

24 July 2018

Dear Editor,

We would like to thank you for the opportunity to resubmit a revised version of the manuscript entitled “Remote land use impacts on river flows through atmospheric teleconnections”. We are grateful for the constructive comments and feedback from the two reviewers, which we believe have helped improve the manuscript in terms of clarity, structure, and substance. The additional analyses of five river basins with relatively low river flows, as suggested by one of the reviewers, particularly help shed light on the role and relevance of accounting for moisture recycling in water scarce basins.

In response to the comments of reviewer #1 and #2, we have made the following revisions to the manuscript:

- Throughout the manuscript
  - Replaced subscript  $_{pot}$  with  $_{pv}$  to stand for potential vegetation scenario. (In response to Reviewer #2 comment on P4 L20.).
  - Corrected cross-references where necessary (In response to Reviewer #2 comment on P4 L20.).
- Introduction
  - P.2: clarified how this study builds upon previous studies. (In response to Reviewer #1 comment 2, and Reviewer #2 comment on P2 L31.)
- Methods
  - Moved most of the methods descriptions from Supplementary Information to the main manuscript, including moving figure illustrating the model coupling procedure (now Fig. 1) to the main manuscript. (In response to Reviewer #1 comment 3.)
  - Added figure of land-use change scenarios and improved colour scheme (the new Fig. 2), as well as caption. (In response to Reviewer #1 comment 6, and Reviewer #2 comment on P6 Figure 1a.)
  - Improved the explanation of the water balance equation  $Q = P - E$  (In response to Reviewer #2 comment on P3 L14.).
- Results
  - Moved maps of actual hydrological flows and precipitation recycling ratio to the main manuscript (new Fig. 3) from the SI.
  - Moved relative changes in hydrological flows to the main manuscript (new Fig. 4) from the SI. Added absolute change in precipitation recycling ratio (Fig. 4e) for completeness. Clarified that river flow changes refer to “river flows changes at outlet”. (In response to Reviewer #2 comment on P6 Figure 1b-d, and P6 L1-8)
  - Added literature sources of the now Fig. 5 directly in the figure, so readers no longer need to consult the SI for this information. (In response to Reviewer #2 comment on P7 Figure 2.)
  - Added new figure (Fig. 6) that shows the effect of TMR on river flows globally, and description of this figure in the text. (This addition was not explicitly requested by the reviewers, nor included in the earlier authors’ comments. Nevertheless, during the implementation of the revisions, we found it to partly address Reviewer #2 main suggestion 3.)

- Added five basins (Zambezi, Niger, Yellow, Huai, Colorado) to Fig. 7 and added a sentence in the figure caption to notify readers that there are two axes in the figure. (In response to Reviewer #2 main suggestion 2, comment on P8, and P8 Fig3.)
- Moved figures of precipitation- and evaporationsheds and their changes (new Fig. 8 and 9) of four basins (Amazon, Yangtze, Yenisei, Niger) from the SI to the main paper, and added descriptions in the text accordingly.
- Moved figure of nation influence in river flows (new Fig 10) of all 15 basins from the SI to the main paper, and added descriptions in the text accordingly.
- Discussion
  - Added discussion of the study's limitations in terms of circulation, ocean feedback, and dams to the Limitations section. (In response to Reviewer #1 comment 9, Reviewer #2 comment on P10 L 21ff, and unnumbered comment on atmospheric circulation and dams.)
  - Moved substantial information regarding alternative methods to investigate land-rainfall-river flow feedback from the SI to the Limitations subsection. (In response to Reviewer #1 comment 4.)
  - Added recent references.
- Supplementary Information
  - Information moved from SI to the main manuscript as specified above. (Also in response to Reviewer #1 comment 8, and Reviewer #2 main suggestion 1.)
  - Explained the limitation of the GRDC dataset in the caption of Fig. S2. (In response to Reviewer #2 comment on P4 L15 and S8)
  - Ordered the SI figures and tables in the order they are mentioned in the main manuscript. (Previously, there were two sections in SI: "Figures and Tables" and "Methods".)
  - Added Fig. S1 that explains the change in parameterisation approach in STEAM for this study. (Not suggested by reviewers, but we found it relevant during the revision process.)
  - Added Table S3 that provides an overview of other studies' estimates of land-use change induced increases and decreases in evaporation in absolute and relative terms. (In response to Reviewer #1 comment 1.)
  - Added Fig. S3 that illustrates the land-cover and land-use scenarios used in the study. (In response to Reviewer #1 comment 6.)
  - Added information of four of the new additional five basins (Zambezi, Yellow, Huai, Colorado) to Fig. S4-S7. (In response to Reviewer #2 main suggestion 2, and comment on P8.)

In a few instances, the reviewers made suggestions we did not agree with or assume are misunderstandings, as we also explained in the authors' comments posted in Interactive Discussion. In summary, these include:

- Reviewer #1 comment 5 posed the question whether "there are sensitivity tests to achieve" the convergence requirements values used for determining the number of model coupling iterations. We clarified in our earlier author's comment that the convergence criterion can be seen as "a sort of sensitivity test in itself". We hope that the current revisions in the Methods section and improved explanation of our model coupling procedure will make this point clearer.

- Reviewer #1 comment 7 suggests including changes in P and E on the bar chart plots, which are already included (Fig. 7), and therefore did not require any revisions.
- Reviewer #1 comment 10 suggest showing surface temperature changes, which is not simulated in this study.

Full details of the revisions are shown in the marked-up manuscript.

We hope that the revised manuscript is now suitable for publication in HESS. Thank you for your consideration.

Sincerely yours,

Lan Wang-Erlandsson on behalf of the authors

# Remote land use impacts on river flows through atmospheric teleconnections

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**Abstract.** The effects of land-use change on river flows have usually been explained by changes within a river basin. However, land-atmosphere feedback such as moisture recycling can link local land-use change to modifications of remote precipitation, with further knock-on effects on distant river flows. Here, we look at river flow changes caused by both land-use change and water use within the basin, as well as modifications of imported and exported atmospheric moisture. We show that in some of the world's largest basins, precipitation was influenced stronger by land-use change occurring outside than inside the basin. Moreover, river flows in several non-transboundary basins was considerably regulated by land-use changes in foreign countries. We conclude that regional patterns of land-use change and moisture recycling are important to consider in explaining runoff change, integrating land and water management, and informing water governance.

## 1 Introduction

River flows ( $Q$ ) are fundamental for ecosystems, nutrient transport, hydropower, navigation, and human well-being (Oki and Kanae, 2006). Land-use change (LUC) has been suggested to be the most important driver of both past (Piao et al., 2007; Sterling et al., 2012) and future (Betts et al., 2015; Milly et al., 2005) changes in river flows ( $\Delta Q$ ). Central to the analysis of  $Q$  is the river basin unit, and estimates of  $\Delta Q$  from LUC often assume that impacts occur exclusively within a basin (Gerten et al., 2008; Piao et al., 2007; Rost et al., 2008a, b; Sterling et al., 2012). Water governance is strongly focused on frameworks such as the Integrated River Basin Management (IWRM) and largely assumes that there is no land-atmosphere feedback, even in discussions of spatial misfit between institutions and hydrological realities (Hoekstra, 2010; Giordano et al., 2015). In fact, land-atmosphere feedbacks are not incorporated in most recent literature on a wide range of topics of relevance for water management, such as virtual water (Dalin et al., 2017), the freshwater planetary boundary (Rockström et al., 2009; Steffen et al., 2015), water scarcity (Mekonnen and Hoekstra, 2016), relative role of climate and LUC for water flows (Zheng et al., 2016), and land acquisition impacts on water (Johansson et al., 2016; Rulli et al., 2012).



However, studies on land-atmosphere interactions clearly shows that changes in land surface properties can considerably influence precipitation ( $P$ ) and  $Q$  through land-atmosphere feedback, sometimes well beyond the local scale (Badger and Dirmeyer, 2016; Garcia et al., 2016; Avissar and Werth, 2005). For example, general circulation model simulations suggest that complete deforestation of Central Africa may decrease February  $P$  by 35 % in the Great Lakes region (Avissar and Werth, 2005), and irrigation in India may support up to 40 % of the  $P$  in some arid regions in Eastern Africa (de Vrese et al., 2016).

5 Under a business-as-usual deforestation scenario,  $Q$  in the Xingu river basin in the Amazon was found to increase by 10-12 % without land-atmosphere feedback, and decrease by 30-36 % when such feedback was taken into account (Stickler et al., 2013). Furthermore, statistical analyses of observed data suggests that irrigation in the US high plains enhances downwind  $Q$  (Kustu et al., 2011) and coupled regional climate modeling shows that irrigation in the California Central Valley can be linked to about 30 % increase in Colorado  $Q$  (Lo and Famiglietti, 2013). At the global scale,  $\Delta Q$  from future climate and

10 LUC scenarios changed from decrease to increase by considering land-atmosphere feedback and by closing the water balance (Arnell and Lloyd-Hughes, 2012; Betts et al., 2015)(Betts et al., 2015).

Land-atmosphere interactions can influence  $Q$  through thermal layer processes, terrestrial moisture recycling (TMR), and circulation perturbation (Goessling and Reick, 2011). First, thermal layer processes refer to the boundary layer and mesoscale circulation perturbation that may lead to a change in total terrestrial evaporation ( $E$ ) and can locally lead to both positive and

15 negative  $P$  responses (Guillod et al., 2015; Seneviratne et al., 2010; Koster et al., 2003). Local forest clearing has for example been shown to enhance  $P$  in downwind areas due to turbulence changes (Khanna et al., 2017; Saad et al., 2010). Second, TMR refers to the process of terrestrial  $E$  returning to land as  $P$  and is underpinned by the mass conservation of water (Brubaker et al., 1993). TMR is often the dominating land-atmosphere process at the regional to continental scale (D’Almeida et al., 2007; Spracklen et al., 2012; Lawrence and Vandecar, 2014; Tuinenburg, 2013). About 40 % of global terrestrial  $P$  (van der

20 Ent et al., 2014) originates from terrestrial  $E$  and the average distance traveled in the atmosphere is 500-5000 km (van der Ent and Savenije, 2011) – a distance likely to exceed the size of most river basins. Lastly, large-scale atmospheric circulation perturbation allow extreme LUC (e.g., complete tropical deforestation) to impact  $P$  in geographically remote regions and continents in unexpected ways (Avissar and Werth, 2005; Badger and Dirmeyer, 2016; Garcia et al., 2016; Lawrence and Vandecar, 2014). Monsoon regions are particularly sensitive to circulation perturbation, and irrigation may for example reduce

25  $P$  by weakening the monsoon onset (Tuinenburg, 2013).

The previous studies that illustrated the importance of remote LUC for basin  $P$  and  $Q$ , did not ~~systematically assess global effects of LUC~~ examine the effect of taking moisture recycling into account for estimating LUC effects on  $Q$  ; or explore and attribute to influence from different nations, nor analysed the interplay between LUC within and outside the river basin. These effects are, however, important to disentangle since they can have profound water governance implications

30 ~~(for e.g., for for example riparian water rights and transboundary river basin treaties ). While there has been discussions of governance implications of land-atmosphere interactions purely based on atmospheric moisture fluxes between nation states (Keys et al., 2017; Dirmeyer et al., 2009; Ellison et al., 2017), no studies have quantified the magnitude of LUC impacts on  $P$  or  $Q$ , despite its high relevance for international water law and governance~~ (Keys et al., 2017; Dirmeyer et al., 2009; Ellison et al., 2017).

Thus, there is a missing interdisciplinary bridge between understanding the role of land-atmosphere feedback over large distances and its importance for water governance at the basin scale.

This study aims to (i) investigate the potential impacts of human LUC on  $Q$  worldwide accounting for TMR, (ii) disentangle the relative influence on  $Q$  from within- and extra-basin LUC, (iii) attribute potential human LUC impacts on  $Q$  to nation states, and (iv) discuss the potential implications for water governance. We focus on the TMR effect because it is transparent, closes the water balance, and explicitly links changes in land and water geographically. Given these advantages, similar TMR approaches have in recent years been used to ~~analyze-analyse~~ unexplored relations, e.g., LUC impacts of crop yields (Bagley et al., 2012), self-amplifying forest die-back from TMR changes (Zemp et al., 2017), and vulnerability to LUC induced reductions in  $P$  (Keys et al., 2016; Miralles et al., 2016). For a comparison of different methods for ~~analyzing-analysing~~ LUC impacts on  $Q$ , see Table S1.

## 2 Methods

### 2.1 ~~Modeling~~Modelling

#### 2.1.1 ~~Hydrological modelling~~

We used the process-based hydrological model Simple Terrestrial Evaporation to Atmosphere Model (STEAM) (Wang-Erlandsson et al., 2014) to simulate water fluxes based on land cover and land use. STEAM partitions evaporation into five fluxes: vegetation interception, floor interception, transpiration, soil moisture evaporation, and open-water evaporation. STEAM uses the Penman-Monteith equation (Monteith, 1965) to estimate potential evaporation, the Jarvis-Stewart equation (Stewart, 1988) to compute stomatal resistance, and Jolly's growing season index (function of minimum temperature, soil moisture content, and daylight) to describe phenology (Jolly et al., 2005). STEAM operates at ~~1.5° x 1.5°~~ 1.5° x 1.5° and a 3 hour resolution. Based on the long term water balance, mean annual river flow ( $Q$ ) is assumed to approximately equal the difference between mean annual precipitation ~~and evaporation. Minor modifications in land parametrization for this study and comparison against observed~~ ( $P$ ) and evaporation ( $E$ ), i.e.,  $Q = P - E$ . STEAM was validated in previous studies (Wang-Erlandsson et al., 2014, 2016) and compared well with recent observation based analyses of evaporation partitioning by land-cover type (Wei et al., 2017). Modifications from the original version of STEAM (Wang-Erlandsson et al., 2014, 2016) includes: (1) update of land-use classification, parameterisation, and parametrisation approach (Table S2, and Fig. S1), (2) use of a temperature threshold of 0 °C for differentiating snowfall from rainfall, and (3) differences in input data (i.e., root zone storage capacity, land surface map, precipitation data source as described in Data). Evaluation against runoff data is shown in ~~Supplementary-Information-Fig. S2.~~ Simulated land-use change effects on evaporation increase and decrease are compared with literature values in Table S3 and found to be in the conservative range. With the study period being 2000–2013, the years 1995–1999 were used as spin-up for STEAM.

