Regional scale hydrodynamic modeling of the river-floodplain-reservoir continuum

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Abstract

River floodplains and reservoirs interact throughout a basin drainage network, defining a coupled human-water system with multiple feedbacks. Recent modeling developments have aimed to improve the representation of such processes at regional to continental scales. However, most large-scale hydrological models adopt simplified lumped reservoir schemes, where an offline routine is run with inflows estimated by the model, with limited consideration of the complementarity between floodplains and reservoirs on attenuating floods at regional scale. This paper presents a novel approach that fully couples river-floodplainreservoir hydrodynamic and hydrological models, significantly improving the representation of reservoir dynamics and operation in the river-flood plain-reservoir continuum at large scale and across multiple dam cascades. The model is applied to the Paraná River Basin with explicit simulation of 31 large dams and river hydraulic variables at basin scale. Three types of reservoir bathymetry representation are compared, from lumped to distributed methods, combined with three reservoir operation schemes and varying degrees of input data requirement within two parameterization scenarios (global and regional setups). The operation schemes were more relevant than the reservoir bathymetry representation to estimate downstream flows and water levels. While the data-driven operation scheme, based on linear regressions between observed water levels and dam outflows, provided the best estimates of both active storage and discharges, the more generic operation reasonably estimated discharges and peak attenuation, albeit not as accurately for active storage. The global parameterization of reservoir operation resulted in poorer performance compared to the regional-based one, but it satisfactorily modeled discharge and peak attenuation. Regarding the reservoir bathymetry representation, a basin scale comparison of the lumped and distributed schemes indicated the inability of the former to represent backwater effects. This was further corroborated by validating the longitudinal water level profile of Itaipu dam with ICES at satellite altimetry data. Finally, the model was used to show the complementarity between floodplains and reservoirs on attenuating floods at regional scale. Large scale models should move beyond offline coupling strategies, and include regional-based, data-driven reservoir operation schemes together with a distributed representation of reservoir bathymetry into river-floodplain hydraulic schemes. This will largely improve the estimation of river discharges, water levels and flood storage, and thus the model ability to represent the regional scale river-floodplain-reservoir continuum.

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22 Abstract

23 River floodplains and reservoirs interact throughout a basin drainage network, defining 24 a coupled human-water system with multiple feedbacks. Recent modeling developments 25 have aimed to improve the representation of such processes at regional to continental 26 scales. However, most large-scale hydrological models adopt simplified lumped 27 reservoir schemes, where an offline routine is run with inflows estimated by the model, 28 with limited consideration of the complementarity between floodplains and reservoirs 29 on attenuating floods at regional scale. This paper presents a novel approach that fully 30 couples river-floodplain-reservoir hydrodynamic and hydrological models, significantly 31 improving the representation of reservoir dynamics and operation in the river-32 floodplain-reservoir continuum at large scale and across multiple dam cascades. The 33 model is applied to the Paraná River Basin with explicit simulation of 31 large dams 34 and river hydraulic variables at basin scale. Three types of reservoir bathymetry 35 representation are compared, from lumped to distributed methods, combined with three 36 reservoir operation schemes and varying degrees of input data requirement within two 37 parameterization scenarios (global and regional setups). The operation schemes were 38 more relevant than the reservoir bathymetry representation to estimate downstream 39 flows and water levels. While the data-driven operation scheme, based on linear 40 regressions between observed water levels and dam outflows, provided the best 41 estimates of both active storage and discharges, the more generic operation reasonably 42 estimated discharges and peak attenuation, albeit not as accurately for active storage. 43 The global parameterization of reservoir operation resulted in poorer performance 44 compared to the regional-based one, but it satisfactorily modeled discharge and peak 45 attenuation. Regarding the reservoir bathymetry representation, a basin scale 46 comparison of the lumped and distributed schemes indicated the inability of the former

47 to represent backwater effects. This was further corroborated by validating the 48 longitudinal water level profile of Itaipu dam with ICESat satellite altimetry data. 49 Finally, the model was used to show the complementarity between floodplains and 50 reservoirs on attenuating floods at regional scale. Large scale models should move 51 beyond offline coupling strategies, and include regional-based, data-driven reservoir 52 operation schemes together with a distributed representation of reservoir bathymetry 53 into river-floodplain hydraulic schemes. This will largely improve the estimation of 54 river discharges, water levels and flood storage, and thus the model ability to represent 55 the regional scale river-floodplain-reservoir continuum.

56

57 **1 Introduction**

58

59 Reservoirs are important infrastructure for energy production, flood control, flow 60 regulation and water supply, among other uses (Lehner et al., 2011). Their construction 61 and operation have also led to major socio-environmental concerns (Grill et al., 2019; 62 Nilsson, 2005; Poff et al., 2010; Richter and Thomas, 2007), and efforts to improve 63 storage allocation (Almeida et al., 2019; Ho et al., 2017; Schmitt et al., 2019). 64 Reservoirs, however, operate in basins with a continuum in the river system connecting 65 it to rivers and floodplains, through which human societies and ecosystems interact with 66 dynamic two-way feedbacks (Di Baldassarre et al., 2013; Pande and Sivapalan, 2017; 67 Viglione et al., 2014). At regional to continental scales, this river-floodplain-reservoir 68 continuum is associated to a complex relationship among surface water processes. For 69 example, in the La Plata River Basin in South America, dozens of large reservoirs have 70 been built since the 1950's interacting with complex wetland systems as the Pantanal,

71 Esteros del Iberá and Paraná floodplains (Minotti, 2018). In the basin, human society 72 has settled around floodplains for centuries, leading to a fully coupled human-water 73 system (Doyle and Barros, 2011; Lee et al., 2018). As the use and development of the 74 floodplain by society evolve, there is an increasing need to better understand the 75 hydrodynamic interactions in this river-floodplain-reservoir continuum, so we can better 76 design and operate water systems to cope with human and ecosystem demands 77 considering hydrological uncertainty, processes and current and future environmental 78 changes.

