AUTHORS' RESPONSE TO REVIEWS

REFEREE 1

We would like to thank Dr E. N. Müller for her invaluable feedback and constructive comments. We have taken many of the comments and suggestions on board and we will incorporate them in the final paper. In the following, we provide a point by point response to her comments. For the sake of clarity we first repeat the reviewer's comments (in italic) and then provide our response.

GENERAL COMMENTS

Comment 1: *The overall structure of the manuscript is not well composed. Both abstract and introduction do not give the right impression on the content of the article. It is not explained which hydrological ecosystem services exist and why four specific ones have been chosen for this study.*

Reply 1: We have taken on board your useful suggestions and we will revise the abstract and the introduction sections accordingly.

Comment 2: *Description of the study area lacks some critical information on type of land-use, specifically what kind of rain-fed crops and why kind of rice (wet or dry rice cultivation in a savannah ecosystem?), what kind of current water supply for domestic use, groundwater conditions, status of current land degradation, erosion, nutrient pollution of surface and groundwater resources. Without this background information, it appears difficult to understand the necessity to analyse corresponding ecosystem services.*

Reply 2: We will provide additional information on the study area especially current water supply for domestic use, land use and the type of rainfed crops. We shall include these in the revised version.

Comment 3: *Substantial lack in the application of the SWAT model: the given parameterisation data are not detailed enough, calibration and validation data on a monthly basis appears not to be sufficient to assess daily soil moisture patterns for the crop water supply, testing data for sediment and nutrient fluxes are not sufficient to say anything about model performance; not information are given on how the groundwater recharge is set up, which appears essential to evaluate groundwater abstractions of private households.*

Reply 3: We have provided the detailed parametrisation data in the attached supplement. It provides information on how the groundwater recharge as well as other hydrologic processes was set up. We will also include this table in the final paper. We fully agree that daily calibration and validation is more appropriate for assessing daily soil moisture patterns of

crop water supply. With our approach the initial set up was carried out with ArcSWAT (which is based on the HRU discretization). This setup was then calibrated and validated on a daily time-step using 12 years of daily streamflow data from 11 different monitoring stations within the watershed. This calibration and validation was carried out SUFI-2 optimization algorithm of SWAT-CUP. The calibrated and validated input parameter sets from the HRU-based setup were transferred to the grid-based setup of the SWAT Landscape model using the SWATgrid interface. The main process distinction of the grid-based SWAT Landscape model from the conventional SWAT model is the use of a spatially distributed topographic index to partition flow into landscape and channel components. This is determined by a parameter called drainage density factor (Rathjens et al., 2014). During simulation of the grid-based SWAT Landscape model, only this parameter was manually calibrated to make sure that model performance was still satisfactory. It was this manual re-calibration that was carried out with monthly streamflow data instead of daily data and the results included in the manuscript. We will re-evaluate the grid-based model with daily streamflow data and include this in the final paper instead of the monthly results.

Concerning sediment and nitrogen fluxes, we acknowledge that the lack of validation introduces uncertainties with regards to model results. Data needs especially for calibration and validation of sediment and nitrogen fluxes for a region such as ours have always been a great challenge. We will therefore include a subsection "SWAT modelling uncertainties and limitations" in the Discussion section to discuss these inadequacies and limitations.

Comment 4: *Section 3.2.3 on water purification and denitrification appears to be wrong. There are no information given on fertiliser rates of the agricultural fields; nutrient leaching through groundwater recharge and nutrient leaching into surface water resources are not considered, and it is not clear how the denitrification rate is linked to water supply.*

Reply 4: Whereas we acknowledge that nutrient leaching through groundwater recharge and also into surface water resources are important components of water purification, we disagree with your opinion that using denitrification as an indicator of water purification is wrong. Ecosystem services are benefit dependent (Boyd and Banzhaf, 2007) and the selection of indicators for quantification and mapping are driven first and foremost by the benefits in question. In our study area, groundwater provides over 90% of the total household drinking water consumption that was why we focussed solely on water purification of groundwater resources and not that of surface water resources. The benefit of groundwater use by households is a function of its quantity and quality, however, the quantity and quality aspects are provided by different ecosystem services. Therefore, for water purification, our approach focussed on soil denitrification and looked at the quality aspect of groundwater provision. We understand that nitrate leaching through groundwater recharge may be another indicator of water purification. However, for this study and within the framework of ecosystem services and accounting, we argue that denitrification is a more appropriate indicator because it determines the quantities (after plant uptake) of nitrate available for leaching into groundwater systems (Jahangir et al., 2012). In order words, the rate of nitrate leaching into groundwater systems is dependent on soil denitrification. In addition, the challenge in terms of quantifying ecosystem services is the identification of management relevant indicators that can be enhanced through management interventions to augment the service production. Soil denitrification can be a management practice to reduce nitrate leaching into groundwater

systems. For example, Kramer et al. (2006) observed that organic farming supports more active and efficient denitrifier communities leading to a considerable reduction in nitrate leaching as compared to conventional farming. We acknowledge and pointed out (page 3489 line 28 and page 3490 line 1 to 5) that denitrification as modelled by the SWAT model is simplified. The SWAT model does not explicitly simulate microbial processes and dynamics but rather it simulates the ecohydrological conditions suitable for denitrification to occur (Boyer et al., 2006). Therefore, there is an inherent assumption of spatial homogeneity with regards to microbial species composition and quantities. These methodological challenges notwithstanding, for our study denitrification is very suitable indicator of water purification.

In Table 4 (page 3516), we provided the total amount of nitrogen/nitrate added to each SEAU either through fertilizer application or atmospheric deposition. Fertilizer use is increasing in the region and high fertilizer inputs is associated with crops such as maize, rice and cotton (Bossa et al., 2012). To simulate these high fertilizer inputs, we used the auto-fertilization routine in the SWAT model. Fertilizer (NPK and/or Urea) was applied automatically whenever the nitrogen stress fell below a specified level.

Comment 5: *The ecosystem service function to assess soil erosion control is oversimplified: does Equation 7 include intercropping?*

Reply 5: The focus of this part of our paper was to model and quantify the role of land/vegetation cover in soil conservation or controlling soil erosion. Empirical distinction of ecosystem service flow and ecosystem service capacity of regulating services such as soil erosion control has always been a challenge. However, approaches as such that used in our study are not uncommon in ecosystem services assessments (e.g. Leh et al., 2013;Terrado et al., 2014). In page 3500 line 15 to 29 and page 3501 line 1 to 16, we believe we have adequately discussed this challenge. Because we simulated only a single crop type (maize) in all agricultural fields Equation 7 does not include intercropping.