### 30 2.1.2 Moisture tracking

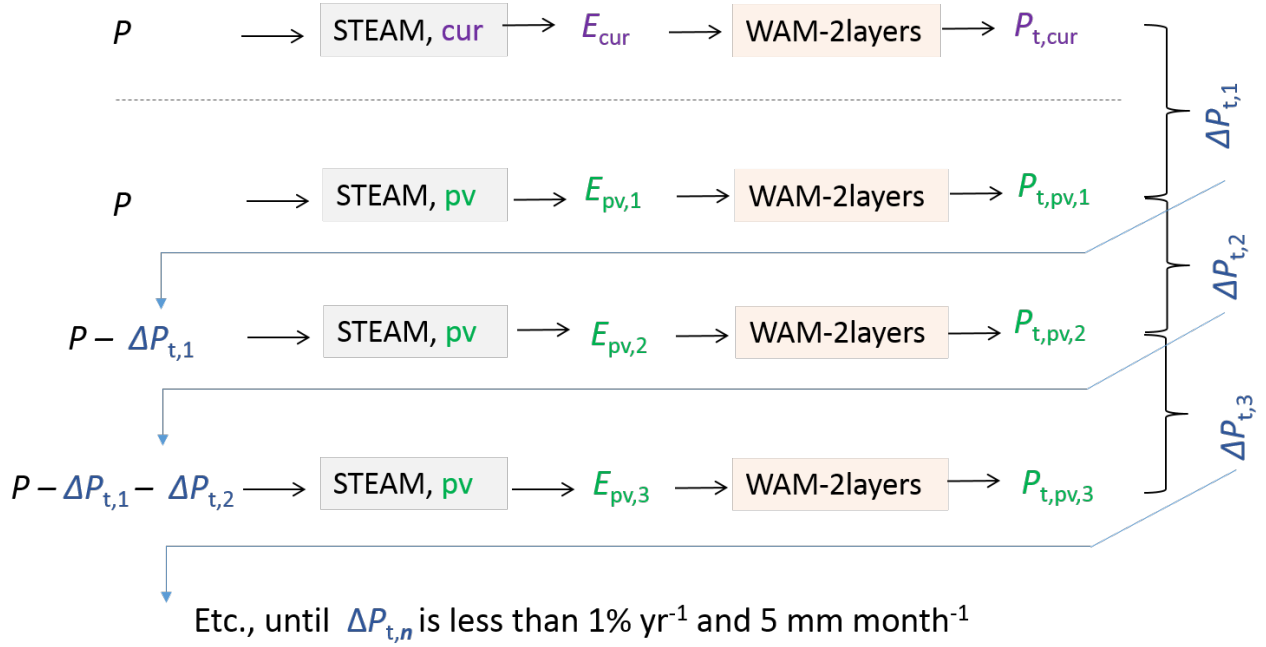
Atmospheric moisture is tracked using the Eulerian moisture tracking scheme Water Accounting Model-2 layers (WAM-2layers) (van der Ent, 2014; van der Ent et al., 2014). WAM-2layers tracks atmospheric moisture from zero pressure to surface pressure in two layers. Within the layers, atmosphere is assumed to be well-mixed. WAM-2layers tracks vapor flows by applying the water balance. For example, the following equation is used to track where evaporation from a given region falls as precipitation (i.e., forward tracking):

$$\frac{\partial S_{\text{tracked}}}{\partial t} = \frac{\partial (S_{\text{tracked}}u)}{\partial x} + \frac{\partial (S_{\text{tracked}}v)}{\partial y} + E_{\text{tracked}} - P_{\text{tracked}} \pm F_{\text{vertical,tracked}} \quad (1)$$

where  $S_{\text{tracked}}$  is the tracked atmospheric storage in an atmospheric column in one layer,  $t$  is time,  $u$  and  $v$  are wind components in the  $x$  zonal and  $y$  meridional direction,  $E_{\text{tracked}}$  is tracked evaporation entering and  $P_{\text{tracked}}$  is precipitation exiting an atmospheric column and layer, and  $F_{\text{vertical,tracked}}$  is the tracked vertical moisture transport between the two layers. An analogous equation is used for tracking the source of precipitation to a given region (i.e., backward tracking). The spatial resolution of WAM-2layers is  $1.5^\circ$  and input data are linearly interpolated to the 15 minute time step to maintain numerical stability. WAM-2layers has been employed previously for analysing atmospheric moisture transport over terrestrial areas (Keys et al., 2012, 2016) and validated against other types of moisture tracking algorithms (van der Ent et al., 2013). We used the MATLAB (The MathWorks, 2014) version of WAM-2layers, but a Python version is also openly available on Github (van der Ent, 2016). With the study period being 2000–2013, the year 1999 is used as spin-up in forward tracking in WAM-2layers, and 2014 is used as spin-up for backward tracking in WAM-2layers.

### 2.1.3 Coupling of the moisture tracking scheme and the hydrological model

Hydrological flows in the current land-use scenarios is simply represented by current data and simulation. To obtain  $E$  and  $P$  under potential land cover is obtained through a coupled model simulation. We use STEAM is coupled with WAM-2layers by (1) simulating present day  $E$  output from STEAM in STEAM and forward tracking terrestrial  $E$  with WAM-2layers, and iteratively adjust the current meaning that the  $E_{\text{tracked}}$  is equal to all evaporation from terrestrial surfaces, i.e., not belonging to the oceans, (2) simulating  $E$  in STEAM based on present day  $P$  and potential land cover, and forward track the fate of terrestrial  $E$  with WAM-2layers, (3) calculating the change in  $P_{\text{tracked}}$ , (4) updating the present day  $P$  forcing to STEAM with the changes in  $P_{\text{tracked}}$ , and (5) simulating  $E$  in STEAM based on updated  $P$  with terrestrial origin obtained by forward tracking continental moisture in and potential land cover, and forward tracking the fate of terrestrial  $E$  with WAM-2layers (SI Materials and Methods), see Fig. 1. Steps 3–5 are iterated until the annual  $P$  differences between model iterations converges after about four simulations: change is below 1 % and the monthly  $P$  change is below  $5 \text{ mm month}^{-1}$  in every grid cell, which in our case ultimately resulted in four iterations in total. This procedure assumes that land-use induced changes in terrestrial  $E$  will result in proportional changes in  $P$  with terrestrial origin.



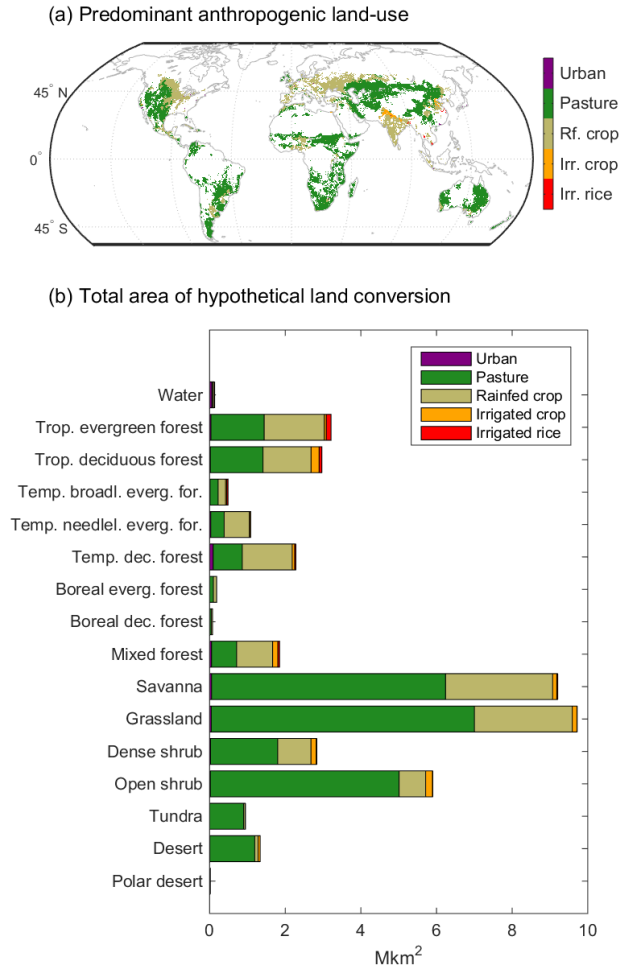
**Figure 1.** Model coupling schematic. Model coupling between STEAM and WAM-2layers based on current land use and potential vegetation scenarios.  $P$  stands for current precipitation,  $E$  stands for evaporation. Subscript t stands for terrestrial origin, pv denotes simulation with potential vegetation, cur denotes simulation with current land use, and  $n$  stands for the number of iterations.

## 30 2.2 Data

### 2.2.1 Land data

Land use and land cover data input to STEAM are based on the Ramankutty potential land-cover (Ramankutty and Foley, 1999) and current land-use scenarios (Ramankutty et al., 2008) for consistency. We further added permanent wetlands, permanent snow or ice, and urban or built-up areas from the Land Cover Type Climate Modeling Grid (CMG) MCD12C1 International Geosphere Biosphere Program (IGBP) land classification created from Terra and Aqua Moderate Resolution Imaging Spectroradiometer (MODIS) data (Friedl et al., 2010) for the year 2005. Monthly irrigated rice and irrigation non-rice crops were obtained from the data set of Monthly Irrigated and Rainfed Crop Areas around the year 2000 (MIRCA2000) V1.1. (Portmann et al., 2010)(~~SI Materials and Methods~~). The urban and irrigated areas were only added to the current land cover map. In this merging procedure, MODIS is allowed to override the Ramankutty datasets, and MIRCA2000 is allowed to override the Ramankutty map as long as it does not extend over the cropland areas. The scenarios used are shown in Fig. S3 and the land-use change is illustrated in Fig. 2.

The root zone storage capacity map is based on a climate-observation based root zone storage capacity ( $S_R$ ) (Wang-Erlandsson et al., 2010) from satellite and energy balance based evaporation, gauge-based precipitation, and modelled irrigation. The best performing



**Figure 2.** Changes in land use resulting from the replacement of the potential vegetation scenario with the current land-use map. Changes in a, land use (current land use is shown, with grid cells without major land-use change masked out), b, total area of difference between potential vegetation (y-axis) and current land-use map (colour legend).

Gumbel normalised root zone storage capacity ( $S_{R,CRU-SM,merged}$ ) was used. Root zone storage capacity for both current and potential land-cover and land-use scenarios were constructed from mean of land-cover type and Köppen-Geiger climate class (Kottek et al., 2006). The mean root zone storage capacity of single land-cover types was used only in places where the combination of land-cover type and climate zone that exists in the potential land-cover scenario did not exist in the current land-use map.

## 5 2.2.2 Meteorological forcing and runoff data

Meteorological data used in WAM-2layers and STEAM, except for land precipitation, were taken from the Earth Retrospective Analysis Interim (ERA-I) from the European Centre for Medium-Range Weather Forecasts (ECMWF) (Dee et al., 2011). ERA-I meteorological forcing to STEAM are: snowmelt, temperature at 2 m height, dew point temperature at 2m height, wind speed (meridional and zonal vectors) at 10 m height, incoming shortwave radiation, and net longwave radiation. In addition, ERA-I evaporation data were used to downscale calculated daily potential evaporation in STEAM to the 3 hour time step. ERA-I model level forcings used in the WAM-2layers are specific humidity, and wind speed at 6 hourly resolution, spanning from zero to surface pressure. Moreover, 3 hourly ocean evaporation is taken from ERA-I. The Modern-Era Retrospective analysis for Research and Applications (MERRA) reanalysis has in a previous study been used as input to WAM-2layers for comparison and generated similar persistent moisture recycling patterns, except in South America where differences arise due to underestimation of precipitation in MERRA (Keys et al., 2014). Precipitation forcing for WAM-2layers and STEAM both come from the state-of-the-art product Multi-Source Weighted-Ensemble Precipitation (MSWEP V1) (Beck et al., 2017) that was specifically created for hydrological modeling. The use of MSWEP as forcing for STEAM resulted in runoff estimates that compare well to observed runoff data (Figure S8Fig. S2). All meteorological forcing data cover temporally 1995–2014, whereas runoff data represent long term average. The years 1995–1999 were used as spin-up for STEAM, 1999 is used as spin-up in forward tracking in WAM-2layers, and 2014 is used as spin-up for backward tracking in WAM-2layers. 2014.

Runoff data used for benchmarking were taken from the composite (observed river discharge consistent with water balance model) from the Global Runoff Data Centre (GRDC) (Fekete et al., 2002). The separate GRDC water balance model runoff fields are included in the comparison for reference (Fig. S2). Where available in the literature, we also compare our simulated river flows to more reliable and recent discharge data in individual basins.

The spatial coverage of all data used is 57°S–79.5°N latitudes at 1.5° x 1.5° resolution. MSWEP originally at 0.25° and GRDC runoff at 0.5° were aggregated to 1.5° resolution by simple averaging.

## 2.3 Analyses

### 2.3.1 Changes in hydrological flows

River flow change without TMR ( $\Delta Q_{\text{noTMR}}$ ) is

$$25 \quad \Delta Q_{\text{noTMR}} = (P_{\text{cur}} - E_{\text{cur}}) - (P_{\text{cur}} - E_{\text{pot,1pv,1}}) \quad (2)$$

where  $P_{\text{cur}}$  is current day precipitation data from MSWEP,  $E_{\text{cur}}$  is current day evaporation based on STEAM simulation, and  $E_{\text{pot,1pv,1}}$  results from STEAM simulation in the potential vegetation scenario and forced with current day precipitation (Fig. S81). River flow change after accounting for TMR ( $\Delta Q$ ) is

$$\Delta Q = (P_{\text{cur}} - E_{\text{cur}}) - (P_{\text{pot,4pv,4}} - E_{\text{pot,5pv,5}}) \quad (3)$$

30 where  $P_{\text{pot},4}$   $P_{\text{pv},4}$  is the converged precipitation (i.e., meeting the convergence requirement of mean annual precipitation change  $< 1\%$  and monthly precipitation change  $< 5$  mm/month in every grid cell) achieved at the fourth iterative coupling between STEAM and WAM-2layers, and  $E_{\text{pot},5}$   $E_{\text{pv},5}$  is the evaporation under the potential vegetation scenario simulated in STEAM with precipitation forcing  $P_{\text{pot},4}$   $P_{\text{pv},4}$ .

5 ~~Precipitation change in a basin that originates from extra-basin evaporation ( $\Delta P_{\text{import}}$ ) is defined as the change in~~ tracked basin precipitation ( $\Delta P_{\text{tracked,basin}}$ ) occurring ~~outside~~ outside the river basin boundaries ~~, whereas change in internally~~ ~~recycled precipitation ( $\Delta P_{\text{basin-recycling}}$ ) is defined as is referred to as  $\Delta P_{\text{import}}$ , whereas~~  $\Delta P_{\text{tracked,basin}}$  originating from ~~within~~ within the basin boundary is referred to as  $\Delta P_{\text{basin-recycling}}$ . Internally recycled evaporation ( $\Delta E_{\text{basin-recycling}}$ ) corresponds to  $\Delta P_{\text{basin-recycling}}$  and all other basin evaporation change is considered exported ( $\Delta E_{\text{basin,recycling}}$ ).

