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Florence Habets, Jérôme Molenat, Nadia Carluer, Olivier Douez, Delphine Leenhardt

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The cumulative impacts of small reservoirs on hydrology: a review

Highlights

- The number of small dams is still increasing and is approaching 39 dams per square kilometre
- Small dams lead to a decrease in annual stream discharge of $13\% \pm 8\%$
- Cumulative impacts cannot be estimated using simple indicators
- Cumulative impacts are difficult to estimate and are most often quantified from modelling
- The lack of information on small reservoir characteristics is a real shortcoming for properly estimating their cumulative impacts

Email addresses: florence.habets@upmc.fr (Habets Florence), jerome.molenat@inra.fr (Molénat Jérôme)

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The cumulative impacts of small reservoirs on hydrology: a review

Habets Florencea,∗, Molénat Jérômeb,∗, Carluer Nadiaa, Douez Olivierd, Leenhardt Delphinee

a CNRS/Sorbonne Université, UMR 7619 Metis, Paris, France
b UMR LISAH, Univ Montpellier, INRA, IRD, Montpellier SupAgro, Montpellier, France
c Irstea, UR RiverLy, centre de Lyon-Villeurbanne, 69625 Villeurbanne, France
d BRGM, Bordeaux, France
eToulouse Univ, INRA, INPT, INP EI PURPAN, AGIR, Castanet Tolosan, France

Abstract

The number of small reservoirs has increased due to their reduced cost, the availability of many favourable locations, and their easy access due to proximity. The cumulative impacts of such small reservoirs are not easy to estimate, even when solely considering hydrology, which is partially due to the difficulty in collecting data on the functioning of such reservoirs. However, there is evidence indicating that the cumulative impacts of such reservoirs are significant.

The aim of this article is to present a review of the studies that address the cumulative impacts of small reservoirs on hydrology, focusing on the methodology and on the way in which these impacts are assessed.

Most of the studies addressing the hydrological cumulative impacts focused on the annual stream discharge, with decreases ranging from 0.2% to 36% with a mean value of 13.4% ±8% over approximately 30 references. However, it is shown that similar densities of small reservoirs can lead to different impacts on stream discharge in different regions. This result is probably due to the hydro-climatic conditions and makes defining simple indicators to provide a first guess of the cumulative impacts difficult. The impacts also vary in time, with a more intense reduction in the river discharge during the dry years than during the wet years. This finding is certainly an important point to take into consideration in the context of climate change.

Two methods are mostly used to estimate cumulative impacts: i) exclusively data-based methods and ii) models. The assumptions, interests and shortcomings of these methods are presented. Scientific tracks are proposed to address the four main shortcomings, namely the estimation of the associated uncertainties, the lack of knowledge on reservoir characteristics and water abstraction and the accuracy of the impact indicators.

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1. Introduction

Large reservoirs have strong impacts on hydrology at regional to global scales. Indeed, it was estimated that such large reservoirs have led to a global runoff decrease of approximately 2% (Biemans et al., 2011), to a sea level decrease of approximately 30 mm (Chao et al., 2008), and that they store a volume equivalent to approximately 10% of the natural annual soil storage capacity at the global scale (Zhou et al., 2016). However, these studies did not consider the impacts of smaller reservoirs on hydrology. Downing (2010) found that small ponds and lakes (smaller than 0.1 km$^2$) cover a larger area and are more numerous than large reservoirs and that approximately 10% of them are constructed reservoirs.

When considered individually, each reservoir may modify its local and remote environment. The cumulative impacts of many reservoirs in a catchment are the modifications induced by a set of reservoirs (or reservoir network) taken as a whole. The cumulative impacts are not necessarily the sum of individual modifications because reservoirs may be inter-dependent, such as cascading reservoirs along a stream course. Cumulative impacts are not the simple addition of individual impacts: they can develop via an additive or incremental process, a supra-additive process (where the cumulative effect is greater than the sum of the individual effects) or an infra-additive process (where the cumulative effect is less than the sum of the individual effects). The total impact is therefore equal to the sum of the impacts of the developments and to interaction effects. Indeed, addressing the cumulative impacts implies covering different spatial and temporal scales (Canter and Kamath, 1995) and having a reference state (McCold and Saulsbury, 1996). The cumulative impacts of small reservoirs on sediment transport, biochemistry, ecology and greenhouse gas emissions have been studied (Berg et al., 2016; Mbaka and Wanjiru Mwaniki, 2015; Downing, 2010; Poff and Zimmerman, 2010; St. Louis et al., 2000), as have the impacts of such reservoirs on hydrology (Nathan and Lowe, 2012; Fowler et al., 2015). The reported impacts are generally strong but present a large variation.

Estimating the cumulative impacts of systems of small reservoirs on a given basin has become an issue as their number increases (for instance, a 3% increase per year in the US (Berg et al., 2016)). This trend may persist because these systems are often considered to be a technique to...
adapt to climate change (van der Zaag and Gupta, 2008). Indeed, small reservoirs are mainly
used to store water during the wet season to support water use during the dry season, particularly
for irrigation and livestock in rural areas (Wisser et al., 2010; Nathan and Lowe, 2012); to store
water during storms to prevent flooding; or to store sediments in check dams to reduce erosion
and muddy flood risks. Because the part of the global population that will experience water
scarcity is projected to increase with climate change and because the intensity of storm events is
also projected to simultaneously increase (Pachauri et al., 2014), there is increasing pressure to
construct small reservoirs (van der Zaag and Gupta, 2008; Thomas et al., 2011).

However, an uncontrolled development of such small reservoirs may increase the water re-
source problem in both quantitative and qualitative ways. Thus, water managers are seeking some
indicators that would help to determine optimal networks of small reservoirs in terms of storage
capacities and in terms of locations and management. Consequently, in France, the Ministry of
the Environment requested a joint scientific assessment to collect useful information/knowledge
and tools to provide local stakeholders with such indicators and methods to assess the cumulative
impacts of small reservoirs. This request led to a review covering biochemistry, ecology, hydrol-
ogy and hydromorphology (Carluer et al., 2016). In this paper, a full review of the cumulative
impacts of small reservoirs on hydrology is presented because the hydrological impact will a
fect the other impacts. Although there is no accepted definition of small reservoirs, it is commonly
accepted that the storage capacities of such reservoirs are below 1 million m$^3$, as stated by Ayalew
et al. (2017) and Thomas et al. (2011). This review does not extend to the very small reservoirs
of few hundreds of m$^3$ that can be used for water harvesting (Lasage and Verburg, 2015).

First, a synthesis of the quantification of the impacts at the basin scale is presented, and the
ability of some conventional descriptors to be used as indicators is studied. Then, the various
ways in which small reservoirs can impact the water cycle are presented, along with the methods
that are used in the literature to estimate the cumulative impacts of such numerous and not always
well-known structures. These results are then discussed, addressing the uncertainties, long-term
trends, and impacts on other biochemical, ecological and social components.

2. Evidence of the impacts of small reservoirs on hydrology

From the literature review, the cumulative impacts of small reservoirs on hydrology are most
often estimated from the annual discharge, low flows and floods. There is a general consensus
that sets of small reservoirs lead to a reduction in the flood peaks (Frickel, 1972; Galea et al.,
2005; Nathan and Lowe, 2012; Thompson, 2012; Ayalew et al., 2017) of up to 45%, particularly
since some reservoirs are constructed as stormwater retention ponds (Fennessey et al., 2001;
Del Giudice et al., 2014). However, over-topping flooding or dam failure can result in large
floods (Ayalew et al., 2017), which may lead to casualties including death (Tingey-Holyoak,
2014). Such failures can be more frequent for small dams than for larger dams due to the lack of
adapted policies, which may lead to a lack of maintenance and a tendency to store excess water
to secure production (Pisaniello, 2010; Camnasio and Becciu, 2011; Tingey-Holyoak, 2014).

The low flows are also frequently reported to decrease when a set of small reservoirs is
present in a basin (Neal et al., 2000; O’Connor, 2001; Hughes and Mantel, 2010; Nathan and
Lowe, 2012; Thompson, 2012) with a large spread (0.3 to 60%), although the water stored can
occasionally be used to sustain a low flow (Thomas et al., 2011). The majority of studies have
focused on the annual stream discharge, reporting a decrease in the mean annual discharge that
ranges from 0.2% (Hughes and Mantel, 2010) to 36% (Meigh, 1995). On average, in approxi-
mately 30 references, the decrease in the mean annual discharge reaches 13.4% ±8% (Figure 1
and Appendix Table A.1).