Comment 6: *In the conclusion it was suggested that such a tool is appropriate for decision-making in water and land management, an idea which I strongly reject. Although it is relatively easy to set up a SWAT simulation run for larger catchments, it is extremely difficult to obtain spatially and temporally correct representations of the underlying (eco)hydrological processes and interactions. Without appropriate, high-resolution and high-quality testing data, SWAT results may be used in relative terms, but not in absolute, hence not for decision-making.*

Reply 6: Yes, it is true that for larger catchments, it is extremely difficult to obtain spatially and temporally correct representations of the underlying hydrological processes and interactions. To achieve this, there is the need for multi-site calibration at different spatial scales with a sufficient length of time-series of data (Santhi et al., 2008). We believe that for the hydrologic components we have adequately fulfilled these requirements. Calibration and validation were not carried out at the watershed outlet only. We used 12 years of daily streamflow data from 11 different monitoring stations that had varying spatial scale (from 297 km^2 to 10,046 km^2) as described in Table 3 in Page 3515. In addition, three quantitative statistics that measure different aspects of model performance were used for model evaluation with satisfactory results. Model results to support decision-making are always associated with a certain degree of uncertainty and the quantitative statistics provide a measure of these uncertainties. Using this tool as a decision-support such as for spatial prioritisation and

conservation to augment and sustain service production should be carried out always with these uncertainties in mind.

SPECIFIC COMMENTS

Comment 1: *P. 3480, line 12: in the contrary: discretisation type does have a significant effect on aggregated hydrological attributes.*

Reply 1: We base this statement on Arnold et al. (2010)

Comment 2: *P. 3480, line 20: not sure what is meant what 'ecohydrological models should do' – I would rather give a generic definition about the capacities of ecohydrological models, which processes and interactions with biogeochemical and vegetation processes are included, which models are available etc.*

Reply 2: The statement highlights the importance of landscape spatial interaction in ecohydrological models that are to be used to support ecosystem accounting. This is important for simulating the impact of upstream land and water management on downstream processes. However, in some models including the standard SWAT model currently in use, this process is not incorporated (Arnold et al., 2010;Bosch et al., 2010;Volk et al., 2007). The impact of upstream land management and uses downstream cannot be simulated if they are not in the same HRU. We however, will reframe this section in the final paper to make this point clearer.

Comment 3: *P. 3481, line 6: which type of stakeholder? Not sure why this is mentioned here.*

Reply 3: We will revise this section accordingly.

Comment 4: *P. 3481, line 12: there should be a longer section in the introduction on hydrological ecosystem services and here a explanation, why these ones have been selected*

Reply 4: We will revise this section accordingly.

Comment 5: *P. 3481, line 15: refer to other studies who used SWAT in Benin, e.g. C. Hiepe, 2008, PhD thesis at Uni Bonn, or the studies by Bossa et al. 2012.*

Reply 5: We will revise this section accordingly.

Comment 6: *P. 3481, line 24-25: land degradation has not been mentioned before – is that an important feature of the catchment? If yes, where is it addressed (potentially in the erosion control service)? Competition for scarce water resources (e.g. agricultural versus domestic use) should be mentioned earlier in the introduction. Section 3.1.1: this should include a short description of the model processes, spatial and temporal scales and resolution etc.*

Reply 6: We will revise this section accordingly.

Comment 7: *P. 3483, line19: not sure how raster cells of 500x500 metres will solve SWAT's original problem of matter routing, this resolution is still too coarse to capture many*

heterogeneities regarding land-use, slope, hillslope-river connections and connectivity patterns

Reply 7: There is always a trade-off between increased spatial detail and run-time efficiency in the SWAT model (Arnold et al., 2010;Notter et al., 2012). We acknowledge that 500m by 500m grid cell resolution may not be the best, however, we believe that we have sufficiently discussed this modelling challenge in Page 3498 line 21 to 25 and in Page 3499 line 1 to 18. We would also like to point out that the problem of matter routing in the SWAT model during the land phase of the hydrological cycle is not because of the coarseness of a spatial resolution or because of the spatial discretization type but mainly because of the lack of spatial interaction between HRUs. It is this lack of spatial interaction that the modified SWAT model, SWAT Landscape (Arnold et al., 2010;Volk et al., 2007) used in this study seeks to address. Therefore, even though 500m by 500m may be relatively coarse it does not affect matter routing because of the use of SWAT Landscape model.

Comment 8: *P. 3485 line 5: information on the extraction rates from shallow aquifers and streams are required.*

Reply 8: We will include this data in the revised version.

Comment 9: *P. 3486, line 5: The approach to re-calibrate and re-validate manually the grid-based simulations of the model requires justification. Section 3.2: this section belongs to the introduction section*

Reply 9: The main process distinction of the grid-based SWAT Landscape model from the conventional SWAT model is the use of a spatially distributed topographic index to partition flow into landscape and channel components. This is determined by a parameter called drainage density factor (Rathjens et al., 2014). It was this parameter that was re-calibrated manually to make sure that model performance of the grid-based SWAT Landscape model was satisfactory. We will revise Section 3:2 accordingly.

Comment 10: *P. 3487, line 2: more information are required on the agricultural system, especially on the rice plantations: why kind of rice is grown, - I assume upland, rainfed rice, - is this adequately included in the SWAT parameterisation?*

Reply 10: For our study, rice cultivation is simulated only in inland valley lowlands (which are the most dominant agro-ecology for rice cultivation in our study area). A map of inland valleys in our study area obtained from field surveys was used for the spatial delineation. The pothole function was used to simulate these lowlands are rice paddy fields.

Comment 11: *P. 3489, line 3: water consumption per capita has to be given.*

Reply 11: We will include this data in the revised version.

Comment 12: *Section 3.2.3: water purification appears to be the wrong wording, also check comments in the general comment section above.*

Reply 12: Please see Reply 4 under General comments

Comment 13: *L 3496, line 11: does the model state that soil erosion is currently not a problem? Is there any other evidence?*

Reply 13: We will revise this section and provide supporting evidence.