79 Regional to continental scale hydrological-hydrodynamic models provide a unique 80 opportunity to address these needs. While recent advances in large scale modeling have 81 improved our capability to simulate river floods at both 1D and 2D dimensions (Bates et 82 al., 2018; Fleischmann et al., 2020; Neal et al., 2012; Paiva et al., 2013; Schumann et 83 al., 2013; Trigg et al., 2016; Yamazaki et al., 2011), most studies on reservoir 84 simulation have focused on representing dam storage and operation (i.e., a water 85 management model) within simpler hydrological models, with less physically based 86 flow routing methods (Droppers et al., 2020; Hanasaki et al., 2018; Yassin et al., 2019). In the studies by Mateo et al. (2014) and Pokhrel et al. (2018), for instance, a 87 88 hydrodynamic model was run (offline) with observed or simulated dam outflows at the 89 grid cell related to the dam, in order to estimate alterations in downstream flooding. 90 Difficulties for detailed reservoir simulation included the unknown bathymetry and 91 specific dam operation at very large scales, and are also a challenge.

92 Only recently the hydrodynamics of dam cascades were explicitly included into 93 regional hydrodynamic models (Shin et al., 2019; Fleischmann et al., 2019a), aiming at 94 representing the river-floodplain-reservoir system with a fully-coupled approach that 95 allowed a distributed representation of variables as discharges, water levels and flood

96 extent and storage in human-altered systems. As the representation of these processes 97 improves, a better and integrated assessment of basin water resources and floods 98 becomes possible. Examples include the understanding of the relative role of 99 floodplains and reservoirs on flood attenuation (Fleischmann et al., 2019a), more 100 detailed simulation of evaporation (Shin et al., 2019), and understanding of the reservoir 101 influence on local climate (Degu et al., 2011; Hossain et al., 2012). The representation 102 of the reservoir dynamics itself and associated backwater effects and flooding in 103 upstream areas, and simulation of carbon cycle and phytoplankton dynamics (e.g., lake 104 emissions and degassing or downstream emissions; Bierkens et al. (2015)), can also 105 benefit from such modeling systems. Ultimately, these tools will provide an important 106 basis towards a fully coupled and distributed human-water modeling system within 107 hyperresolution Earth system models (Nazemi & Wheater, 2015; Pokhrel et al., 2016; 108 Wood et al., 2011), adopting detailed grids and daily temporal resolution (Gutenson et 109 al., 2020; Zajac et al., 2017).

110 Regarding the representation of dam operation in regional to global models, there 111 have been major improvements since pioneering studies by Hanasaki et al. (2006) and 112 Haddeland et al. (2006), which have been used and adapted for many studies (Adam et 113 al., 2007; van Beek et al., 2011; Biemans et al., 2011; Pokhrel et al., 2012; Shin et al., 114 2019; Wisser et al., 2010). Since these first generic algorithms, data-driven reservoir 115 operation schemes are now feasible, while optimization methods have also been 116 developed, involving storage, outflow and inflow observations (Solander et al., 2016; 117 Wu and Chen, 2012; Yassin et al., 2019), and downstream water or energy demands 118 (Haddeland et al., 2006). All these developments highlight the ongoing necessity to 119 better estimate actual reservoir operation in order to achieve hyperresolution models 120 that are locally relevant. Cross-scale comparisons among different approaches, from

simpler, globally-based, to more complex, regionally-derived setups, can yield
meaningful insights on the ways forward, especially using regionally set up models
(Fleischmann et al., 2019c; Nazemi & Wheater, 2015; Trigg et al., 2016).

124 The need for a fully-coupled approach was explicitly highlighted by some authors 125 (Fleischmann et al., 2019a; Shin et al., 2019). However, the benefits of representing the 126 reservoir dynamics fully coupled within the river-floodplain-reservoir continuum 127 processes with a distributed approach, over the traditional lumped and offline 128 representation, remains a knowledge gap in the field for large scale models. The extent 129 to which simple, level-pool reservoir simulations (i.e., lumped) may lead to similar 130 results as distributed (i.e., dynamic), more complex ones, is not yet understood. Finally, 131 a correct reservoir operation also needs to be incorporated to improve the understanding 132 of the human feedback in the river-floodplain-reservoir continuum.