### 2.3.2 Country influence on changes in river flows

10 The influence on river flow change in river basin b from country c without considering TMR ( $I_{\text{b,c,noTMR}}$ ) is:

$$I_{\text{b,c,noTMR}} = |\Delta E_{\text{b,c}}|. \quad (4)$$

where  $\Delta E_{\text{b,c}}$  is evaporation change in the part of river basin b located in country c. The influence on river flow change in basin b from country c with consideration of TMR ( $I_{\text{b,c,TMR}}$ ) is:

$$I_{\text{b,c,TMR}} = |\Delta E_{\text{b,c,export}}| + |\Delta P_{\text{b,c,import}}| \quad (5)$$

15 where  $\Delta E_{\text{b,c,export}}$  is the evaporation change exported from the part of basin b located in country c, and  $\Delta P_{\text{b,c,import}}$  is the precipitation change imported to basin b from country c.

Influences from countries below 5 % of total influences in a specific basin ( $I_{\text{b,c,noTMR}} < 0.05 \times \sum I_{\text{b,c,TMR}}$ ) were lumped into the category "Other".

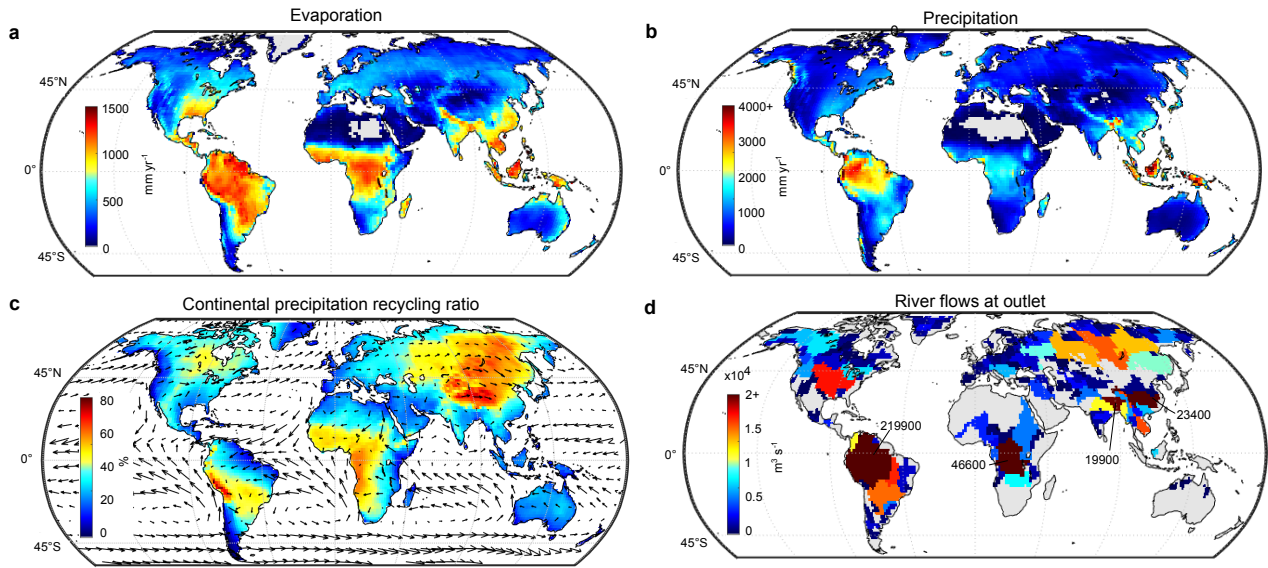
## 3 Results

### 20 3.1 LUC impacts on global water flows

Our results show that human LUC (from potential land cover to current land use) (Fig. ~~1a2~~) has led to reductions in  $E$  and  $P$ , and to increases in  $Q$ , in ~~most regions~~ large parts of the world (Fig. ~~1b-d~~ 4b-d).  $E$  has decreased primarily in Southwest China, Europe, West Africa, south of Congo, and southeast South America resulting from substantial pasture and agricultural expansion (Ramankutty et al., 2008). Following prevailing wind directions (Fig. ~~3e3c~~), subsequent  $P$  has decreased in all  
25 tropical regions, in South Central China, eastern US, and Europe.

Nevertheless, in some areas,  $E$  increased due to incremental irrigation ~~—~~ notably in India, the Western US, Northeast China, and in the Middle East (Fig. ~~1b4a,b~~). Due to the combination of heavy irrigation in India and orography,  $P$  has increased substantially along the Himalaya mountain ridge (Fig. ~~1e4b,c~~). Weak increases in  $P$  are observed in other downwind regions:





**Figure 3.** Differences between the current land-use and the potential land cover scenario. Current mean annual hydrological flows 2000–2013. Changes in **a**, land-use (only shifts in grid-cell dominant land-use types are shown), Current evaporation simulated by STEAM, **b**, mean annual evaporation–current precipitation (changes  $< \pm 20 \text{ mm yr}^{-1}$  are in white; MSWEP data), **c**, mean annual–current continental precipitation recycling ratio (changes  $< \pm 4 \text{ mm yr}^{-1}$  are in white; i.e., precipitation with terrestrial origin divided by total precipitation:  $P_{\text{tracked}}/P$ ) where arrows show average winds in the lower atmosphere, and **d**, mean annual–current river flow (changes  $< \pm 200 \text{ m}^3 \text{ yr}^{-1}$  are in white) at outlet based on  $P-E$ . Note that changes in river flows—Values below about 0.5 % of maximum display value are aggregated to river-basin level and shown in  $\text{m}^3 \text{ s}^{-1}$  grey. Changes in hydrological flows are based on model simulations with meteorological input data covering the years 2000–2013.

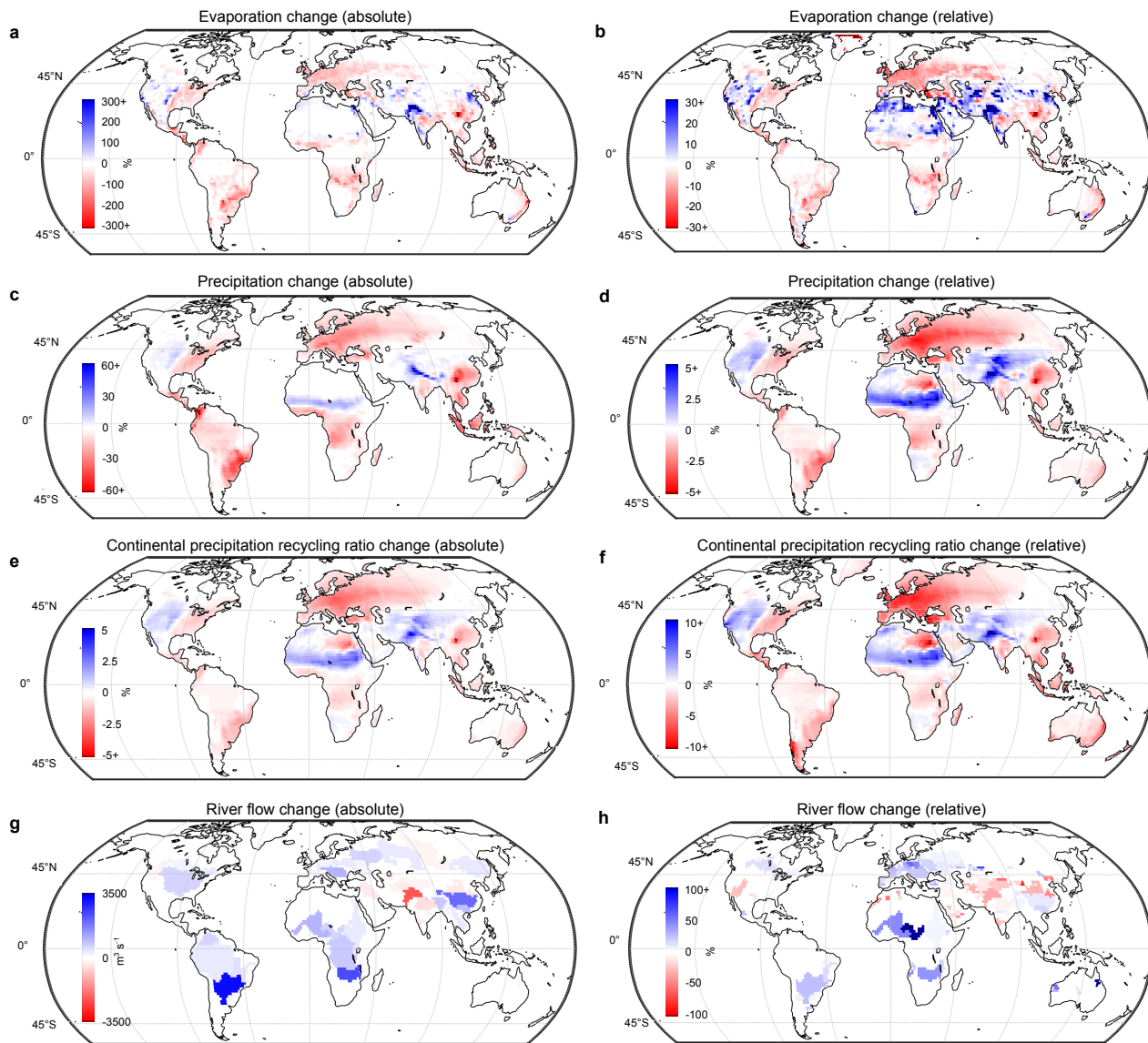
the Sahel (i.e., downwind irrigation areas along the Nile) and in the Western US. Continental precipitation recycling ratios are modified — with some exceptions — in a similar pattern as  $P$  (Fig. 4e,f). Large  $\Delta Q$  are seen in the La Plata basin in South America, the Zambezi in Southern Africa, the Yangtze in China, and the Indus in North India (Fig. 1d). Figure S1 shows the corresponding 4g), and relative changes in hydrological flows.  $Q$  are large in for example the Colorado basin in the US, the Odra basin in eastern Europe, and Lake Chad river basin in Africa (Fig. 4g).

### 5 3.2 The role of TMR for $\Delta Q$

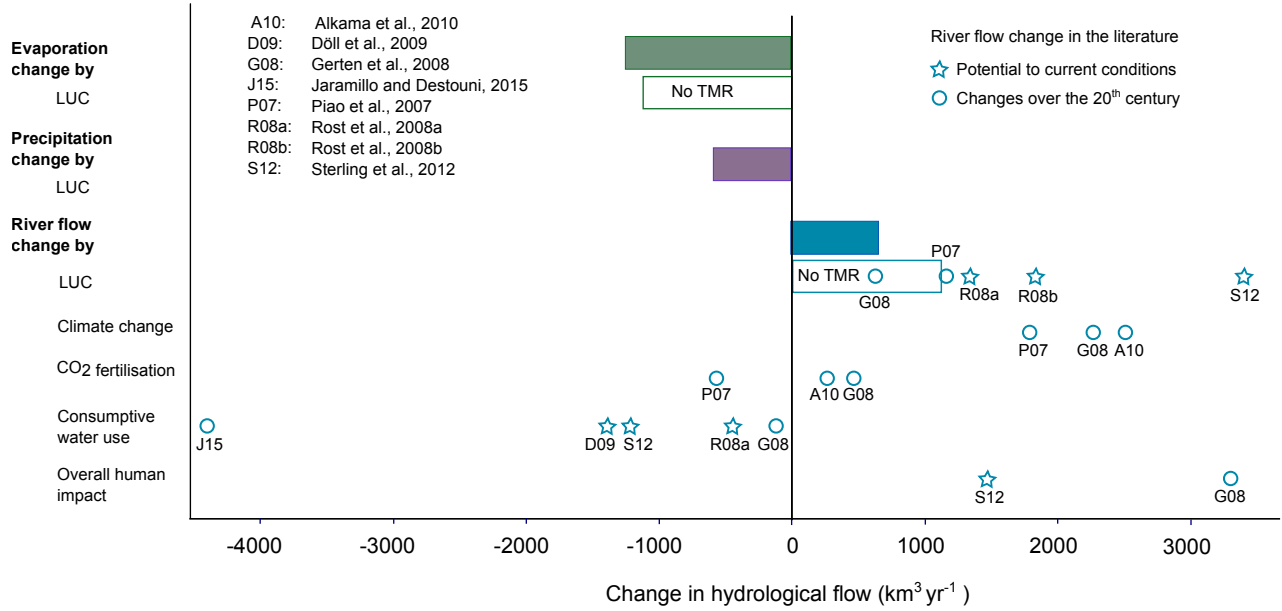
In aggregate (Fig. 25), when accounting for TMR, LUC changed global terrestrial  $E$  by  $-1251$ – $-1251 \text{ km}^3 \text{ yr}^{-1}$  ( $-1.8$ – $-1.8 \%$  from  $69,211 \text{ km}^3 \text{ yr}^{-1}$ ),  $P$  by  $-586$ – $-586 \text{ km}^3 \text{ yr}^{-1}$  ( $-0.5$ – $-0.5 \%$  from  $107,800 \text{ km}^3 \text{ yr}^{-1}$ ), and  $Q$  by  $664 \text{ km}^3 \text{ yr}^{-1}$  ( $+1.7$ – $+1.7 \%$  from  $38,589 \text{ km}^3 \text{ yr}^{-1}$ ). The estimated changes to  $Q$  tend to fall in the conservative end of previous estimates (Sterling et al., 2012) (Gerten et al., 2008; Piao et al., 2007; Rost et al., 2008a, b; Sterling et al., 2012) (Fig. 2).

5) However, recent research (Jaramillo and Destouni, 2015) suggests that consumptive water use is severely underestimated in earlier studies (e.g., Döll et al., 2009; Sterling et al., 2012).  $\Delta Q$  with TMR corresponds to the difference between  $\Delta E$  and





**Figure 4.** Land-use change induced changes in hydrological flows (current land-use - potential vegetation scenario): **a**, absolute change in evaporation, **b**, relative change in evaporation, **c**, absolute changes in precipitation, **d**, relative change in precipitation, **e**, absolute change in continental precipitation recycling ratio (i.e., precipitation with terrestrial origin divided by total precipitation  $P_{\text{tracked}}/P$  and converted to the unit of percent), **f**, relative change in continental precipitation recycling ratio, **g**, absolute change in river flow at outlet, and **h**, relative change in river flows at outlet.