Comment 14: *P. 3501, line 1-2: This sentence needs re-writing: are you referring to the temporal scaling of water fluxes, i.e. quick runoff of water originating from high-intensity storms versus low intensity rain and its capacity to increase soil moisture? How is this related to crop and land management?*

Reply 14: We will revise this section accordingly.

REFEREE 2

We would like to thank the reviewer for his/her invaluable feedback and constructive comments. We have taken many of the comments and suggestions on board and we will incorporate them in the final paper. In the following, we provide a point by point response to the comments. For the sake of clarity we first repeat the reviewer's comments and then provide our response.

Comment 1: *The topic of this paper "accounting of hydrological ecosystem services" is interesting. However, the topic is a rather complex one, as it combines hydrological modelling of different processes at relatively large scales, the translation of the hydrological states and flows into hydrological ecosystem services and subsequently a meaningful way of accounting for the whole of hydrological ecosystem services. In my opinion in many parts of this research the treatment of certain processes or the calculation of hydrological ecosystem services was rather too simplified. I worry about the application of a spatially explicit model with so many parameters when in various aspects data and processes have to be so strongly simplified: can you still justify the use of such a model?*

Reply 1: Many approaches have been used in modelling and quantifying hydrological ecosystem services. Some of them are specifically tailored towards quantifying hydrological ecosystem services whereas others build on existing hydrologic models (Vigerstol and Aukema, 2011). There are methodological challenges as well as advantages associated with each of them. For ecosystem accounting (which is an emerging field that provides a systematic framework to link ecosystems to economic activities (EC et al., 2013;Edens and Hein, 2013;Obst et al., 2013), one of the essential issues yet to be resolved is the incorporation of hydrological ecosystem services into ecosystem accounts for environmental management, monitoring and decision-making . This necessitates that several methodological challenges associated with modelling hydrological ecosystem services must be addressed. The challenges include appropriate decisions regarding model process inclusion, spatial heterogeneity, physical and mathematical representation, temporal resolution, and model accuracy. In order to address these issues, we considered several hydrological models out there and selected the SWAT model to be more appropriate for this purpose**.** The SWAT model has been tested extensively for different applications and in different landscapes and climates with satisfactory results (Gassman et al., 2007). Its wide applicability, process representation, spatial explicitness and extensibility make it a suitable tool to support modelling and accounting for hydrological ecosystem services. We understand your issue with over-parametrization, which is not limited to the SWAT model but other complex models as well. However, model complexity, process inclusion, operational feasibility and explicitness are some of the trade-offs that must be made.

Comment 2: *The English is generally fine, however the structure of the article is not very good and in some cases the explanations are fuzzy and therefore not clear despite correct English / grammar. I personally find the article a bit difficult to read and would recommend the authors to work on the structure and style so that it may improve the possible impact of the article.*

Reply 2: We will restructure and rewrite some sections of the paper to make it clearer, concise and coherent.

Comment 3: *I did not find which parameters of the model were used to calibrate the model with and how this calibration was performed, I think that is important information.*

Reply 3: We agree that calibration parameters are important information that must be included in the paper. We have provided them here for your perusal. We will also include this table in the final paper. Calibration of the HRU-based SWAT model was carried out using the SUFI-2 optimization algorithm in SWAT-CUP (Abbaspour et al., 2008). Streamflow calibration was first undertaken and followed by sediment and nutrient calibration respectively. These calibrated and validated parameter sets were then transferred to the grid-based setup of the SWAT Landscape model. This setup uses a spatially distributed topographic index to partition flow into landscape and channel components. The parameter that controls the flow separation is drainage density factor (Rathjens et al., 2014). This parameter was further calibrated manually in the grid-based SWAT Landscape setup.

Comment 4: *The calibration was performed for streamflow as well as sediment and nitrogen load, but validation only for streamflow. Based on the strongly simplified model, I expect the validation results for the sediment and nitrogen load to be very poor. It would be good to show these validation results too, as you present results from the model for these ecosystem services and it would be good to have some measure of how reliable these model outcomes are.*

Reply 4: As we pointed out in p 3485 line 16 to 17, due to the absence of a long time-series of data, we were only able to calibrate the nitrogen and sediment loads. We were not able to validate this part of the model. Whereas this increases the uncertainties associated with model results for nitrogen and sediment loads, this approach is not uncommon. In the absence of a long time-series of data either only calibration or validation is undertaken (Gassman et al., 2007). In the final paper, we will include a subsection under Discussions to discuss model parameter uncertainties and model limitations.

Comment 5a: *- the result of this simplified approach of estimating water purification based on saturation would mean that with a higher amount of saturation there is a larger denitrification, while I would expect that in case of a higher saturation in the soil there is a stronger groundwater recharge and more potential flow of pollutants to groundwater. So is denitrification really the best way to estimate the purification?*

Reply 5a: We acknowledge that denitrification as modelled by the SWAT model is simplified and we explicitly stated these model simplifications and limitations in our paper (p 3489 line 28 and p 3490 line 1 to 5). Our approach models the spatiotemporal distribution of ecohydrological conditions necessary for denitrification to occur. Namely, soil moisture content, soil temperature, presence of a carbon source and nitrate availability. Furthermore, we would like to state that the assumption that higher saturation leads to higher groundwater recharge and in turn a higher potential flow of pollutants to groundwater is not necessarily true. Denitrification controls the amount of nitrates available in the soil profile for leaching into groundwater systems (Jahangir et al., 2012). Therefore, higher denitrification rates brought on by soil saturation reduce nitrate leaching even though groundwater recharge may be high. Fine-textured soils (high clay), which are more prone to saturation and consequently

denitrification, are generally less susceptible to nitrate leaching than sandy-textured soils because water permeability and hence groundwater recharge is much lower. Because of this, a key practice for reducing leaching losses is to minimize the amount of nitrate present in the soil at any given time either through denitrification or other means. For example, Kramer et al. (2006) observed that organic farming supports more active and efficient microbial denitrifier communities leading to a considerable reduction in nitrate leaching as compared to conventional farming.