133 This study brings a novel contribution to these gaps with an improved 134 representation of the river-floodplain-reservoir interactions within hydrologic-135 hydrodynamic models, followed by a broad analysis of the continuum of hydraulic 136 variables basin-wide considering the reservoir operation effects. The contributions of 137 this study address three main research questions: (i) what are the differences between 138 simulating lumped and distributed reservoir bathymetry in coarse-scale, online coupled 139 hydrologic-hydrodynamic models, in terms of different variables as water levels, 140 discharge, flood extent and storage, and evapotranspiration? (ii) how do generic and 141 more data-driven reservoir operation schemes differ in terms of hydrological variables 142 estimation in regional scale models? And (iii) what is the relative role of floodplains 143 and reservoirs on flood attenuation in large-scale anthropized systems, considering 144 basin-wide hydrological processes? To answer these questions, new modeling 145 approaches to improve reservoir representation and operation are proposed and tested in

146	a ~950,000 km ² watershed (Upper Paraná River Basin, Brazil). Different simulation
147	scenarios are performed to assess the dynamics of reservoirs in terms of complexity of
148	bathymetry representation (lumped to distributed) and reservoir operation (from generic
149	to data-driven approaches, and from globally to regionally derived parameterizations).
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151	2 Methods
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153	2.1 Study area: The Upper Paraná River Basin
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155	The Upper Paraná River Basin (Figure 1) was selected as a study area given (i) its
156	large number of dam cascades (in parallel and in series); (ii) the existence of large
157	floodplains both upstream and downstream of dam cascades; and (iii) the availability of
158	observed daily time series of dam inflows, outflows and storage from the Brazilian
159	National Water Agency (ANA). Those are desirable characteristics to address in large
160	scale modeling of river-floodplain-reservoir systems.
161	The Paraná River is formed by the confluence of Grande and Paranaíba rivers in

162 Brazil, with major tributaries being the Tietê, Paranapanema, Ivaí and Iguaçu rivers, all 163 in its left margin. The Upper Paraná River Basin has a drainage area of ~950,000 km² 164 and it is among those with the largest hydropower installed capacity in the world (and 165 almost 50% of the installed hydropower capacity of Brazil), including the Itaipu dam 166 (ID 18 in Figure 1) which is one of the largest dams in the world (Itaipu, 2018). Large 167 cities as São Paulo and Brasília (Brazil Federal Capital) are located within the basin, 168 which holds a population of nearly 70 million people. The wet period usually occurs from November/January to May/June (Agostinho et al., 2000), with average annual 169

rainfall of about 1400 mm (Boulanger et al., 2005). There are contrasting hydroclimatic
regions throughout the basin, with the northern (southern) regions presenting a seasonal
(non-seasonal) precipitation regime.

173 The basin has 86 large dams (> 30 MW) in operation, with an installed capacity of 174 48,083 MW (ANEEL, 2020) (Figure 1). There are also 500 (58) small (large) proposed, planned or under construction dams, related to 5,643 (3,909) MW of installed capacity, 175 176 respectively (ANEEL, 2020). Sixty-two large dams are currently connected to the 177 Brazilian National Interconnected System (SIN), with 32 run-of-river and 30 flow 178 regulation dams (ONS, 2020), which are coordinately operated with other power 179 sources (e.g., thermal and wind) to generate and distribute energy to the whole system 180 minimizing costs (Marques & Tilmant, 2013). Most dams are also operated with 181 multiple uses, such as flood control, water supply and navigation. Overall, there is a 182 large hydrological alteration at the basin scale due to reservoir operation (Santos, 2015). 183 The extensive floodplains throughout the basin, as the 230 km reach in the Paraná 184 mainstem between Porto Primavera and Itaipu dams, harbor important ecosystem 185 services (Agostinho et al., 2001, 2008; Baumgartner et al., 2018).



Figure 1. The Upper Paraná River Basin within South America, and the 31 simulated (green circles) and other non-simulated Brazilian dams (red: large dams – installed capacity > 30 MW; grey: small dams – capacity between 3 and 30 MW). The total number of large dams is 86 (green and red dots). Rivers of interest are labeled.

192 2.2 Hydrological and hydrodynamic representation of the river-floodplain 193 reservoir continuum

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The MGB model ("Modelo de Grandes Bacias" in Portuguese, an acronym meaning "Model of Large Basins") (Collischonn et al., 2007; Pontes et al., 2017) is used to implement and test the proposed representation of the river-floodplain-reservoir continuum and reservoir operation. It is a semi-distributed, hydrological-hydrodynamic model developed to simulate large-scale basins. This model is chosen given its proven track record of simulation in several other river basins at different scales, from regional to continental domains (Siqueira et al., 2018). First, the original MGB modeling approach is presented, followed by the proposed improved representation of thereservoirs and their hydraulic and hydrodynamic relationships with the continuum.

204 In MGB's representation, the basin is divided into unit-catchments of equal river 205 lengths, and within each the model simulates vertical hydrological processes as 206 evapotranspiration, soil water infiltration and runoff generation (from surface, subsurface and groundwater reservoirs) (Figure 2). Local runoff is added as a lateral 207 208 boundary condition to the drainage network, and a hydrodynamic routing is performed 209 to simulate river, floodplains and reservoirs' surface water dynamics. Soil and 210 vegetation model parameters are defined for each Hydrologic Response Units (HRU's) 211 within a given sub-basin, and the HRU's are derived from a combination of soil and 212 vegetation maps. Evapotranspiration is computed with the Penman-Monteith equation 213 for soil/vegetated areas, and Penman equation for flooded areas (i.e., assumed as open 214 water). A dynamic two-way feedback between the hydrologic and hydrodynamic 215 modules is also considered, by which floodplain water can infiltrate into the unsaturated 216 soil, and evapotranspiration/open water evaporation and runoff generation are 217 dynamically computed considering the surface flooded fraction at each time step. More 218 details on the hydrological model are presented in Supplementary Material S2 and in 219 Collischonn et al. (2007), Pontes et al. (2017) and Siqueira et al. (2018). Recent MGB 220 applications in the Paraná basin with the simpler, Muskingum-Cunge flood routing 221 scheme were performed in Fleischmann et al., (2019b) and Quedi & Fan, (2020).



