congo are less 500governed by channel connectivity because flooded areas in the Congo 501502are broadly distributed and do not have abundant floodplain channels as in the Amazon. 503

Although we assumed the contribution of soil moisture and ground-504water variation to the total storage change is negligible compared to 505that of the surface water, further studies are needed to accurately deter-506mine the portions of the soil moisture and groundwater changes that 507account for the total changes in the water balance, compared to the 508channel and wetland discharges. The HRR hydrologic and hydraulic 509model (Beighley et al., 2009, 2011) and the CaMa-Flood macro-scale 510511 floodplain model (Yamazaki et al., 2011) can help us determine not only those portions, but also simulate the wetland storage changes in 512 the Congo to compare with our results. 513

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References

	Adler, R. F., Huffman, G. J., Chang, A., Ferraro, R., Xie, P., Janowiak, J., et al. (2003). The	515
	version 2 global precipitation climatology project (GPCP) monthly precipitation	516
	analysis (1979–present). Journal of Hydrometeorology, 4, 1147–1167.	517
	Alsdorf, D., Han, SC., Bates, P., & Melack, J. (2010). Seasonal water storage on the Amazon	518
	floodplain measured from satellites. <i>Remote Sensing of Environment</i> , 114, 2448–2456.	519
	Beighley, R. E., Eggert, K. G., Dunne, I., He, Y., Gummadi, V., & Verdin, K. L. (2009). Sim-	520
	ulating hydrologic and hydraulic processes throughout the Amazon River Basin.	521
	Point Processes, 25, 1221–1255.	522
	satellite derived precipitation datasets using the Hillslope River Routing (HRR)	524
	model in the Congo River Basin Hydrological Processes doi:10.1002/hyp. 8045	525
	(published online).	526
	Bettadpur, S. (2007). CSR Level-2 processing standards document for product Release	527
	04. GRACE 327–742, Center for Space Research, University of Texas at Austin.	528
	Bwangoy, JR. B., Hansen, M. C., Roy, D. P., de Grandi, G., & Justice, C. O. (2009). Wet-	529
	land mapping in the Congo Basin using optical and radar remotely sensed data and	530
	derived topographical indices. Remote Sensing of Environment, 114, 73-86.	531
	Duan, X. J., Guo, J. Y., Shum, C., & van der Wal, W. (2009). On the postprocessing removal of	532
	correlated errors in GRACE temporal gravity held solutions. <i>Journal of Geodesy</i> , 83,	533
	1095-1100.	034 595
	for determining mass changes over land and ocean using CRACE data. <i>Conhysical</i>	000 536
	Journal International doi:10.1111/i1365-246X 2010.04534 x	537
	Han, SC., Shum, C., Jekeli, C., & Alsdorf, D. (2005). Improved estimation of con-	538
	tinental water storage change from GRACE. Geophysical Research Letters, 32,	539
	doi:10.1029/2005GL022382.	540
	Hansen, M. C., Roy, D. P., Lindquis, E., Adusel, B., Justice, C. O., & Altstatt, A. (2008). A method	541
	for integrating MODIS and Landsat data for systematic monitoring of forest cover and	542
	change in the Congo Basin. Remote Sensing of Environment, 112, 2495–2513.	543
	Julig, H., Alsdori, D., Lee, H., Higg, M., & Fewtrell, I. (2010). Hydrogeomorphic modeling of the Congo interfluxial wetlands	044 545
	ACII fall meeting abstract	546
	Jung H Hamski I Durand M Alsdorf D Hossain F Lee H et al (2010) Character-	547
	ization of complex fluvial systems using remote sensing of spatial and temporal	548
	water level variations in the Amazon, Congo, and Brahmaputra Rivers. Earth Sur-	549
	face Processes and Landforms, 35, 294–304.	550
	Laporte, N., Justice, C., & Kendall, J. (1995). Mapping the dense humid forest of Cameroon	551
	and Zaire using AVHRR satellite data. International Journal of Remote Sensing, 16,	552
	1127–1145.	553
	Laraque, A., Mahe, G., Orange, D., & Marieu, B. (2001). Spatiotemporal variations in hy-	554
	arological regimes within Central Arrica during the XXth Century. Journal of Hydrol-	000 556
	Lee H Durand M Jung H Alsdorf D Shum C & Sheng Y (2010) Characterization	557
	of surface water storage changes in Arctic lakes using simulated SWOT measure-	558
	ments. International Journal of Remote Sensing, 31, 3931–3953.	559
	Lehner, B., Verdin, K., & Jarvis, A. (2008). New global hydrography derived from space-	560
	borne elevation data. Eos, 89, 93–94.	561
_		

H. Lee et al. / Remote Sensing of Environment xxx (2011) xxx-xxx

- 562McCollum, J. R., Gruber, A., & Ba, M. B. (2000). Discrepancy between gauge and satellite estimates of rainfall in equatorial Africa. Journal of Applied Meteorology, 39, 666–679. 563
- 564565
- Melack, J. M., & Engle, D. (2009). An organic carbon budget for an Amazon floodplain lake. Verhandlungen Internationale Vereinigung Limnologie, 30, 1179–1182.
 Nicholson, S. E., Some, B., McCollum, J., Nelkin, E., Klotter, D., Berte, Y., et al. (2003). Val-idation of TRMM and other rainfall estimates with a high-density gauge dataset for 566 567 West Africa. Part I: Validation of GPCP rainfall product and pre-TRMM satellite and blended products. Journal of Applied Meteorology, 42, 1337–1354.
 Pavlis, N. K., Holmes, S. A., Kenyon, S. C., & Factor, J. K. (2008, April). An earth gravita-tional model to degree 2160: EGM2008. Vienna, Austria: EGU-2008 (also see). 568 569
- 570571http://earth-info.nga.mil/GrandG/wgs84/gravitymod/egm2008/index.html 572
- Swenson, S., & Wahr, J. (2006). Post-processing removal of correlated errors in GRACE data. Geophysical Research Letters, 33, L08402, doi:10.1029/2005GL025285. 573574
- 589

- Tapley, B. D., Bettadpur, S., Ries, J., Thompson, P., & Watkins, M. (2004). GRACE mea- 575 surements of mass variability in the Earth system. *Science*, 305, 503–505. 576
- Verdin, K. L., & Verdin, J. P. (1999). A topological system for delineation and codification 577
- of the earth's river basins. Journal of Hydrology, 218, 1–12. 578 Wahr, J., Molenaar, M., & Bryan, F. (1998). Time variability of the earth's gravity field: 579 Hydrological and oceanic effects and their possible detection using GRACE. Journal 580
- Hydrological and oceanic effects and then possible detection using Greek, Journal 500 of Geophysical Research, 103, 30205–30229.
 Williams, E. R., & Satori, G. (2004). Lightning, thermodynamics and hydrological comparison of the two tropical continental chimneys. Journal of Atmospheric and Solar-583 Terrestrial Physics, 66, 1213–1231, doi:10.1016/j.jastp. 2004.05.015.
- Yamazaki, D., Kanae, S., Kim, H., & Oki, T. (2011). A physically based description of 585 floodplain inundation dynamics in a global river routing model. *Water Resources* 586 *Research*, 47, W04501, doi:10.1029/2010WR009726. 587
 - 588