Figure 1: Left: Distribution of the estimated annual stream discharge decrease attributed to reservoir networks. The
distribution is established based on 20 values. Right: Impact on the annual discharge estimated during wet, median
and dry years. Each bar corresponds to a different catchment. The estimations are from the following references: a:

The right part of Figure 1 shows that the impacts on annual flows are not constant from year
to year but tend to be lower during the wet years and two times greater than the median impact
in the driest years. This result is very important because it indicates that even without changing
the small reservoir network, its impacts will change in the context of climate change: it may
decrease in areas that will become wetter but may increase in areas that will become drier.

One key issue in estimating the cumulative impacts is understanding how such impacts are
related to the reservoir network, i.e., the level at which the basin is equipped with small dams
to avoid over-equipping the basin, with consequences in terms of economy and ecology. Having
a single indicator or a set of indicators capable of estimating the cumulative impacts of small
reservoirs on the mean annual discharge would be helpful to most water management agencies.

Based on the estimated values collected in the literature, a preliminary analysis was performed
to determine whether some easy-to-access properties of the reservoir network could be used as
indicators. For this purpose, we collected the main characteristics of the basins and of their small
reservoir network from the available studies and attempted to connect them to the impacts on the
mean annual discharge. We used the reservoir’s density, expressed as the number of reservoirs
per square kilometre or as the volume stored per square kilometre, and the mean precipitation
or the mean discharge in the basin. The results presented in Figure 2 show that none of these
characteristics are able to be used as indicators for such contrasted basins as the ones found in
the literature. Indeed, within a narrow range of specific discharge or precipitation, the decrease
of the annual discharge varies a lot and can not be correlated to the density of reservoir network.

A more regional-scale view could be useful to attempt to disentangle different types of cli-
mate or use. However, according to the sample of available studies, only a continental-scale anal-
ysis was possible. It appears from these figures that the general characteristics present a wider
spread between continents than within a given continent, even if the results are from different
studies. For instance, the specific discharge is low in Australia, the density is low in Africa, and
the storage volume tends to be important in America. However, even within a continent, these
characteristics are not sufficiently well linked to the impacts to be reliably used as indicators.

This result occurs because the cumulative impacts of reservoir networks rely on a large num-
ber of factors: the hydrological processes occurring in each reservoir, the water management
(water abstraction rate and timing, water uptakes from and releases to the river), the reservoir
characteristics, the reservoir network geometry, and the connectivity of each reservoir to the
stream drainage network. These points are detailed below.
3. How do small reservoirs impact hydrology?

Small reservoirs have an impact on hydrology because they affect the natural water cycle that would occur without reservoirs. To understand how networks of small reservoirs impact river flow at the basin scale, it is necessary to understand the functioning of a single reservoir, how it can have an impact on the river flow and why the impact varies in time and from one reservoir to another.

3.1. Water balance of a small reservoir

Figure 3 presents the various terms of the water balance of the reservoir. From a general perspective, the reservoir water balance can be expressed by the following equation:

\[ \text{Reservoir Water Balance} = \text{Inflow} - \text{Outflow} - \text{Storage Change} - \text{Losses} + \text{Infiltration} \]
Figure 3: Water balance of a small reservoir and its main drivers. The components of the water balance are indicated by large arrows: inputs can be inflows, such as upstream runoff, lateral surface runoff, and direct precipitation; outputs can be outflows, abstraction, seepage and evaporation.

\[
\frac{dV}{dt} = Q_{in} + P + GW_{in} - Q_{out} - E - S - Q_{abs}
\]  
(1)

Here, \(dV\) is the water volume variation \([m^3]\) over the period \(dt\) \([s]\), \(Q_{in}\) is the stream inflow to the reservoir \([m^3/s]\), \(Q_{out}\) is the outflow from the reservoir \([m^3/s]\), \(E\) is the evaporation rate \([m^3/s]\), \(P\) is the precipitation rate \([m^3/s]\), \(S\) is the seepage rate \([m^3/s]\), \(GW_{in}\) is the groundwater inflow \([m^3/s]\) and \(Q_{abs}\) is the water abstraction \([m^3/s]\).

Inflow can have 4 sources: i) the upstream flow, which depends on the way in which the reservoir is connected to the river (Section 3.3); ii) the surface runoff from the area directly drained by the reservoir along its bank; iii) the intercepted precipitation; and iv) a groundwater inflow, although none was reported in the literature review.

Outflux includes outflow (downstream flow) and water abstraction, as well as evaporation and seepage losses from the reservoir. Outflow is defined as the downstream flow due to reservoir release. Abstraction corresponds to the water uptake, often by pumping, for human use (irrigation, livestock watering, and so forth). Seepage flow may occur as water infiltration through the reservoir bed or through or below the dam.

All these fluxes can vary considerably from one reservoir to another. For instance, abstraction...
can be the main output, especially for farm reservoirs. However, it can also be null, such as in storm water or check dam reservoirs. Section 6.3 discusses how abstraction can be estimated at the basin scale.

Water losses are present for every type of reservoir, but with a large spread of intensity, ranging from the main outflux to negligible ones. The next section focuses on these losses and on how they can be estimated.

3.2. Losses from small reservoirs

3.2.1. Seepage

Seepage (also called percolation flux) may be particularly important to consider for small reservoirs because most of these reservoirs are built with earthen dams. The seepage rate depends on the hydraulic head gradient between the reservoir and the underlying aquifer (or unsaturated zone) or dam wall, as well as on the hydraulic conductivities of the aquifer and reservoir bed material.

Although seepage is a loss at the reservoir scale, the water is not lost and is mostly diverted. Indeed, infiltration tanks, encountered especially in Asia, are built to favour infiltration through the reservoir bed to increase the groundwater recharge. In this way, a larger part of the monsoon flow is stored in the groundwater while avoiding the evaporation loss from reservoirs during the dry season (Glendenning et al., 2012). However, when dams are intended to store water over the long term, seepage is considered as a loss. In such cases, impervious layers of clay or geomembrane (Alonso et al., 1990; Yiasoumi and Wales, 2004) are used to reduce seepage, but their efficiency decreases with age. Thus, irrespective of the intended function of the reservoir, it is rather important to estimate the seepage rate from the reservoir because it determines its efficiency for storing water (then, a low seepage rate is expected) or within the groundwater (then, a high seepage rate is expected). In the literature, estimations of the seepage rate were based on water balance approaches constrained by local observations of the precipitation, potential evaporation and reservoir’s water level (Culler, 1961; Kennon, 1966; Sukhija et al., 1997; Singh et al., 2004; Boutefeha et al., 2015), as well as on additional observations of the soil moisture and piezometric heads (Shinogi et al., 1998; Antonino et al., 2005; Massuel et al., 2014), environmental tracers (Sukhija et al., 1997), or more frequently on modelling approaches (Zammouri and Feki, 2005; Boisson et al., 2014; Jain and Roy, 2017).

Figure 4 presents some estimations of the seepage and evaporation losses from the literature.
under different hydroclimatic contexts and for reservoirs built for various purposes. Most estimated seepage values are greater than 5mm/day on average in the studied periods, and thus, the seepage rate appears to be higher than the evaporation rate. However, most of the values found in the literature are from percolation tanks, i.e. from dams built to promote a rapid infiltration of the runoff during the wet season to recharge the water table. For the other types of dams, the estimations can be lower: less than 1mm/day for Culler (1961) in the US and up to 6.2mm/day for Shinogi et al. (1998) over a 6-month period in a basin in Brazil. Fowler et al. (2012, 2015) consider that hillslope dams in Australia are not efficient for storing water if the seepage rate is greater than 5mm/day.

When the cumulative impacts are considered, both the seepage rate and the seepage fate are important. In the case of infiltration into the dam wall, the seepage water might flow downstream in the river, and thus, the seepage flux might not be lost at the scale of the river basin. An illustration of such a process was provided by Kennon (1966), who observed that ephemeral rivers have become permanent after the implementation of dams built to prevent erosion (see Section 4.1.1), and by those studies that include groundwater recharge from dam seepage (Ramireddy et al., 2000; Barber et al., 2009; Smout et al., 2010; Shinde et al., 2010; Perrin et al., 2012). Therefore, seepage fluxes from each reservoir should not be aggregated to estimate the loss at the basin scale and thus for the estimation of the cumulative impacts of small dams on hydrology.