Figure 2. Overview of the MGB model structure: unit-catchment water and energy balance (middle panel), river-floodplain-reservoir hydraulic routing (upper right), and types of reservoir simulation (bottom panel).

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228 2.2.1 MGB model river-floodplain hydrodynamic routing

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The local inertia explicit method proposed by Bates et al. (2010) is adopted within MGB to simulate 1D flow propagation along the drainage network. This method is a simplification of Saint-Venant equations, neglecting the convective acceleration term from the momentum equation, which has been proven satisfactory to represent flood wave transport along rivers at both 1D and 2D dimensions (Getirana et al., 2017; Neal et al., 2012; Siqueira et al., 2018; Yamazaki et al., 2013). Within MGB, floodplains are represented as storage units, i.e., they are ineffective areas without active flow, river-floodplain water exchange is instantaneous, and water surface elevation is assumed the same along the river-floodplain system within a given unit-catchment (Paiva et al., 2011). Channel cross sections are assumed rectangular, as typically adopted in large scale hydraulic modeling (Paiva et al., 2013; Trigg et al., 2009).

The flux between two adjacent unit-catchments is computed with the discretizedmomentum Equation 1:

243
$$Q_{out,i}^{t+\Delta t} = \frac{Q_i^t - gB_i\Delta th_i S_i}{1 + \frac{g\Delta t |Q_i^t| n^2}{B_i(h_i)^{7/3}}}$$
Equation 1

+

Where $Q_{out,i}^{t+\Delta t}$ is the discharge at unit-catchment *i* at time $t + \Delta t$, *n* is the Manning's coefficient, h_i the flow depth between unit-catchments *i* and i + 1, S_i the water surface level slope, Δt the model time step, B_i the flow width, and *g* the gravitational acceleration.

248 The continuity equation can be approximated for each unit-catchment river reach249 as:

250
$$\frac{v_i^{t+\Delta t} - v_i^t}{\Delta t} = \sum Q_{in,i}^{t+\Delta t} - Q_{out,i}^{t+\Delta t} + Q_{local} + P_i - E_i$$
 Equation 2

251 Where *V* is the stored volume in unit-catchment i, $\sum Q_{in,i}^{t+\Delta t}$ the sum of inflows from 252 upstream unit-catchments, Q_{local} the locally generated runoff, *P* the precipitation over 253 flooded areas (i.e. river reach surface area plus flooded floodplain or reservoir area), and 254 *E* the flooded area open water evaporation computed with Penman equation. Once the unit-catchment volume is updated with (2), water level in the unitcatchment is estimated from its level-volume relationship (hypsometric curve). For stages below bank elevation, this is derived from the channel cross section. For stages above bank elevation, it represents the floodplain topography, and it is obtained with a GIS pre-processing step that computes flooded areas associated to increments in Height Above Nearest Drainage values (HAND; Rennó et al. (2008)) extracted from the SRTM Digital Elevation Model (DEM) (Siqueira et al., 2018).

Effective hydraulic parameters that are required for each river reach are channel bed elevation, cross section bankfull width and depth, and Manning roughness coefficient. Bed elevation is derived for each unit-catchment from the average DEM river network pixels (Siqueira et al., 2018) subtracted by bankfull depth. The hydrodynamic routing time step is determined by the Courant-Friedrichs-Levy condition with an additional multiplier parameter for ensuring model numerical stability (Bates et al., 2010; Yamazaki et al., 2011).

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270 2.3 Reservoir routing271

Two main aspects differ the proposed improved reservoir representation from the original MGB river-floodplain routing scheme. First, at the unit-catchment corresponding to the dam location, the momentum equation (1) is replaced by the dam outflow equation (i.e., it is set as an internal boundary condition), which is based on simple spillway or outlet works equations, or on more complex reservoir operation derived from actual dam operational data. 278 Second, the reservoir storage (and bathymetry) is represented in MGB by 279 adjusting the level-volume relationship in the unit-catchments located within the 280 reservoir lake, originally extracted from a DEM. If the dam did not exist during the 281 DEM acquisition date, the storage is already represented in the level-volume 282 relationships, and thus no correction is necessary. In such cases, it is only required to 283 define the dam outflow equation. On the other hand, if the reservoir already existed, the 284 DEM will likely miss the storage representation (depicting a flat lake area instead). This 285 demands additional bathymetry information (e.g., reservoir level-storage relationships) 286 to correct the model. These two main aspects were added to the MGB framework by 287 Fleischmann et al. (2019a). Open water evaporation and direct precipitation on lake are 288 considered in the same way as for floodplain areas.

In the improved reservoir representation proposed in this paper, this scheme is further developed by comparing different types of reservoir bathymetry representation (Section 2.4) and operation (Section 2.5), which are detailed in the next sections.

