

Towards ecosystem accounting: a comprehensive approach to modelling multiple hydrological ecosystem services

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Abstract

Ecosystem accounting is an emerging field that aims to provide a consistent approach to analysing environment-economy interactions. One of the specific features of ecosystem accounting is the distinction between the capacity and the flow of ecosystem services. Ecohydrological modelling to support ecosystem accounting requires considering among others physical and mathematical representation of ecohydrological processes, spatial heterogeneity of the ecosystem, temporal resolution, and required model accuracy. This study examines how a spatially explicit ecohydrological model can be used to analyse multiple hydrological ecosystem services in line with the ecosystem accounting framework. We use the Upper Ouémé watershed in Benin as a test case to demonstrate our approach. The Soil Water and Assessment Tool (SWAT), which has been configured with a grid-based landscape discretization and further enhanced to simulate water flow across the discretized landscape units, is used to simulate the ecohydrology of the Upper Ouémé watershed. Indicators consistent with the ecosystem accounting framework are used to map and quantify the capacities and the flows of multiple hydrological ecosystem services based on the model outputs. Biophysical ecosystem accounts are subsequently set up based on the spatial estimates of hydrological ecosystem services. In addition, we conduct trend analysis statistical tests on biophysical ecosystem accounts to identify trends in changes in capacity of the watershed ecosystems to provide service flows. We show that the integration of hydrological ecosystem services in an ecosystem accounting framework provides relevant information on ecosystems and hydrological ecosystem services at appropriate scales suitable for decision-making.

1 Introduction

Ecosystem accounting provides a systematic framework to link ecosystems to economic activities (Boyd and Banzhaf, 2007; Maler et al., 2008; EC et al., 2013; Edens and Hein, 2013; Obst et al., 2013). Specifically, ecosystem accounting aims to integrate the concept of ecosystem services in a national accounting context as described in UN et al. (2009). There is increasing interest in ecosystem accounting as a new, comprehensive tool for environmental monitoring and management (Obst et al., 2013). The recently released System of Environmental-Economic Accounting (SEEA)-Experimental

1 Ecosystem Accounting guideline (EC et al., 2013) provides guidelines for setting up both biophysical
2 and monetary ecosystem accounts. Biophysical accounting for ecosystem services forms the basis for
3 monetary accounting.

4 Ecosystem services are the contributions of ecosystems to human welfare (TEEB, 2010;EC et al.,
5 2013). Hydrological ecosystem services, specifically, are the contributions to human benefits produced
6 by the effects of terrestrial ecosystem components on freshwater as it moves through the landscape.
7 Terrestrial ecosystem components directly modify different attributes (such as quantity, quality,
8 location and timing) of various ecohydrological processes resulting in augmentation or degradation of
9 these processes (Brauman et al., 2007). Factors such as the presence of beneficiaries (Boyd and
10 Banzhaf, 2007), spatiotemporal accessibility (Fisher et al., 2009), and management pressure (Schröter
11 et al., 2014) then determine if the ecohydrological processes constitute hydrological ecosystem
12 services. Hydrological ecosystem services are diverse and can be broadly classified into five
13 categories; improvement of extractive water supply, improvement of in-stream water supply, water
14 damage mitigation, provision of water related cultural services, and water-associated supporting
15 services (Brauman et al., 2007). Production of these services underlies water and food security and the
16 protection of human lives and properties.

17 Biophysical accounting for hydrological ecosystem services allows for the organisation and analysis of
18 biophysical data on these services at different spatial and temporal scales suitable for the development,
19 monitoring and evaluation of public policy (EC et al., 2013). Biophysical accounting also allows for
20 the distinction between the flow of hydrological ecosystem services and the capacity of watershed
21 ecosystems to provide service flows (EC et al., 2013). Service flow is the contribution in space and
22 time of an ecosystem to either a utility function (e.g. private household) or a production function (e.g.
23 crop production) that leads to a human benefit, whereas service capacity is a reflection of ecosystem
24 condition and extent at a point in time, and the resulting potential to provide service flows (EC et al.,
25 2013;Edens and Hein, 2013). For hydrological ecosystem services, high service capacity areas and
26 high service flow areas may occur in different points or areas in space (Fisher et al., 2009) making the
27 need for their empirical distinction and separate spatial characterization crucial for land and watershed
28 management.

29 Many approaches have been used for modelling, mapping and quantifying hydrological ecosystem
30 services (e.g. Le Maitre et al., 2007;Naidoo et al., 2008;Liquete et al., 2011;Maes et al., 2012;Notter et
31 al., 2012;Willaarts et al., 2012;Leh et al., 2013;Liu et al., 2013;Terrado et al., 2014 for an overview).
32 For ecosystem accounting, however, key aspects requiring further research include the modelling of
33 hydrological ecosystem services with adequate spatiotemporal detail and accuracy at aggregated
34 scales, distinguishing between service capacity and service flow, and linking ecohydrological
35 processes (and ecosystem components) to the supply of dependent hydrological ecosystem services.
36 Addressing these issues requires the consideration of among others physical and mathematical
37 representation of ecohydrological processes, spatial heterogeneity of ecosystems, temporal resolution,
38 and required model accuracy (Guswa et al., 2014). Adequate representation of spatial heterogeneity of
39 biophysical environment in ecohydrological models is crucial in ecosystem accounting because spatial
40 units form the basic focus of measurement similar to functions of economic units in national
41 accounting (EC et al., 2013). In addition, if ecosystem accounting is to provide reliable information for
42 the assessment of integrated policy responses at the landscape level, then physical and mathematical
43 representation of model processes should be based on scientific consensus (Vigerstol and Aukema,
44 2011). Furthermore, model results should be accurate and model uncertainties should be understood
45 and reported (Seppelt et al., 2011;Martínez-Harms and Balvanera, 2012). Finally, ecohydrological
46 modelling for ecosystem accounting necessitates the use of continuous simulation watershed models
47 that are able to capture short and long-term temporal variability in ecohydrological processes.

48 Our objective is to present a spatially explicit modelling approach aligned with an ecosystem
49 accounting framework to map and quantify the capacities and the flows of multiple hydrological
50 ecosystem services. We use the Soil Water and Assessment Tool (SWAT), which has been configured
51 with a grid-based landscape discretization and further enhanced to simulate water flow across the
52 discretized landscape units, to simulate the watershed ecohydrology. The model is calibrated and

1 validated and indicators consistent with the ecosystem accounting framework are used to map and
2 quantify the capacities and the flows of multiple hydrological ecosystem services based on the model
3 outputs. Biophysical ecosystem accounts are subsequently set up based on the spatial estimates of
4 hydrological ecosystem services. We use the Upper Ouémé watershed in Benin as a test case to
5 demonstrate our approach. This case-study area was selected because of a relatively high data
6 availability (Judex and Thamm, 2008;AMMA-CATCH, 2014). It is also a microcosm of rural sub-
7 Saharan Africa, where large sections of the population depend on smallholder rainfed agriculture for
8 their livelihood, where groundwater is the major source of drinking water, and where rapid population
9 growth and increasing land use change are prevalent. The hydrological ecosystem services we model
10 and account for are crop water supply, household water supply (groundwater supply and surface water
11 supply), water purification, and soil erosion control. We select these four services because they are
12 critical to food and water security for the population. Agriculture is the major source of income and
13 livelihood in the watershed and is predominantly rainfed. Furthermore, groundwater is the major
14 source of household water use (for both drinking and non-drinking purposes).

17 **2 Description of case-study area**

18 The Upper Ouémé watershed as depicted in Fig. 1 is located in central Benin covering an area of
19 approximately 14 500 km². The natural vegetation is a mosaic of savannah woodland and small forest
20 islands. The Upper Ouémé forest reserve is the major protected forest area in the watershed with an
21 approximate area of 2 420 km². Smallholder rainfed agriculture is the major economic activity and is
22 supported by climatic conditions that are characterized by a unimodal rainfall season from May to
23 October of about 1 250 mm per year. Maize, rice, yam, cassava and millet are some of the important
24 food crops cultivated in this area with cotton being the major cash crop. These crops are
25 predominantly cultivated using rainfed agriculture. The irrigation sector is relatively poorly
26 developed. Rice is mostly cultivated in inland valley lowlands due to their higher water availability,
27 lower soil fragility and higher fertility compared to upland areas (Giertz et al., 2012;Rodenburg et al.,
28 2014). Fertilizer use is increasing in the region and high fertilizer inputs are associated with crops such
29 as maize, rice and cotton (Bossa et al., 2012). An estimated average of 100 – 250 kg ha⁻¹ of fertilizer
30 (NPK + Urea) is applied to cotton, rice and maize (Bossa et al., 2012). With a population of about 400
31 000, there is low demographic density (28 inhabitants km⁻²) in the watershed (Judex and Thamm,
32 2008). However, the population is growing rapidly (about 4% per annum) due to migrants coming
33 from different parts of the country and other neighbouring countries to farm. Rapid population growth
34 has caused the expansion of agricultural areas and led to both deforestation and increasing scarcity of
35 agricultural land (Judex and Thamm, 2008) accompanied by increasing soil degradation due to
36 shortening of the fallow period (Giertz et al., 2012). It has been estimated that there will be nearly
37 complete deforestation in some parts of the Upper Ouémé watershed assuming a 6% per annum
38 expansion of agricultural areas (Orekan, 2007). Conversion of savannah woodland and forests for crop
39 cultivation is mainly through slash and burn techniques. In addition, the population obtain about 90%
40 of their drinking water needs directly from groundwater, with about 5% from small lakes, ponds and
41 rivers collectively referred to in this study as surface water (Judex and Thamm, 2008).

1 **3 Methods**

2 **3.1 Modelling framework**

3 **3.1.1 Model selection**

4 In order to address modelling challenges regarding model process inclusion, spatial heterogeneity,
5 physical and mathematical representation, temporal resolution, and model accuracy, we considered
6 several watershed models and selected the SWAT model (Arnold et al., 1998) to be most appropriate
7 for this study. The SWAT model has a comparative advantage in integrated assessment modelling of
8 ecohydrological interactions that underpin hydrological ecosystem services provision (Vigerstol and
9 Aukema, 2011). The SWAT model is a physically based, ecohydrological model that simulates the
10 impact of land use and land management practices on water, sediments and agricultural chemicals in
11 large complex watersheds (Neitsch et al., 2009). It is a continuous simulation watershed model
12 operated at a daily time-step. In the SWAT model, a watershed can be spatially discretized using three
13 approaches. They are grid cells, representative hillslopes, and hydrologic response units (HRUs)
14 (Arnold et al., 2013). The HRU-based discretization is the most popular and most geographic
15 information system interfaces are set up to use this discretization (e.g. ArcSWAT). Each HRU is a
16 lumped area within a subwatershed that is comprised of unique land cover, soil and management
17 combinations (Neitsch et al., 2009). The hydrological cycle is divided into two phases. The first is the
18 land phase which controls the amount of water, sediment, nutrient and pesticide loadings to the main
19 channel in each subwatershed. Land phase processes include; weather, hydrology (canopy storage,
20 infiltration, evapotranspiration, surface runoff, lateral subsurface flow, return flow) plant growth,
21 erosion, nutrients and management operations (Neitsch et al., 2009). Surface runoff, lateral flow and
22 return flow from the land phase are then routed through the channel network of the watershed to the
23 outlet in the second phase called the routing phase. This phase also includes processes such as
24 sediment and nutrient routing (Neitsch et al., 2009).