### 3.2.2. Evaporation

Unlike seepage, evaporation fluxes from each reservoir should be aggregated at the basin scale. The impact of the reservoirs on the evaporation losses is then the difference between the evaporation from the land cover that was present prior to the dams being built and the evaporation from the reservoirs. Such estimations are not straightforward, particularly because the heat storage of the water body affects the surface energy flux (Assouline et al., 2008; McMahon et al., 2013). This storage partly depends on the temperature of the water columns, which is impacted by the depth of the dams (although in opposite ways depending on the references (Girard, 1966; Martínez Alvarez et al., 2007; Magliano et al., 2015) due to the associated change in the free water area); on the water circulation within the reservoir (which is also impacted by the reservoir's management); and on the interaction with the edges, which can be rather close for small reservoirs and that affects the wind velocity and the advection of air humidity (Figure 3). Several methods were used to provide estimations of the evaporation from small reser-
Figure 4: Estimation of the seepage loss and the evaporation flux of small reservoirs on a seasonal to annual basis. Two types of reservoirs are distinguished: infiltration reservoirs and other types of reservoirs. The values are taken from the articles cited in this section.

Martínez Alvarez et al. (2007) proposed a relationship between the small reservoir evaporation loss and the Class A pan evaporation that varies according to the reservoir’s depth and area and that varies in time (from 86 to 94%).

Several estimations of the small reservoir evaporation loss based on meteorological data were proposed (de Bruin, 1978; Martínez Alvarez et al., 2007; McJannet et al., 2013; McMahon et al., 2013; Morton, 1983). Benzaghta and Mohamad (2009), Martínez Alvarez et al. (2008) and Craig (2008) found that the evaporation losses from reservoirs can be very important at the regional scale and have an important economic impact.

Several techniques might help reduce evaporation from reservoirs: casual chemical treatment
to modify the albedo or form a monolayer film, completely or partially covering the reservoirs,
managing the reservoir edges to reduce wind speed, and optimizing the use of the water in reser-
voir networks based on the temperature of the water in the reservoirs (Barnes, 2008; Lund, 2006;
Assouline et al., 2011; Martínez-Alvarez and Maestre-Valero, 2015; Gallego-Elvira et al., 2011;
Carvajal et al., 2014; Reca et al., 2015). However, such techniques are not yet widely used and
are not considered in the existing cumulative impact studies.

3.3. Connection to the stream

By itself, the connection of the reservoir to the stream is key to understanding the impacts
of the reservoir on the river flow. Indeed, this connection will impact both the inflow and the
outflow. Small reservoirs can collect all the upstream flow (Figure 5-a for a hillslope reservoir
or dam situated on the stream with no minimum flow) or only a part of the flow (reservoir
with minimum flow by-pass, Figure 5-b, which allows maintaining a minimum flow, or dam
situated in diversion Figure 5-c since in this case, the reservoir can not fill as long as inflow
does not exceed some thresholds). In the case that all the upstream flows are collected, the
downstream outflow will primarily depend on the level of the spill and on the reservoir water
storage. Following “fill-and-spill” (Deitch et al., 2013), downstream discharge occurs only when
the reservoir is fully filled; conversely, as long as the reservoir has not reached its capacity,
downstream discharge is null. Therefore, it is possible to have periods with no downstream flow
while upstream flow exists, such as for hillslope reservoirs and check dams. Such reservoirs have
strong impacts on the intensity and the duration of low flows. In particular, the resumption of
flow in the fall can be significantly delayed. In the case of diversion or a minimum flow bypass
reservoir, a downstream flow is ensured when the upstream flow is non-zero. If the reservoir is
located in diversion, then the filling period of the reservoir can be managed such that the reservoir
may have no impact on the river flow during parts of the year, which may allow preserving
the ecological function of the river. This management can also be adapted to the hydrological
situation of each year. The reservoirs built mainly to favour groundwater recharge can have
all types of connections with the river; however, it appears that most of them are built directly
in the river stream, thus collecting all the upstream flows (Shinogi et al., 1998; Siderius et al.,
2015). Depending on the respective inflow and abstraction dynamics, cumulative abstraction
may exceed the reservoir storage capacity, as illustrated in Figure 5 for example, for which the
abstractions from the reservoirs reach 105 to 120% of the maximum storage capacity.
Figure 5: Illustration of 3 different connections between the river and the reservoir and its consequences in terms of river flow. Inflow, outflow and abstraction are accumulated weekly values, whereas storage is a weekly value. They are all expressed as a fraction of the maximum storage. Abstractions in the reservoirs reached 105 to 120% of the maximum storage capacity. a) Hillslope reservoir is managed as a fill and spill, with a weak and irregular inflow. b) The minimum flow bypass ensures that a minimum outflow occurs as long as inflow is present. c) The reservoir in diversion is expected to fill up as soon as the inflow reaches a given minimum flow or depending on management practices.

4. Methods to estimate the cumulative impacts of small reservoirs on hydrology

Quantifying the cumulative impacts of small reservoirs has been conducted using a variety of methods, all of them requiring data and observations. Two classes of methods can be distinguished: i) the methods exclusively based on the analysis of observed data and ii) the methods based on hydrological modelling.

4.1. Exclusively data-based methods

4.1.1. From observation of selected reservoirs to estimation of cumulative impacts

This approach was mainly used in early works performed from the 50s to the early 70s in the US (Kennon, 1966; Culler, 1961; Frickel, 1972) and in Brazil (Dubreuil et al., 1968; Dubreuil and Girard, 1973; Molle, 1991). In light of these pioneering works, it can be observed that the
cumulative impacts on hydrology have been a scientific and water management issue for a long time.

Despite some differences in the methodology among these studies, they all aimed at quantifying single reservoir hydrologic functioning from the monitoring of a sample of reservoirs. Losses were estimated using a mass balance of the sampled representative reservoirs based at least on the monitoring of the water level, inflows and outflows of the reservoirs. These early studies initially made the assumption that cumulative reservoir impacts were the sum of the impact of each reservoir following an aggregation process. However, the main outcome of these studies was to show that this assumption was not valid. Indeed, Culler (1961) and Kennon (1966) found that the seepage was a significant loss for the sampled reservoirs but contributed to downstream flow. Therefore, interactions between reservoirs and hydrologic compartments, especially the stream, were identified very early as processes to be taken into consideration to reliably estimate the cumulative impacts.

4.1.2. Statistical analyses of the observed discharge

The idea is to connect the detected changes in the statistical properties of river discharge time series with the evolution of the reservoir network within the basin. In doing so, the details of each reservoir functioning are not taken into consideration. To our knowledge, this type of study based solely on observations was only performed by Galea et al. (2005). A study based on a 30-year river discharge time series of two French catchments showed no stationarity break in summer, while a break was shown in winter, i.e., during the filling period (Galea et al., 2005).

One difficulty of such statistical analyses is discriminating the specific impact of small reservoirs from those of land use and land cover (LULC) evolution or of climate change (CC). Reservoir development occurred over decades, a sufficiently long period to be sensitive to LULC modifications (such as agricultural intensification or crop modification) and CC. To overcome this issue, Schreider et al. (2002) compared the observed river flows with simulated ones obtained using the observed atmospheric forcing, but without any explicit representation of the small dams in the models. The IHACRES rainfall-runoff model, a dynamic, lumped parameter model, was used to simulate stream flow with parameters calibrated considering periods before the development of reservoirs. They found significant decreasing trends in the observed discharge of basins that had a development of farm dam capacity, and they were able to attribute these trends to non-climatic stressors since such trends were not simulated with a reservoir-free basin.
4.1.3. Paired-catchment experiment

A paired-catchment experiment is an approach already used in hydrology for quantifying the impact of LULC changes from a comparative analysis of stream flows monitored in two contrasted catchments (see, for instance, Brown et al. (2005) for a review in forest hydrology). Thompson (2012) is, to our knowledge, the only study using this approach to compare stream flows from two adjacent and similar catchments, one without a reservoir and the second with three small reservoirs. From an 18-month monitoring, annual stream flow was estimated to be lower by 40% in the catchment with 3 reservoirs than in the ”no-reservoir” catchment (Thompson, 2012). Although the experiment found differences in the specific discharge, the full comparison of the water balance remained difficult. The main shortcoming of Thompson’s approach is that catchment properties (soils, lithology, land cover, topography, and so forth) were spatially heterogeneous over a short distance, making deciphering the stream flow differences difficult. Furthermore, indirect reservoir impacts on land use, such as the cattle grazing around the reservoir in Thompson’s case study, can also modify stream flow. The study would have benefited from following the classic approach used in paired-catchment experiments, implying a calibration period where both catchments are monitored, followed by a period when one of the catchments is subjected to land use change (reservoir building) and the other remains as a control. However, building a reservoir network over a large area is generally difficult for practical and financial reasons. Consequently, such an approach has never been utilized to our knowledge.