Comment 5b: *And though I understand that purification is a service is it giving us the right signal? In case 80% purification takes place in a location with a high load and high flow to groundwater, there may still be much more nutrient flow to groundwater than in another location with only 20% purification but a rather low load and maybe only very low groundwater recharge. So the service is high but the problem is still remaining much stronger in the first case.*

Reply 5b: Ecosystem services are the contributions of ecosystems to benefits used in economic and other human activities (EC et al., 2013). The benefit may be realized directly from the biophysical outcome or may be realized only after economic and human inputs have been combined with the biophysical outcome. Therefore, using your example of 80% purification taking place in a location with a high absolute load and high absolute nitrate leaching; we argue that in such a place, denitrification removing 80% of nitrates from the system and therefore increasing the groundwater quality is a great example of an ecosystem service to the dependent population. Even though 20% leaching is high in absolute terms and therefore problematic, the 80% purification would have reduced the level of economic or human input that would otherwise have had to be added in order to make groundwater suitable for human consumption.

Comment 6: *I think the reference to Galloway et al 2003 is not very appropriate here, I would expect something more specific for the region, how is fertilizer use in this region? Is it really so widespread in these catchments or is it spatially very variable how much fertilizer is applied? What kind of data did you use for the nitrogen input?*

Reply 6: We will change the reference to Galloway et al 2003 to a more appropriate one such as Bossa et al. (2012). Fertilizer use is increasing in the region and high fertilizer inputs are associated with crops such as maize, rice and cotton (Bossa et al., 2012). An estimated average of 100–250 kg ha^{-1} of fertilizer (NPK + Urea) is applied to cotton, rice and maize (Bossa et al., 2012). To simulate these high fertilizer inputs, we used the auto-fertilization routine in the SWAT model. Fertilizer (NPK and/or Urea) was applied automatically whenever the nitrogen stress fell below a specified level.

Comment 7: *If I understand the description well, all crops except for rice are simulated with the parameters for maize and have a standard growing period of 103 days. I do understand that it is difficult to obtain spatially explicit data, but in case the spatial parameterization is not possible, can you such a fully spatially explicit model?*

Reply 7: We want to clarify that crop growth parameters for maize were not used to simulate all crops. For our study area, we simulated only two crops, maize in croplands (which were

referred to in our manuscript as uplands) and rice in inland valleys. For our simulations, we assumed that all croplands are used for only maize cultivation with the growing period of 103 days (which is the most common crop in our study area) This assumption was made because of the lack of spatially explicit data on the types of crops grown on agricultural fields. Rice on the other hand was grown in inland valleys.

Comment 8: *I think to calculate the soil water capacity for plants as an average moisture content over the full growing period is also really quite critical. Whether the soil is middle moist for the full time or really dry at the beginning and wet at the end or the other way around has a huge influence on the yield, so I think it is not right to define the service capacity as an average moisture content over the growing period.*

Reply 8: Whereas we agree that the timing of soil water availability also influences the yield we do not agree with the assertion that defining service capacity as an average soil moisture content over the growing period is wrong. Timing of availability and actual quantities of soil moisture are all important factors that affect crop yield. The challenge then is to select a single indicator that can capture all these important factors. Even though the indicator we used aggregates soil moisture availability across the growing period, it provides a consistent basis for comparison between service capacity and service flow.

Comment 9: *In the case of soil erosion I find the definition of service capacity as the maximum potential soil loss in absence of vegetation is strange. I would expect to see the maximum potential reduction in soil loss produced by the vegetation of the area and the actual to be that which occurs under the actual soil cover by that vegetation. The sentence " It reveals the soil conservation capacity of a particular vegetation cover type." (p 3492, l 18-19) is not even true in case the capacity is calculated as the maximum soil loss in absence of vegetation cover.*

Reply 9: Yes, it is true that service capacity should be the maximum potential reduction in soil loss produced by the vegetation cover of the study area. This maximum potential reduction in soil loss (maximum potential soil retention) can be said to be equal to the maximum potential soil loss. The maximum potential soil loss is modelled assuming there is no vegetation cover (Leh et al., 2013;Terrado et al., 2014). For example, for a specified spatial unit, if the maximum potential soil loss in the absence of the vegetation cover is estimated as 2tons/ha/yr then it indicates that the potential reduction in soil loss due to the vegetation cover cannot be greater than 2tons/ha/yr. Therefore, the maximum potential reduction in soil loss can be said to be equal to the maximum potential soil loss. We also agree that the sentence "it reveals the soil conservation capacity of a particular vegetation cover type (p 3491, l18-119)" can be misleading. We will rewrite this whole methodological section to remove any ambiguity and to make our point clearer.

Comment 10: *Table 3: it should be possible to read every table on itself without looking through the text so please define the abbreviations. Table 4: GP is only specified in the header but nowhere in the table.*

Reply 10: Thank you for pointing out these inconsistencies to us. We shall correct them accordingly.

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List of Major Changes

1. Abstract has been rewritten
2. A significant portion of the Introduction section has been rewritten
3. Additional information has been provided in the Section 2 (Description of study area)
4. Some of the subtitles under Section 3 (Methods) have been changed.
5. Additional information on the SWAT model processes has been provided under Section 3.1.1
6. Section 3.1.2 has been completely rewritten.
7. Under Section 3.1.4, the data sediment and nitrogen loads were also validated. Under Section 3.2.1 the indicator for service capacity of crop water supply was changed completely. The parametrization data has also been added and can be found in Table 3.
8. Under Section 3.2.3, the text was rewritten and more specific references provided.
9. Under Section 3.2.4, the text for service capacity of soil erosion control was rewritten.
10. Section 3.3 has been shortened.
11. Under Section 4.1, the results for daily calibration and validation are presented instead of monthly calibration and validation. Results for validation of sediment and nitrogen loads are also presented.
12. Under Section 4.2, the results for service capacity of crop water supply were changed because of the change in methodology. Furthermore, the presentation of results for soil erosion control was revised. The results are the same on the text was changed.
13. Under Section 4.3, trend test revealed no trend in changes in service capacity of crop water supply in all SEAUs.
14. A new subsection titled "Model uncertainties and limitations" has been added under Section 5.
15. Section 5.2 has been shortened.
16. Some sections of section 6 have been rewritten.
17. New references have been added and some deleted.
18. A new table, Table 3 has been added.

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.