25 **3.1.2 Model modification**

26 The SWAT model used in this study had two major modifications; the first one was a model process
27 modification whereas the second one was a modification of the spatial discretization scheme. The
28 process modification involved the incorporation of a landscape routing sub-model that simulates
29 surface water, lateral and groundwater flow interactions across discretized landscape units. This sub-
30 model was developed and incorporated into the standard SWAT model by Volk et al. (2007) and
31 Arnold et al. (2010). The modified model, SWAT Landscape model, addresses an inherent weakness
32 in the standard SWAT model. The standard SWAT model uses an HRU-based discretization and
33 transported water, sediment, nutrient and pesticide loadings from upstream HRUs are routed directly
34 into stream channels bypassing downstream HRUs (Gassman et al., 2007; Volk et al., 2007; Arnold et
35 al., 2010; Bosch et al., 2010). Therefore, the impact of management of upstream HRUs on downstream
36 HRUs cannot be sufficiently assessed. This weakness is a result of the lack of spatial interactions
37 among different HRUs in the land phase of the hydrological cycle (Neitsch et al., 2009). The SWAT
38 Landscape model addresses this weakness by using a constant flow separation ratio to partition
39 landscape and channel flow in each HRU (Arnold et al., 2010). The channel flow portion is routed
40 through the stream network whereas the landscape flow portion is routed from upstream HRUs to
41 downstream HRUs.

42 The second modification was a change from the HRU-based spatial discretization scheme of the
43 standard SWAT model to a grid-based landscape discretization scheme. We set up the SWAT
44 Landscape model with this grid-based landscape discretization using SWATgrid (Rathjens and Oppelt,
45 2012). The grid-based setup of the SWAT Landscape model uses a modified topographic index to
46 estimate spatially distributed proportions of landscape and channel flow (Rathjens et al., 2014), unlike
47 the HRU-based setup which uses a constant flow separation ratio. A new parameter called the drainage
48 density factor controls the spatially distributed flow separation in the SWATgrid setup (Rathjens et al.,
49 2014). This parameter can be adjusted during calibration. For this study, the grid-based setup of the

1 SWAT Landscape model was used to delineate the watershed into spatially interacting grid cells. Flow
2 paths were determined from the DEM and the digital landscape analysis tool TOPAZ (Garbrecht and
3 Martz, 2000) and runoff from a grid cell flowed to one of eight adjacent cells (Rathjens et al., 2014). A
4 detailed description of the two modifications can be found in Arnold et al. (2010) and Rathjens et al.
5 (2014).

6 **3.1.3 Model input data**

7 A combination of spatial and non-spatial input data from a variety of sources were used to set up the
8 model. The spatial input data are described in Table 1. A 30m digital elevation model (DEM) was
9 obtained from the National Aeronautics and Space Administration (NASA) ASTER Global Digital
10 Elevation Map to generate stream network, watershed configurations and to estimate topographic
11 parameters. Land cover and soil maps were obtained from the “Integrated Approach to Efficient
12 Management of Scarce Water Resources in West Africa” (IMPETUS) project database (Judex and
13 Thamm, 2008). The land cover map had been derived from classification of LANDSAT-7 ETM+
14 satellite image. Gridded daily precipitation data were obtained from the “African Monsoon and
15 Multidisciplinary Analysis–Coupling the Tropical Atmosphere and the Hydrological Cycle” (AMMA-
16 CATCH) database (AMMA-CATCH, 2014) and gridded temperature data were obtained from Climate
17 Research Unit (CRU) TS 3.21 database (Jones and Harris, 2013). Data on groundwater and surface
18 water household consumption (including drinking and non-drinking purposes) were obtained from the
19 IMPETUS project database. These had been derived from national census and household surveys in
20 about 200 towns and communities within the watershed (INSAE, 2003;Hadjer et al., 2005;Judex and
21 Thamm, 2008). For our study area, per capita groundwater consumption was 19 litres per day per
22 person and per capita surface water consumption was 14 litres per day per person (INSAE,
23 2003;Hadjer et al., 2005).

24 **3.1.4 Model configuration and performance evaluation**

25 The initial model setup was carried out with the ArcSWAT interface, which is based on an HRU
26 configuration. This was essential for generating input data for the grid-based configuration.
27 Simulations of the HRU-based SWAT model were conducted for the period 1999 to 2012. The first
28 two years (1999 and 2000) served as warm-up period for the model to assume realistic initial
29 conditions. Potential evapotranspiration was modelled with the Hargreaves method (Hargreaves et al.,
30 1985) and water transfers for households were modelled as constant extraction rates from shallow
31 aquifers (groundwater extractions) and streams (surface water extractions). The Soil Conservation
32 Service curve number approach was used to model surface runoff and daily curve number value was
33 calculated as a function of plant evapotranspiration (Neitsch et al., 2009). The HRU-based SWAT
34 model was first calibrated and validated with streamflow data before calibration and validation of
35 sediment and nitrogen loads. A split-time calibration and validation technique was carried out on the
36 HRU-based model using the Sequential Uncertainty Fitting (SUFI-2) optimization algorithm of the
37 SWAT-Calibration and Uncertainty Program (Abbaspour et al., 2008). For calibration and validation
38 of streamflow, we used daily observed streamflow data from 11 monitoring stations within the
39 watershed. These stations had drainage areas of varying spatial scale to capture watershed-scale and
40 subwatershed-scale ecohydrological processes. Calibration was mostly from 2001 to 2007 and
41 validation was from 2008 to 2011. To evaluate transport of sediments and nutrients, the model was
42 further calibrated with weekly measured sediment and organic nitrogen load data. Two years of data
43 (2008 to 2009) were available from a single monitoring station, Beterou station. Sediment and organic
44 nitrogen load data for 2008 were used for the calibration whereas data for 2009 were used for the
45 validation.

46 The calibrated and validated input parameter sets from the HRU-based setup were transferred to the
47 grid-based setup of the SWAT Landscape model using the SWATgrid interface (Rathjens and Oppelt,
48 2012). Given the computational resources and time required to run a grid-based setup of the SWAT
49 Landscape model at a higher spatial resolution (e.g. 1 ha) for a relatively large watershed such as the
50 Upper Ouémé (Arnold et al., 2010;Rathjens and Oppelt, 2012), we resampled the DEM, soil and land

cover data to a resolution of 500m × 500m. The resampling allowed for a balance between computational efficiency during model simulation and maintenance of accurate spatial representation of landscape patterns. Grid-based simulations of the SWAT Landscape model were conducted for the period 1999 to 2012. The first two years served as model warm-up period. The grid-based setup of the SWAT Landscape model was then calibrated manually by adjusting only the drainage density factor parameter. The full calibrated parameter values are listed in Table 3. Three quantitative statistics recommended by Moriasi et al. (2007) were selected to evaluate model performance: Nash-Sutcliffe efficiency (NSE), percent bias (PBIAS), and ratio of the root mean square error to the standard deviation of measured data (RSR). Nash-Sutcliffe efficiency is a normalized statistic that determines the relative magnitude of the residual variance compared to the measured data variance (Nash and Sutcliffe, 1970); PBIAS measures the average tendency of the simulated data to be larger or smaller than their observed counterparts (Gupta et al., 1999); RSR standardizes root mean squared error using the observations standard deviation (Moriasi et al., 2007).

3.2 Spatial assessment of hydrological ecosystem services

For each hydrological ecosystem service, two appropriate indicators were selected to model service flow and service capacity. Computations were made for each grid cell enabling the model to reflect spatial differences in service flow and in service capacity. The selected hydrological ecosystem services and their service flow and service capacity indicators are shown in Table 2.

3.2.1 Crop water supply

An important hydrological ecosystem service input to crop production in rainfed agricultural systems is the provision of plant available water by ecohydrological processes that affect the soil water balance (Pattanayak and Kramer, 2001; IWMI, 2007; Zang et al., 2012). Crop water stress is a major limitation to crop production in rainfed agricultural systems (IWMI, 2007). We modelled service flow in croplands, which is referred to in this study as upland agricultural fields, and in inland valley lowlands (Rodenburg et al., 2014). Whereas inland valley lowlands in the study area are predominantly used for rice cultivation, the land cover input data did not differentiate the types of crops grown in upland agricultural fields. For our simulations we assumed that all upland agricultural fields were used for only maize cultivation (which is the most common crop in our study area in terms of extent of cultivated land area). For maize cultivation, the growing period (GP), i.e. the time-period between crop establishment and harvesting, was 103 days whereas GP for rice cultivation was 123 days. For both maize and rice, crop establishment was in the month of June. Service flow of crop water supply was modelled as the total number of days during a growing period in which there was no water stress (i.e., days when the total plant water uptake was sufficient to meet maximum plant water demand). Service flow depends on the specific type of crop cultivated. This approach is based on the model output variable, daily water stress, and is a modification of Notter et al. (2012). For each day, the model used Eq. (1) to compute water stress for a given grid cell, j (Neitsch et al., 2009). After model simulation, service flow was computed using Eq. (2).

$$W_{strs, j} = 1 - T_{act, j} / T_{max, j}, \quad (1)$$

where W_{strs} is daily water stress, T_{act} is plant water uptake or actual transpiration (mm), and T_{max} is maximum plant water demand or maximum transpiration (mm).

$$S_f, j = N(d_1, d_2, \dots, d_n | W_{strs} = 0)_j, \quad (2)$$

where S_f is the service flow (days GP⁻¹), N is the number of days d_1 to d_n , when W_{strs} was zero.

Service capacity on the other hand was modelled as the total number of days in a year when the sum of actual evapotranspiration and the amount of residual moisture added to the soil profile equalled or exceeded potential evapotranspiration. For a given spatial unit, this gives an indication of the number

1 of days when potentially there will be no crop water stress irrespective of the crop type to be
 2 cultivated. This approach has management relevance. Our approach was based on the commonly used
 3 method, FAO (1978) and FAO (1983), for determining the length of growing period in rainfed
 4 agricultural systems. Unlike that method where moisture supply was based on precipitation, moisture
 5 supply in our approach was based on simulated spatiotemporal soil moisture dynamics. We used this
 6 approach for our study because at the local scale terrestrial ecosystem components have very little
 7 effect on precipitation attributes such as quantity, location, timing etc. For a given day, the SWAT
 8 model used Eq. (3) to compute water balance. From the water balance components we computed the
 9 total available soil moisture and subsequently calculated potential water stress using Eq. (4). Service
 10 capacity of crop water supply was then computed using Eq. (5).

$$\Delta SW_i = \sum_{i=1}^n (R_{day} - Q_{surf} - ET_a - W_{seep} - Q_{gw}), \quad (3)$$

11 where ΔSW is the amount of residual moisture added to the soil profile on day i (mm); n is number of
 12 days in the year; R_{day} is the amount of precipitation (mm); Q_{surf} is the amount of surface runoff (mm);
 13 ET_a is actual evapotranspiration (mm); W_{seep} is percolation exiting soil profile (mm); Q_{gw} is return flow
 14 (mm).

$$15 \quad W_{pstrs} = 1 - [(\Delta SW + ET_a)/ET_p] \quad \text{if } \Delta SW > 0, \quad (4)$$

16 W_{pstrs} is potential daily water stress; ΔSW is the amount of residual moisture added to the soil profile
 17 (mm); ET_a is actual evapotranspiration (mm); ET_p is potential evapotranspiration (mm).

$$18 \quad S_c = N(d_1, d_2, \dots, d_n \mid W_{pstrs} \leq 0), \quad (5)$$

19 where S_c is service capacity (days yr⁻¹); N is the number of days d_1 to d_n in a year when potentially
 20 there will be no water stress.