4.2. Modelling approaches

Modelling is the most widely used approach for studying and quantifying the cumulative impacts of small reservoirs. Although various modelling approaches have been developed, all are based on the coupling of the small reservoir water balance model with a quantitative method to estimate stream inflow into the small reservoirs. Three of the main model components are detailed below: i) the small reservoir water balance model, ii) the quantitative method used to quantify inflow to reservoirs, and iii) the spatial representation of the reservoir network. The inflow quantification method and the spatial representation of the reservoir have to be consistent and are thus intrinsically dependent. A spatially distributed representation of reservoirs requires being able to estimate the spatial distribution of stream flow to estimate the upstream inflow to each reservoir. Conversely, an aggregated estimation of stream flow over a sub-basin or over the full catchment leads to the reservoir network representation being aggregated on the same
domain.

Most of the reviewed studies focused on assessing the impacts of reservoirs used for irrigation or livestock watering on stream flow. In such cases, the impacts are quantified by comparing the catchment stream flow simulation with and without reservoirs, except for the TEDI model, as we will see in Section 4.2.2. The exceptions to modelling approaches dedicated to stream flow impacts are those aiming at assessing the impacts on groundwater. These approaches mostly focus on infiltration tanks, for which part of the stored volume recharges the aquifer. In such cases, only the impacts on the aquifer due to the loss from the reservoirs are represented, either without simulation of the groundwater (Martín-Rosales et al., 2007; Hughes and Mantel, 2010), with a simplified representation of the aquifer (Smout et al., 2010; Shinde et al., 2010; Perrin et al., 2012), or even more seldom, with a 2-D hydrogeological model (Ramireddygari et al., 2000; Barber et al., 2009).

4.2.1. Reservoir water balance model

Reservoir water balance models rely on equation (1). Most small reservoir water balance models take into account the evaporation and abstraction, for which temporal estimation is rarely well known and is often an important point (see Section 5.3) (Table 1). When seepage is taken into account, it is considered only as infiltration to groundwater. Ignoring seepage is justified by the small expected rates (Hughes and Mantel, 2010) or by the lack of information on the process (rate, timing, and driving factor Günther et al. (2004)) and by the fact that seepage flux can contribute to downstream flow. To simulate the reservoir water mass balance, downstream discharge is simulated considering that reservoirs operate with the technique of "fill-and-spill" (Section 3.3, unless a conservation flow is taken into account (Table 1). Reservoir inflow is simulated by different approaches, as presented in the next section.

4.2.2. Reservoir inflow quantification

In most modelling approaches, upstream inflow is provided by a catchment hydrological model simulating the water balance (WB), or the energy and water balance (EVB), in the upstream catchment and the routing of the flow downstream (Table 1). Existing catchment hydrological models are used in the modelings, reflecting the diversity of current hydrological models. Such models need atmospheric forcing and some information on the land cover, soil and topography, unless the model parameters are calibrated without any data on the physiographic characteristics. Two models, TEDI and Deitch (Table 1), developed an alternative and pragmatic
method based on using observed discharge time series as input to the model. In doing so, the TEDI and Deitch models do not belong to any current modelling approaches. Using observed discharge at available river gauges implies being able to successively i) disentangle the natural flow from the anthropogenic flow and ii) distribute the observed discharge along the reservoir networks. To achieve the first step, Deitch et al. (2013) used historical gauged discharge measured prior to the reservoir pre-development period. The discharge was then spatially distributed according to the drainage area of each reservoir and the spatial distribution of the average annual rainfall. The propagation of stream water was then operated from the most upstream reach to the catchment outlet by considering the water volume intercepted in each reservoir. The cumulative impact of reservoirs is then classically the difference between simulated discharge and the gauged discharge. In TEDI, Nathan et al. (2005) used the observed discharge of the period of interest. The inflow in each reservoir is calculated from the observed catchment discharge assuming a proportionality with the reservoir catchment area. The outflow from every reservoir is transferred directly to the outlet. It is then considered that the obtained cumulative impact corresponds to twice the simulated impact of the reservoir network because the gauge discharge already includes the impacts of existing reservoirs.

4.2.3. Reservoir spatial representation

How the reservoir network is represented from a spatial perspective varies from one model to another. The spatial representation of the reservoir network can be classified into the following three types (see Figure 6, Table 1).

- In the spatially aggregate approach, all the reservoirs in a catchment (in Table 1, A for aggregation on sub-catchments and A* for aggregation on a grid cell) are represented in the form of a single equivalent, or composite, reservoir.

- The statistical representation constitutes a refinement of the aggregate representation (Figure 6-B). The reservoir network is represented in the model in an aggregated way by grouping reservoirs into a finite number of classes. Some hydrological connections between several of these classes may be represented (S in Table 1).

- The spatially explicit representation consists of representing every reservoir (Figure 6-C).
<table>
<thead>
<tr>
<th>Model</th>
<th>spatial representation</th>
<th>time step</th>
<th>Processes included</th>
</tr>
</thead>
<tbody>
<tr>
<td>CRU&lt;sup&gt;i&lt;/sup&gt;</td>
<td>A</td>
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<tr>
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<td>month/day</td>
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<tr>
<td>WaterCAST&lt;sup&gt;c&lt;/sup&gt;</td>
<td>S</td>
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<td>D</td>
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<tr>
<td>CHEAT&lt;sup&gt;b&lt;/sup&gt;</td>
<td>D</td>
<td>month</td>
<td>OBS</td>
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<tr>
<td>Deitch et al.&lt;sup&gt;e&lt;/sup&gt;</td>
<td>D</td>
<td>day</td>
<td>OBS</td>
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<td>spill</td>
</tr>
<tr>
<td>PITMAN&lt;sup&gt;f&lt;/sup&gt;</td>
<td>A</td>
<td>month</td>
<td>WB</td>
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<td>spill</td>
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Table 1: Main processes in reservoir water balance model, as well as temporal and spatial representations of reservoirs in numerical models. Spatial representation can be the following (see Figure 6): A: aggregate representation by catchment (or A* by grid in grid-based models), S: statistical representation, or D: distributed representation. Inflow to the reservoirs can be derived from OBS: observations, WB: water balance, or EWB: energy and water balance. Outflow is computed either based on spill (above a water level or volume in the reservoir) and/or taking into account a conservation flow (CF).<sup>a</sup>: Nathan et al. (2005),<sup>b</sup>: Nathan et al. (2005),<sup>c</sup>: Cetin et al. (2009),<sup>d</sup>: Guntner et al. (2004),<sup>e</sup>: Deitch et al. (2013),<sup>f</sup>: Hughes and Mantel (2010),<sup>g</sup>: Perrin et al. (2012),<sup>h</sup>: Habets et al. (2014),<sup>i</sup>: Tarboton and Schulze (1991),<sup>j</sup>: Ramireddygari et al. (2000),<sup>k</sup>: Payan et al. (2008),<sup>l</sup>: Ragab et al. (2001),<sup>m</sup>: Shinogi et al. (1998), Jayatilaka et al. (2003), and<sup>n</sup>: Zhang et al. (2012)
<table>
<thead>
<tr>
<th>Scheme</th>
<th>Principles</th>
</tr>
</thead>
</table>
| Catchment | Reservoirs characteristics  
- Reservoirs (•) along the stream network (—)  
- Each reservoir has an upslope drainage area (the dashed areas for 2 examples) |
| Aggregative | Reservoir network represented by an equivalent single reservoir  
The simulated catchment discharge, \( Q_{\text{sup}} \), is the sum of \( aQ_{\text{arr}} \), the non-intercepted simulated stream discharge and \( Q_{\text{trans}} \), the downstream flow from the equivalent reservoir simulated by the mass balance reservoir model. \((1-a)\) is the catchment areal fraction of the cumulative upslope drainage area of all the reservoirs |
| Statistical | Reservoir network represented by classes of equivalent reservoirs. Each class corresponds to a given range of reservoir capacity.  
\( Q_{\text{arr}} \), the catchment discharge, is distributed over each reservoir class by simple rules. The outflow from one reservoir class is propagated through the downstream classes by simple rules of cascading |
| Distributed | Each reservoir represented along with its upstream flow (\( Q_{\text{arr}} \))  
A single mass balance model, applied to each reservoir, simulated the downstream flow of the reservoir (\( Q_{\text{out(r)}} \)), which is propagated downstream through the stream network |

Figure 6: Spatial representation of reservoir network in models used to quantify cumulative reservoir hydrologic impacts.
**Aggregate representation**

In the aggregate representation (Figure 6-A), the characteristics of the equivalent reservoir (capacity and surface area) are obtained by aggregating single reservoir characteristics. The main interest of the aggregate representation is to require only global information about the reservoirs and their characteristics. In fact, the spatial density of reservoirs within a catchment can be large, greater than 10 reservoirs/km\(^2\) in some cases (Nathan et al., 2005), and an exhaustive inventory of all reservoirs along with their characteristics is out of reach. Rather, a global estimation of reservoirs and their characteristics may be approximated from simple rules of spatial extrapolation (cf. Habets et al. (2014)). For instance, to estimate the inflow into the equivalent reservoir, it is necessary to determine the contributive catchment. It can be a fraction of the catchment area (Tarboton and Schulze, 1991; Hughes and Mantel, 2010) that can be estimated from the sum of the drainage area of all reservoirs or depending on the cumulative reservoir area (Habets et al., 2014).