Comment [CD1]: Change in wording to make it concise and clearer

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

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

Comment [CD2]: Additional information

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

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

Comment [CD3]: New information on hydrological ecosystem services and ecosystem accounting

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

Our objective is to present a spatially explicit modelling approach aligned with an ecosystem accounting framework to map and quantify the capacities and the flows of multiple hydrological ecosystem services. We use 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, to simulate the watershed ecohydrology. The model is calibrated and

validated and 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. We use the Upper Ouémé watershed in Benin as a test case to demonstrate our approach. This case-study area was selected because of a relatively high data availability (Judex and Thamm, 2008;AMMA-CATCH, 2014). It is also a microcosm of rural sub-Saharan Africa, where large sections of the population depend on smallholder rainfed agriculture for their livelihood, where groundwater is the major source of drinking water, and where rapid population growth and increasing land use change are prevalent. The hydrological ecosystem services we model and account for are crop water supply, household water supply (groundwater supply and surface water supply), water purification, and soil erosion control. We select these four services because they are critical to food and water security for the population. Agriculture is the major source of income and livelihood in the watershed and is predominantly rainfed. Furthermore, groundwater is the major source of household water use (for both drinking and non-drinking purposes).

> **Comment [CD4]:** Change in wording and additional information on the choice of hydrological ecosystem services

2   Description of case-study area

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

> **Comment [CD5]:** Additional information

> **Comment [CD6]:** New information on study area

> **Comment [CD7]:** New information on study area

3   Methods

3.1   Modelling framework

3.1.1   Model selection

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

3.1.2   Model modification

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

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

4

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

3.1.3 Model input data

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

3.1.4 Model configuration and performance evaluation

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

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

Comment [CD12]: New information on the modified SWAT model used.

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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).

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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).

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$$W_{strs, j} = 1 - T_{act, j}/T_{max,j} \,, \tag{1}$$

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).

$$S_{f, j} = N \,(d_1, d_2,..., d_n \mid W_{strs} = 0)_j \,, \tag{2}$$

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

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

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

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

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

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

$$S_c = N(d_1, d_2, \ldots, d_n \mid W_{pstrs} \leq 0), \tag{5}$$

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 there will be no water stress.

> **Comment [CD21]:** New methodology for service capacity of crop water supply

3.2.2 Household water supply

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

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

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

3.2.3 Water purification

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

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

$$N_{dn} = 0 \qquad \text{if } \gamma_{sw} < \gamma_{sw,\,thr} \; , \tag{7}$$

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

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

$$\text{DN}_{eff} = \left(N_{dn} / N_{total} \right) \cdot 100 \; , \tag{8}$$

8

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

3.2.4 Soil erosion control

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

$$SD_{rtd} = S_{yld,\ pot} - S_{yld}, \qquad\qquad (9)$$

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

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

3.3 Accounting for hydrological ecosystem services

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

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

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

4 Results

4.1 SWAT Landscape model calibration and validation results

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

Comment [CD28]: New results presented based on daily calibration and validation of the model. New results also for calibration and validation of sediment and nitrogen transport processes

4.2 Spatial patterns of hydrological ecosystem services

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

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Comment [CD30]: New results for service capacity of crop water supply in upland agricultural fields

Comment [CD31]: New results for service capacity of crop water supply in inland valleys.

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

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

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

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

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Comment [CD33]: Same results. Only change in wording

4.3   Biophysical ecosystem accounts

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

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

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

5   Discussion

5.1   Model uncertainties and limitations

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

For larger watersheds, it is also extremely difficult to obtain spatially and temporally correct representations of the underlying ecohydrological processes and interactions. To achieve this, there is the need for multi-site calibration at different spatial scales with a sufficient length of time-series of data to capture high and low flow years, annual, seasonal and monthly variations (Santhi et al., 2008). Whereas the use of 11 years of daily streamflow data from 11 monitoring stations in the Upper Ouémé watershed reduces the uncertainties of modelled results, data for calibration and validation of sediment and nitrogen loads may not have been sufficient to enable the model to more accurately represent sediment and nitrogen transport processes. In evaluating model performance of sediment and nitrogen transport processes, we used only one year of data from a single monitoring station. Without multi-site calibration and validation, there remain large uncertainties in modelled results of sediment and transport processes at different spatial scales. In addition without long term temporal validation, there remain large uncertainties in the ability of the model to capture annual variability in these transport processes. Even with sufficient length of time series of multi-site data for calibration and validation, the problem of parameter non-uniqueness inherent in complex watershed models such as the SWAT model also introduces a degree of uncertainty in modelled results. Parameter non-uniqueness refers to the reproduction of similar observed ecohydrological signals by different input parameter sets. Therefore even for so called calibrated and validated SWAT models, there is always a degree of uncertainty introduced by parameter non-uniqueness. To limit this non-uniqueness and consequently reduce parameter uncertainty requires the use of comprehensive data on different fluxes, loads and ecohydrological processes such as crop yield, evapotranspiration (Abbaspour et al., 2008) that are most of the time not readily available.

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

> **Comment [CD35]:** New section

5.2 Lessons for ecosystem accounting

In ecosystem accounting, detailed and accurate land cover and land use data are important. Apart from their use as inputs in modelling ecosystem services, land cover classes are also used as ecosystem accounting units based on which ecosystem services are aggregated (Remme et al., 2014;Schröter et al., 2014). A single lumped land cover class for agricultural areas or croplands (be it as model input data or ecosystem accounting units) may be suitable when modelling and accounting for other ecosystem services (Remme et al., 2014;Schröter et al., 2014). However, when modelling and accounting for crop water supply, land cover and land use data with detailed and spatially disaggregated information on the types of crops grown in agricultural areas is needed. This is because different crops have different water requirements (Allen et al., 1998). In rainfed agricultural systems, crop water supply is the major limitation to crop production and is the main factor responsible for low yields in the seasonally dry and semiarid tropics and subtropics (Shaxson and Barber, 2003). However, in many of these regions, land cover and land use data with this level of detail are currently not available. Obtaining such information is complicated by the small plot sizes and cropping patterns varying from year to year. Our study area was no exception. Despite these constraints, the lack of

detailed data reduces the accuracy and reliability of modelled results of service flow of crop water supply. In our study area, this limitation resulted in the simulation of only a single crop type in upland agricultural areas. Therefore, the results for service flow of crop water supply should be interpreted in the context of the crop simulated. However, because methodologies such as (Allen et al., 1998) have been used extensively to compute the water requirements of various crops, our approach serves as a reference or baseline from which the service flow of crop water supply of a spatial unit could be estimated if a crop other than maize is grown.