21 **3.2.2 Household water supply**

22 This hydrological ecosystem service refers to the amount of water extracted before treatment for
 23 household consumption (drinking and non-drinking purposes) (EC et al., 2013). This measurement
 24 boundary excluded other sources of water (e.g. tap water) where economic agents or inputs (e.g. water
 25 treatment facilities) were used to modify the state of the water resources before household
 26 consumption. We acknowledge that inflows to reservoirs of water distribution and processing facilities
 27 that deliver tap water can be considered as a hydrological ecosystem service. However, we excluded
 28 this from our study. This is because in our study area, the population obtain about 90% of their
 29 drinking water needs from groundwater, with about 5% from small lakes, ponds and rivers collectively
 30 referred to in this study as surface water (Judex and Thamm, 2008). A distinction was made between
 31 service capacity and service flow from groundwater, and service capacity and service flow from
 32 surface water.

33 To model service flow from groundwater and surface water, data on water consumption per capita,
 34 village population and water access for about 200 communities within the watershed were used. These
 35 data had been extracted from the 2002 national census (INSAE, 2003) and from household surveys in
 36 the study area (Hadjer et al., 2005). The data represented household water consumption (including
 37 drinking and non-drinking purposes) and lacked information on the actual points of extraction.
 38 Therefore, in modelling the service flow, we assumed that there is a positive spatial correlation
 39 between points of consumption and points of extraction. Furthermore, to estimate village population
 40 from 2003 to 2012, we applied a 4% per annum growth rate (Judex and Thamm, 2008). Water
 41 consumption per capita, however, was kept constant. A population density grid was created using
 42 ArcGIS Kernel Density function (ESRI, 2012) and multiplied by water consumption per capita to
 43 estimate the amount of water consumed per grid cell. The amount consumed per grid cell then gives an
 44 indication of the amount extracted per grid cell.

1 The ecosystem's capacity to support groundwater extractions was modelled as groundwater recharge,
 2 which is the total amount of water entering the aquifers within a specified time-step (e.g. month or
 3 year) (Arnold et al., 2013). The ecosystem's capacity to support surface water extractions, however,
 4 was modelled as the water yield. Water yield is the net amount of water contributed by a grid cell to
 5 the river network within a specified time-step (Arnold et al., 2013). Both groundwater recharge and
 6 water yield are model output variables.

7 3.2.3 Water purification

8 In the Upper Ouémé watershed, fertilizer application is increasing and high fertilizer inputs is
 9 associated with crops such as maize, rice and cotton (Bossa et al., 2012). Increasing fertilizer
 10 application can lead to contamination of groundwater and surface water resources through nutrient
 11 leaching. This poses serious environmental and health risks to beneficiaries of these systems (Tilman
 12 et al., 2002; Wolfe and Patz, 2002). In our study area, groundwater provides over 90% of the total
 13 household water consumption. Water purification is, therefore, an essential ecosystem service in the
 14 Upper Ouémé watershed that increases the quality of groundwater for human consumption as well as
 15 other purposes. One of the challenges in terms of quantifying hydrological ecosystem services is the
 16 identification of management relevant indicators that can be enhanced through management
 17 interventions to augment the service production. For this study, we used soil denitrification as an
 18 indicator of this hydrological ecosystem service. Soil denitrification controls the rate of nitrate
 19 leaching by determining the quantities (after plant uptake) of nitrate available for leaching into
 20 groundwater systems (Jahangir et al., 2012). For example, Kramer et al. (2006) observed that organic
 21 farming supports more active and efficient denitrifier communities leading to a considerable reduction
 22 in nitrate leaching as compared to conventional farming. In this study, the SWAT Landscape model
 23 was used to simulate the complete nitrogen cycle and service flow was estimated directly as the rate of
 24 denitrification, a model output variable. We should emphasize that the SWAT Landscape model does
 25 not explicitly simulate microbial processes and dynamics but rather it simulates the ecohydrological
 26 conditions suitable for denitrification to occur (Boyer et al., 2006). The model, therefore, computes
 27 denitrification as a function of soil moisture content, soil temperature, presence of a carbon source and
 28 nitrate availability using Eq. (6) and Eq. (7) (Neitsch et al., 2009).

$$29 \quad N_{dn} = \text{NO}_3 \cdot \left(1 - \exp \left[-\beta_{dn} \cdot \gamma_{tmp} \cdot C_{org} \right] \right) \quad \text{if } \gamma_{sw} \geq \gamma_{sw, thr} , \quad (6)$$

$$30 \quad N_{dn} = 0 \quad \text{if } \gamma_{sw} < \gamma_{sw, thr} , \quad (7)$$

31 where N_{dn} is the amount of nitrogen lost through denitrification (kg ha^{-1}), NO_3 is the amount of nitrate
 32 in the soil (kg ha^{-1}), β_{dn} is the rate coefficient for denitrification, γ_{tmp} is the nutrient cycling temperature
 33 factor, γ_{sw} is the nutrient cycling water factor, $\gamma_{sw,thr}$ is the threshold value of nutrient cycling water
 34 factor for denitrification to occur, C_{org} is the amount of organic carbon (%). The values of β_{dn} and $\gamma_{sw,thr}$
 35 are user defined values and were adjusted during calibration; β_{dn} was 1.4 and $\gamma_{sw,thr}$ was 1.1.

36 Service capacity was estimated as the denitrification efficiency, which in this study was computed
 37 using Eq. (8). When the ecohydrological conditions required for denitrification are present, the rate of
 38 denitrification (service flow) is determined by the amount of nitrate available in the soil. Unlike other
 39 land cover types (which only receive nitrogen or nitrates from wet deposition or from overland flow),
 40 cropland areas receive additional nitrogen or nitrates through fertilizer application. Therefore, for a
 41 given grid cell, denitrification efficiency determines the proportion of the total nitrate that is
 42 denitrified. As a measure of service capacity, denitrification efficiency gives an indication of the
 43 suitability of a spatial unit for denitrification.
 44

$$45 \quad \text{DN}_{\text{eff}} = (N_{dn}/N_{\text{total}}) \cdot 100 , \quad (8)$$

1 where DN_{eff} is the denitrification efficiency (%), N_{dn} is the amount of nitrogen lost through
2 denitrification in the time-step (kg ha^{-1}), N_{total} is the total amount of nitrogen available (e.g. through
3 fertilizer application, wet deposition etc.) in the time-step (kg ha^{-1}).

4 **3.2.4 Soil erosion control**

5 Controlling soil erosion in the watershed has numerous benefits including maintaining soil fertility,
6 preventing river sedimentation, and downstream water quality. There are inherent physical soil and
7 landscape properties such as soil erodibility and slope that affect soil erosion (Williams, 1975).
8 However, we focussed on the role of vegetation cover in controlling soil erosion. Service flow was
9 modelled as the actual reduction in soil loss produced by the existing vegetation cover and was
10 computed using Eq. (9).

$$11 \quad SD_{\text{rtd}} = S_{\text{yld, pot}} - S_{\text{yld}}, \quad (9)$$

12 where SD_{rtd} is the reduction in soil loss produced by the existing vegetation cover ($\text{metric tons ha}^{-1}$),
13 $S_{\text{yld, pot}}$ is the maximum potential soil loss in the absence of vegetation cover ($\text{metric tons ha}^{-1}$), and S_{yld}
14 is the soil loss under prevailing vegetation cover and land management practices ($\text{metric tons ha}^{-1}$).
15 Both $S_{\text{yld, pot}}$ and S_{yld} were computed with the Modified Universal Soil Loss Equation (Williams, 1975)
16 incorporated in the SWAT Landscape model.

17 For service capacity of soil erosion control, we used the maximum potential reduction in soil loss
18 produced by the vegetation cover as an indicator. This maximum potential reduction in soil loss
19 (maximum potential soil retention) can be said to be equal to the maximum potential soil loss. For
20 example, for a specified spatial unit, if the maximum potential soil loss in the absence of the
21 vegetation cover is estimated as $2 \text{ metric tons ha}^{-1} \text{ yr}^{-1}$ then it indicates that the potential reduction in
22 soil loss due to the vegetation cover cannot be greater than $2 \text{ metric tons ha}^{-1} \text{ yr}^{-1}$. The maximum
23 potential soil loss was modelled assuming there was no vegetation cover (e.g. Leh et al., 2013; Terrado
24 et al., 2014).

26 **3.3 Accounting for hydrological ecosystem services**

27 Biophysical ecosystem accounts are the basis for monetary accounting and were set up in accordance
28 with SEEA-Experimental Ecosystem Accounting guidelines (EC et al., 2013). We defined 11
29 Subwatershed Ecosystem Accounting Units (SEAU) as the spatial scales of aggregation. We set up
30 annual biophysical service capacity and service flow accounts for each SEAU. The 11 SEAU were
31 defined from a total of 44 subwatersheds based on the drainage areas of streamflow monitoring
32 stations within the watershed. The monitoring stations are listed in Table 4. The 44 subwatersheds
33 were delineated from the ASTER Global Digital Elevation Map as part of the initial model setup with
34 ArcSWAT. Some monitoring stations with smaller drainage areas were nested within those with larger
35 drainage areas. In such cases the SEAU was defined as the drainage area of the nested monitoring
36 station because we wanted to set up spatially disaggregated accounts. Large drainage areas of other
37 monitoring stations had nested subwatersheds within them that were ungauged. In these cases also, the
38 SEAU was defined as the nested subwatershed. For each SEAU, the spatial estimates of service
39 capacity-load per grid cell ($500\text{m} \times 500\text{m}$) and service flow-load per grid cell ($500\text{m} \times 500\text{m}$) that had
40 been computed in Sect. 3.2 were then aggregated.

41 A key motivation for ecosystem accounting is to provide information for tracking changes in
42 ecosystems and linking those changes to economic and other human activities (EC et al., 2013). Trend
43 analysis statistical tests were conducted on the total annual values (or total seasonal values for crop
44 water supply) of service capacity accounts in each SEAU. Trend analysis determines if the changes in
45 service capacity over time are due to random variability or statistically significant and consistent
46 changes. This was conducted using the non-parametric Mann-Kendall test for trend. The Mann-
47 Kendall test for trend statistically determines if there is a monotonic upward or downward trend of a

1 variable over time. A trend was detected if temporal variation in service capacity was statistically
2 significant at 5% significance level (P-value < 0.05). If a trend was detected, the Mann-Kendall
3 statistic and Sen's slope estimator were calculated. The Mann-Kendall statistic is a measure of the
4 strength and direction of a trend, whereas Sen's slope estimator is a measure of the magnitude of a
5 trend.
6
7

8 **4 Results**

9 **4.1 SWAT Landscape model calibration and validation results**

10 Table 4 shows the statistical results of the model calibration and validation and Fig.2 and Fig.3 show
11 the graphical results. There are no established absolute criteria for judging model performance. For
12 this study, we used the criteria recommended by Moriasi et al. (2007). A watershed model is said to be
13 performing satisfactorily if the NSE > 0.50 and RSR < 0.70, PBIAS within the range -25 to 25 for
14 streamflow, -55 to 55 for sediment, and -70 to 70 for nutrients. At different spatial scales (e.g. Affont-
15 Pont, 1 172 km²; Igbomakoro, 2 309 km²; Beterou, 10 046 km²), the model simulated hydrological
16 processes satisfactorily as shown in Fig. 2. Seven out of 11 stations recorded NSE values greater than
17 0.5 during model validation of streamflow. Even though the NSE values for some monitoring stations
18 were less than 0.5, all but one were greater than 0.0, indicating that the simulated streamflow was still
19 a better predictor than the mean of the observed values. Monitoring stations with larger drainage areas
20 recorded higher NSE values than stations with smaller drainage areas. The PBIAS values in Table 4
21 show the level of bias in simulated streamflow. A negative PBIAS value indicates model
22 overestimation whereas a positive PBIAS value indicates model underestimation. The validation
23 results show that the model largely underestimated streamflow at upstream stations and overestimated
24 it downstream. The RSR results show varying levels of residual variation indicating the level of errors
25 in simulated streamflow as compared to observed streamflow. The closer the RSR value is to zero, the
26 lower the level of residual variation in simulated streamflow. During model validation, five stations
27 recorded RSR values lower than 0.7. For sediment and nitrogen transport processes, the model
28 performed satisfactorily. The statistical and graphical results of sediment load and organic nitrogen
29 load during calibration and validation are shown in Fig. 3 and Table 4.
30