The aggregate representation leads to obtaining a simulation of the hydrological cumulative impacts of reservoirs at the catchment, grid-cell or sub-catchment outlet but intrinsically does not allow simulating the cumulative impacts along the river network from the head to the outlet, unless the sub-catchments are small, which is often not the case because the size of the sub-catchment is often determined by the availability of river gauges. Furthermore, this representation may not reflect the different responses of the various reservoirs in terms of key processes (evaporation, infiltration, operations, and so forth (Zhang et al., 2012)).

**Statistical representation**

The statistical representation is a trade-off between the other two representations. It considers that information about the location and characteristics of reservoirs, particularly of small- and medium-sized reservoirs, cannot be exhaustively available. It also relies on the assumption that reservoir connectivity may play a role in the cumulative impacts. The reservoir network is represented by classes of reservoirs determined following reservoir water capacity (Güntner et al., 2004; Nathan et al., 2005; Lowe et al., 2005)) and also reservoir drainage area (Zhang et al., 2012). Each class is represented as a single equivalent reservoir. Güntner et al. (2004) and Zhang et al. (2012) used a coupled sequential and parallel scheme to represent the upstream-downstream connectivity of different water reservoir classes in the catchment.

As a main advantage, the statistical representation has to consider the diversity of key reser-
voir processes, which can be variable from one reservoir to another but quite homogeneous in reservoirs of similar sizes. In this way, it overcomes one of the main shortcomings of the aggregate representation. Evaporation, for example, depends on the water column height and circulation within the reservoir, which is expected to depend on reservoir size (cf. Section 3.2). Connectivity to the network—reservoirs and rivers—and operation rules may also be different depending on the reservoir function, which also depends on the reservoir size. Another advantage of the statistical representation is being computationally faster than the fully distributed one because fewer reservoir mass balances have to be computed and water transfers between reservoirs are simplified. The main shortcoming is that it does not obtain distributed simulations of the hydrological impacts of reservoirs; particularly, the cumulative impacts along the full river network cannot be simulated.

**Distributed representation**

A distributed representation of the reservoir is the only way to explicitly represent the interactions between reservoirs by considering the outflow from one reservoir as a contribution to the inflow of the downstream one and the interactions between reservoirs and hydrological compartments (river, soil, and aquifer) by estimating the impacts of each reservoir on its connected river reach or/and aquifer. Indeed, two dams with similar characteristics may have different impacts according to their location along the stream network, mostly because the inflow is not the same. The interest in a spatially explicit representation is in quantifying and understanding the local hydrologic impact at a river reach scale and the cumulative impacts along the river network (Deitch et al., 2013). Quantifying local hydrologic impacts may be particularly relevant to water quality, ecological disturbance or morphogenesis evolution. In a spatially explicit representation, water inflow into every single reservoir as stream discharge and lateral surface runoff has to be known or estimated. To our knowledge, only Shinogi et al. (1998) and Smout et al. (2010) have performed catchment hydrologic modelling to obtain these estimations, with an application to relatively simple case studies characterized by few reservoirs. Other reported case studies using spatially explicit representations used observed-stream-discharge-based models (Nathan et al., 2005; Cetin et al., 2009; Deitch et al., 2013).

A spatially explicit representation relies on the availability of exhaustive information about reservoir location, characteristics, water uses and topology, which are rarely available over large areas. This point constitutes a main shortcoming of the approach, as addressed in Section 4.1.1.
Furthermore, it can be expected that uncertainties in the local information, added to the uncertainty in estimated spatial discharge and individual reservoir water balances, can skew the local simulated impacts, and by propagation, the cumulative impacts. This could alleviate the theoretical interest in the spatially explicit representation. Acknowledging the lack of information and the difficulty to obtain it exhaustively, statistical representations and aggregate representations are considered as pragmatical solutions and used in most modelling studies.

5. How to obtain access to the information needed on small reservoirs?

5.1. What type of data?

Stream discharge time series, at one or several points in the catchment, are required data in statistical analyses (Section 4.1.2) and in the TEDI and Deitch models (Nathan et al., 2005; Deitch et al., 2013, section 4.2.1), and such data are also used by the other types of models to calibrate or assess the modelling. Such data are expected to be found in existing databases. Statistical analyses require rather long observation periods for both the discharge and the temporal evolution of the reservoir network to cover contrasted periods. The modelling approaches generally need to collect more data, even if focusing on a shorter time period. These data include atmospheric and physiographic data, as well as the characteristics of each reservoir (or of the aggregated ones), the connection between the reservoirs, and the management of the reservoirs, particularly in terms of abstraction. Table 2 presents some of the most commonly required data on the reservoirs used for such studies.

5.2. Physical and topographical characteristics of small reservoirs

Data on small reservoir characteristics may be collected and stored in databases by stakeholders or state or regional agencies. Although they are often a first base to initiate a study and may prove very useful, such databases are generally incomplete, even for the census of the reservoirs, either because the survey did not include all the existing reservoirs or because the database is not up to date. Moreover, all the needed data are not available. Therefore, to fill the gaps, several methods can be used: i) additional field surveys, ii) remote sensing data (either satellite or aerial images) and related image analysis techniques and iii) empirical relationships to recover one variable according to other properties. In most studies, several methods are combined.

Here, only some indications on the available methods are presented because it is beyond the scope of the present review to fully describe such techniques. Some details can be found
<table>
<thead>
<tr>
<th>Variables</th>
<th>Description</th>
<th>Spatial representation</th>
<th>Access</th>
</tr>
</thead>
<tbody>
<tr>
<td>Number</td>
<td>number of reservoirs in catchment</td>
<td>D/C/S/A</td>
<td>DB/RS/Map</td>
</tr>
<tr>
<td>Location</td>
<td>geographical coordinates</td>
<td>D/C</td>
<td>DB/RS/Map</td>
</tr>
<tr>
<td>River flows</td>
<td>observed discharge at some places of the area under study (m³/s)</td>
<td>D/C/S/A</td>
<td>DB</td>
</tr>
<tr>
<td>Maximum area</td>
<td>area of the free surface water (m²)</td>
<td>D/C/S/A</td>
<td>DB/RS/Map</td>
</tr>
<tr>
<td>Drainage area</td>
<td>upstream basin whose runoff may feed the reservoir(s)</td>
<td>D/C/S/A</td>
<td>DB/ER</td>
</tr>
<tr>
<td>Storage capacity</td>
<td>maximum capacity volume of the reservoir(s) (m³)</td>
<td>D/C/S/A</td>
<td>DB/RS/ER</td>
</tr>
<tr>
<td>Abstraction</td>
<td>volume and timing of water uptake in the reservoir(s)(m³/period)</td>
<td>D/C/S</td>
<td>DB/ER</td>
</tr>
<tr>
<td>River connection</td>
<td>hillslope, across the river course, in diversion</td>
<td>D/S</td>
<td>DB/RS/Map</td>
</tr>
<tr>
<td>Bathymetry</td>
<td>relations between height-water volume-water free water surface area (m)</td>
<td>D/S</td>
<td>DB/RS/ER</td>
</tr>
<tr>
<td>Age</td>
<td>time since building the reservoir (year)</td>
<td>D/S</td>
<td>DB/RS/Map</td>
</tr>
</tbody>
</table>

Table 2: Key variables needed to conduct a cumulative impact study of small reservoirs from the most common (top) to the less used (bottom). Spatialization can be either D: distributed, S: statistical, C: catchment or A: aggregated (see Figure 6). Access to the variables can be from DB: databases, RS: remote sensing (satellite data, aerial images, lidar and so forth), Map: mapping, or ER: empirical relationships (see subsection below). Variables in brown are associated with the management of the reservoir discussed in Section 5.2, whereas the other ones are discussed in Section 5.3.
Field surveys are not often described in the literature because they are quite basic. However, field surveys represent a guaranteed method to locate all the reservoirs on a catchment and to ensure their type of connection to the river. However, this method is time consuming and cannot be used on large areas. The detection of reservoirs is efficient with remote sensing methods based on aerial or satellite images, which allows retrieving both the number and areas of the reservoirs (Chao et al., 2008; Messager et al., 2016). However, very small reservoirs (approximately 100 m²) are still difficult to detect, even with high-resolution aerial images (Carvajal et al., 2014).