292

293 **2.4 Reservoir storage representation**

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To improve reservoir storage representation, three different types of reservoir storage/bathymetry representation are compared: (i) a lumped representation of the reservoir storage, by which all storage is concentrated in one only unit-catchment (associated to the dam location), and a distributed method in which the storage is (ii) equally and (iii) variably split among all unit-catchments that compound the reservoir lake, thus allowing the representation of reservoir dynamics. **Figure 2** (bottom panel) presents the schemes for the three simulation methods. 302 The lumped method (i) ("Lum") consists of concentrating the reservoir stage-303 volume curve (in this study, provided by the Brazilian National Electric System 304 Operator - ONS) on the unit-catchment holding the dam location. This method is 305 analogous to a level-pool routing method, used in simpler reservoir routing schemes and 306 mainly assuming a horizontal water surface along the reservoir. This approaches a 307 dynamic method (ii and iii) if reservoir length is short, depth is large, inflow hydrograph 308 volume is large, and inflow hydrograph time of rise is long (Fread, 1992). Unit-309 catchments along the reservoir lake are considered as a river with rectangular cross 310 section, and the downstream boundary condition at the dam location is considered as a 311 simplified uniform flow (a local average slope was adopted in this case).

The equal bathymetry method (ii) ("Eq") consists of equally distributing the volume through the unit-catchments that composes the reservoir. For each level, the reservoir water surface area is equally distributed to the unit-catchments on the reservoir domain through the stage-area relationship. Thus, all the unit-catchments that compose a reservoir have the same storage capacity.

The variable bathymetry method (iii) ("Var") explicitly simulates the reservoir 317 318 dynamics to improve accuracy in the distribution of reservoir volume across the unit-319 catchments associated to the reservoir lake. Since the DEM measures the surface water 320 level, there is no information on it about the reservoir bathymetry. Thus, the proposed 321 method estimates the stage-area curve below the reservoir water level (RWL) and 322 combines it with the stage-area curve above the RWL to construct the reservoir actual 323 stage-volume curve, which can be later checked against existing data (in this study, 324 provided by the Brazilian National Electric System Operator - ONS). This method has 325 four steps:

326 1) Estimation of the stage-area curve above the RWL in a given unit-catchment.
327 This process is automatically obtained using the DEM information within a unit328 catchment, by counting the number of cells lower than a specific elevation.

2) Estimation of the "original" river bank elevation in every unit-catchment within the reservoir lake. The "original" riverbanks (i.e., in pristine conditions) were inundated by the dam. Thus, the bank elevation of all unit-catchments that compose the reservoir were defined through a linear interpolation between the bank elevation just downstream of the dam, and the one immediately upstream of the reservoir lake (Equation 3).

335
$$Z_i = Z_{down} + (Z_{up} - Z_{down}) \times (\Delta X_{down,i} / \Delta X_{down,up})$$
 Equation 3

Where *Z* represents the bank elevation and ΔX the distance between the river sections. The indices *i*, *down* and *up* represent the sections of the *i*-th unit-catchment within the reservoir, and the sections immediately downstream to the dam and upstream to the reservoir lake, respectively.

340 3) Estimation of the stage-area curve for the levels below RWL: it is assumed 341 that the water surface area below the RWL linearly increases with level, and that the 342 water surface area at the river bank elevation is zero. Thus, the stage-area curve below 343 RWL is a line going from an area equal to zero at the river bank elevation (Z_i) to the 344 first point in the stage-area curve above RWL.

4) Matching the estimated reservoir stage-volume curve with the actual one: the reservoir stage-volume is a table relating reservoir volume (V_0) with level (Z). It can be directly compared to the reservoir stage-area curve built with the combination of all the unit-catchments within the reservoir (hereafter *Res*), which is a table relating area (A_{Res}) to level (Z). In every position j on the stage-area table, a level increment (Z^j – 350 Z^{j-1}) is multiplied by its related reservoir surface water area $((A_{Res}^{j} + A_{Res}^{j-1})/2)$, 351 resulting in an incremental volume (ΔV_{Res}^{j}) . Then, the incremental volume observed on 352 the stage-volume curve related to the level Z^{j} $(\Delta V_{0}^{j} = V_{0}^{j} - V_{0}^{j-1})$ is divided by the 353 calculated incremental volume calculated from the stage-area curve (ΔV_{Res}^{j}) , generating 354 a volume ratio $(VR^{j} = \Delta V_{0}^{j} / \Delta V_{Res}^{j})$. The stage-area curve of each unit-catchment (*i*) 355 within the reservoir is recalculated independently to keep the same incremental volume 356 as the actual stage-volume curve:

357
$$(a_{Res}^{*\,i,j} + a_{Res}^{i,j-1}) = VR^j \times (a_{Res}^{i,j} + a_{Res}^{i,j-1})$$
 Equation 4

358
$$a_{Res}^{*\,i,j} = VR^j \times (a_{Res}^{i,j} + a_{Res}^{i,j-1}) - a_{Res}^{i,j-1}$$
 Equation 5

359 Where $a_{Res}^{i,j}$ is the water surface area of the unit-catchment *i* at the stage-area 360 table position *j* related to the level Z^{j} . The superscript * indicates the recalculated *a* 361 values. Equations 4 and 5 indicate an adjustment on the water surface area in level Z^{j} in 362 order to preserve the incremental volume indicated by the actual stage-volume curve.