31 **4.2 Spatial patterns of hydrological ecosystem services**

32 Water supply by soil moisture is essential to reduce crop water stress in rainfed agricultural systems. If
33 all other factors for crop growth (such as nutrients and temperature) remain constant, then a higher
34 service capacity and higher service flow result in a higher crop yield. Computations of crop water
35 supply were spatially restricted to upland agricultural fields. High service flow indicates the suitability
36 of a spatial unit under assumed maize cultivation whereas high service capacity indicates the potential
37 suitability for crop cultivation irrespective of the crop type and not maize alone. The results of service
38 capacity are indicative of the least number of days during a year crops would not experience water
39 stress. Figure 4 reveals high spatial variability in service capacity and service flow in upland
40 agricultural fields. Mean annual values of service capacity in upland agricultural fields ranged from 51
41 to 146 days yr⁻¹ with a watershed-wide mean of 93 days yr⁻¹ and standard deviation of 24 days yr⁻¹.
42 The spatial distribution of mean annual values of service capacity and service flow in inland valley
43 rice fields are not shown because of their significantly low total area (less than 1 % of total cropland
44 area). Mean annual values of service capacity in inland valleys ranged from 92 to 136 days yr⁻¹ with a
45 watershed-wide mean of 124 days yr⁻¹ and standard deviation of 9 days yr⁻¹. Mean seasonal values of
46 service flow in inland valleys ranged from 67 to 123 days GP⁻¹ with a watershed-wide mean of 117
47 days GP⁻¹ and a standard deviation of 12 days GP⁻¹. Overall, more than 95% (approximately 1 050 ha)
48 of inland valley rice fields recorded mean seasonal values of service flow of at least 90 days whereas

1 less than 25% (approximately 36 000 ha) of upland agricultural areas recorded mean seasonal values
2 of service flow of at least 90 days.

3 The spatial distribution of mean annual values of service capacity and service flow of groundwater
4 supply and surface water supply are shown in Fig. 5 and Fig. 6 respectively. Groundwater is the major
5 source of water for household consumption (drinking and non-drinking purposes) with the service
6 flow (groundwater extraction) significantly higher than service flow of surface water supply (surface
7 water extraction). High service flows of groundwater supply are concentrated in the most populous
8 towns in the watershed. However, service flows in Parakou, which is the most populous city in the
9 watershed, are relatively lower than other areas such as Djougou. This is because the population in
10 Parakou depend mainly on tap water sources. Service capacity of groundwater supply exhibited high
11 spatial variability. High values of service capacity were concentrated in the south-western part of the
12 watershed. For service capacity of surface water supply, Fig. 6 shows areas with a high propensity for
13 generating water yield. These areas, referred to as Hydrologically Sensitive Areas (HSAs) (Agnew et
14 al., 2006), were not peculiar to a particular land cover type. They occurred in almost all land cover
15 types. They occurred more frequently in savannah woodland and shrubland because approximately
16 80% of the total watershed area is either one of this land cover type.

17 Water purification modelled as denitrification is essential to control the quantities of nitrate available
18 for leaching and contaminating groundwater resources (Jarvis, 2000; Jahangir et al., 2012). Service
19 capacity was measured as the percentage of nitrate that is denitrified and service flow was the rate of
20 denitrification. The spatial distribution of mean annual values of service capacity and service flow of
21 water purification are distinctly concentrated in the northern and eastern parts of the watershed with
22 the south-western parts recording zero values (Fig. 7). All barren land cover types also recorded zero
23 values of service capacity and service flow. The zero values recorded are a result of the lack of soil
24 saturation conditions and not the lack of nitrate availability. Soil saturation induces soil anaerobic
25 conditions required for denitrification to take place. In areas where denitrification was recorded, the
26 highest mean annual values of service flow were recorded in inland valley rice fields ($12 \text{ kg ha}^{-1} \text{ yr}^{-1}$)
27 and grasslands ($7 \text{ kg ha}^{-1} \text{ yr}^{-1}$). The highest mean annual values of service capacity were also recorded
28 in grasslands ($55 \% \text{ yr}^{-1}$) and inland valley rice fields ($35 \% \text{ yr}^{-1}$).

29 The spatial distributions of mean annual values of service capacity and service flow of soil erosion
30 control are shown in Fig. 8. High service capacity indicates high potential for reduction in soil loss
31 produced by the vegetation cover. The service flow, however, is a measure of the actual reduction in
32 soil loss under existing vegetation cover. Under existing vegetation cover, mean annual rate of soil
33 loss in the watershed was recorded at $0.01 \text{ metric tons ha}^{-1} \text{ yr}^{-1}$ (standard deviation of 0.02 metric ton
34 $\text{ha}^{-1} \text{ yr}^{-1}$). The mean annual rate of soil loss in the watershed will increase significantly to 0.05 metric
35 $\text{tons ha}^{-1} \text{ yr}^{-1}$ (standard deviation of $0.07 \text{ metric ton ha}^{-1} \text{ yr}^{-1}$) should there be complete loss of the
36 existing vegetation cover. This value, $0.05 \text{ metric tons ha}^{-1} \text{ yr}^{-1}$ (standard deviation of 0.07 metric ton
37 $\text{ha}^{-1} \text{ yr}^{-1}$), can also be interpreted as the maximum potential reduction in soil loss (service capacity) that
38 can be produced by the existing vegetation cover. Under existing vegetation cover and management
39 conditions, however, the actual reduction in soil loss (service flow) was recorded at a watershed-wide
40 mean annual value of $0.04 \text{ metric tons ha}^{-1} \text{ yr}^{-1}$ (standard deviation of $0.07 \text{ metric ton ha}^{-1} \text{ yr}^{-1}$). For
41 both service capacity and service flow, only about 0.04% of the total area of the watershed recorded
42 mean annual values greater than $1 \text{ metric ton ha}^{-1} \text{ yr}^{-1}$. These areas had the steepest slopes, indicating
43 the importance of vegetation cover in soil erosion control in these areas. In forested areas, service flow
44 was equal to service capacity, indicating that overall there was no net soil loss from forested areas.

45 46 **4.3 Biophysical ecosystem accounts**

47 The service capacity (Table 5) and service flow (Table 6) ecosystem accounting tables show the
48 distribution of hydrological ecosystem services across the 11 SEAUs for the most current year of
49 simulation, 2012. The total annual values of service capacity correlated with the spatial extent of an
50 SEAU. Larger SEAUs recorded higher values than smaller SEAUs. However, the mean values for

1 service capacity varied depending on the biophysical environment of an SEAU. For example, whereas
2 the Beterou-Ouest SEAU is the largest, the highest mean service capacity of groundwater supply was
3 recorded in Sarmanga and Terou-Igbomakoro SEAUs. This signifies that the rate of groundwater
4 recharge is highest in Sarmanga and Terou-Igbomakoro SEAUs. The service flow table reveals that
5 the ecohydrological conditions required for denitrification (water purification) do not occur in
6 Aguimo, Terou-Igbomakoro, Terou-Wanou, and Wewe SEAUs. However, a total of 77 000 m³ of
7 groundwater was extracted in Terou-Igbomakoro and Wewe SEAUs in 2012. In Aguimo and Terou-
8 Wanou SEAUs, there is currently no groundwater extraction. For crop water supply, the tables also
9 show the total area of land currently under crop cultivation in each SEAU. Upland agricultural areas
10 provide over 99% of total cropland area. The SEAUs with the largest upland agricultural areas did not
11 necessarily record the highest service flow. For example, the highest service flow was recorded in
12 Sarmanga and Terou-Igbomakoro. This signifies that maize cultivation in these SEAUs is less prone to
13 water stress than in any other SEAU.

14 Temporal analysis of ecosystem accounts makes it possible to track ecosystem changes and measure
15 the degree of sustainability, degradation or resilience. Decreasing capacity of ecosystems to sustain
16 human welfare over time is a measure of ecosystem degradation (EC et al., 2013). Figure 9 shows the
17 results of trend analysis statistical tests of service capacities at the SEAU level. Increasing trends were
18 observed in changes in service capacities of water purification, groundwater supply and surface water
19 supply. For ground water supply, increasing trends were observed in all SEAUs. The results in Fig. 9A
20 are of the five SEAUs with the highest Mann-Kendall statistic. Increasing trend in changes in surface
21 water supply was observed in four SEAUs, whereas increasing trend in changes in water purification
22 was observed in only the Aval-Sani SEAU. No trend was observed in changes in service capacity of
23 crop water supply in both upland agricultural fields and inland valleys in all SEAUs. No trend was
24 also observed in changes in service capacity of soil erosion control in all the SEAUs.

25
26

27 **5 Discussion**

28 **5.1 Model uncertainties and limitations**

29 Model results to support decision-making are always associated with a certain degree of uncertainty.
30 Uncertainty in ecohydrological modelling with SWAT may be from input data, model algorithms,
31 model calibration and validation (parameter non-uniqueness) (Abbaspour et al., 2008). The major
32 input uncertainty in our study was a result of resampling of spatial data from fine spatial resolutions to
33 relatively coarse spatial resolutions in order to increase operational feasibility and computational
34 efficiency of the grid-based SWAT Landscape model. We resampled land use/land cover data, DEM
35 and soil map to a spatial resolution of 500m × 500m. Even though the spatial rigor of ecosystem
36 accounting requires that modelling approaches that maintain adequate landscape spatial heterogeneity
37 are more suitable, decisions on choice of spatial resolution should be made with model computational
38 efficiency and operational feasibility in mind. For the SWAT model (and SWAT Landscape model),
39 increasing spatial detail results in a considerable increase in computing time irrespective of the spatial
40 discretization scheme employed (e.g. Arnold et al., 2010;Notter et al., 2012). In our case-study area,
41 over 1 400 000 grid cells are generated at 1 ha resolution requiring over two days for each simulated
42 year on 2.6Ghz and 8GB RAM. Computer storage capacity for the huge data outputs generated may
43 not also be readily available. We acknowledge that in many regions of the world high-resolution
44 spatial input data may not be available at large spatial scales. However, for the grid-based setup of the
45 SWAT Landscape model, when such high-resolution spatial data are available, it may be necessary to
46 compromise spatial explicitness to achieve operational feasibility. This introduces a certain amount of
47 uncertainty with regards to spatial variation in ecohydrological processes, therefore, such decisions
48 should be made taking into consideration the degree of spatial heterogeneity of landscape features. The
49 need to compromise spatial detail for operational feasibility may limit the applicability of this model
50 configuration for larger watersheds.