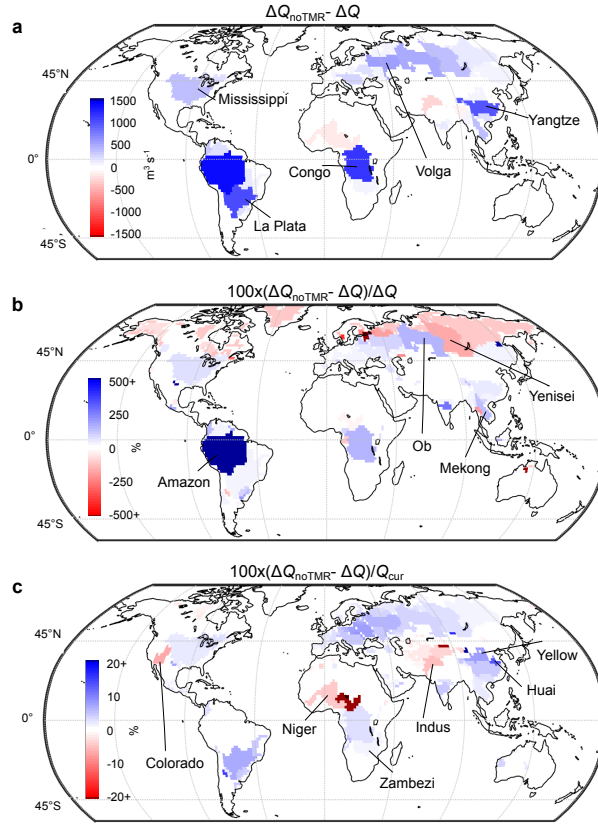
**Figure 5.** Human impact on global hydrological flows. The solid bars show our estimated net change (terrestrial area 131.7 106 km<sup>2</sup> and time period 2000–2013) in evaporation, precipitation and river flows including consideration of terrestrial moisture recycling (TMR). Hollow bars show flow changes without TMR. Circles and stars indicate river flow change estimates from other studies (Table S2S4), where land-use change (LUC) implicitly accounts for consumptive water use. Note that while consumptive water use alone always reduces river flows, other human impacts have both positive and negative influences that are concealed by the global aggregate.

$\Delta P$  change including TMR (Fig. 25, solid bars), whereas  $\Delta Q$  without accounting for TMR simply corresponds to  $\Delta E$  without TMR (Fig. 25, hollow bars).

Including TMR nearly halves the global  $\Delta Q$  estimate. This is because  $E$  returns as  $P$  over land and thus compensates for the initial water 'loss' from the basin. This suggests that previous studies without TMR (e.g., Gerten et al., 2008; Piao et al., 2007; Sterling et al., 2012) may have substantially overestimated the net LUC impacts on  $Q$ . Our estimate of LUC impact on  $Q$  is comparable to slightly larger than some of the estimates of CO<sub>2</sub> fertilization and consumptive water use (i.e., net withdrawal) impacts, but fertilisation (e.g., Alkama et al., 2010; Gerten et al., 2008), but substantially smaller than climate change and overall human impact (e.g., Alkama et al., 2010; Gerten et al., 2008) (Fig. 25).

Our river basin analysis shows that accounting for TMR considerably alters estimates of  $\Delta Q$  (Fig. 3a7a): in the Congo, Volga, and Ob basins,  $\Delta Q$  are reduced by more than half; in the Amazon,  $\Delta Q$  drops from 1630 to 270 m<sup>3</sup> s<sup>-1</sup>; and in the Yenisei, the sign of  $\Delta Q$  is reversed from an increase (150 m<sup>3</sup> s<sup>-1</sup>) to a decrease (-220–220 m<sup>3</sup> s<sup>-1</sup>).

At the basin level, the TMR effect on river flow change is estimated to be the largest in large and relatively wet basins such as the Amazon, Congo, and Yangtze river basins in terms of absolute volumes (Fig. 6a). Not accounting for TMR clearly generates the largest relative deviations in river flow change estimates in the Amazon (i.e.,  $\Delta Q_{\text{noTMR}}$  is approximately five times larger



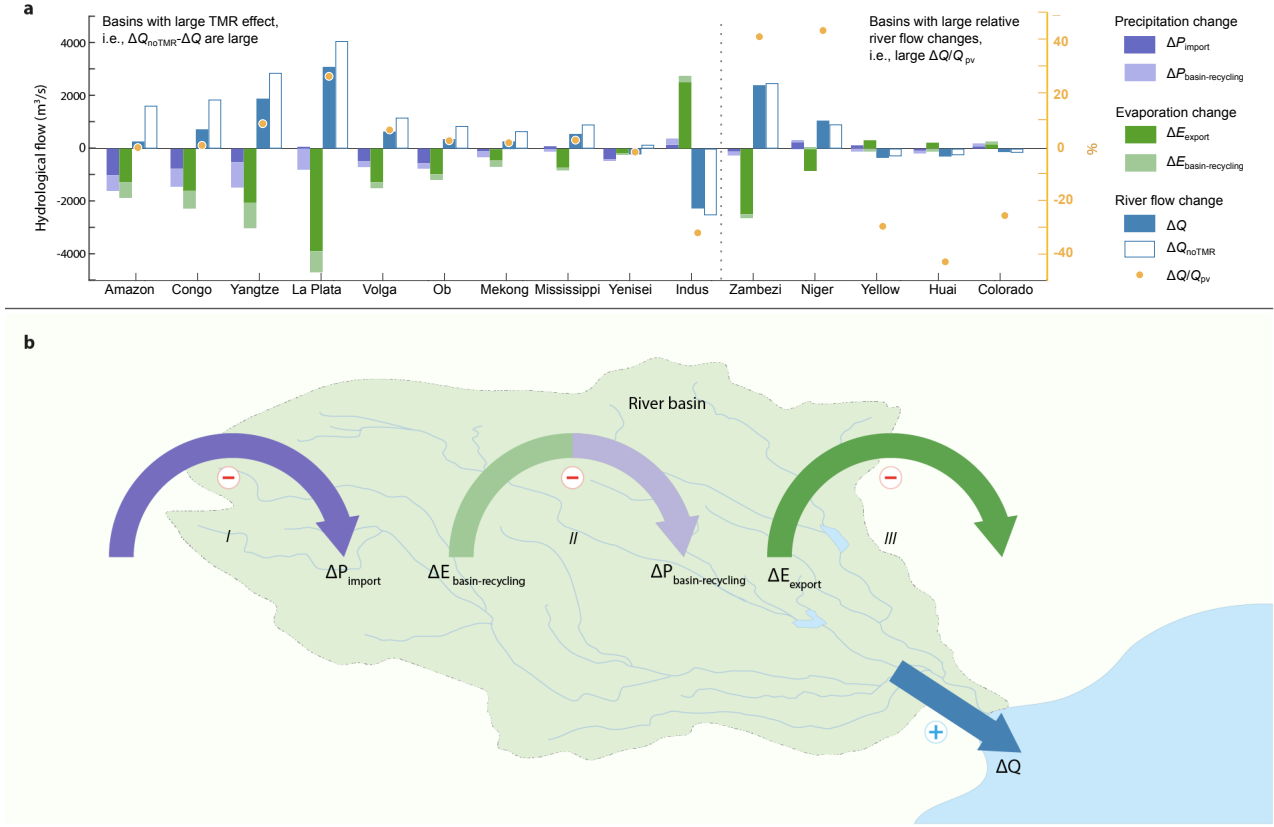
**Figure 6.** The effect of accounting for TMR on river flow change estimates, shown **a**, as absolute difference between river flow change without and with TMR effect, i.e.,  $\Delta Q_{\text{noTMR}} - \Delta Q$ , **b**, as this difference relative river flow change with TMR effect, i.e.,  $(\Delta Q_{\text{noTMR}} - \Delta Q)/\Delta Q$ , and **c**, this difference relative current river flows, i.e.,  $(\Delta Q_{\text{noTMR}} - \Delta Q)/Q_{\text{cur}}$

than  $\Delta Q$ ), and large relative TMR effects are seen in many large basins worldwide, including e.g., Congo ( $\Delta Q_{\text{noTMR}}$  is 150 % higher than  $\Delta Q$ ), Yenisei ( $\Delta Q_{\text{noTMR}}$  is 165 % lower than  $\Delta Q$ ), and Ob ( $\Delta Q_{\text{noTMR}}$  is 140 % higher than  $\Delta Q$ ) river basins (Fig. 6b). The TMR effect relative  $Q_{\text{cur}}$  (Fig. 6c) shows that TMR effect can be important also in more arid basins such as the Colorado, Niger, and the Yellow river.

### 3.3 The interplay between internal and external LUC

- 5 Furthermore, atmospheric moisture does not respect river basin boundaries (Fig 3a-, 7a., and spatial maps in Fig. S2, S38, 9, S4, and S5, S6, and S7). In fact,  $P$  over the basins has been modified more significantly by external than by internal LUC (change in imported precipitation  $\Delta P_{\text{import}} >$  change in internally recycled precipitation  $\Delta P_{\text{basin-recycling}}$ ) in some of the largest basins (Fig. 3a7a). Likewise, internally recycled evaporation changes ( $E_{\text{basin-recycling}}$ ) (Fig. ,3b-, 7b, II) are substantially smaller

than  $\Delta E$  affecting  $P$  elsewhere ( $\Delta E_{\text{basin-recycling}} < \text{change in exported evaporation } \Delta E_{\text{export}}$ ) for all selected river basins (Fig. 3a7a).



**Figure 7.** Changes in hydrological flows at river basin scale. **a**, Changes in hydrological flows by ten of the basins with the largest terrestrial moisture recycling (TMR) effect on river flows ( $Q$ ): (eight basins with increased and two with decreased  $Q$ ). Basins are ordered by maximum absolute TMR effect on  $Q$ , i.e., the difference between  $Q$  and five basins with or without TMR large relative changes in river flows ( $\Delta Q - \Delta Q_{\text{noTMR}}$  two basins with positive  $\Delta Q$  and three with negative  $\Delta Q$ ). **b**, Conceptual figure of hydrological flow changes in a basin. The (-) and (+) in b may be different for different basins, and the (-) and (+) as displayed here are e.g. for example seen in Amazon, Congo, Volga derived from results shown in and Yangtze, see a. In Indus and Yenisei Note also that the figure has two y-axis,  $\text{m}^3 \text{s}^{-1}$  to the (-) left and (+) are reversed: % to the right

Internal moisture recycling (Fig. 3b7b, II) does not affect  $\Delta Q$  directly, only indirectly if  $\Delta P_{\text{basin-recycling}}$  affects subsequent  $\Delta E_{\text{export}}$  under transient change (Fig. 3b7b and Methods). Thus, provided steady-state,  $\Delta Q$  simply corresponds to the difference between  $\Delta E_{\text{export}}$  and  $\Delta P_{\text{import}}$  (Fig. 3a7a). For example,  $\Delta Q$  in the Amazon is very small because the reduced  $\Delta P_{\text{import}}$  is almost entirely offset by reduced  $\Delta E_{\text{export}}$ . In Congo, about half of the within-basin LUC induced  $Q$  increase is counteracted by extra-basin LUC (i.e.,  $\Delta P_{\text{import}} \approx 0.5 \Delta E_{\text{export}}$ ). The effect of TMR on  $\Delta Q$  ( $\Delta Q_{\text{noTMR}} - \Delta Q$ , where

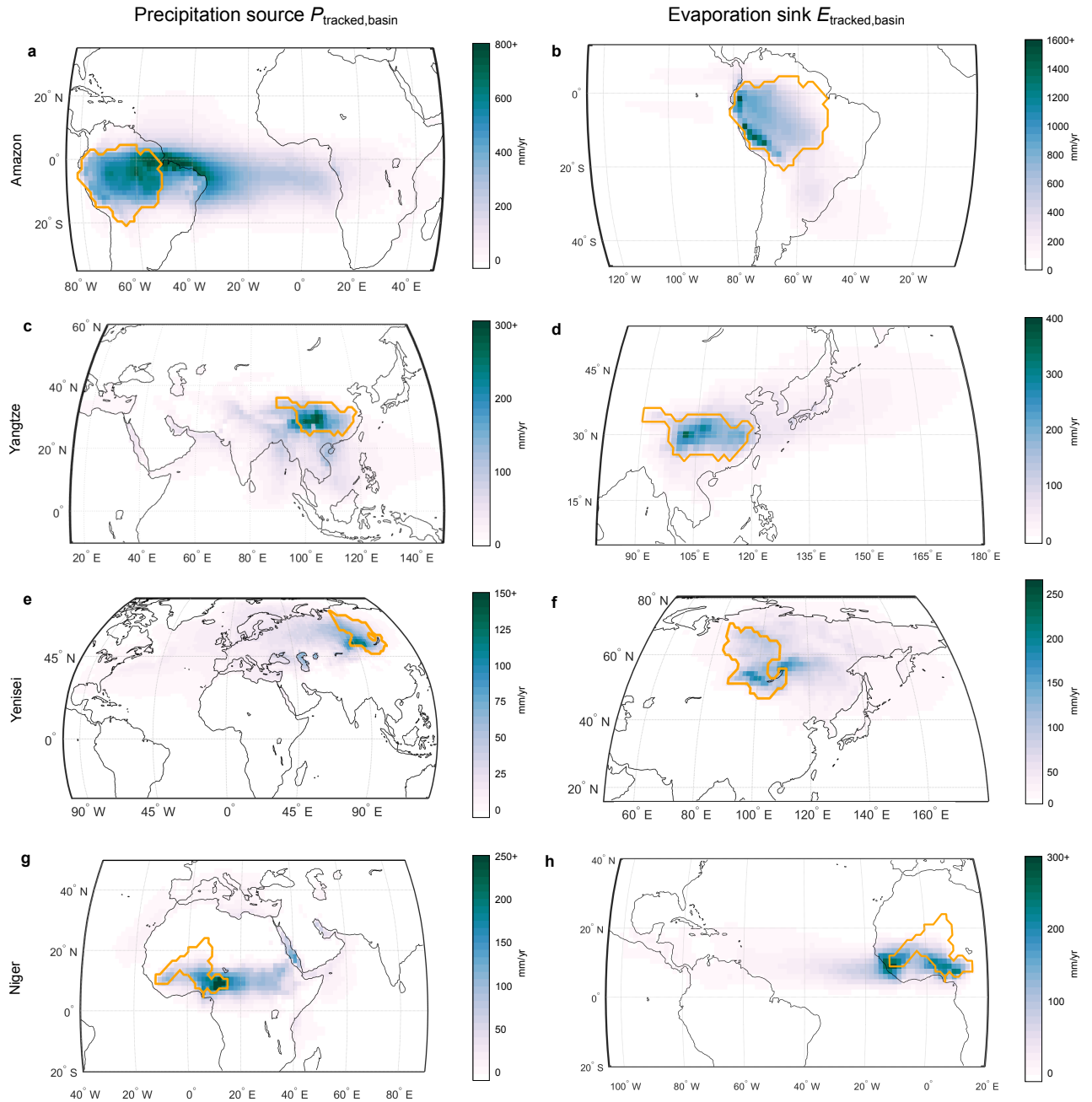
subscript  $\text{noTMR}$  denotes simulation without TMR) corresponds to total  $\Delta P$  (i.e.,  $\Delta P_{\text{import}} + \Delta P_{\text{basin-recycling}}$ ) and any indirect  $\Delta E$  (i.e.,  $\Delta E_{\text{noTMR}} - \Delta E$ , not shown). In Yangtze, the  $\Delta Q$  is mitigated mostly by  $\Delta P_{\text{basin-recycling}}$ . The strong flow reduction in the heavily irrigated Indus, however, is only mildly compensated by TMR (i.e.,  $\Delta P_{\text{import}} \ll \Delta E_{\text{export}}$ ).