Storage volume and bathymetry are more difficult to assess by remote sensing (Gal et al., 2016), whereas uncertainty in the storage volume can lead to important error in impact studies (Hughes and Mantel, 2010; Fowler et al., 2015). Thus, some empirical relationships are most often used. Based on a geometrical analysis of a variety of reservoir shapes, Molle (1991) showed that the relations between the reservoir surface and volume correspond to power laws. The parameters of the laws vary in space, depending on the geomorphological context, but remain generally constant within a given region (Thompson, 2012). Consequently, a common approach is to fit the law parameters from a set of reference reservoirs. The law can then be applied to all reservoirs in the catchment (Malveira et al., 2012; Hughes and Mantel, 2010).

The drainage area of the reservoirs can be derived from digital terrain models. However, this requires having a precise position of the reservoirs to be able to connect them with the correct river reaches to avoid error in the estimation of the upstream drainage area (Hughes and Mantel, 2010). Moreover, the determination of the type of connection between the reservoir and the river is a key point for assessing how the reservoir is filled. For modelling approaches that are not fully distributed, it is possible to use some relationship between the free surface water area (or volume) and the drainage area of the reservoir. Linear (Habets et al., 2014; Nathan et al., 2005) or non-linear (Fowler et al., 2015) relationships have been used. However, these relationships are again often specific to the studied catchment and cannot be generalised to very different contexts.

5.3. Water reservoir management characteristics

Water reservoir management operations refer to how the volume is stored in the reservoir and released from the reservoir either downstream, outflow, or withdrawn for some usage (most often, agricultural use). The type of reservoir-stream connection is an important driver for such
management, as shown in Section 3.3. Information on the connection can be included in some databases managed by stakeholders or regional agencies, particularly where legal regulations exist, for instance, to maintain a conservation flow. However, as stated previously, such databases are often incomplete. Hughes and Mantel (2010) show that it is difficult to obtain this information from remote sensing. Covering all the small reservoirs with a field survey is also difficult; such information is thus likely to be incomplete. This is perhaps the reason why most existing studies do not consider the ability to disconnect the small reservoirs from the stream network or to maintain some minimum flow by some type of diversion canal or low-flow bypass. Some exceptions are the works of Fowler et al. (2009) and Thompson (2012) that considered low-flow bypasses and of Habets et al. (2014) that considered the possibility to disconnect the reservoirs during part of the year (as if they were in diversion) to manage a filling period as required by the regional regulation. However, a limitation is that in these cases, the management operations were supposed to be homogeneous within the basin.

Water abstraction is the most sensitive information needed to infer the cumulative impacts of small reservoirs on hydrology (Hughes and Mantel, 2010; Fowler et al., 2015). However, the abstraction is rarely known, and at best, only an annual estimation of the abstracted water volume is known. To retrieve the temporal evolution of the water abstraction, which of course varies from year to year, several methods are used in the literature, either based on the estimation of the water demand or on the water offer (i.e. the available water volume stored in the reservoirs).

Water demand approaches attempt to quantify the needs associated with irrigating crops and watering livestock. Consumption for watering livestock is considered to be constant throughout the year (Fowe et al., 2015), whereas irrigation is estimated according to the sub-seasonal climate conditions. The water demand of the crop is often calculated on the basis of the crop coefficient Kc, which varies over time, and potential evapotranspiration (PET) (Fernández et al., 2007; Wisser et al., 2010; Biemans et al., 2011; Fowe et al., 2015).

Water offer approaches consider that the abstraction accounts for a given fraction of the total reservoir capacity. This approach is mainly used in Australia (Nathan et al., 2005; Cetin et al., 2009; Fowler et al., 2015). The fraction of the total storage can be obtained through surveys of reservoir owners or occasionally by remote detection (Fowler et al., 2015) and is highly variable depending on usage (irrigation vs. watering livestock) and region. Nathan and Lowe (2012) refers to fractions ranging from 10% to 400%, which implies that the reservoir can be filled
several times within a year. Although rather simple, this method allows considering a seasonal
distribution of the abstraction according to known uses (Cetin et al., 2009). This method can
also be used when no information on the abstractions is available simply by assuming that the
abstraction volume is a given fraction of the storage capacity (Habets et al., 2014; Deitch et al.,
2013).

6. Discussion

6.1. The uncertainty issue

Regardless of the approach (exclusively data-based method or modelling approaches), stream
flow is a crucial variable in any reservoir impact estimation and may be a source of uncertainty
in cumulative impact estimation. The uncertainty arises from uncertain measurements of stream
flow, including the need to transpose data from neighbouring catchments, as well as from time
series that are too short. It can lead to incorrect conclusions in trend analysis within statisti-
cal analyses of time series (Section 4.1.2) and in comparisons of paired-catchment hydrology
(Section 4.1.3).

In modelling approaches, when catchment models are used to simulate inflow to reservoirs
and transfer of reservoir outflow to the outlet, uncertainties in cumulative impact simulations
derive from uncertainties classically associated with catchment hydrologic models, namely, the
model itself (structure and parameters) and the data used to calibrate and validate the model. An
extensive presentation and discussion of these sources of uncertainty are beyond the scope of
the present review and can be found elsewhere (see, for instance, Hingray et al. (2009)). When
observed discharge is used rather than hydrologic catchment models, as in Deitch’s model, in
TEDI or in CHEAT, the simplifications performed to spatialize observed discharge as reservoir
inflow may result in strong errors in reservoir dynamics, in outflow simulation and thus in cu-
mulative impact estimation. The assumption used to aggregate reservoir outflow may also be
another source of uncertainty. To our knowledge, no sensitivity and uncertainty analyses of the
simplifications and assumptions have been performed.

How the reservoirs are accounted for in the models, together with how the hydrological
processes are estimated, are key components of the models. Incorrect representations may lead
to significant uncertainty in the estimation of cumulative impacts. Indeed, processes and factors
that affect reservoir water balance (Section 3.1) and thus cumulative impacts (Section 4.2) are
numerous. In the approaches for quantifying cumulative impacts, choices are made irrespective of the key processes and their representation; seepage, for instance, is often neglected (Table 1). The reservoir network representations (Table 1) in models also vary from one approach to another. The physical, topographic and management characteristics of reservoirs (Table 2) may also have uncertainties due to a lack of information or measurement and survey errors. The uncertainty in the estimation of cumulative impacts is thus a key issue.

A few modelling studies have addressed this issue by conducting sensitivity analyses (Habets et al., 2014; Hughes and Mantel, 2010; Malveira et al., 2012; Nathan et al., 2005). Although incomplete, three preliminary results can be emphasized. a) The effect of the uncertainty on the estimated upstream drainage area of reservoirs on inflow is controversial. On the one hand, it was shown to be a key morphological characteristic. This would have to be expected as the larger the upstream drainage area is, the larger the flow intercepted by reservoirs (Habets et al., 2014; Hughes and Mantel, 2010). On the other hand, the stream flow was shown to not be very sensitive to the reservoir drainage area (Nathan et al., 2005). The hydrologic characteristics (annual flow, monthly flow, and flow duration curves) taken into consideration to evaluate the cumulative impacts may explain the differences between these findings. b) Water management of reservoirs appears to play a dominant role in stream flow reduction. This was clearly shown by Hughes and Mantel, quantifying the key role of water demand uncertainty confirmed by Güntner et al. (2004), stating that "local experience suggests that uncertainty in human withdrawal add the largest uncertainty". c) Nathan et al. (2005) found that for the studied Australian catchments, the spatial representation of reservoirs, especially the topology and the cascading between reservoirs, does not exert a great role on stream flow reduction within the range of reservoir distribution.

From these preliminary conclusions, we highlight in the two following sections the need and the ways to improve knowledge of reservoir characteristics and estimate water abstraction from reservoirs. Uncertainty derived from process representations also deserves a thorough analysis, particularly how reservoir evaporation is quantified and the consequence of neglecting seepage in most of the approaches. It is expected that the sensitivity and uncertainty propagation may be different as functions of the hydrologic characteristics used to assess the cumulative impacts.