1 For larger watersheds, it is also extremely difficult to obtain spatially and temporally correct
2 representations of the underlying ecohydrological processes and interactions. To achieve this, there is
3 the need for multi-site calibration at different spatial scales with a sufficient length of time-series of
4 data to capture high and low flow years, annual, seasonal and monthly variations (Santhi et al., 2008).
5 In our study, the model underestimated streamflow (especially peak flow) at some monitoring stations
6 whereas at other stations it overestimated streamflow. These biases in streamflow estimation lead to
7 error propagation in the other components of the water balance such as soil moisture and actual
8 evapotranspiration. Whereas the use of 11 years of daily streamflow data from 11 monitoring stations
9 in the Upper Ouémé watershed reduces the uncertainties of modelled results, data for calibration and
10 validation of sediment and nitrogen loads may not have been sufficient to enable the model to more
11 accurately represent sediment and nitrogen transport processes. In evaluating model performance of
12 sediment and nitrogen transport processes, we used only two years of data from a single monitoring
13 station. Without multi-site calibration and validation, there remain large uncertainties in modelled
14 results of sediment and transport processes at different spatial scales. In addition, without long term
15 temporal validation, there remain large uncertainties in the ability of the model to capture annual
16 variability in these transport processes. Even with sufficient length of time series of multi-site data for
17 calibration and validation, the problem of parameter non-uniqueness inherent in complex watershed
18 models such as the SWAT model also introduces a degree of uncertainty in modelled results.
19 Parameter non-uniqueness refers to the reproduction of similar observed ecohydrological signals by
20 different input parameter sets. Therefore even for so called calibrated and validated SWAT models,
21 there is always a degree of uncertainty introduced by parameter non-uniqueness. To reduce error
22 propagation, non-uniqueness and consequently reduce parameter uncertainty requires the use of
23 comprehensive data on different fluxes, loads and ecohydrological processes such as crop yield, soil
24 moisture, groundwater level and evapotranspiration (Abbaspour et al., 2008) that are most of the time
25 not readily available.

26
27 For this study, we used soil denitrification as an indicator of water purification service. Quantifying
28 denitrification at watershed and subwatershed scales requires the use of models such as SWAT. It
29 involves the simulation of a complex set of processes controlling denitrification that can broadly be
30 classified as the prerequisite environmental/ecohydrological conditions, and microbial processes and
31 dynamics. The SWAT model, however, provides only simplified representations of the complex set of
32 processes controlling denitrification and modelled estimates of denitrification rates remain highly
33 uncertain (Boyer et al., 2006). The model only simulates the environmental/ecohydrological
34 conditions and does not explicitly simulate microbial processes and dynamics. There is, therefore, an
35 inherent assumption of spatial homogeneity with regards to denitrifier community species
36 composition, quantities and activities across all land use types. Kramer et al. (2006) reported that
37 specific land use and management types (such as organic, integrated and conventional agriculture)
38 enhance or inhibit soil denitrifier activities affecting the rate of denitrification. In the SWAT model,
39 however, spatial variability in denitrification is determined mainly by spatial variability in
40 ecohydrological and abiotic controlling factors.

41 42 **5.2 Lessons for ecosystem accounting**

43 In ecosystem accounting, detailed and accurate land cover and land use data are important. Apart from
44 their use as inputs in modelling ecosystem services, land cover classes are also used as ecosystem
45 accounting units based on which ecosystem services are aggregated (Remme et al., 2014; Schröter et
46 al., 2014). A single lumped land cover class for agricultural areas or croplands (be it as model input
47 data or ecosystem accounting units) may be suitable when modelling and accounting for other
48 ecosystem services (e.g. Remme et al., 2014; Schröter et al., 2014). However, when modelling and
49 accounting for crop water supply, land cover and land use data with detailed and spatially
50 disaggregated information on the types of crops grown in agricultural areas is needed. This is because
51 different crops have different water requirements (Allen et al., 1998). In rainfed agricultural systems,
52 crop water supply is the major limitation to crop production and is the main factor responsible for low

1 yields in the seasonally dry and semiarid tropics and subtropics (Shaxson and Barber, 2003). However,
2 in many of these regions, land cover and land use data with this level of detail are currently not
3 available. Obtaining such information is complicated by the small plot sizes and cropping patterns
4 varying from year to year. Our study area was no exception. Despite these constraints, the lack of
5 detailed data reduces the accuracy and reliability of modelled results of service flow of crop water
6 supply. In our study area, this limitation resulted in the simulation of only a single crop type in upland
7 agricultural areas. Therefore, the results for service flow of crop water supply should be interpreted in
8 the context of the crop simulated. However, because methodologies such as (Allen et al., 1998) have
9 been used extensively to compute the water requirements of various crops, our approach serves as a
10 reference or baseline from which the service flow of crop water supply of a spatial unit could be
11 estimated if a crop other than maize is grown.

12 A key feature of ecosystem accounting is the distinction between service capacity and service flow.
13 The empirical distinction and separate spatial characterisation of service capacity and service flow is
14 essential in understanding the dynamics of service provision and in planning and devising sustainable
15 management options. The distinction is also important for subsequent monetary valuation. Service
16 capacity and service flow should be based on measurable indicators that have policy and management
17 relevance. Indicators must also be able to represent cause-effect relations. For hydrological ecosystem
18 services, selecting single indicators of service capacity that meet the above requirements and that
19 sufficiently reflect ecosystem condition and their potential to provide service flows is difficult. This is
20 because of the non-linear complex interactions among several ecohydrological processes that each
21 relies on a suite of ecosystem components (van Oudenhoven et al., 2012; Villamagna et al., 2013). In
22 this study, the service capacity indicators of crop water supply and household water supply meet the
23 above requirements. For example, Ennaanay (2006) and Yan et al. (2013) reported that changes in
24 land use and other ecosystem components alter the hydrological cycle, affecting patterns of
25 evapotranspiration, infiltration, water retention, groundwater recharge and water yield. However, for
26 services such as water purification and soil erosion control, the capacity indicators presented in this
27 study are derived indicators and not actual physical processes. Such indicators do not convey
28 information regarding key physical processes and therefore may not have management relevance. In
29 such cases, a key question that arises is if the underlying ecosystem components and processes should
30 be weighted and aggregated to produce one composite indicator for service capacity (Edens and Hein,
31 2013). For example, soil erosion control is a function of surface runoff, slope, soil erodibility, cover
32 and management factors, and support practice factors. Weighing and aggregation of ecosystem
33 components and processes to establish a composite indicator for service capacity, however, is not
34 straightforward and is challenging (Weber, 2007; Stoneham et al., 2012).

35

36 **5.3 Implications for watershed and ecosystem management**

37 Three of the key issues critical for watershed management and land use planning in an agricultural
38 watershed such as the Upper Ouémé are nitrate leaching, non-point source pollution and alteration in
39 streamflow regime. Nitrate leaching contaminates groundwater resources (Jarvis, 2000; Jahangir et al.,
40 2012). Agricultural non-point source pollution leads to pollution of river networks (Agnew et al.,
41 2006). Alteration of streamflow regime affects riverine ecological integrity and downstream water
42 availability (Carlisle et al., 2011). Ecosystem accounting and spatial characterization of hydrological
43 ecosystem services capacity and flow provide relevant information to help address these issues in
44 policy-making. Such analyses can reveal high-risk areas (i.e., areas that would be affected from
45 changes or continued trends in watershed ecohydrology) or high service production areas (i.e., areas
46 that are crucial for maintaining water flow downstream). For example, our analyses reveal areas where
47 the ecohydrological conditions required for denitrification do not occur but where there is currently
48 groundwater extraction. These areas are high-risk areas of groundwater contamination from nitrate
49 leaching. More crucially, there is currently crop cultivation in some of these areas. Agricultural
50 intensification in these areas, therefore, will result in higher nitrate leaching and contamination of
51 groundwater resources

1 Furthermore, the grid-based setup of the SWAT Landscape model enabled us to identify HSAs at a
2 finer spatial resolution. Characterization of the spatiotemporal dynamics of HSAs is essential in
3 controlling non-point source pollution and in maintaining streamflow regime. Hydrologically
4 Sensitive Areas have significant impact on key ecohydrological processes affecting interaction and
5 transport of water, sediment, nutrients and pollutants. They also provide key landscape controls on
6 riverine ecosystem integrity including aquatic flora and fauna and downstream water availability and
7 quality. Agricultural intensification in HSAs has a higher potential of generating agricultural non-point
8 source pollution (Agnew et al., 2006). Land use change in these areas can have a more significant
9 impact on the streamflow regime. Such analyses can form the basis for establishing Payment for
10 Ecosystem Services schemes (PES) (Pagiola and Platais, 2007; Turpie et al., 2008). Watershed PES
11 provides financial support to ecosystem management in high service production areas that are of
12 particular relevance downstream (Lopa et al., 2012; Lu and He, 2014). We acknowledge that detailed
13 ecohydrological modelling is only one of the considerations in establishing a watershed PES. Other
14 considerations include transaction costs and ability to pay of downstream water users. However,
15 ecohydrological modelling can be used to support watershed PES schemes by providing a tool for
16 upstream water managers to monitor the provision of hydrological ecosystem services or by
17 identifying high service production areas that are potentially relevant for a new PES.

20 **6 Conclusion**

21 There are various components involved in ecosystem service delivery that need to be measured in
22 order to better understand the full dynamics of service provision and to devise sustainable
23 management options. Key amongst these components is service capacity and service flow. Empirical
24 distinction of service capacity and service flow of ecosystem services is a distinguishing feature of
25 ecosystem accounting and is the basis for monetary accounting. In the case-study area, we have shown
26 that despite the non-linear complex interactions among several ecohydrological processes, it is
27 empirically feasible to distinguish between service capacity and service flow of hydrological
28 ecosystem services. This requires appropriate decisions regarding physical and mathematical
29 representation of ecohydrological processes, spatial heterogeneity of ecosystems, temporal resolution,
30 and required model accuracy. The service flows we modelled are the contributions in time and space
31 of ecosystems to productive and consumptive human activities leading to human benefits, whereas the
32 service capacities we modelled reflect ecosystem condition and extent at a point in time, and the
33 resulting potential to provide service flows. We demonstrated our approach by using a SWAT model,
34 which has been configured with a grid-based landscape discretization and further enhanced to simulate
35 water flow across the discretized landscape units, to map and quantify four hydrological ecosystem
36 services vital to food and water security in the Upper Ouémé watershed in Benin. We set up ecosystem
37 accounting tables for both service capacity and service flow and analysed trends in service capacities.
38 For each hydrological ecosystem service, we were able to identify Subwatershed Ecosystem
39 Accounting Units (SEAUs) where either service capacity or service flow is concentrated. We were
40 also able to identify trends in changes in service capacity of hydrological ecosystem services for some
41 SEAUs. Our approach can be extended and applied to other watersheds because it is based on the
42 SWAT model, which has been tested extensively in different watersheds and landscapes. Our analyses
43 show that integrating hydrological ecosystem services in an ecosystem accounting framework
44 provides relevant information on watershed ecosystems and hydrological ecosystem services at
45 appropriate scales suitable for decision-making.

46
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48 Rathjens developed the grid-based model code; C. Duku performed the simulations and analyses; C.
49 Duku and L. Hein prepared the manuscript with contributions from S. J. Zwart and H. Rathjens.