This process is repeated through all levels $(Z^1 \text{ to } Z^n)$ of the stage-volume curve, modifying the stage-area curve of each unit-catchment within the reservoir.

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366 2.5 Reservoir operation

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The dam release is set as an internal boundary condition of the hydrodynamic model (MGB), by replacing Equation 1 by a dam outflow equation. Three types of operation schemes are compared here, considering two different approaches each: one based on regionally available data, and another with global-based parameterization. The three operation types are representative of different approaches that have been
implemented in state-of-the-art modeling systems, from generic to data-driven ones,
described as follows.

375

- 376 2.5.1 Reservoir operation schemes H06 and H06Glob
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This is a generic, inflow-based operation based on the equation proposed by Hanasaki et al. (2006) and adapted by Shin et al. (2019). This operation considers that the dam outflow is a simple function of the inflow modulated by the dam regulation capacity and the storage at the beginning of each hydrological year. Here, it is used at a daily basis and for hydropower plants, so that it does not take into account downstream water demands for irrigation or other uses. Dam outflow is defined by Equation 6:

384
$$Q(i,t) = R_i K_{i,v} I_m + (1-R_i) I_{t-1}$$
 Equation 6

Where Q(i, t) is the *i*th dam outflow at the time step *t*, R_i a regulation capacity constant that can be calibrated with observations or estimated with Equation 7 (Shin et al., 2019), I_m and I_{t-1} the annual average and dam inflow, respectively, and $K_{i,y}$ the storage fraction at the beginning of the hydrological year (Equation 8). The hydrological year of each dam is defined as the month where the naturalized flow becomes lower than the average (i.e., the beginning of the drawdown season) (Hanasaki et al., 2006).

$$391 R_i = \min(1, \alpha c_i) Equation 7$$

$$392 K_{i,y} = S_{first,y} / \alpha C_i Equation 8$$

393 The term c_i is the ratio between the reservoir maximum storage C_i and the 394 annual average dam inflow ratio ($c_i = C_i/I_m$), $S_{first,y}$ is the storage at the beginning of each hydrological year y, and αC_i is the target storage, where α is a constant set to 0.85 following Hanasaki et al. (2006).

397 Scenario H06 estimates R from a calibration procedure based on regionally 398 available observations, while scenario H06Glob (global) adopts equation (6) for 399 estimating R.

400

401 2.5.2 Reservoir operation schemes 3PT and 3PTGlob (Three-point rule curve)

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403 This is a target storage-and-release-based rule (Yassin et al. (2019), consisting of 404 a three-point rule built upon simple dam characteristic parameters, as minimum and 405 maximum operational levels, and maximum discharges (Error! No se encuentra el 406 origen de la referencia.a). Similar approaches were adopted by Zajac et al. (2017) and 407 Yassin et al. (2019). This operation emulates a reservoir rule curve that is constant 408 throughout the year with outflow as a linear function of water level, guided by three 409 points. The regional approach (scenario 3PT) adopts the following points based on 410 actual dam information (i.e. observations): minimum operational level (for which 411 outflow is zero), average operational level (for which outflow is obtained from the 412 average observed outflow), and maximum design level (associated to the dam design discharge). Supplementary Material S1 presents the adopted parameters for all dams. 413 414 Scenario 3PTGlob (global approach) follows Zajac et al. (2017), and adopts the 415 percentiles 0.1, 0.3 and 0.97 for the minimum (conservative), normal, and maximum 416 (flood) storages, which are associated to the 5th, 30th, and 97th percentiles of 417 naturalized daily discharge, respectively.

418





420 Figure 3. Reservoir operation exemplified for the Jurumirim Dam (ID 1 in Figure 1), for the (a) 421 operation scheme '3PT', considering only three pre-defined points related to dam characteristics 422 (water level and design discharges), (b) 'REG' scheme, with significant regression between 423 observed monthly mean water level and dam outflow obtained for the month of December, and 424 (c) 'REG' scheme, without significant regression for the month of September. The blue dots 425 represent observations, and the lower (upper) extrapolation of the reservoir operation for ranges 426 out of observations is depicted in red (black). See description of the operation schemes in the 427 text (Section 2.5).

429 2.5.3 Reservoir operation schemes REG and REGGlob (Regression-based rule)

430

431 This is also a target storage-and-release-based rule based on a data-intensive 432 approach (¡Error! No se encuentra el origen de la referencia.b and ¡Error! No se 433 encuentra el origen de la referencia.c). Linear regressions are computed between 434 observations of monthly average water levels and dam outflows, so that it emulates a 435 rule curve for each month of the year following the actual operation (Oliveira and 436 Loucks, 1997). A similar operation was investigated by Solander et al. (2016). For each 437 month, positive relationships are adopted as those with Pearson correlation higher than 438 0.4, which is considered satisfactory based on a visual inspection (;Error! No se

encuentra el origen de la referencia.b). For non-positive relationships (¡Error! No se encuentra el origen de la referencia.c), the monthly average discharge was used for all simulated days for a given month. For water levels out of the observed range for a given month, dam characteristics related to minimum operational level (for which outflow is zero), and maximum design level (dam design discharge) were adopted, and linearly interpolated with the observed ranges.