6.2. Improving knowledge of small reservoir characteristics

Estimating the cumulative impacts of small reservoir networks requires obtaining the key physical and geometrical characteristics of networks and reservoirs (Table 2). Unlike large reser-
voirs, the knowledge of the characteristics constitutes a real and specific challenge in consideration of the large number of small reservoirs within a catchment, up to nearly 10 /km$^2$ in some regions (Figure 2). This review shows that a variety of methods, ranging from field surveys to remote sensing, are available. However, uncertainty in the estimation of characteristics can be large and constitutes a difficulty specific to small reservoirs. One way to address this challenge is to choose methods for impact estimation that are minimally sensitive to the lack of information or uncertainty in small reservoir properties. This choice is made, for instance, in the global and statistical representations of reservoir networks used in some modelling approaches. Global indicators, as we investigated in this review, are also a way to overcome a lack of or uncertainty in information about the key characteristics of small reservoirs. However, the development of remote sensing methods and image analysis techniques should help in the future to map and quantify the properties over vast areas while reducing the uncertainties (Zhang et al., 2012). Following this approach, remote sensing may also be a way to derive height-surface area-volume relations (Mialhe et al., 2008). To date, such relationships established in a given region were used for all the reservoirs, while relations may vary from one reservoir to another. The synthesis by Carluer et al. (2016) found that operational studies collect a wealth of data on small dam network properties, data that were rarely used beyond the studies. Therefore, along with improvements in survey and remote sensing methods, one track to improve our capability of estimating small reservoir cumulative impacts also relies on storing and sharing information collected through operational surveys and scientific studies.

6.3. Improving abstraction estimations

When small reservoirs are intended to provide water for agricultural uses (irrigating crops and watering livestock), abstraction is a key parameter in hydrologic reservoir dynamics and in cumulative impacts (Hughes and Mantel, 2010; Nathan and Lowe, 2012). However, the present review shows that current estimations rely on very pragmatic choices and simple methods because existing, readily available information about abstraction is very difficult to obtain in every country. Water abstraction may vary broadly from one reservoir to another. Abstraction rate and timing from a given reservoir result from a complex process including biophysical considerations: crop or livestock demands, availability of reservoir water and also of other water resources (river and groundwater). Social and economic considerations are also at stake: water abstraction resulting from an agronomic strategy developed by farmers, involving crop yield and profit tar-
gets, also related to water resource sharing between water users. Abstractions at least depend on
laws or regulations fixing water use restrictions and downstream water release rates and timings
from reservoirs for other water functions.

Two ways may enhance abstraction estimations. In many countries, farmers have to declare
to water management agencies or state services the abstraction volumes and occasionally the
timing from their own reservoirs. Storing this information through database systems and making
it available would allow obtaining a precise estimation of where and when water is withdrawn
from reservoirs. Empirical relations relating the characteristics of reservoirs with crop or animal
needs could be one way to estimate and spatialize the water abstraction from small reservoirs
more accurately than the current simple and pragmatic methods. Another way would be to take
advantage of the agronomic state of the art in terms of crop management strategies. Decision rule
models are available to simulate and predict tillage, sowing, fertilization, hoeing, irrigation, crop
protection, and harvesting periods. Such models could be coupled to hydrologic models, allow-
ing estimating the impacts of agricultural land use strategy in a reservoir-equipped catchment on
stream flow and other water compartments. As an example, the MAELIA platform proposes a
framework to couple such crop models and decisional models with the SWAT hydrologic model
(Thérond et al., 2014).

6.4. Impact indicators

Simple indicators of cumulative impacts are needed by stakeholders and water management
actors. The challenge is the design of the reservoir system and particularly the identification of
sensitive areas where no other reservoir should be built, and even where some reservoirs should
be removed, while other areas could benefit from the construction of new reservoirs to increase
the available water resource. From a scientific perspective, this operational need consists of first
analysing whether cumulative impacts can be derived from properties of reservoir networks or
others. Our analysis shows that there is no relationship between the hydrological impact rates
and some simple network density indicator (Figure 2). The analysis was performed based on
data collected from worldwide studies involving a large range of hydrological, climatic, geolo-
 logical, pedological, and land use contexts. Catchment hydrological functioning, particularly
runoff temporal and spatial variability, must be a key factor in the impact process, although indi-
cators only based on reservoir properties do not account for. This point is clearly supported
by the variability of impacts for a given catchment depending on wet and dry years (Figure 1).
Furthermore, reservoir management (abstraction rate and time, outflow by water release, including minimal outflow when relevant and connection to the stream) is also another key factor in the impacts. The large number of factors involved in the cumulative impacts makes the search for a universal indicator a never-ending quest. Instead, one research track would be to develop regional indicators based on regional analysis of the cumulative impacts. Within areas of homogeneous hydrology, soil occupation, and standardized water management operations, indicators of reservoir network properties may be more relevant than at the global scale. Following this approach, Hughes and Mantel (2010) proposed and explored for a few catchments the relevancy of an indicator integrating the annual water demand for small reservoirs, a measure of stream flow temporal variability and the mean contributing area of reservoirs. They found a correlation between the indicator and the annual mean flow decrease. Another important point would also be to differentiate between exploited and non-exploited reservoirs, considering the role of reservoir management on the impacts.

7. Conclusion

In this study, we investigated the cumulative impacts of small reservoirs on water resources from a quantitative aspect only. Although the reviewed studies agree that the main impacts of small reservoirs are a decrease in the river discharges and peak flow due to water abstraction from the reservoirs and water losses, the intensity of this decrease can vary considerably and is not easy to anticipate with various types of indicators. Impacts on low flow and river regime can vary from basin to basin due to the many types of reservoirs and their different uses. It was shown that a key issue with studying the cumulative impacts of small reservoirs is the lack of data on the properties and usage of the small reservoirs, which leads the various studies to adapt their strategy to address this ill-defined problem by using assumptions to simplify the estimation of these characteristics.

However, this review focused only on some aspects of the impacts of small reservoirs. Indeed, the numerous small reservoirs also impact sediment transfer, hydromorphology, biodiversity, and biochemistry. Although the literature on such topics associated with small reservoirs is not vast, these aspects were reviewed by Carluer et al. (2016). From this review, it appears that a fine spatial and temporal estimation of the hydrological impact may be required to assess these other impacts. The lack of data on some characteristics of the small reservoirs is also challenging. Even
with such difficulties, it is assumed that small reservoirs have a large impact on sediment trapping (Yang et al., 2011) and river channel (Petts and Gurnell, 2005). The impacts on some biochemical components can accumulate according to the discontinuity distance (Bergkamp et al., 2000). The impacts on biodiversity (especially fishes) from large reservoirs are rather well known (Poff and Zimmerman, 2010); thus, the question is now how to reduce the major impacts by removing the most impacting reservoirs (Poff and Hart, 2002; Doyle et al., 2005; Grantham et al., 2014). There is no doubt that the question of removing small reservoirs should also be extended to attempt to reduce the other types of impacts, including quantitative hydrological impacts.

Socioeconomic impacts are also very important to consider since it is often the key driver to build reservoirs. It was shown in India that large reservoirs can have some drawbacks for the neighbouring population (Duflo and Pande, 2007). However, this impact can be reduced by the presence of small reservoirs that are having positive socioeconomic impacts on the local population (Blanc and Strobl, 2013; Acheampong et al., 2014). Lasage et al. (2015), for instance, focus on the social benefit of small sand reservoirs to secure water access in the context of climate change.

Indeed, it is rather important to consider the long-lasting life of the reservoir (more than 50 years) since this means an impact in the long term, but also within a changing climate. As stated in the introduction, there is increasing pressure to build reservoirs, partly to adapt to climate change. The global impacts of small reservoirs on hydrology are already estimated to be 5% of the mean discharge and 44% of the low flow (Wisser et al., 2010), although the impacts can vary in space and season (Wanders and Wada, 2015). Moreover, there is an increasing number of studies that show that water management can aggravate the duration of droughts, particularly where the development of water use was not controlled and for longer droughts (Van Loon et al., 2016; He et al., 2017; Lin et al., 2017). It is thus important to integrate in new projects the cumulative impacts of the reservoir network in the basin, as well as its ability to evolve in time according to the hydrologic conditions due to global change.