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11 **Tables**
 12

13 **Table 1.** Description of spatial input data of the Upper Ouémé watershed for the SWAT Landscape model

Data type	Description	Resolution	Source
Topography	ASTER Digital Elevation Model (DEM)	30m	NASA
Land use/ land cover	Classified LANDSAT-7 ETM+ image	28.5m	IMPETUS
Soil types	Soil map and associated parameters derived from geological maps and field surveys	1:200,000	IMPETUS
Precipitation	Gridded daily precipitation data (1999 to 2012)	25km	AMMA-CATCH
Temperature	Gridded monthly average minimum and maximum temperatures (1999 to 2012)	50km	CRU TS 3.21
Household water consumption	Groundwater and surface water extractions	(village level)	IMPETUS

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1 **Table 2.** Overview of selected hydrological ecosystem services and associated service flow and service capacity
 2 indicators (GP is growing period)

Hydrological ecosystem service	Service flow indicator	Service capacity indicator
1. Crop water supply	Total number of days during the growing period in which there was no water stress (days GP ⁻¹)	Total number of days in a year when the sum of actual evapotranspiration and the amount of residual moisture added to the soil profile equalled or exceeded potential evapotranspiration. (days yr ⁻¹)
2. Household water supply		
a. Groundwater supply	Amount of groundwater extracted (m ³ ha ⁻¹ yr ⁻¹)	Groundwater recharge (m ³ ha ⁻¹ yr ⁻¹)
b. Surface water supply	Amount of surface water extracted (m ³ ha ⁻¹ yr ⁻¹)	Water yield (m ³ ha ⁻¹ yr ⁻¹)
3. Water purification	Rate of denitrification (kg ha ⁻¹ yr ⁻¹)	Denitrification efficiency (% denitrified)
4. Soil erosion control	Reduction in soil loss (metric tons ha ⁻¹ yr ⁻¹)	Maximum potential reduction in soil loss (metric tons ha ⁻¹ yr ⁻¹)

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1 **Table 3.** Description of calibrated parameter values of the SWAT Landscape model. Superscript **a** indicates that
2 the fitted values depended on the land cover type. Superscript **b** indicates that this parameter was used only in
3 the calibration of the grid-based SWAT Landscape model. Subscript **v_** indicates that the parameter value is
4 replaced by the fitted value. Subscript **r_** indicates the parameter value is multiplied by (1 + the fitted value).

Parameter name	Description	Fitted values
r_CN2	Initial SCS runoff curve number for moisture condition II	(from -0.2 to -0.05) ^a
v_RCHRG_DP	Deep aquifer percolation fraction	0.2
v_GW_REVAP	Groundwater re-evaporation coefficient	0.18
v_GWQMN	Threshold depth of water in the shallow aquifer required for return flow to occur	1000
v_REVAPMN	Threshold depth of water in the shallow aquifer for re-evaporation or percolation to the deep aquifer to occur	500
v_SURLAG	Surface runoff lag coefficient	0.12
r_SOL_AWC	Available water capacity of the soil	0.1
v_ESCO	Soil evaporation compensation factor	(from 0.001 to 0.2) ^a
v_EPCO	Plant uptake compensation factor	(from 0.1 to 1) ^a
v_USLE_P	USLE equation support practice factor	0.13
v_USLE_C	Minimum value of USLE C factor for water erosion applicable to the land cover	(from 0.038 to 0.45) ^a
v_NPERCO	Nitrate percolation coefficient	0.2
v_N_UPDIS	Nitrogen uptake distribution parameter	70
v_SDNCO	Denitrification threshold water content	1.1
v_CDN	Denitrification exponential rate coefficient	1.4
v_DD ^b	Drainage density factor which affects the flow separation ratio	7.5

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1 **Table 4.** Calibration and validation results for streamflow, sediment and organic nitrogen loads (Prefix H___
2 indicates results for streamflow calibration and validation; prefix s___ indicates results for sediment load
3 calibration N___ indicates results for organic nitrogen load calibration). NSE is Nash-Sutcliffe efficiency, PBIAS
4 is percent bias, and RSR is ratio of the root mean square error to the standard deviation of measured data.

Monitoring stations	Drainage area (km ²)	Calibration			Validation		
		NSE	PBIAS	RSR	NSE	PBIAS	RSR
Upstream stations							
H_Affon-Pont	1 172	0.69	27.0	0.56	0.62	15.9	0.62
H_Aval-Sani	760	0.70	12.0	0.55	0.64	7.8	0.60
H_Bori	1 608	0.65	-24.7	0.59	-0.49	-121.4	1.22
H_Tebou	522	0.47	43.5	0.72	0.58	20.3	0.65
Downstream stations							
H_Beterou	10 046	0.85	5.7	0.39	0.78	-17.8	0.47
H_Barerou	2 128	0.71	20.8	0.54	0.72	-22.7	0.53
H_Cote-238	3 040	0.69	3.5	0.56	0.68	-18.4	0.56
H_Igbomakoro	2 309	0.76	11.3	0.49	0.71	-4.0	0.54
H_Sarmanga	1 334	0.48	23.2	0.72	0.44	17.2	0.75
H_Aguiimo	394	0.25	-20.9	0.87	0.12	-60.1	0.94
H_Wewe	297	0.42	21.6	0.76	0.42	-6.5	0.76
s_Beterou	10 046	0.45	6.9	0.74	0.83	2.55	0.42
N_Beterou	10 046	0.50	47.4	0.71	0.55	56.3	0.67

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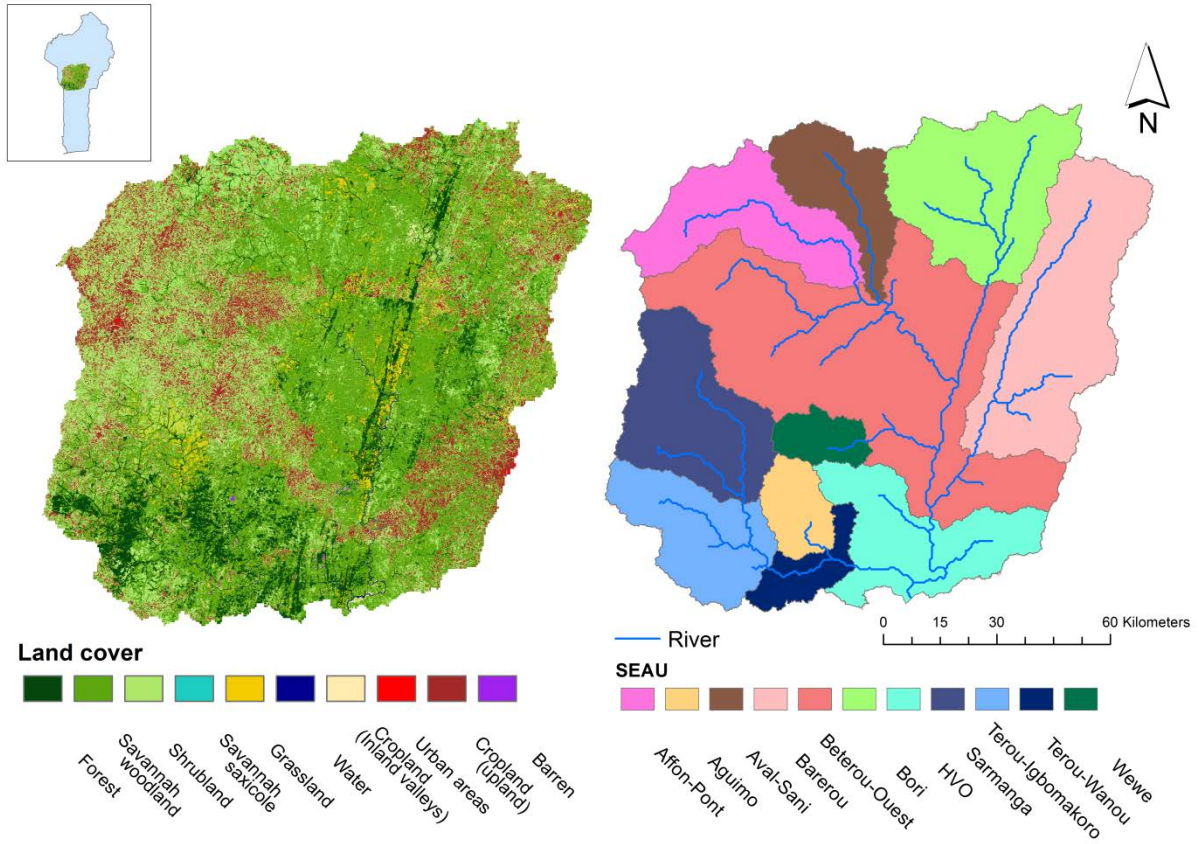
1 **Table 5.** Biophysical ecosystem account for service capacity at the SEAU level in the Upper Ouémé watershed in 2012 (SD is standard deviation)

Subwatershed Ecosystem Accounting Unit (SEAU)	Hydrological ecosystem service											
	Crop water supply				Household water supply				Water purification		Soil erosion control	
	Upland agricultural areas		Inland valley rice fields		Groundwater supply		Surface water supply		Total N added (10 ³ kg)	(% N denitrified)	Total (10 ³ metric tons yr ⁻¹)	Mean (SD) (kg ha ⁻¹ yr ⁻¹)
	Area (10 ³ ha)	Mean (SD) (days yr ⁻¹)	Area (ha)	Mean (SD) (days yr ⁻¹)	Total (10 ⁶ m ³ yr ⁻¹ recharge)	Mean (SD) (10 ³ m ³ ha ⁻¹ yr ⁻¹ recharge)	Total (10 ⁶ m ³ yr ⁻¹ water yield)	Mean (SD) (10 ³ m ³ ha ⁻¹ yr ⁻¹ water yield)				
Affon-Pont	20.2	103 (28)	200	149 (4)	121	26 (26)	624	133 (90)	2 719	30	5.2	44 (74)
Aguimo	0.3	95 (34)	0	0	58	37 (28)	255	161 (112)	589	0	2.6	65 (102)
Aval-Sani	4.0	100 (32)	0	0	114	38 (20)	458	151 (89)	1 370	36	3.5	45 (85)
Barerou	33.4	104 (29)	100	118 (18)	244	29 (20)	1 328	156 (101)	4 707	11	18.5	87 (101)
Beterou-Ouest	54.3	103 (29)	425	135 (16)	615	38 (30)	2 526	155 (99)	8 550	19	22.9	56 (102)
Bori	12.0	109 (30)	0	0	185	29 (20)	1 082	168 (97)	3 138	26	6.4	40 (79)
HVO	7.0	104 (32)	50	112 (33)	206	43 (40)	638	133 (80)	1 953	15	8.3	69 (93)
Sarmanga	9.7	106 (24)	175	139 (12)	304	57 (37)	809	152 (82)	2 382	13	4.5	34 (42)
Terou-Igbomakoro	4.0	108 (25)	50	138 (2)	222	57 (36)	591	151 (93)	1 561	0	4.9	51 (91)
Terou-Wanou	0.8	91 (30)	25	96 (0)	73	54 (22)	170	126 (58)	514	0	2.5	74 (90)
Wewe	4.1	96 (30)	75	131 (20)	48	40 (33)	213	177 (117)	638	0	1.8	61 (182)
Total	149.8		1 100		2 190	—	8 694	—	28 121	—	81.1	—

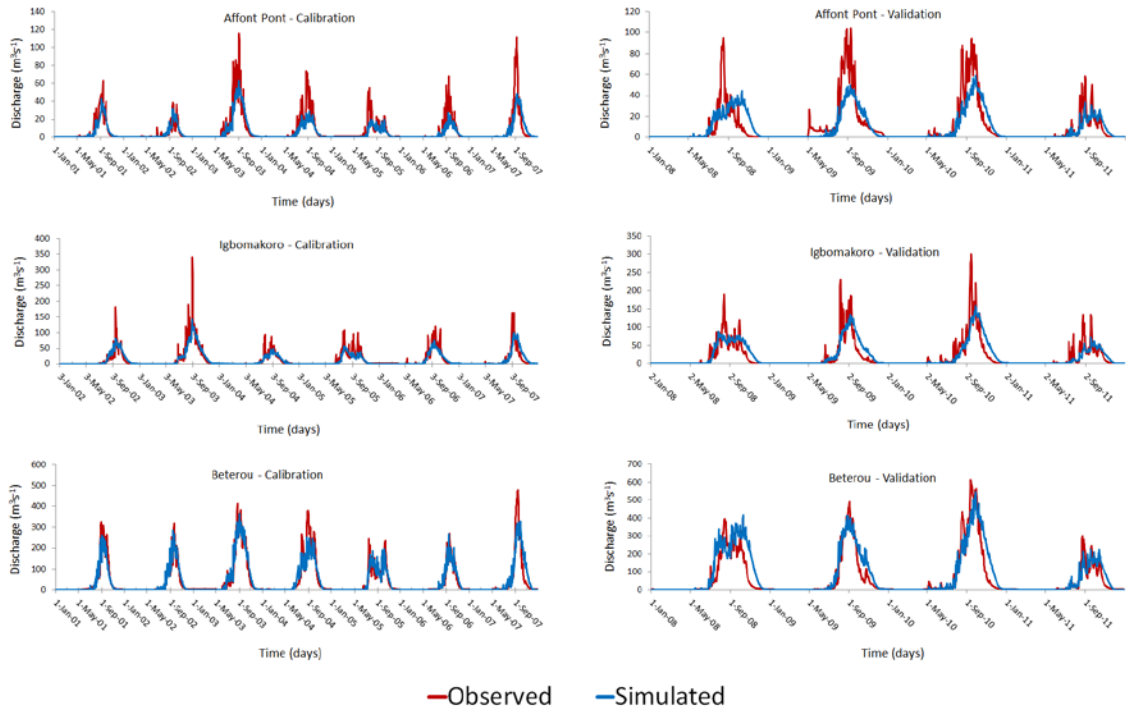
1 **Table 6.** Biophysical ecosystem account for service flow at the SEAU level in the Upper Ouémé watershed in 2012 (GP is length of growing period between crop
 2 establishment and harvest; Upland agricultural areas had a GP of 103 days; Inland valley rice fields had a GP of 123 days; SD is standard deviation)