The pattern of overlapping precipitation- and evaporationsheds illustrated in Fig. 7b and moderated by wind directions can also be clearly seen in the basin specific precipitation- and evaporationshed maps (Fig. 8). In the Amazon (Fig. 8a,b), the moisture arrives from the east, is stopped up by the Andes, and changes direction towards the southeast. The hotspot of precipitation source and sink within the Amazon basin do not overlap, with major moisture providing spots located along the northeastern border and the major moisture receiving spots located along the Andes in the west. In the Yangtze (Fig. 8c,d), the moisture comes from a large area in the south, and leaves in the direction of Japan in a relatively narrow band. In the Yenisei (Fig. 8e,f), the moisture follows the westerlies, coming in straight from the west, and leaving straight towards the east. In the Niger (Fig. 8g,h), the moisture is mostly supplied from the east from terrestrial areas, and flows towards the west into the Atlantic. For precipitation- and evaporationsheds of other basins, see Fig. S4 and S6 respectively.

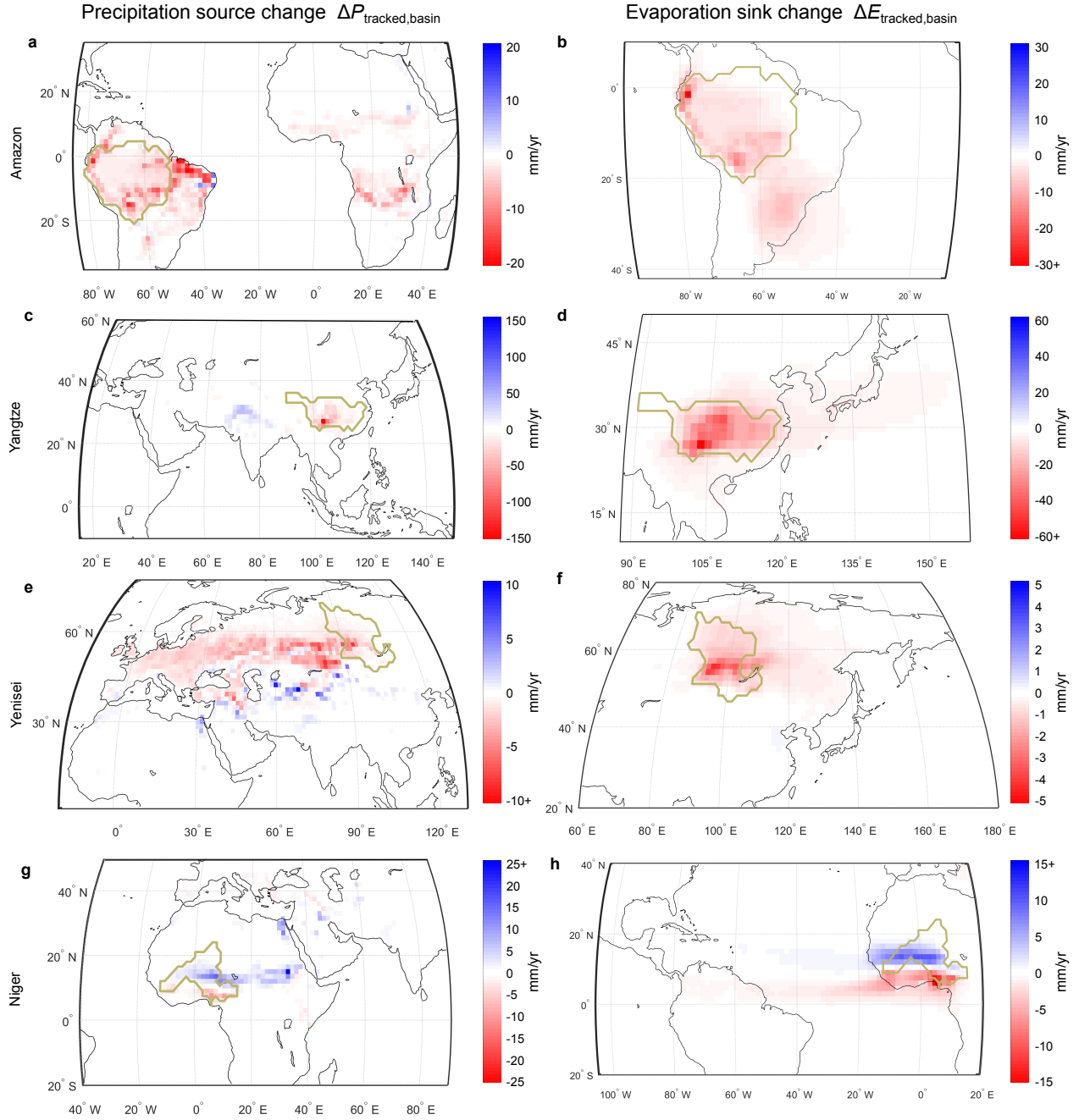
While changes in precipitation- and evaporationsheds are conditioned by the original moisture flows, the resulting pattern ultimately depend on the distribution of LUC-induced hydrological change (compare Fig. 8 and 9). For example, although the Amazon precipitationshed is weak over Africa (Fig. 8a), the precipitationshed change is in fact relatively strong there due to strong LUC-induced hydrological change (Fig. 9a). In other cases, aggregated changes in Fig. 7 hide spatially heterogeneous increases and decreases in moisture flows. For example, agricultural activities and irrigation in India, the Sahel, and regions around the Nile increase moisture supply to the Yangtze, Yenisei, and Niger basins and offset deforestation induced moisture supply decrease elsewhere (Fig. 9c,e,g). For changes in precipitation- and evaporationsheds of other basins, see Fig. S5 and S7 respectively.

### 3.4 Attributing influence on $\Delta Q$ to nations

Typically, TMR attributes LUC influence on  $Q$  (methods described in Sect. 2.3.2) to a larger number of nations than when only basin boundaries are considered (Fig. 4 and Fig. S6). For example, in 10). In several of the studied basins, (such as the Amazon, Congo, Volga, Ob, Yenisei, and Niger basins, see Fig. 10a,b,e,f,i,l), the share of nations contributing less than 5 % to  $\Delta Q$  originates from as far away as Africa if considering TMR (Fig. 4b-e). Because of the large LUC-induced  $E$  reductions, the African influence on Amazon grows considerably when TMR is considered. In some cases, nations not considered a key influencer of  $\Delta Q$ , in fact influence  $\Delta Q$  is comparable to within-basin influence (SI Methods), more than 5 % when TMR is accounted for: in Mekong, India is only an important influencer (10 % influence) when TMR is considered (Fig. 10g); in Yenisei, Mongolia falls below 5 % influence, while Kazakhstan (11 %) and China (6 %) climb considerably in influence (Fig. 10i); and in Niger basin, Sudan/South Sudan (8 %) and Niger (5 %) replace Ivory Coast and Guinea as important influencers (Fig. 10l). Notably, basins geographically confined within one nation can be influenced by LUC taking place in foreign nations. This is for example the case in Yangtze, where simulated Yellow, and Huai, where irrigation in India increases the basin's basins'  $P$  (Fig. 4g)-10d,m,n). The TMR lead to limited difference in nation influence only in the North American basins (Fig. 10h,o) and La Plata (Fig. 10c).

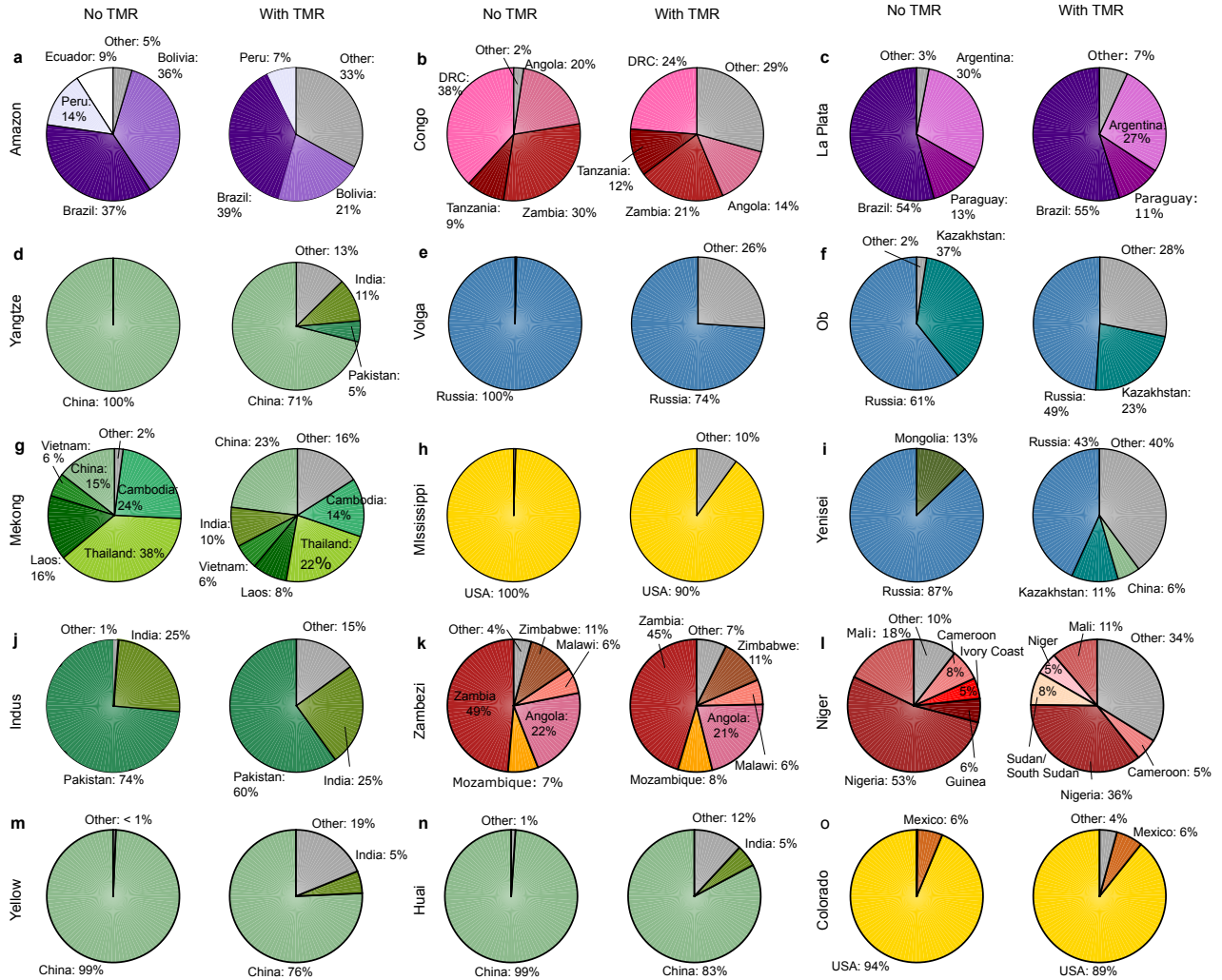


**Figure 8.** Mean annual precipitation sources and evaporation sinks for selected river basins (boundaries in orange).



**Figure 9.** Impacts of human land-use change on mean annual precipitation source (i.e.,  $\Delta P_{\text{import}} + \Delta P_{\text{basin-recycling}}$ ) and evaporation sinks (i.e.,  $\Delta E_{\text{export}} + \Delta E_{\text{basin-recycling}}$ ) for selected river basins (boundaries in dark yellow).





**Figure 10.** Nation influence on river flow change depending on consideration of whether terrestrial moisture recycling (TMR) is taken into account in **a-b**, Amazon and **e-f**, Yangtze the 15 studied basins. Without considering TMR, the nation influence on river flow change originates entirely from within-basin country-wise evaporation change (**a, e**) within the basin. By considering With consideration of TMR, the nation influence **b, f** to river flow change is considered as the sum of absolute imported precipitation change (precipitation source outside encircled basin boundaries changes in **e** precipitation import and **g**) and the sum of absolute exported changes in evaporation change export (evaporation sink outside encircled basin boundaries in **d** and **h**) (Methods). Precipitation sources (**e, g**) and evaporation sinks (**d, h**) within basin boundaries are recycled. Single country contribution below 5 % is bundled into the category "Other".



## 4 DiscussionsDiscussion

### 4.1 Interplay between TMR and LUC

At the global scale,  $\Delta Q$  as a response to LUC can be almost halved by taking TMR into account (Fig. 25). However, these effects vary widely by regions. While the TMR effects are negligible in some basins, remote LUC can compensate the majority of the impact on  $Q$  from local LUC in other basins (e.g., Amazon, Fig. 3a7a) and even propose new transboundary relationships (e.g., Yangtze, Fig. 4e10d). From a TMR perspective, the impact on  $Q$  from within-basin LUC depends on the  $\Delta E$  exported from the basin as much as the  $\Delta P$  imported to the basin.

Our analysis shows the importance of considering LUC on par with TMR to identify anthropogenic influence on water resources, beyond analyses of pure moisture exchanges (Dirmeyer et al., 2009; Keys et al., 2017). While Africa does not constitute a major moisture source of Amazonian  $P$  (7 % of all Amazon  $P$ , 13 % of Amazon  $P$  with continental origin, see also Fig. S2a8a), the spatial extent of  $\Delta E$  from LUC was sufficient to elevate the relative importance of African LUC for Amazonian  $\Delta Q$  (28 % of Amazon  $\Delta P$ , see also Fig. 4e10a). Similarly, India is not identified as a major moisture source of Yangtze (see Fig. S2e and Wei et al. (2012)8c and Wei et al. (2012)), but has about 10 % influence on Yangtze  $\Delta Q$  (Fig. 4f and 4g10d).