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

5.3 Implications for watershed and ecosystem management

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

Furthermore, the grid-based setup of the SWAT Landscape model enabled us to identify HSAs at a finer spatial resolution. Characterization of the spatiotemporal dynamics of HSAs is essential in controlling non-point source pollution and in maintaining streamflow regime. Hydrologically Sensitive Areas have significant impact on key ecohydrological processes affecting interaction and

transport of water, sediment, nutrients and pollutants. They also provide key landscape controls on riverine ecosystem integrity including aquatic flora and fauna and downstream water availability and quality. Agricultural intensification in HSAs has a higher potential of generating agricultural non-point source pollution (Agnew et al., 2006). Land use change in these areas can have a more significant impact on the streamflow regime. Such analyses can form the basis for establishing Payment for Ecosystem Services schemes (PES) (Pagiola and Platais, 2007;Turpie et al., 2008). Watershed PES provides financial support to ecosystem management in high service production areas that are of particular relevance downstream (Lopa et al., 2012;Lu and He, 2014). We acknowledge that detailed ecohydrological modelling is only one of the considerations in establishing a watershed PES. Other considerations include transaction costs and ability to pay of downstream water users. However, ecohydrological modelling can be used to support watershed PES schemes by providing a tool for upstream water managers to monitor the provision of hydrological ecosystem services or by identifying high service production areas that are potentially relevant for a new PES.

6   Conclusion

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

Comment [CD36]: Change in wording

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Tables

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

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})

Comment [CD43]: New indicator

Comment [CD44]: Change in wording

Comment [CD45]: Change in wording

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4

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

5

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Comment [CD46]: New table added to manuscript

Table 4. Calibration and validation results for streamflow, sediment and organic nitrogen loads (Prefix $_H$__ indicates results for streamflow calibration and validation; prefix $_S$__ indicates results for sediment load calibration $_N$__ indicates results for organic nitrogen load calibration). NSE is Nash-Sutcliffe efficiency, PBIAS 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
$_H$_Aguimo	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

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

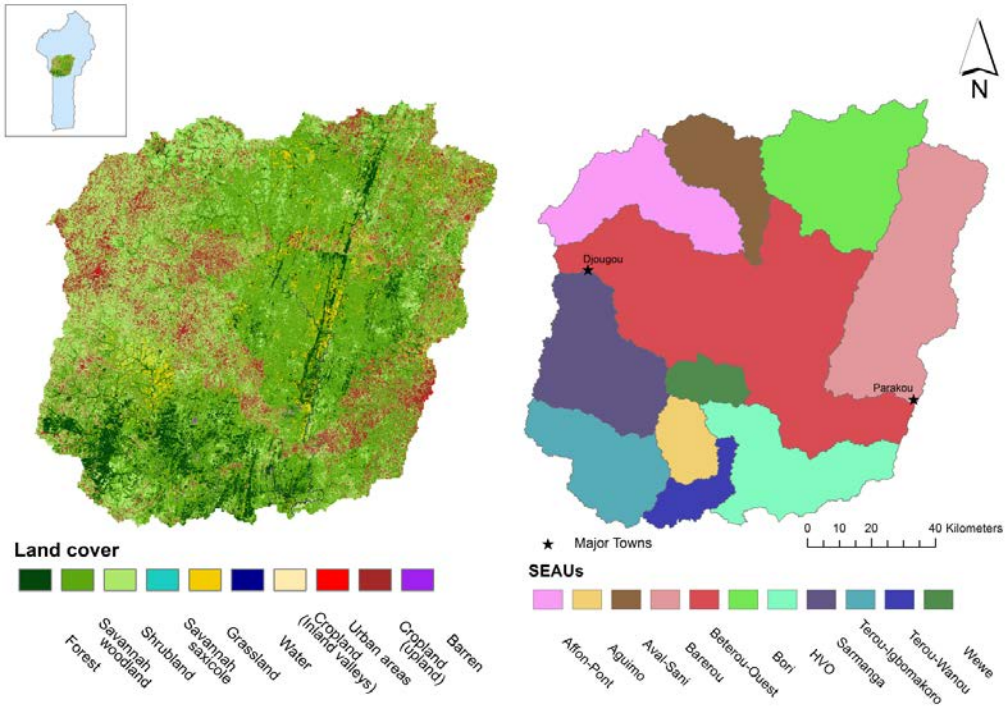| Subwatershed Ecosystem Accounting Unit (SEAU)	Hydrological ecosystem service											
	Crop water supply				Household water supply				Water purification		Soil erosion control	
	Upland agricultural areas		Inland valley rice fields		Groundwater supply		Surface water supply					
	Area (10^3 ha)	Mean (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)	Total N added (10^3 kg)	(% N denitrified)	Total (10^3 metric tons yr^{-1})	Mean (SD) (kg ha^{-1} yr^{-1})
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	—

24

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)

Comment [CD51]: New information

| 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				
	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)	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})
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	—