Subwatershed Ecosystem Accounting Unit (SEAU)	Hydrological ecosystem service									
	Crop water supply				Household water supply		Water purification		Soil erosion control	
	Upland agricultural areas		Inland valley rice fields		Groundwater	Surface water	Total (10 ³ kg N yr ⁻¹ denitrified)	Mean (SD) (kg ha ⁻¹ yr ⁻¹ denitrified)	Total (10 ³ metric tons yr ⁻¹)	Mean (SD) (kg ha ⁻¹ tons yr ⁻¹)
	Area (10 ³ ha)	Mean (SD) (days GP ⁻¹)	Area (ha)	Mean (SD) (days GP ⁻¹)	Total (10 ³ m ³ yr ⁻¹ water extracted)	Total (10 ³ m ³ yr ⁻¹ water extracted)				
Affon-Pont	20.2	59 (30)	200	123 (0)	123	65	810	6.9 (10)	4.4	38 (67)
Aguimo	0.3	52 (35)	0	—	0	0	0	0.0 (0)	2.3	58 (92)
Aval-Sani	4.0	64 (29)	0	—	8	0.2	498	6.5 (7)	3.2	42 (81)
Barerou	33.4	63 (31)	100	107 (17)	510	64	503	2.4 (6)	15.9	75 (92)
Beterou-Ouest	54.3	59 (31)	425	115 (22)	1 124	219	1 613	4.0 (9)	18.9	46 (90)
Bori	12.0	65 (32)	0	—	196	30	815	5.1 (7)	5.4	33 (67)
HVO	7.0	56 (32)	50	88 (35)	71	37	297	2.5 (5)	7.0	59 (79)
Sarmanga	9.7	69 (34)	175	119 (8)	532	66	317	2.3 (5)	4.0	30 (39)
Terou-Igbomakoro	4.0	69 (34)	50	123 (0)	95	36	0	0.0 (0)	4.4	45 (85)
Terou-Wanou	0.8	45 (35)	25	92 (0)	0	0	0	0.0 (0)	2.2	65 (83)
Wewe	4.1	63 (31)	75	107 (23)	41	41	0	0.0 (0)	1.5	51 (178)
Total	149.8	—	1 100	—	2 700	558.2	4 853	—	69.2	—

1 **Figures**
 2



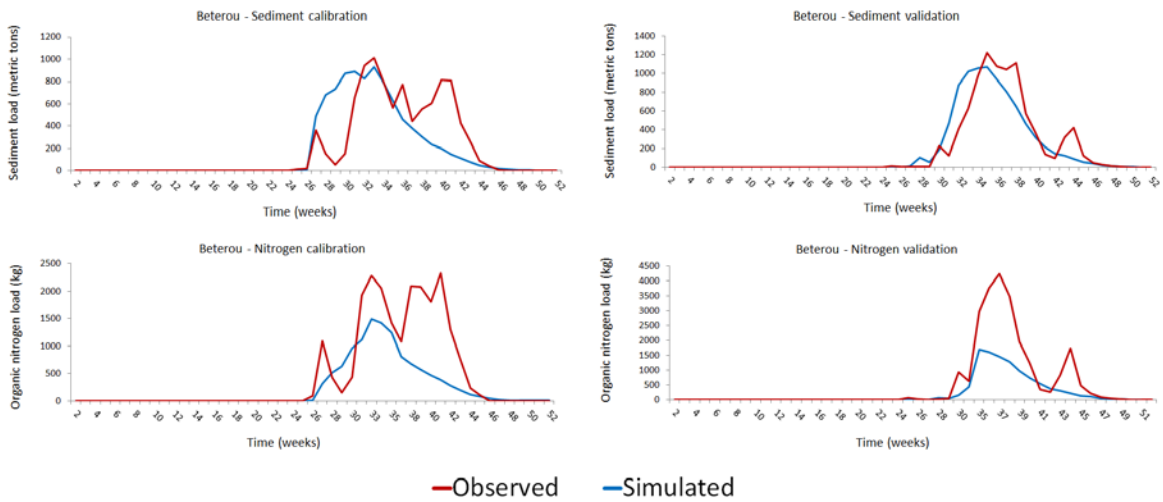
3
 4 **Figure 1.** Land cover and Subwatershed Ecosystem Accounting Units (SEAUs) of the Upper Ouémé
 5 watershed. Land cover data adapted from Judex and Thamm (2008)
 6



1

2 **Figure 2.** Comparing simulated and observed streamflow for three monitoring stations with varying
 3 drainage areas; Affont-Pont, 1 172 km²; Igbomakoro, 2 309 km²; Beterou, 10 046 km²

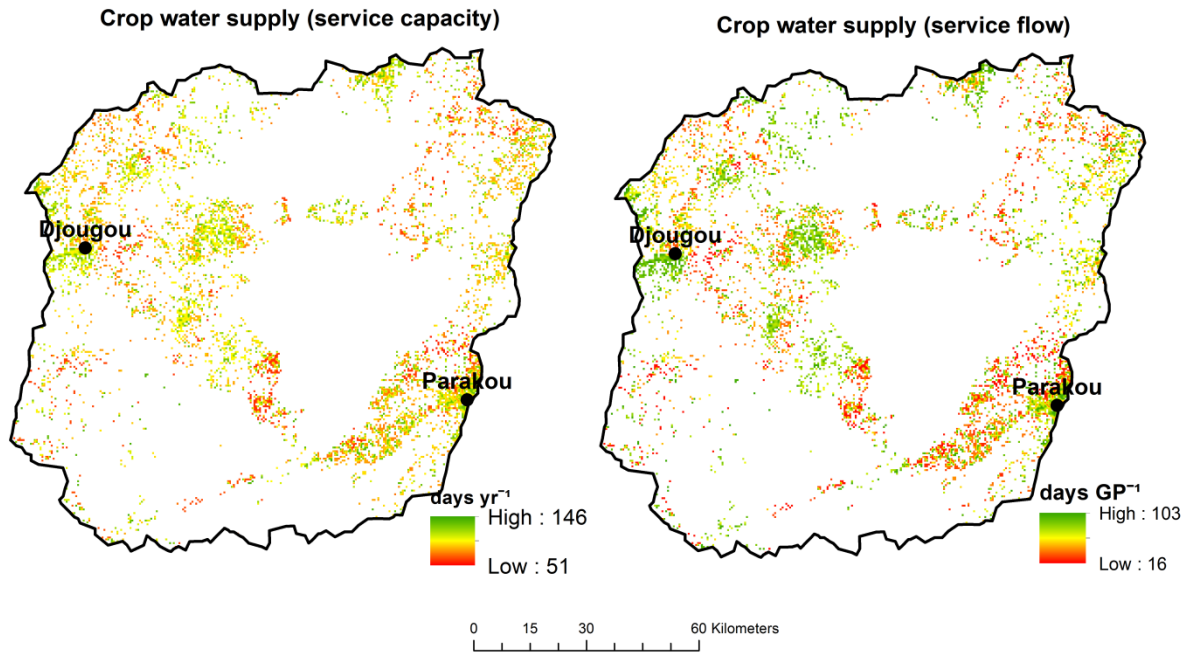
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6 **Figure 3.** Comparing simulated and observed sediment loads and organic nitrogen loads during
 7 calibration and validation at Beterou monitoring station for the period 2008 to 2009

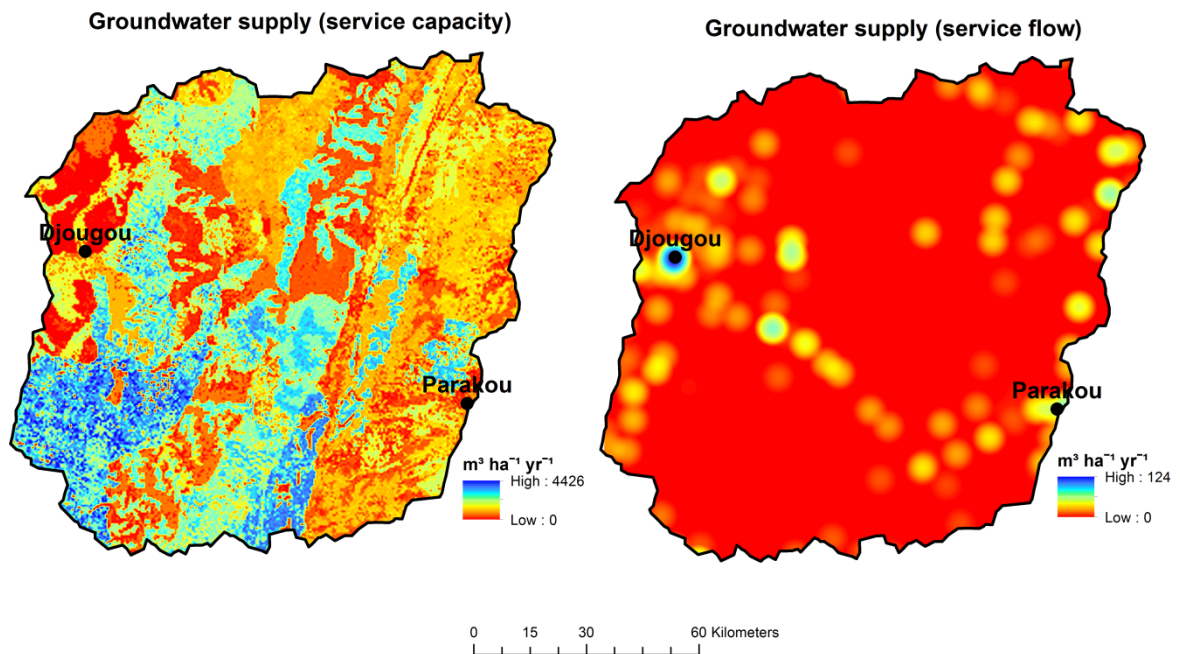
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2 **Figure 4.** Spatial distribution of mean annual values of service capacity and mean seasonal values of
 3 service flow of crop water supply in upland agricultural areas in the Upper Ouémé watershed from the
 4 year 2001 to 2012 (GP indicates growing period).