The global approach (scenario REGGlob) adopts long term outflow average instead of monthly regressions, making the operation similar to the standard operating policy (SOP) (Draper and Lund, 2004), considering the long term streamflow as the demand.

449

450 **2.6 Model application in the Upper Paraná River Basin**

451

452 The model was applied to the Upper Paraná River Basin with daily time step for 453 the period 1st Jan 1979 to 31st Dec 2015 (35 years + 1 spin-up year). It was run with in-454 situ daily precipitation from 2030 gauges from the following institutions: Brazilian 455 National Water Agency (ANA), Water Resources Agency of Argentina 456 (BDHI) (http://bdhi.hidricosargentina.gov.ar/) and National Meteorological and Hydrological Service of Paraguay (DMH) (https://www.meteorologia.gov.py/). Details 457 458 on precipitation data interpolation to model units are provided in Supplementary 459 Material S2. Long term climate averages from 195 stations of the Brazilian National 460 Institute of Meteorology (INMET, available at <<u>http://www.inmet.gov.br/></u>) were used 461 to compute evapotranspiration.

462 Drainage network and unit-catchments (total of 9625 units with 10 km long river 463 reaches) were derived from the 90 m Hydrosheds SRTM DEM (Lehner et al., 2008) 464 with the IPH-HydroTools GIS toolkit (Siqueira, et al., 2016a). Hydrologic Response 465 Units (HRU's) were used to define homogeneous regions for the rainfall-runoff 466 parameters, and were derived from the South America HRU map developed by Fan et 467 al. (2015). Model parameters related to soil, vegetation and river hydraulics (bankfull 468 width and depth from geomorphic relationships, and Manning's roughness coefficient) 469 are further discussed in Supplementary Material S2.

470 The model was calibrated (validated) for the period 1990-2010 (1980-1990) with 471 143 in-situ discharge gauges from ANA considering the pristine scenario (i.e., without 472 reservoirs). Naturalized flows from ONS were considered for gauges downstream of 473 dams. Supplementary Material S2 presents details on the model adjustment, including 474 performance metrics and simulated hydrographs. Overall, the model satisfactorily 475 represented natural discharges basin-wide, with 78% (79%) of the evaluated gauges 476 with Nash-Sutcliffe (Log Nash-Sutcliffe) metric > 0.6 for the validation period, and 477 42% of the gauges with the absolute value of bias < 10%.

478 The 30 regulation dams within SIN were considered, in addition to Itaipu dam (a 479 run-of-river dam but very relevant in terms of size and energy production) (Figure 1). 480 For simplicity, all other run-of-river reservoirs were not considered in the simulations, 481 since our focus was on dams with regulation capacity. To properly address basin-wide 482 flow regulation, the dams were only considered after their year of inauguration, so that 483 the model simulated the dam first filling. The effects of reservoirs were not used for 484 model calibration, but only considered for the scenarios presented in the following 485 Section 2.7.

488 2.7 Experimental design

489

490 A total of 12 simulation scenarios were run, considering the different reservoir 491 bathymetry representation and reservoir operation schemes (Table 1). The performance 492 of a given reservoir simulation was first assessed in terms of discharge and active 493 storage for all dams. Observed time series of active storage and dam outflows were 494 obtained from ANA (https://www.ana.gov.br/sar/). The hydrodynamics was assessed in 495 terms of the water surface elevation longitudinal profile at Itaipu dam, by comparing 496 simulations with satellite altimetry estimates from the ICES at mission (Schutz et al., 497 2005). ICESat carries a LiDAR sensor and has a maximum inter-track distance of 30 498 km and a repeat cycle of 91 days. A basin scale assessment was also made by 499 computing, for each river reach, the root mean squared deviation (RMSD) between 500 simulated water levels under scenarios Lum, Eq and Var (Table 1).

501 Model performance for discharge was assessed with Nash-Sutcliffe (NSE), 502 normalized root mean squared error (NRMSE) of peak discharges, and relative errors in 503 high $(Q_{10}, i.e., discharge that is exceeded 10\% of the time)$ and low flows (Q_{90}) . For 504 reservoir active storage, NRMSE and Pearson correlation metrics were adopted. Finally, 505 the average peak attenuation for each dam was assessed by first computing the 506 discharge reduction between dam inflow and outflow for each of the dam's maximum 507 annual events, followed by estimation of the average of the annual values. The 508 simulated peak attenuation was compared to the observed one with the NRMSE metric.

In addition, the role of the online coupling between hydrology and hydrodynamicprocesses was tested by performing tests with and without coupling in Section 3.1. The

511 simulation without coupling was performed by considering that evapotranspiration only 512 occurs from the non-flooded soil/vegetation system, i.e., reservoir open water 513 evaporation is not considered into the evapotranspiration computation.

514 Since there is high uncertainty on the estimation of the active flow width (*B* in 515 Equation 1) along the reservoir when it is simulated in a distributed way (i.e., reservoir 516 storage types Eq and Var; Section 2.4), hydrographs are presented in Section 3.1 517 considering two types of computation: (i) adopting the original channel width estimated 518 from geomorphic relationships, and (ii) considering the active width for unit-catchments 519 within reservoirs as the unit-catchment flooded area divided by its length (10 km), i.e., 520 considering that there is active flow along the whole reservoir cross section area.