Figures

3

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

6
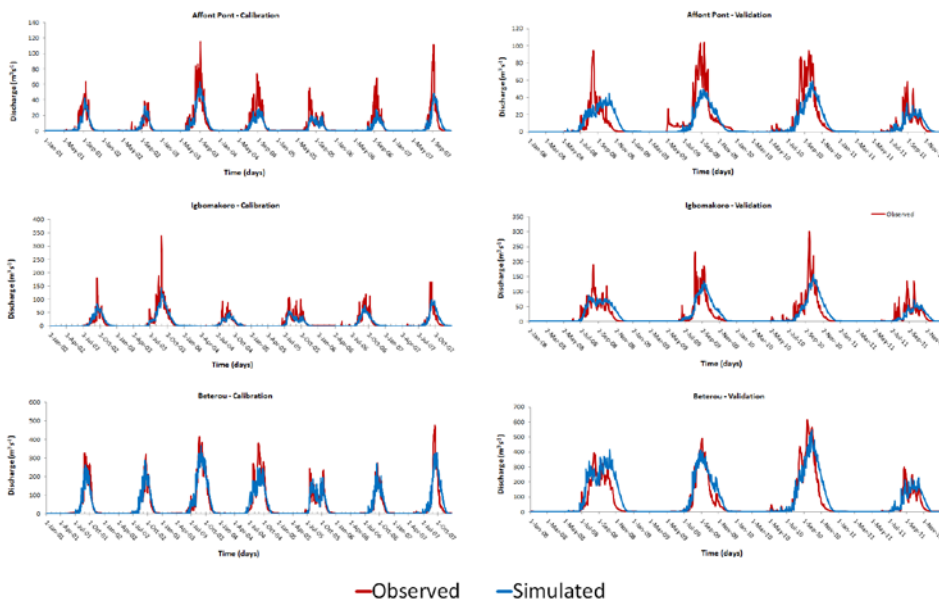
**Figure 2.** Comparing simulated and observed streamflow for three monitoring stations with varying drainage areas; Affont-Pont, 1 172 km^2; Igbo, 2 309 km^2; Beterou, 10 046 km^2
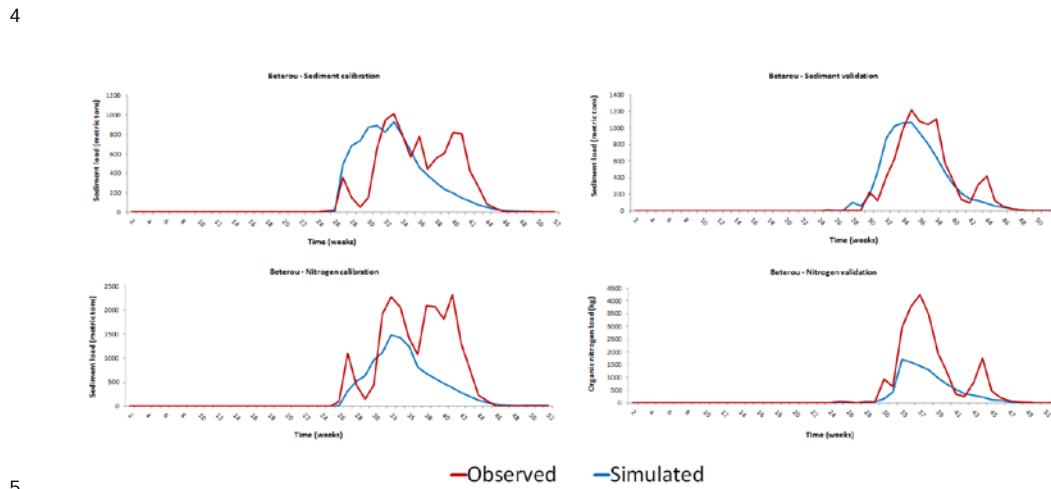
Figure 3. Comparing simulated and observed sediment loads and organic nitrogen loads during calibration and validation at Beterou monitoring station for the period 2008 to 2009
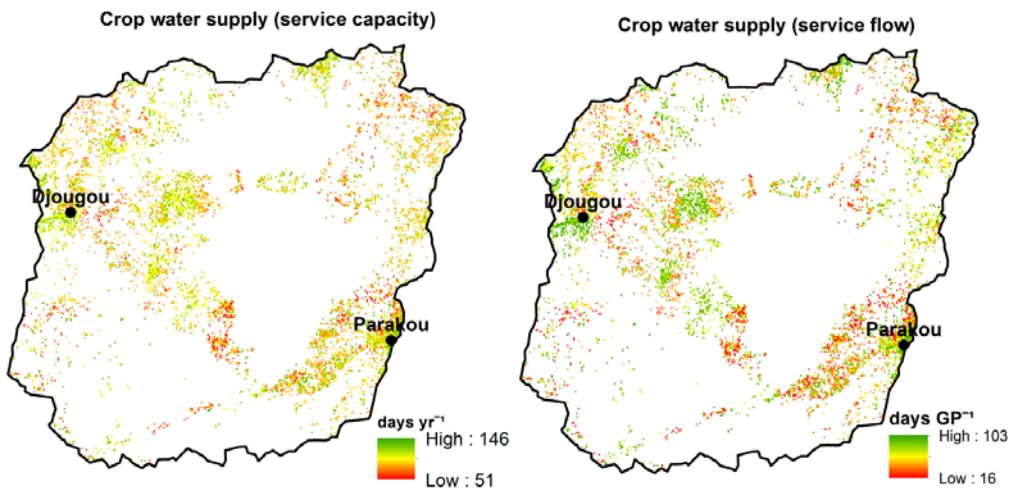
Figure 4. Spatial distribution of mean annual values of service capacity and mean seasonal values of service flow of crop water supply in upland agricultural areas in the Upper Ouémé watershed from the year 2001 to 2012 (GP indicates growing period).