### 4.2 Potential governance relevance

Our results indicate that both precipitation- and evaporationsheds of river basins are relevant governance units. Previous studies of TMR for water management (Berger et al., 2014; Keys et al., 2017) have emphasized the importance of considering the  $P$  source region, i.e., the precipitationshed (Keys et al., 2012), which was introduced as a concept analogue to watershed for water resources management. This study finds that the evaporationshed (van der Ent and Savenije, 2013), i.e., the  $E$  sink region, is just as important when considering changes to  $Q$ .

LUC impacts  $Q$  through TMR in different ways depending on how precipitationshed, river basin, and evaporationshed are aligned. For example, where an evaporationshed has a limited overlap with river basin boundaries, reforesting a river basin may lead to unexpectedly large reductions in  $Q$ , if considerable deforestation simultaneously occurs in the precipitationshed outside the river basin.

The magnitude of TMR effects from remote LUC on  $Q$  can be comparable to managed water flows. For example, Yangtze River provides 36 % of the country's surface water resources, and is subject to two of the world's most ambitious water engineering projects: the Three Gorges Dam and the South-to-North Water Diversion (CWRC, 2017). The overall TMR effect on mean annual LUC-induced  $\Delta Q$  is here estimated at  $980 \text{ m}^3 \text{ s}^{-1}$  in the Yangtze basin, and the mean annual moisture change imported to the basin from foreign countries is estimated at  $1,110 \text{ m}^3 \text{ s}^{-1}$  (Fig. 4g9c). As comparison, the  $Q$  difference between a normal and a dry year is about  $300 \text{ m}^3 \text{ s}^{-1}$  and the total amount of water to be transferred from the Yangtze through the South-to-North Water Diversion is aimed to be  $1420 \text{ m}^3 \text{ s}^{-1}$  (NSBD, 2011). Seasonal and interannual flow variability is a major challenge facing the Yangtze, and future research in the seasonal LUC influence and interaction with the monsoon system is

needed. Note, however, that our estimates are associated with parameter sensitivity (see Fig. S7S9) and large uncertainties as discussed in the Limitations.

We note that the relevance of considering TMR governance depends on future LUC. The simulated  $\Delta Q$  in this paper follows from a rather extreme LUC scenario (from potential to current land-use). The current LUC in this study is 15 million km<sup>2</sup> cropland and 28 million km<sup>2</sup> pasture conversion (Ramankutty et al., 2008). As comparison, models used in the Intergovernmental Panel on Climate Change (IPCC) Fourth Assessment Report (AR4) estimated cropland changes from ~~-1.2 to +12~~ -1.2 to +12 million km<sup>2</sup> between 2000 and 2050 (IPCC, 2007). A more recent multi-model comparison range cropland conversion until 2050 from ~~-1 to +8.5~~ -1 to +8.5 million km<sup>2</sup> across different scenarios (Schmitz et al., 2014). In total, the potential land for agricultural conversion has been estimated at 17 million km<sup>2</sup> (Schmitz et al., 2014). Thus, future LUC can be considerable, and potential TMR impacts on  $Q$  will be dependent on the type and geographical distribution of LUC, as well as dependent on prevailing winds, hydroclimate, and orography.

### 4.3 Limitations

~~Our TMR analyses should be seen as an inquiry to better understand the relative importance of local and remote LUC effects on  $Q$  from a water balance perspective, rather than an exact prediction. Our~~ In interpreting our results, it should be noted that our approach only accounts for the TMR effects. The frequency or intensity of  $P$  ~~(Medvigy et al., 2011)~~ are assumed to remain unaffected by thermal layer processes or circulation perturbation, which may introduce a bias in the quantitative estimates of hydrological flows under water limited conditions (i.e., semi-arid regions and temperate region during summertime) ~~(Medvigy et al., 2011)~~. Furthermore, vegetation response to  $\Delta P$  is not simulated, such as forest dieback from increased fire risk under drying conditions. ~~Our analyses concern mean annual  $\Delta Q$ , and can be considered conservative in the sense that seasonal signals are expected to be much stronger.~~ Human modification of  $Q$  through dams and climate change (Haddeland et al., 2014) are also not considered in this study. In addition, the land-use change over land may affect above ocean processes mainly through modification of the energy balance and circulation in monsoon regions, which we do not account for. Changes in fresh water discharge to the oceans might have implications for ocean circulation and climate, as studies of for example river discharge to the Arctic Ocean showed (Peterson, 2002, 2006). However, moisture recycling's buffering effect (which mitigates river flow changes), should have a mitigating effect on the ocean's response to fresh water inflow. Otherwise, precipitation over ocean can influence ocean salinity (IPCC, 2013) and precipitation patterns over land can be influenced by sea surface temperature (Xie et al., 2010), but we consider this outside the scope of our study and likely to be of minor importance for the research questions that we address. Our TMR analyses should, thus, be seen as an inquiry to better understand the relative importance of local and remote LUC effects on  $Q$  from a water balance perspective, rather than an exact prediction. Nevertheless, due to the inevitable recycling of moisture in the global hydrological cycle, uncertainties in the magnitude will unlikely affect our key conclusions that upwind extra-basin LUC can be essential for  $Q$ .

The magnitude of our estimated  $\Delta P$  (Fig. 25) and  $\Delta Q$  from LUC is ~~also~~ conservative in comparison to the literature (Spracklen and Garcia-Carreras, 2015). ~~Thus, despite the uncertainties of~~ For example, a meta-analysis of 96 different general circulation models (GCM) and regional climate model (RCM) deforestation simulations showed that under 10 % conversion

of Amazon forest to pasture or soybean production, the inter-quartile range of rainfall change in the Amazon basin is 0 to  $-4$  % (Spracklen and Garcia-Carreras, 2015). In comparison, the magnitude of change, we are confident that upwind extra-basin LUC can be essential for  $Q$ . STEAM-WAM2layers approach with change from potential to current land-use change (i.e., 8.8 % deforestation extent in the Amazon), causes a rainfall reduction of 0.4 % in the Amazon and thus falls in the conservative range. In addition, our analyses concern mean annual  $\Delta Q$ , and can also be considered conservative in the sense that seasonal signals are expected to be much stronger.

The limitations of our methods should also been seen in light of the strengths and limitations of alternative methods for studying hydrological LUC effects, see Table S1. The most complex and coupled modelling approaches account for the highest number of feedback processes. However, the high degree of freedom in GCMs also contributes to the high sensitivity of precipitation to initial conditions and the low signal-to-noise ratios. For example, a scenario replacing natural with present-day land cover only detected a significant response in less than 5 % of all grid cells in a single model analysis (Findell et al., 2007) and less than 5 % in non-perturbed grid cells across seven different models (Pitman et al., 2009). Regional deforestation scenarios generate higher ratios of significant results near the source of change, albeit noise remains high in distant regions (Werth and Avissar, 2002). The challenges in simulating precipitation due to cloud formation, aerosol representation, and inherent uncertainties in circulation response (Aloysius et al., 2016; Koren et al., 2012; Shepherd, 2014), and non-closure of water balance in semi-coupled modelling approaches (Bring et al., 2015) also contribute to a high model dependence in estimates of river flow change from LUC (Kundzewicz et al., 2007). Thus, the sign, magnitude, and location of impacts vary widely among models (Aloysius et al., 2016; Pitman et al., 2009). Observation-based methods relate presence of vegetation or irrigation to precipitation or river flows using statistical methods, often in combination with moisture tracking to determine the geographical origin of rainfall (DeAngelis et al., 2010; Kustu et al., 2010, 2011; Spracklen et al., 2012). Limitations of this type of methods include variations in data quality, challenges in isolating effects of land-use from climate variability, and difficulties establishing causation from correlation (Matin and Bourque, 2015). Key elements missing in all approaches including our own are socio-economic dynamics and landscape resilience, which are complex issues currently explored in experimental model settings (Nitzbon et al., 2017; Reyer et al., 2015).

#### 4.4 Future research outlook

Research of land-use change dynamics and moisture recycling is becoming increasingly detailed, looking at e.g., the role of transpiration for moisture recycling during dry periods (Wang-Erlandsson et al., 2014; van der Ent et al., 2014; Staal et al., 2018), studying the interplay with groundwater use and recharge (Keune et al., 2018), and identifying influential source areas within a basin (Weng et al., 2018; Staal et al., 2018). Nevertheless, a key challenge for considering TMR effects in water governance is the modeling uncertainties and inherent variabilities associated with land-atmosphere feedback processes. The most complex modeling approaches account for the highest number of feedback processes. However, the sign, magnitude, and location of impacts vary widely even among state-of-the-art climate models (Pitman et al., 2009; Aloysius et al., 2016). Key future improvements in climate models' ability to simulate  $\Delta P$  from LUC will contribute to the governability of TMR. In-depth examination of differences in model simulation of  $P$  (e.g., the ongoing Precipitation Driver Response Model Intercomparison

Project (Myhre et al., 2017)) is one step in this direction. Tracking moisture in coupled climate models could further help identify causes for simulated differences in atmospheric and hydrological outputs. Key elements missing in current research on LUC effects on hydrological flows include socio-economic dynamics and landscape resilience, which are complex issues currently explored in experimental model settings (Nitzbon et al., 2017; Reyer et al., 2015). In the meantime, ~~no-regret~~ "no-regret" policies in river basin management, where TMR objectives align with other aims can potentially be explored in conjunction with LUC scenarios that include TMR effects.

## 5 Conclusions

We ~~analyzed~~ analysed the potential impact of human LUC on  $Q$  worldwide through TMR, and separately looked at the remote and local LUC effects of relevance to water governance. Despite the river basin being the standard unit in water governance and water resources management, we find that  $\Delta Q$  are ultimately dependent on the modifications in both incoming  $P$  and outflowing  $E$ . For example, where extra-basin LUC affects basin  $P$  more strongly than within-basin LUC, reforesting a river basin may lead to unexpectedly large reductions in  $Q$  if deforestation simultaneously occurs in  $P$  source regions outside the river basin. Therefore, we emphasize the necessity of considering both the origin of basin  $P$  as well as the fate of basin  $E$  for management of local water resources. Further, we suggest the potential need for transboundary governance of river basins where extra-basin LUC is important for  $\Delta Q$ . International governance arrangements of teleconnected LUC influence could be needed, even for river basins that today are not considered transboundary. We conclude that consideration of TMR is essential for understanding  $Q$  modifications and managing water resources in a rapidly changing and tele-coupled world (Liu et al., 2013) facing increasing pressure on both land (Schmitz et al., 2014) and water (Mekonnen and Hoekstra, 2016). Further research in both climate modeling and water governance strategies is needed to internalize land-atmosphere interactions in future water resources considerations.

*Code and data availability.* The moisture tracking scheme Water Accounting Model-2 layers (WAM-2layers) in Python code can be obtained from GitHub (<https://github.com/ruudvdent/WAM2layersPython>). Earth Retrospective Analysis Interim (ERA-I) meteorological data can be obtained from the European Centre for Medium-Range Weather Forecasts (ECMWF) (<http://apps.ecmwf.int/datasets/data/interim-full-daily/levtype=sfc/>). The Multi-Source Weighted-Ensemble Precipitation (MSWEP) data can be downloaded from <http://www.gloh2o.org/>. The Ramankutty potential land data come from <https://nelson.wisc.edu/sage/data-and-models/global-potential-vegetation/index.php>. The current cropland and pasture map can be obtained from EarthStat (<http://www.earthstat.org/data-download/>). Land Cover Type Climate Modeling Grid (CMG) MCD12C1 International Geosphere Biosphere Program (IGBP) land classification created from Terra and Aqua Moderate Resolution Imaging Spectroradiometer (MODIS) data can be downloaded at <https://modis.gsfc.nasa.gov/data/dataproduct/mod12.php>. Monthly irrigated rice and irrigation non-rice crops were obtained from the data set of Monthly Irrigated and Rainfed Crop Areas around the year 2000 (MIRCA2000) V1.1. and can be downloaded at <http://www.uni-frankfurt.de/45218031>.

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*Competing interests.* The authors declare that they have no conflict of interest.

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## References

- Alkama, R., Kageyama, M., and Ramstein, G.: Relative contributions of climate change, stomatal closure, and leaf area index changes to 20th and 21st century runoff change: A modelling approach using the Organizing Carbon and Hydrology in Dynamic Ecosystems (ORCHIDEE) land surface model, *Journal of Geophysical Research Atmospheres*, 115, D17 112, <https://doi.org/10.1029/2009JD013408>, 2010.
- 5 Aloysius, N. R., Sheffield, J., Saiers, J. E., Li, H., and Wood, E. F.: Evaluation of historical and future simulations of precipitation and temperature in central Africa from CMIP5 climate models, *Journal of Geophysical Research: Atmospheres*, 121, 130–152, <https://doi.org/10.1002/2015JD023656>, 2016.
- Arnell, N. W. and Lloyd-Hughes, B.: Global-scale impacts of climate change at different levels of forcing, AVOID project report AV/WS2/D1/R35, Tech. rep., [www.avoid.uk.net/2012](http://www.avoid.uk.net/2012), 2012.
- 10 Avissar, R. and Werth, D.: Global hydroclimatological teleconnections resulting from tropical deforestation, *Journal of Hydrometeorology*, 6, 134–145, 2005.
- Badger, A. M. and Dirmeyer, P. A.: Remote tropical and sub-tropical responses to Amazon deforestation, *Climate Dynamics*, 46, 3057–3066, <https://doi.org/10.1007/s00382-015-2752-5>, 2016.
- Bagley, J. E., Desai, A. R., Dirmeyer, P. A., and Foley, J. A.: Effects of land cover change on moisture availability and potential crop yield in the world’s breadbaskets, *Environmental Research Letters*, 7, 014 009, <https://doi.org/10.1088/1748-9326/7/1/014009>, 2012.
- 15 Beck, H. E., van Dijk, A. I. J. M., Levizzani, V., Schellekens, J., Miralles, D. G., Martens, B., and de Roo, A.: MSWEP: 3-hourly 0.25 degree global gridded precipitation (1979-2015) by merging gauge, satellite, and reanalysis data, *Hydrology and Earth System Sciences*, 21, 589–615, <https://doi.org/10.5194/hess-21-589-2017>, 2017.
- Berger, M., van der Ent, R., Eisner, S., Bach, V., and Finkbeiner, M.: Water accounting and vulnerability evaluation (WAVE): considering atmospheric evaporation recycling and the risk of freshwater depletion in water footprinting., *Environmental science & technology*, 48, 4521–8, <https://doi.org/10.1021/es404994t>, 2014.
- 20 Betts, R. A., Golding, N., Gonzalez, P., Gornall, J., Kahana, R., Kay, G., Mitchell, L., and Wiltshire, A.: Climate and land use change impacts on global terrestrial ecosystems and river flows in the HadGEM2-ES Earth system model using the representative concentration pathways, *Biogeosciences*, 12, 1317–1338, <https://doi.org/10.5194/bg-12-1317-2015>, 2015.
- 25 Bring, A., Asokan, S. M., Jaramillo, F., Jarsjö, J., Levi, L., Pietroń, J., Prieto, C., Rogberg, P., and Destouni, G.: Implications of freshwater flux data from the CMIP5 multimodel output across a set of Northern Hemisphere drainage basins, *Earth’s Future*, 3, 206–217, <https://doi.org/10.1002/2014EF000296>, 2015.
- Brubaker, K. L., Entekhabi, D., and Eagleson, P. S.: Estimation of continental precipitation recycling, *Journal of Climate*, 6, 1077–1089, 1993.
- 30 CWRC: Changjiang Water Resources Commission of the Ministry of Water Resources, <http://www.cjw.gov.cn/zjzx/cjyl/>, 2017.
- Dalin, C., Wada, Y., Kastner, T., and Puma, M. J.: Groundwater depletion embedded in international food trade, *Nature*, 543, 700–704, <https://doi.org/10.1038/nature21403>, 2017.
- D’Almeida, C., Vorosmarty, C. J., Hurtt, G. C., Marengo, J. A., Dingman, S. L., and Keim, B. D.: The effects of deforestation on the hydrological cycle in Amazonia: a review on scale and resolution, *International Journal of Climatology*, 27, 633–647, 2007.
- 35 de Vrese, P., Hagemann, S., and Claussen, M.: Asian Irrigation, African Rain: Remote Impacts of Irrigation, *Geophysical Research Letters*, 43, 3737–3745, <https://doi.org/10.1002/2016GL068146>, 2016.