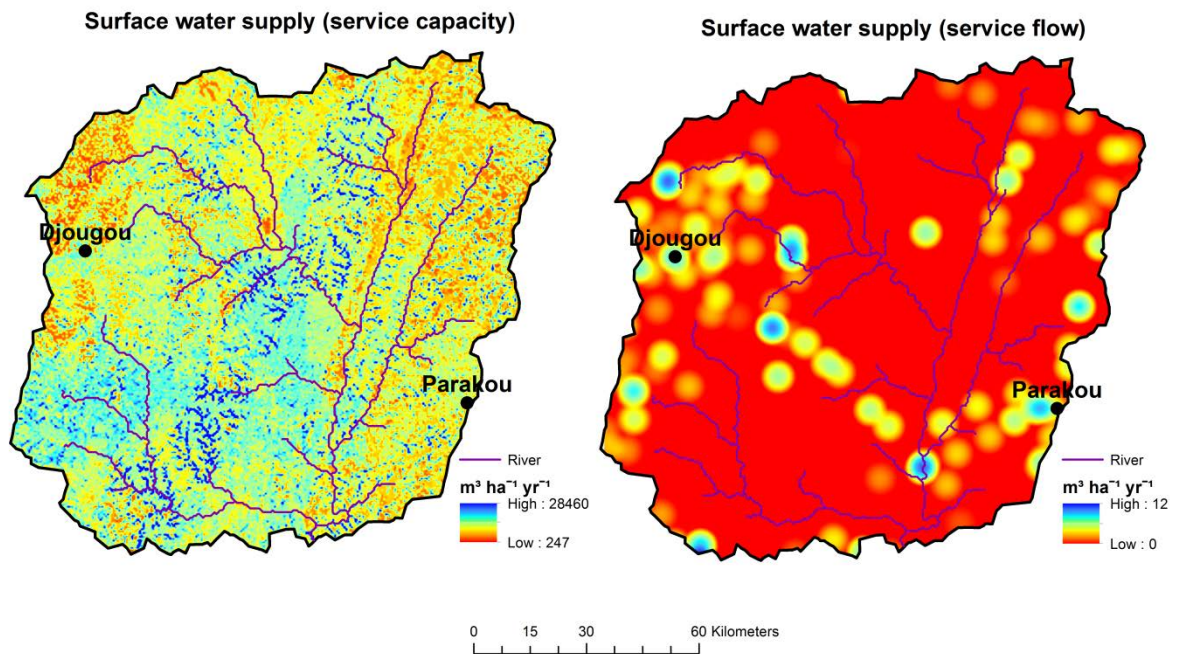
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6

7 **Figure 5.** Spatial distribution of mean annual values of service capacity and service flow of
 8 groundwater supply in the Upper Ouémé watershed from the year 2001 to 2012.

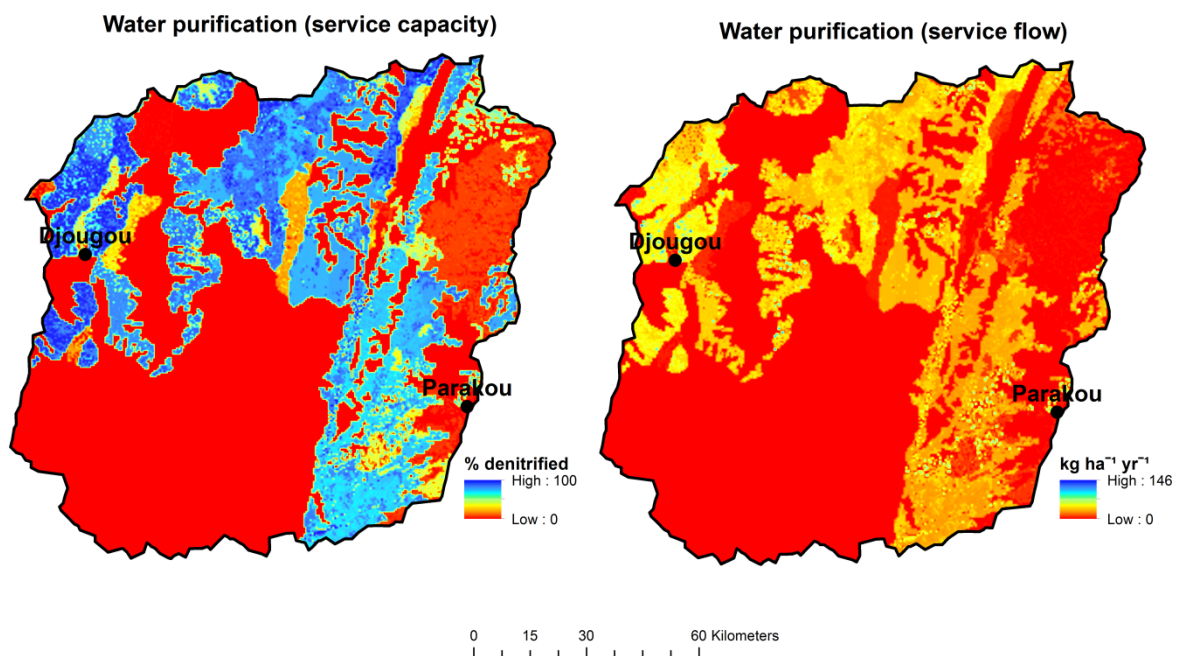
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2 **Figure 6.** Spatial distribution of mean annual values of service capacity and service flow of surface
 3 water supply in the Upper Ouémé watershed from the year 2001 to 2012.

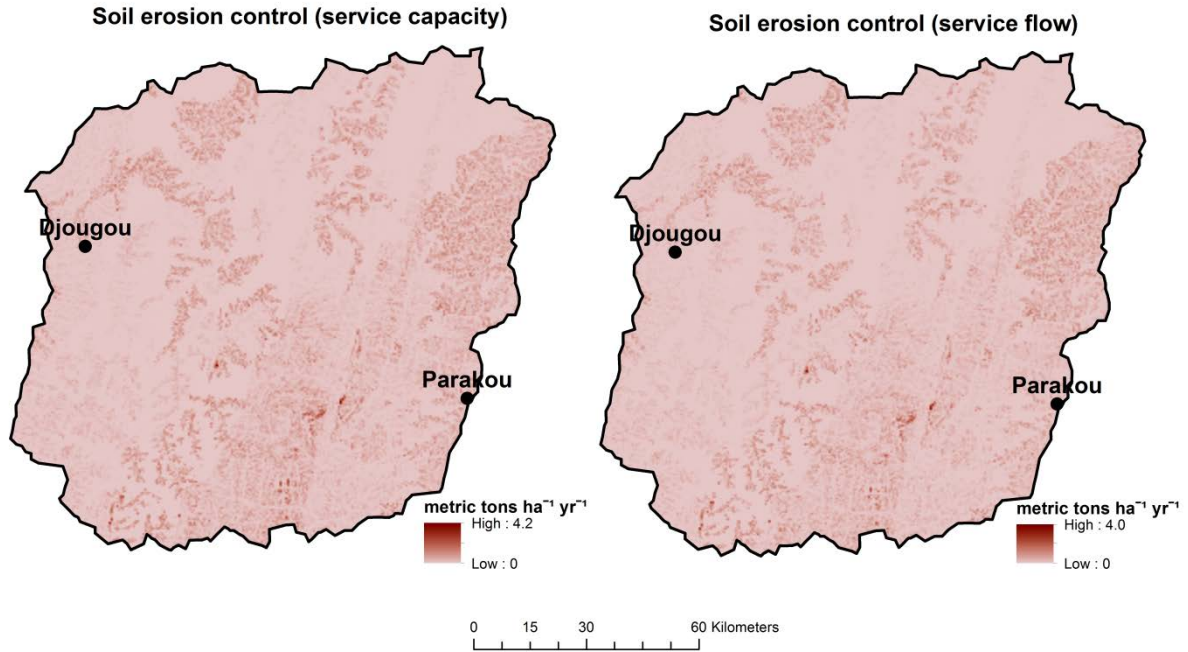
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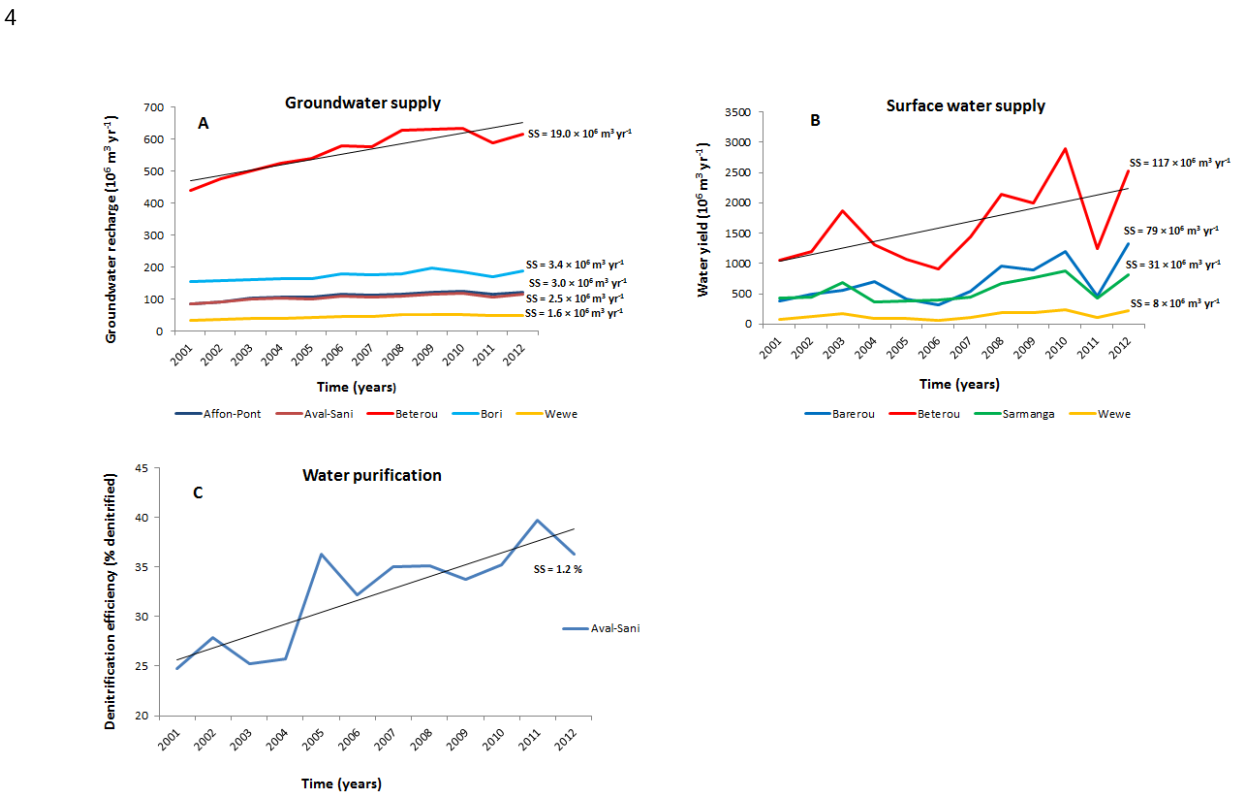
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6 **Figure 7.** Spatial distribution of mean annual values of service capacity and service flow of water
 7 purification in the Upper Ouémé watershed from the year 2001 to 2012.

8



1
 2 **Figure 8.** Spatial distribution of mean annual values of service capacity and service flow of soil
 3 erosion control in the Upper Ouémé watershed from the year 2001 to 2012.



5
 6 **Figure 9.** Trends in service capacity of hydrological ecosystem services at the SEAU level in
 7 the Upper Ouémé watershed (SS is Sen's Slope estimator, which is a measure of the
 8 magnitude of change of a trend). For each graph, a single trend line is drawn solely to
 9 illustrate the direction of trend.

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