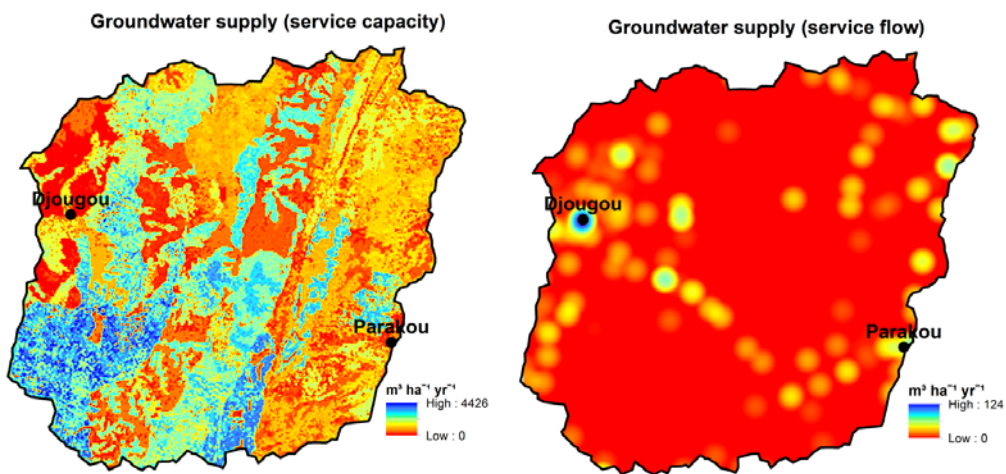
Figure 5. Spatial distribution of mean annual values of service capacity and service flow of groundwater supply in the Upper Ouémé watershed from the year 2001 to 2012.
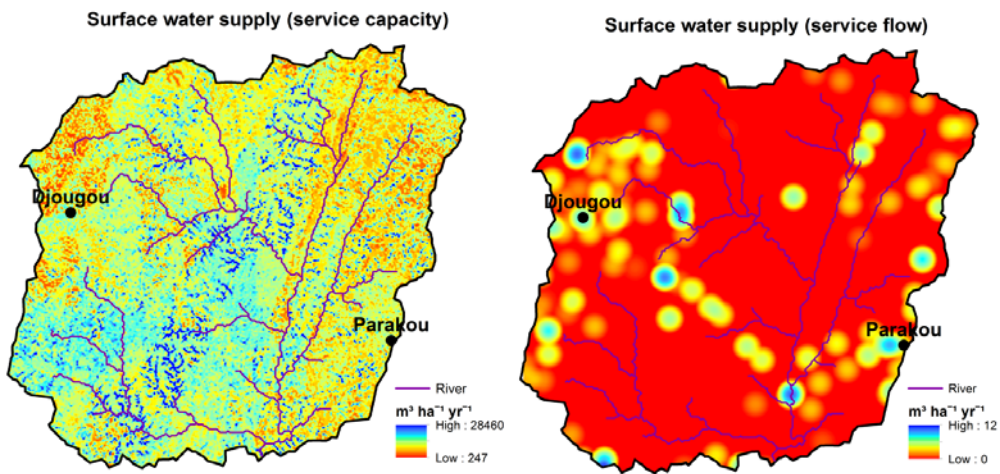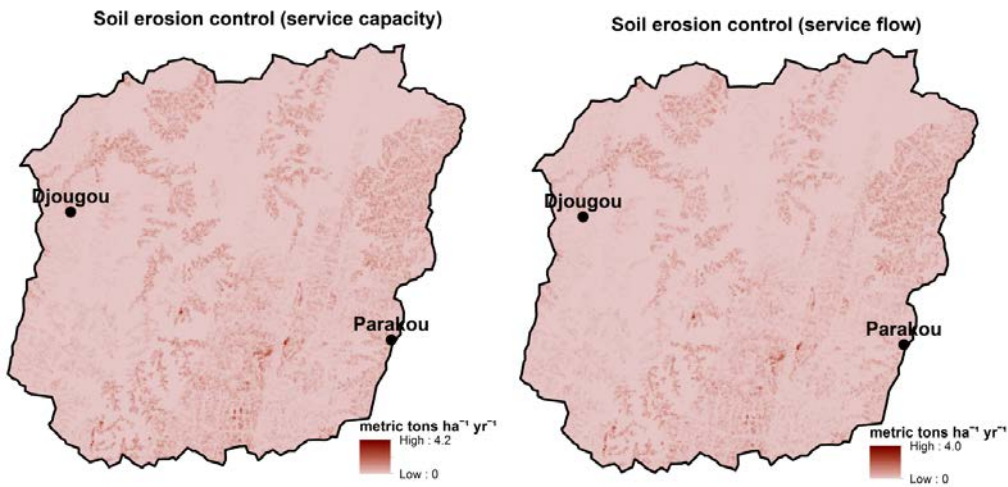
Figure 6. Spatial distribution of mean annual values of service capacity and service flow of surface water supply in the Upper Ouémé watershed from the year 2001 to 2012.
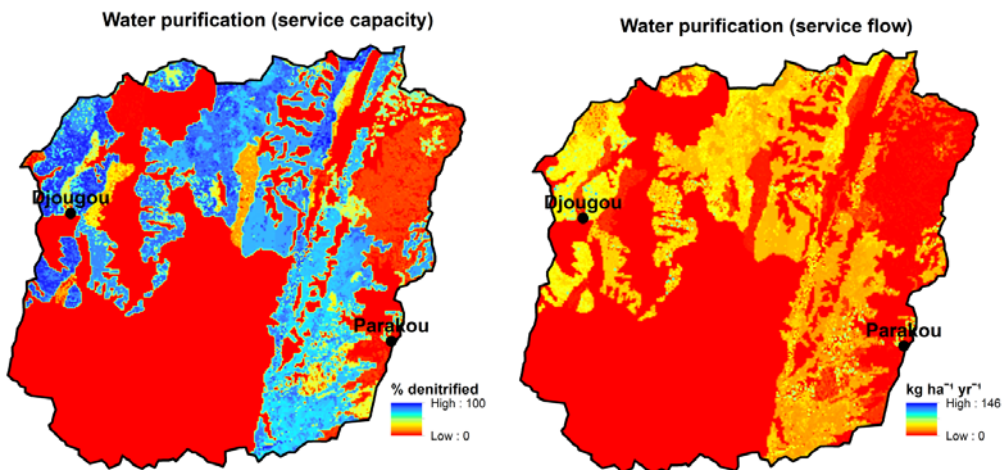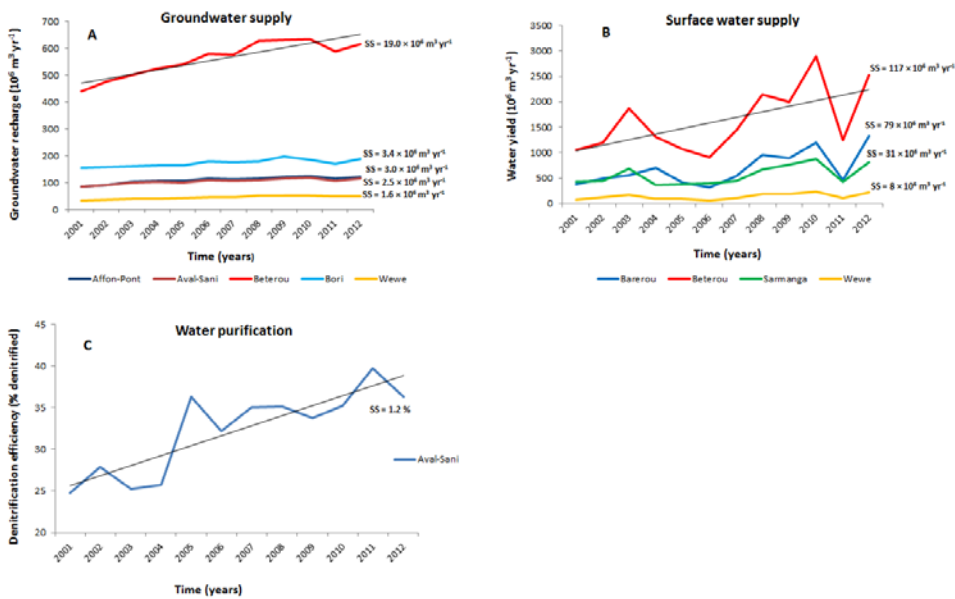

Figure 7. Spatial distribution of mean annual values of service capacity and service flow of water purification in the Upper Ouémé watershed from the year 2001 to 2012.

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

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