- DeAngelis, A., Domínguez, F., Fan, Y., Robock, A., Kustu, M. D., and Robinson, D.: Evidence of enhanced precipitation due to irrigation over the Great Plains of the United States, *Journal of Geophysical Research*, 115, D15 115, <https://doi.org/10.1029/2010JD013892>, 2010.
- Dee, D., Uppala, S., Simmons, A. J., Berrisford, P., Poli, P., Kobayashi, S., Andrae, U., Balmaseda, M., Balsamo, G., Bauer, P., Bechtold, P., Beljaars, A. C. M., van de Berg, L., Bidlot, J., Bormann, N., Delsol, C., Dragani, R., Fuentes, M., Geer, A., Haimberger, L., Healy, S., Hersbach, H., Hólm, E., Isaksen, I., Kållberg, P., Köhler, M., Matricardi, M., McNally, A., Monge-Sanz, B., Morcrette, J.-J., Park, B.-K., Peubey, C., de Rosnay, P., Tavolato, C., Thépaut, J.-N., and Vitart, F.: The ERA-Interim reanalysis: configuration and performance of the data assimilation system, *Quarterly Journal of the Royal Meteorological Society*, 137, 553–597, <https://doi.org/10.1002/qj.828>, 2011.
- 5 Dirmeyer, P. A., Brubaker, K. L., and DelSole, T.: Import and export of atmospheric water vapor between nations, *Journal of Hydrology*, 365, 11–22, <https://doi.org/10.1016/j.jhydrol.2008.11.016>, 2009.
- Döll, P., Fiedler, K., and Zhang, J.: Global-scale analysis of river flow alterations due to water withdrawals and reservoirs, *Hydrology and Earth System Sciences*, 13, 2413–2432, <https://doi.org/10.5194/hess-13-2413-2009>, 2009.
- 10 Ellison, D., Morris, C. E., Locatelli, B., Sheil, D., Cohen, J., Murdiyarso, D., Gutierrez, V., van Noordwijk, M., Creed, I. F., Pokorny, J., Gaveau, D., Spracklen, D. V., Tobella, A. B., Ilstedt, U., Teuling, A. J., Gebrehiwot, S. G., Sands, D. C., Muys, B., Verbist, B., Springgay, E., Sugandi, Y., and Sullivan, C. A.: Trees, forests and water: Cool insights for a hot world, *Global Environmental Change*, 43, 51–61, <https://doi.org/10.1016/j.gloenvcha.2017.01.002>, 2017.
- 15 Fekete, B. M., Vörösmarty, C. J., and Grabs, W.: High-resolution fields of global runoff combining observed river discharge and simulated water balances, *Global Biogeochemical Cycles*, 16, 15–1–15–10, <https://doi.org/10.1029/1999GB001254>, 2002.
- Findell, K. L., Shevliakova, E., Milly, P. C. D., Stouffer, R. J., Findell, K. L., Shevliakova, E., Milly, P. C. D., and Stouffer, R. J.: Modeled Impact of Anthropogenic Land Cover Change on Climate, *Journal of Climate*, 20, 3621–3634, <https://doi.org/10.1175/JCLI4185.1>, 2007.
- Friedl, M. a., Sulla-Menashe, D., Tan, B., Schneider, A., Ramankutty, N., Sibley, A., and Huang, X.: MODIS Collection 5 global land cover: Algorithm refinements and characterization of new datasets, *Remote Sensing of Environment*, 114, 168–182, <https://doi.org/10.1016/j.rse.2009.08.016>, 2010.
- 20 Garcia, E. S., Swann, A. L. S., Villegas, J. C., Breshears, D. D., Law, D. J., Saleska, S. R., and Stark, S. C.: Synergistic eco-climate teleconnections from forest loss in different regions structure global ecological responses, *PLoS ONE*, 11, e0165042, <https://doi.org/10.1371/journal.pone.0165042>, 2016.
- 25 Gerten, D., Rost, S., von Bloh, W., and Lucht, W.: Causes of change in 20th century global river discharge, *Geophysical Research Letters*, 35, L20 405, <https://doi.org/10.1029/2008GL035258>, 2008.
- Giordano, M., Suhardiman, D., and Peterson-Perlman, J.: Do hydrologic rigor and technological advance tell us more or less about transboundary water management?, *International Environmental Agreements: Politics, Law and Economics*, 16, 815–831, <https://doi.org/10.1007/s10784-015-9297-2>, 2015.
- 30 Goessling, H. F. and Reick, C. H.: What do moisture recycling estimates tell us? Exploring the extreme case of non-evaporating continents, *Hydrology and Earth System Sciences*, 15, 3217–3235, <https://doi.org/10.5194/hess-15-3217-2011>, 2011.
- Guillod, B. P., Orlowsky, B., Miralles, D. G., Teuling, A. J., and Seneviratne, S. I.: Reconciling spatial and temporal soil moisture effects on afternoon rainfall., *Nature communications*, 6, 6443, <https://doi.org/10.1038/ncomms7443>, 2015.
- Haddeland, I., Heinke, J., Biemans, H., Eisner, S., Flörke, M., Hanasaki, N., Konzmann, M., Ludwig, F., Masaki, Y., Schewe, J., Stacke, T., Tessler, Z. D., Wada, Y., and Wisser, D.: Global water resources affected by human interventions and climate change., *Proceedings of the National Academy of Sciences of the United States of America*, 111, 3251–6, <https://doi.org/10.1073/pnas.1222475110>, 2014.

- Hoekstra, A. Y.: The Global Dimension of Water Governance: Why the River Basin Approach Is No Longer Sufficient and Why Cooperative Action at Global Level Is Needed, *Water*, 3, 21–46, <https://doi.org/10.3390/w3010021>, 2010.
- IPCC: Mitigation of Climate Change, Contribution of Working Group III to the Fourth Assessment Report of the Intergovernmental Panel on Climate Change, in: *Climate Change 2007*, edited by Metz, B., Davidson, O., Bosch, P., Dave, R., and Meyer, L., Cambridge University Press, Cambridge, UK, 2007.
- 5 IPCC: The Physical Science Basis. Contribution of Working Group I to the Fifth Assessment Report of the Intergovernmental Panel on Climate Change, in: *Climate Change 2013*, edited by Stocker, T., Qin, D., Plattner, G.-K., Tignor, M., Allen, S., Boschung, J., Nauels, A., Y. Xia, V. B., and Midgley, P., September 2013, p. 1535, Cambridge University Press, Cambridge, United Kingdom and New York, NY, USA, <https://doi.org/10.1017/CBO9781107415324>, 2013.
- Jaramillo, F. and Destouni, G.: Local flow regulation and irrigation raise global human water consumption and footprint, *Science*, 350, 1248–1251, <https://doi.org/10.1126/science.aad1010>, 2015.
- 10 Johansson, E. L., Fader, M., Seaquist, J. W., and Nicholas, K. A.: Green and blue water demand from large-scale land acquisitions in Africa., *Proceedings of the National Academy of Sciences*, p. 201524741, <https://doi.org/10.1073/pnas.1524741113>, 2016.
- Jolly, W. M., Nemani, R., and Running, S. W.: A generalized, bioclimatic index to predict foliar phenology in response to climate, *Global Change Biology*, 11, 619–632, <https://doi.org/10.1111/j.1365-2486.2005.00930.x>, 2005.
- 15 Keune, J., Sulis, M., Kollet, S., Siebert, S., and Wada, Y.: Human Water Use Impacts on the Strength of the Continental Sink for Atmospheric Water, *Geophysical Research Letters*, 45, 4068–4076, <https://doi.org/10.1029/2018GL077621>, 2018.
- Keys, P. W., van der Ent, R. J., Gordon, L. J., Hoff, H., Nikoli, R., and Savenije, H. H. G.: Analyzing precipitation sheds to understand the vulnerability of rainfall dependent regions, *Biogeosciences*, 9, 733–746, <https://doi.org/10.5194/bg-9-733-2012>, 2012.
- Keys, P. W., Wang-Erlandsson, L., and Gordon, L. J.: Revealing Invisible Water: Moisture Recycling as an Ecosystem Service., *PloS one*, 11, e0151993, <https://doi.org/10.1371/journal.pone.0151993>, 2016.
- 20 Keys, P. W., Wang-Erlandsson, L., Gordon, L. J., Galaz, V., and Ebbesson, J.: Approaching moisture recycling governance, *Global Environmental Change*, 45, 15–23, <https://doi.org/10.1016/j.gloenvcha.2017.04.007>, 2017.
- Khanna, J., Medvigy, D., Fueglistaler, S., and Walko, R.: Regional dry-season climate changes due to three decades of Amazonian deforestation, *Nature Climate Change*, 7, 200–204, <https://doi.org/10.1038/nclimate3226>, 2017.
- 25 Koren, I., Altaratz, O., Remer, L. A., Feingold, G., Martins, J. V., and Heiblum, R. H.: Aerosol-induced intensification of rain from the tropics to the mid-latitudes, *Nature Geoscience*, 5, 118–122, <https://doi.org/10.1038/ngeo1364>, 2012.
- Koster, R. D., Suarez, M. J., Higgins, R. W., and Van den Dool, H. M.: Observational evidence that soil moisture variations affect precipitation, *Geophysical Research Letters*, 30, 1–4, <https://doi.org/10.1029/2002GL016571>, 2003.
- Kottek, M., Grieser, J., Beck, C., Rudolf, B., and Rubel, F.: World Map of the Köppen-Geiger climate classification updated, *Meteorologische Zeitschrift*, 15, 259–263, <https://doi.org/10.1127/0941-2948/2006/0130>, 2006.
- 30 Kundzewicz, Z. W., Mata, L. J., Arnell, N. W., Doll, P., Kabat, P., Jimenez, B., Miller, K. A., Oki, T., Scen, Z., and Shiklomanov, I. A.: Freshwater resources and their management, in: *Climate Change 2007: Impacts, Adaptations, and Vulnerability. Contribution of Working Group II to the Fourth Assessment Report of the Intergovernmental Panel on Climate Change*, edited by Parry, M., Canziani, O., Palutikof, J., van der Linden, P., and Hansen, C., pp. 173–210, Cambridge University Press, Cambridge, UK, 2007.
- 35 Kustu, M. D., Fan, Y., and Robock, A.: Large-scale water cycle perturbation due to irrigation pumping in the US High Plains: A synthesis of observed streamflow changes, *Journal of Hydrology*, 390, 222–244, <https://doi.org/10.1016/j.jhydrol.2010.06.045>, 2010.