Acknowledgments

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Appendix A.
<table>
<thead>
<tr>
<th>Ref</th>
<th>Basin</th>
<th>A</th>
<th>LU</th>
<th>Dam</th>
<th>P</th>
<th>PET</th>
<th>Q</th>
<th>VD</th>
<th>ND</th>
<th>M</th>
<th>Impact</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>CHECK DAMS</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Martinez Alvarez et al. (2007)</td>
<td>ESP, Sierra Gador</td>
<td>320</td>
<td>Low vegetation</td>
<td>CD</td>
<td>400</td>
<td>900</td>
<td>9.3</td>
<td>0.6</td>
<td>0.3</td>
<td>A</td>
<td>Increase groundwater recharge</td>
</tr>
<tr>
<td>Xu et al. (2013)</td>
<td>CHN, Yanhe</td>
<td>7,725</td>
<td>Grassland</td>
<td>CD</td>
<td>505</td>
<td></td>
<td></td>
<td>0.8</td>
<td></td>
<td>OBS</td>
<td>Erosion impact</td>
</tr>
<tr>
<td>Ayalew et al. (2017)</td>
<td>USA, Soap Ck</td>
<td>660</td>
<td>FL, LS, FP</td>
<td></td>
<td>62</td>
<td>0.2</td>
<td>D</td>
<td></td>
<td></td>
<td></td>
<td>Reduce peak flow by 20% to 70%</td>
</tr>
<tr>
<td>Frickel (1972)</td>
<td>USA, Willow Ck</td>
<td>1,400</td>
<td>Grazing, conservation, farming</td>
<td>FL</td>
<td>320</td>
<td></td>
<td>42</td>
<td>0.1</td>
<td></td>
<td>OBS</td>
<td>-18% AD + reduction of peak discharge</td>
</tr>
<tr>
<td>Kemmon (1966)</td>
<td>USA, Sandstone Ck</td>
<td>221</td>
<td>FL, LS</td>
<td>635</td>
<td></td>
<td>76</td>
<td>120</td>
<td>0.1</td>
<td></td>
<td>OBS</td>
<td>-12% AD</td>
</tr>
<tr>
<td>Ramírez et al. (2000)</td>
<td>USA, Wet Walnut Ck</td>
<td>4,100</td>
<td>FL, LS</td>
<td>510</td>
<td></td>
<td>108</td>
<td></td>
<td></td>
<td></td>
<td>A</td>
<td>Decrease in runoff and piezometric level</td>
</tr>
<tr>
<td><strong>FLOOD CONTROL DAMS</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Carvajal et al. (2014)</td>
<td>ESP, Almería</td>
<td>7</td>
<td>I</td>
<td>250</td>
<td></td>
<td>68</td>
<td>39.5</td>
<td>WB</td>
<td></td>
<td></td>
<td>Collecting rainwater from the roofs of the greenhouses and covering small dams could reduce external water needs by 53%.</td>
</tr>
<tr>
<td>Martínez Alvarez et al. (2008)</td>
<td>ESP, Segura</td>
<td>3,774</td>
<td>I</td>
<td>1850</td>
<td>78</td>
<td>3.7</td>
<td>WB</td>
<td></td>
<td></td>
<td>Regional evaporation losses of small dams represents 27% of the domestic water use in a 2 million inhabitants region</td>
<td></td>
</tr>
<tr>
<td>Withner and Girard (1973)</td>
<td>BRA, Sítia</td>
<td>1,790</td>
<td>I</td>
<td>700</td>
<td>2,100</td>
<td>72</td>
<td>34</td>
<td>OBS</td>
<td></td>
<td>-11% to -24% AD</td>
<td></td>
</tr>
<tr>
<td>Cueto et al. (2005)</td>
<td>FRA, Séoule</td>
<td>463</td>
<td>I</td>
<td></td>
<td>14</td>
<td>0.3</td>
<td>OBS</td>
<td>stat</td>
<td></td>
<td></td>
<td>Water flow reduced by 31%</td>
</tr>
<tr>
<td>Schreider et al. (2002)</td>
<td>AUS, Yass R.</td>
<td>388</td>
<td>LLS</td>
<td></td>
<td>53.1</td>
<td>17</td>
<td>5.8</td>
<td></td>
<td></td>
<td>-9.5% AD + 1.5M increase in farm dam storage corresponded to a 2 to 3 Ml decrease in streamflow annual yield</td>
<td></td>
</tr>
<tr>
<td>Thompson (2012)</td>
<td>NZL, upper Tupiko R</td>
<td>0.7</td>
<td>LS</td>
<td>656.5</td>
<td>653</td>
<td></td>
<td>216</td>
<td>16.6</td>
<td>4.3</td>
<td>OBS</td>
<td>Pair -80% AD + change in flow regime</td>
</tr>
</tbody>
</table>

Table A.1 continues on next page
<table>
<thead>
<tr>
<th>Hughes and Mantel (2010)</th>
<th>ZAF, H10A Bread R.</th>
<th>Deciduous fruit orchards</th>
<th>LLS</th>
<th>234</th>
<th></th>
<th>500 to 1 000</th>
<th>1 650</th>
<th>154</th>
<th>0.1</th>
<th>0.6</th>
<th>A</th>
<th>-17% AD</th>
</tr>
</thead>
<tbody>
<tr>
<td>ZAF, H10B Bread R.</td>
<td>Deciduous fruit orchards</td>
<td>LLS</td>
<td>162</td>
<td>273</td>
<td>0.1</td>
<td>0.5</td>
<td>A</td>
<td>-21.5% AD</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>ZAF, H10C Bread R.</td>
<td>Deciduous fruit orchards</td>
<td>LLS</td>
<td>260</td>
<td>564</td>
<td>0.1</td>
<td>0.8</td>
<td>A</td>
<td>-27% AD</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>ZAF, H10D Bread R.</td>
<td>Deciduous fruit orchards</td>
<td>LLS</td>
<td>97</td>
<td>2 054</td>
<td>A</td>
<td>-18% AD</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>ZAF, X21F</td>
<td>Stock grazing</td>
<td>LLS, LS</td>
<td>397</td>
<td>760</td>
<td>1 400</td>
<td>109</td>
<td>0.02</td>
<td>0.5</td>
<td>A</td>
<td>-16% AD</td>
<td></td>
<td></td>
</tr>
<tr>
<td>ZAF, D52A</td>
<td>Stock grazing</td>
<td>L, LS</td>
<td>378</td>
<td>320</td>
<td>1 900</td>
<td>13</td>
<td>0.02</td>
<td>0.1</td>
<td>A</td>
<td>-35%</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

| Perrin et al. (2012)    | IND, Gavel        | Semi-arid scrubland, rainfed crop, irrigated rice | I | 812 | 1 800 | A | Evaporation loss is dominant. Tank infiltration represents 43% of the groundwater recharge on average, 54% AD during dry year and 32% AD during wet year |
|-------------------------|-------------------|-------------------------------------------------|---|-----|-------|-----|-----------|--------|--------|--------|---------|
| Tambor and Schulz (1991)| ZAF, Midmar      |                                                  | 912 | 952 | 110 | A | -6% AD |
| Tabeti et al. (2014)    | FRA, Layon        | Maize, vineyards, I | 930 | 660 | 1 475 | 1.5 | 1.4 | Ag | -9% AD |
| Meigh (1995)            | BWA, Garo-bone   |                                                  | 5 983 | 7.6 | 6.5 | 0.05 | As | -25% AD |
|                         | BWA, Bokau       |                                                  | 5 70 | 3.7 | 1  | 0.03 | As | -13% AD |
|                         | BWA, Shakte      |                                                  | 3 650 | 29  | 0.04 | 0.004 | As | -0.2% AD |
| O’Connor (2001)         | ZAF, Limpopo R.  | Riparian woodland                                | 1 992 | 377 | 2 050 | 3.5 | 0.04 | Anat | The many small farm dams reduce flow during critical dry years to levels causing dieback of some vegetation |

| Cetin et al. (2009)     | AUS, Campaspe R. |                                                  | 4 000 | 1 350 | 74 | 11 | 3.2 | S | 7% |
| Fowler et al. (2015)    | AUS, Stringybark Ck |                                                  | 73  | 1 650 | 116 | 43.2 | 7.7 | S | -22% AD |
| Günther et al. (2004)   | BRA, Upper Jaguaribe | Woodland, cattle farming crops (bean, maize)     | 24 200 | 700 | 2 300 | 285 | 0.2 | S | -21% AD |
| Nathan et al. (2005)    | AUS, Avoca R.     |                                                  | 77  | 580 | 70 | 6.7 | 2.8 | S | The natural flow is closed to the observed one and there is small |

Table A.1 continues on next page
The natural flow is underestimated by the statistical model compared to the distributed model, which lead to underestimation of the impact of small dams by a factor of 2.
the watercatchment modelling framework. Modelling and Simulation Society of Australia and New Zealand and International Association for Mathematics and Computers in Simulation, Cairns, Australia, pp 3478–3484


Craig IP (2008) Loss of storage water through evaporation with particular reference to arid and semi-arid zone pastoralism in australia. Report, NCEA publication, University of Southern Queensland, Australia


Grantham TE, Viers JH, Moyle PB (2014) Systematic screening of dams for environmental flow assessment and imple-