- Kustu, M. D., Fan, Y., and Rodell, M.: Possible link between irrigation in the U.S. High Plains and increased summer streamflow in the Midwest, *Water Resources Research*, 47, <https://doi.org/10.1029/2010WR010046>, 2011.
- Lawrence, D. and Vandecar, K.: Effects of tropical deforestation on climate and agriculture, *Nature Climate Change*, 5, 27–36, <https://doi.org/10.1038/nclimate2430>, 2014.
- Liu, J., Hull, V., Batistella, M., DeFries, R., Dietz, T., Fu, F., Hertel, T. W., Izaurralde, R. C., Lambin, E. F., Li, S., Martinelli, L. A.,  
5 McConnell, W. J., Moran, E. F., Naylor, R., Ouyang, Z., Polenske, K. R., Reenberg, A., de Miranda Rocha, G., Simmons, C. S., Verburg, P. H., Vitousek, P. M., Zhang, F., and Zhu, C.: Framing Sustainability in a Telecoupled World, *Ecology and Society*, 18, 26, <https://doi.org/10.5751/ES-05873-180226>, 2013.
- Lo, M.-h. and Famiglietti, J. S.: Irrigation in California ' s Central Valley strengthens the southwestern U . S . water cycle, *Geophysical Research Letters*, 40, <https://doi.org/10.1002/GRL.50108>, 2013.
- 10 Matin, M. A. and Bourque, C. P.-A.: Relating seasonal dynamics of enhanced vegetation index to the recycling of water in two endorheic river basins in north-west China, *Hydrology and Earth System Sciences*, 19, 3387–3403, <https://doi.org/10.5194/hess-19-3387-2015>, 2015.
- Medvigy, D., Walko, R. L., and Avissar, R.: Effects of deforestation on spatiotemporal distributions of precipitation in South America, *Journal of Climate*, 24, 2147–2163, <https://doi.org/10.1175/2010JCLI3882.1>, 2011.
- Mekonnen, M. M. and Hoekstra, A. Y.: Four billion people facing severe water scarcity, *Science Advances*, 2, e1500323,  
15 <https://doi.org/10.1126/sciadv.1500323>, 2016.
- Milly, P. C. D., Dunne, K. A., and Vecchia, A. V.: Global pattern of trends in streamflow and water availability in a changing climate, *Nature*, 438, 347–350, <https://doi.org/10.1038/nature04312>, 2005.
- Miralles, D. G., Nieto, R., McDowell, N. G., Dorigo, W. A., Verhoest, N. E., Liu, Y. Y., Teuling, A. J., Dolman, A. J., Good, S. P., and Gimeno, L.: Contribution of water-limited ecoregions to their own supply of rainfall, *Environmental Research Letters*, 11, 124007,  
20 <https://doi.org/10.1088/1748-9326/11/12/124007>, 2016.
- Monteith, J. L.: Evaporation and environment, in: *Symp Soc Exp Biol*, vol. 19, chap. The State, pp. 205–234, Cambridge University Press, Swansea, 1965.
- Myhre, G., Forster, P. M., Samset, B. H., Odnebrog, Sillmann, J., Aalberg, S. G., Andrews, T., Boucher, O., Faluvegi, G., Fläschner, D., Iversen, T., Kasoar, M., Kharin, V., Kirkevåg, A., Lamarque, J. F., Olivé, D., Richardson, T. B., Shindell, D., Shine, K. P., Stjern, C. W.,  
25 Takemura, T., Voulgarakis, A., and Zwiers, F.: PDRMP: A precipitation driver and response model intercomparison project-protocol and preliminary results, *Bulletin of the American Meteorological Society*, 98, 1185–1198, <https://doi.org/10.1175/BAMS-D-16-0019.1>, 2017.
- Nitzbon, J., Heitzig, J., and Parltz, U.: Sustainability, collapse and oscillations of global climate, population and economy in a simple World-Earth model, *Environmental Research Letters*, <https://doi.org/10.1088/1748-9326/AA7581>, 2017.
- NSBD: South-to-North Water Diversion, <http://www.nsbd.gov.cn/zx/english/>, 2011.
- 30 Oki, T. and Kanae, S.: Global hydrological cycles and world water resources., *Science*, 313, 1068–1072, <https://doi.org/10.1126/science.1128845>, 2006.
- Peterson, B. J.: Increasing River Discharge to the Arctic Ocean, *Science*, 298, 2171–2173, <https://doi.org/10.1126/science.1077445>, 2002.
- Peterson, B. J.: Trajectory Shifts in the Arctic and Subarctic Freshwater Cycle, *Science*, 313, 1061–1066, <https://doi.org/10.1126/science.1122593>, 2006.
- 35 Piao, S., Friedlingstein, P., Ciais, P., de Noblet-Ducoudré, N., Labat, D., and Zaehle, S.: Changes in climate and land use have a larger direct impact than rising CO<sub>2</sub> on global river runoff trends., *Proceedings of the National Academy of Sciences of the United States of America*, 104, 15 242–15 247, <https://doi.org/10.1073/pnas.0707213104>, 2007.

- Pitman, A. J., de Noblet-Ducoudré, N., Cruz, F. T., Davin, E. L., Bonan, G. B., Brovkin, V., Claussen, M., Delire, C., Ganzeveld, L., Gayler, V., van den Hurk, B. J. J. M., Lawrence, P. J., van der Molen, M. K., Müller, C., Reick, C. H., Seneviratne, S. I., Strengers, B. J., and Voldoire, A.: Uncertainties in climate responses to past land cover change: First results from the LUCID intercomparison study, *Geophysical Research Letters*, 36, 1–6, <https://doi.org/10.1029/2009GL039076>, 2009.
- Portmann, F. T., Siebert, S., and Döll, P.: MIRCA2000—Global monthly irrigated and rainfed crop areas around the year 2000: A new high-resolution data set for agricultural and hydrological modeling, *Global Biogeochemical Cycles*, 24, 1–24, <https://doi.org/10.1029/2008GB003435>, 2010.
- Ramankutty, N. and Foley, J. A.: Estimating historical changes in global land cover: Croplands historical have converted areas, *Global Biogeochemical Cycles*, 13, 997–1027, 1999.
- Ramankutty, N., Evan, A. T., Monfreda, C., and Foley, J. A.: Farming the planet: 1. Geographic distribution of global agricultural lands in the year 2000, *Global Biogeochemical Cycles*, 22, 1–19, <https://doi.org/10.1029/2007GB002952>, 2008.
- Reyer, C. P. O., Brouwers, N., Rammig, A., Brook, B. W., Epila, J., Grant, R. F., Holmgren, M., Langerwisch, F., Leuzinger, S., Lucht, W., Medlyn, B., Pfeifer, M., Steinkamp, J., Vanderwel, M. C., Verbeeck, H., and Vilella, D. M.: Forest resilience and tipping points at different spatio-temporal scales: Approaches and challenges, *Journal of Ecology*, 103, 5–15, <https://doi.org/10.1111/1365-2745.12337>, 2015.
- Rockström, J., Steffen, W., Noone, K., Persson, A., Chapin III, F. S., Lambin, E. F., Lenton, T. M., Scheffer, M., Folke, C., Schellnhuber, H. J., Nykvist, B., de Wit, C. A., Hughes, T., van der Leeuw, S., Rodhe, H., Sörlin, S., Snyder, P. K., Costanza, R., Svedin, U., Falkenmark, M., Karlberg, L., Corell, R. W., Fabry, V. J., Hansen, J., Walker, B., Liverman, D., Richardson, K., Crutzen, P., and Foley, J. A.: Planetary boundaries: Exploring the safe operating space for humanity, *Ecology and Society*, 14, 2009.
- Rost, S., Gerten, D., Bondeau, A., Lucht, W., Rohwer, J., and Schaphoff, S.: Agricultural green and blue water consumption and its influence on the global water system, *Water Resources Research*, 44, W09405, <https://doi.org/10.1029/2007WR006331>, 2008a.
- Rost, S., Gerten, D., and Heyder, U.: Human alterations of the terrestrial water cycle through land management, *Advances in Geosciences*, 18, 43–50, <https://doi.org/10.5194/adgeo-18-43-2008>, 2008b.
- Rulli, C. M., Savioli, A., and Odorico, P. D.: Global land and water grabbing, *Proceedings of the National Academy of Sciences*, 110, 892–897, <https://doi.org/10.1073/pnas.1213163110/-DCSupplemental>, [www.pnas.org/cgi/doi/10.1073/pnas.1213163110](http://www.pnas.org/cgi/doi/10.1073/pnas.1213163110), 2012.
- Saad, S. I., Da Rocha, H. R., Silva Dias, M. A. F., and Rosolem, R.: Can the deforestation breeze change the rainfall in Amazonia? A case study for the BR-163 highway region, *Earth Interactions*, 14, 1–25, <https://doi.org/10.1175/2010EI351.1>, 2010.
- Schmitz, C., van Meijl, H., Kyle, P., Nelson, G. C., Fujimori, S., Gurgel, A., Havlik, P., Heyhoe, E., D’Croz, D. M., Popp, A., Sands, R., Tabeau, A., van der Mensbrugghe, D., von Lampe, M., Wise, M., Blanc, E., Hasegawa, T., Kavallari, A., and Valin, H.: Land-use change trajectories up to 2050: Insights from a global agro-economic model comparison, *Agricultural Economics (United Kingdom)*, 45, 69–84, <https://doi.org/10.1111/agec.12090>, 2014.
- Seneviratne, S. I., Corti, T., Davin, E. L., Hirschi, M., Jaeger, E. B., Lehner, I., Orlowsky, B., and Teuling, A. J.: Investigating soil moisture–climate interactions in a changing climate: A review, *Earth-Science Reviews*, 99, 125–161, <https://doi.org/10.1016/j.earscirev.2010.02.004>, 2010.
- Shepherd, T. G.: Atmospheric circulation as a source of uncertainty in climate change projections, *Nature Geoscience*, 7, 703–708, <https://doi.org/10.1038/ngeo2253>, 2014.
- Spracklen, D., Arnold, S. R., and Taylor, C. M.: Observations of increased tropical rainfall preceded by air passage over forests, *Nature*, 489, 282–285, <https://doi.org/10.1038/nature11390>, 2012.

- Spracklen, D. V. and Garcia-Carreras, L.: The impact of Amazonian deforestation on Amazon basin rainfall, *Geophysical Research Letters*, 42, 9546–9552, <https://doi.org/10.1002/2015GL066063>, 2015.
- Staal, A., Tuinenburg, O. A., Bosmans, J. H., Holmgren, M., Van Nes, E. H., Scheffer, M., Zemp, D. C., and Dekker, S. C.: Forest-rainfall cascades buffer against drought across the Amazon, *Nature Climate Change*, 8, 539–543, <https://doi.org/10.1038/s41558-018-0177-y>, 2018.
- 5 Steffen, W., Richardson, K., Rockstrom, J., Cornell, S. E., Fetzer, I., Bennett, E. M., Biggs, R., Carpenter, S. R., de Vries, W., de Wit, C. A., Folke, C., Gerten, D., Heinke, J., Mace, G. M., Persson, L. M., Ramanathan, V., Reyers, B., and Sorlin, S.: Planetary boundaries: Guiding human development on a changing planet, *Science*, 347, 1259 855, <https://doi.org/10.1126/science.1259855>, 2015.
- Sterling, S. M., Ducharne, A., and Polcher, J.: The impact of global land-cover change on the terrestrial water cycle, *Nature Climate Change*, 3, 385–390, <https://doi.org/10.1038/nclimate1690>, 2012.
- 10 Stewart, J.: Modelling surface conductance of pine forest, *Agricultural and Forest Meteorology*, 43, 19–35, [https://doi.org/10.1016/0168-1923\(88\)90003-2](https://doi.org/10.1016/0168-1923(88)90003-2), 1988.
- Stickler, C. M., Coe, M. T., Costa, M. H., Nepstad, D. C., McGrath, D. G., Dias, L. C. P., Rodrigues, H. O., and Soares-Filho, B. S.: Dependence of hydropower energy generation on forests in the Amazon Basin at local and regional scales., *Proceedings of the National Academy of Sciences of the United States of America*, 110, 9601–6, <https://doi.org/10.1073/pnas.1215331110>, 2013.
- 15 The MathWorks, I.: MATLAB 8.4, 2014.
- Tuinenburg, O. A.: Atmospheric Effects of Irrigation in Moonsoon Climate: The Indian Subcontinent, Ph.D. thesis, Wageningen University, 2013.
- van der Ent, R. J.: A new view on the hydrological cycle over continents, Ph.D. thesis, Delft University of Technology, 2014.
- van der Ent, R. J.: WAM2layersPython, <https://github.com/ruudvdent/WAM2layersPython>, 2016.
- 20 van der Ent, R. J. and Savenije, H. H. G.: Length and time scales of atmospheric moisture recycling, *Atmospheric Chemistry and Physics*, 11, 1853–1863, <https://doi.org/10.5194/acp-11-1853-2011>, 2011.
- van der Ent, R. J. and Savenije, H. H. G.: Oceanic sources of continental precipitation and the correlation with sea surface temperature, *Water Resources Research*, 49, n/a–n/a, <https://doi.org/10.1002/wrcr.20296>, 2013.
- van der Ent, R. J., Tuinenburg, O. A., Knoche, H. R., Kunstmann, H., and Savenije, H. H. G.: Should we use a simple or complex model for moisture recycling and atmospheric moisture tracking?, *Hydrology and Earth System Sciences*, 17, 4869–4884, <https://doi.org/10.5194/hess-17-4869-2013>, 2013.
- 25 van der Ent, R. J., Wang-Erlandsson, L., Keys, P. W., and Savenije, H. H. G.: Contrasting roles of interception and transpiration in the hydrological cycle – Part 2: Moisture recycling, *Earth System Dynamics*, 5, 471–489, <https://doi.org/10.5194/esd-5-471-2014>, 2014.
- Wang-Erlandsson, L., van der Ent, R. J., Gordon, L. J., and Savenije, H. H. G.: Contrasting roles of interception and transpiration in the hydrological cycle – Part 1: Temporal characteristics over land, *Earth System Dynamics*, 5, 441–469, <https://doi.org/10.5194/esd-5-441-2014>, 2014.
- 30 Wang-Erlandsson, L., Bastiaanssen, W. G. M., Gao, H., Jägermeyr, J., Senay, G. B., van Dijk, A. I. J. M., Guerschman, J. P., Keys, P. W., Gordon, L. J., and Savenije, H. H. G.: Global root zone storage capacity from satellite-based evaporation, *Hydrology and Earth System Sciences*, 20, 1459–1481, <https://doi.org/10.5194/hess-20-1459-2016>, 2016.
- 35 Wei, J., Dirmeyer, P. A., Bosilovich, M. G., and Wu, R.: Water vapor sources for Yangtze River Valley rainfall: Climatology, variability, and implications for rainfall forecasting, *Journal of Geophysical Research*, 117, D05 126, <https://doi.org/10.1029/2011JD016902>, 2012.

- Wei, Z., Yoshimura, K., Wang, L., Miralles, D. G., Jasechko, S., and Lee, X.: Revisiting the contribution of transpiration to global terrestrial evapotranspiration, *Geophysical Research Letters*, 44, 2792–2801, <https://doi.org/10.1002/2016GL072235>, 2017.
- Weng, W., Luedeke, M., Zemp, D., Lakes, T., and Kropp, J.: Aerial and surface rivers: Downwind impacts on water availability from land use changes in Amazonia, *Hydrology and Earth System Sciences*, 22, 911–927, <https://doi.org/10.5194/hess-22-911-2018>, 2018.
- Werth, D. and Avissar, R.: The local and global effects of Amazon deforestation, *Journal of Geophysical Research*, 107, 8087, <https://doi.org/10.1029/2001JD000717>, 2002.
- 5 Xie, S. P., Deser, C., Vecchi, G. a., Ma, J., Teng, H. Y., and Wittenberg, a. T.: Global Warming Pattern Formation: Sea Surface Temperature and Rainfall, *Journal of Climate*, 23, 966–986, <https://doi.org/10.1175/2009jcli3329.1>, 2010.
- Zemp, D. C., Schleussner, C.-F., Barbosa, H. M. J., Hirota, M., Montade, V., Sampaio, G., Staal, A., Wang-Erlandsson, L., and Rammig, A.: Self-amplified Amazon forest loss due to vegetation-atmosphere feedbacks, *Nature Communications*, 8, 14681, <https://doi.org/10.1038/ncomms14681>, 2017.
- 10 Zheng, J., Sun, G., Li, W., Yu, X., Zhang, C., Gong, Y., and Tu, L.: Impacts of land use change and climate variations on annual inflow into the Miyun Reservoir, Beijing, China, *Hydrology and Earth System Sciences*, 20, 1561–1572, <https://doi.org/10.5194/hess-20-1561-2016>, 2016.