521 Finally, the developed regional scale hydrodynamic model was used to investigate 522 the relative role of floodplains and reservoirs on flood attenuation. This was carried out 523 following the approach by Fleischmann et al. (2019a), where the model was run with 524 three river/floodplain scenarios: (i) pristine flow scenario (naturalized flow; with 525 floodplains but without reservoirs); (ii) without both floodplains and reservoirs, where 526 cross sections were assumed always rectangular and thus disregarding floodplain 527 topography; and (iii) with both floodplains and reservoirs. The role of reservoirs on 528 flood attenuation was estimated by computing the peak attenuation between scenarios 529 (i) and (iii) for the maximum flood event of each simulation year. The role of 530 floodplains was similarly computed, but considering the difference between scenarios 531 (ii) and (i). Table 1 summarizes the model runs.

533 Table 1

534 *Reservoir simulation scenarios.*

Reservoir	Storage	Scenario	Operation datails
operation*	representation**		Operation detans
H06	Lum	H06Lum	
H06	Eq	H06Eq	R calibrated
H06	Var	H06Var	_
H06Glob	Var	H06GlobVar	R estimated as min(1, αc_i)
3PT	Lum	3PTLum	Minimum, normal and maximum levels
3PT	Eq	3PTEq	and outflows derived from dam
3PT	Var	3PTVar	characteristics
3PTGlob	Var	3PTGlobVar	Minimum, normal and maximum levels
			and normal and maximum outflows
			estimated as simple percentiles as
			proposed by Zajac et al. (2017)
REG	Lum	REGLum	Monthly linear regressions between level
REG	Eq	REGEq	and outflows; months with low
REG	Var	REGVar	correlation use monthly average outflow
			instead
REGGlob	Var	REGGlobVar	Annual average outflow used for all
			months

535

* Reservoir operation: H06 (based on Hanasaki et al. (2006)), 3PT (three-point rule curve), REG (regression-based

537 operation). "Glob" refers to the global parameterization of each operation type.

538 ** Reservoir storage representation types: Lumped (Lum), Equal bathymetry (Eq), Variable bathymetry (Var).

539

- 541 **3 Results**
- 542
- 543 **3.1** Effects of reservoir storage representation
- 544

545 This section presents the results and differences in the reservoir dynamics 546 according to the storage representation approaches (Lum, Eq, Var). While all schemes 547 yielded similar estimates of dam outflows, as exemplified for a few dams in Figure 4a, 548 some key differences were identified. In some cases, the lumped method led to a 549 discharge attenuation in relation to the other two methods (smaller and delayed peaks). 550 This is due to the typical approach adopted on large-scale hydrological modeling in the 551 lumped (offline) simulation, i.e., using as inflows the simulated discharges at the dam 552 location, instead of computing the inflows as the modeled flows at the river reaches 553 close to the most upstream reservoir lake area. This approach causes the flood wave to 554 be routed along the reservoir as it was a river reach, adding artificial routing along the 555 drainage network. Hence, for lumped model applications, it is best to select all 556 tributaries that drain into the reservoir (along with direct lake inputs) and consider it as 557 the dam inflow. This effect was clearer for Itaipu dam, which is located in the lower part 558 of the basin, integrating the effects of all upstream dams and having a long reservoir.

559 Our results show that simpler, lumped reservoir models can simulate downstream 560 discharge similarly to dynamic and distributed ones (**Figure 4**a). However, the lumped 561 method fails to represent backwater effects when compared to the distributed methods 562 (Equal and Variable bathymetry) (**Figure 5**a). High deviation among Lum and Var 563 scenarios (RMSD > 10 m for some reservoirs) occurs for most reaches upstream from 564 dams. 565 As a validation experiment, ICES at satellite altimetry data were used to assess the 566 simulated profile of water surface elevation along Itaipu reservoir under the three 567 different storage schemes (Figure 5b). The lumped method is unable to simulate it 568 properly, and the slope in the upstream part of the lake was better represented with the 569 Var method. An intermediate behavior was obtained with the Eq method. Although this 570 method considers the reservoir to behave as a large box with horizontal water level, the 571 lake is assumed as connected to the rest of the drainage network, and thus the method is 572 capable to represent backwater.

573 On the other hand, the actual level in the lake area closer to the dam is more 574 dependent on the dam operation, and its simplification led to higher errors in the 575 estimated Itaipu reservoir storage. For instance, the low water level in Oct/2003 (lake 576 level closest to the dam at 216.6 m; **Figure 5**b) is related to the low simulated active 577 storage (around 14 km³; see **Figure 6**d in next section), while actual values were around 578 219.3 m for level and 18.1 km³ for storage. Itaipu is also a large dam (~170 km long), 579 and its lake is composed of many unit-catchments (which are 10 km long).

580 The hydrographs presented in **Figure 4**a also compare different ways of 581 representing flow width in the distributed reservoir simulation (bold lines), as well as 582 scenarios with and without reservoir open-water evaporation (i.e., not considering an 583 online hydrologic-hydrodynamic coupling; dashed lines). Downstream discharges had 584 mostly similar values, indicating a low sensitivity to both flow width conceptualization 585 and the online coupling scheme.

Regarding evapotranspiration estimates, differences among Lum, Eq and Var scenarios would arise if reservoir flooded areas were largely divergent, but this difference was relatively small in comparison to other model uncertainties. Our estimations of reservoir evaporation rates are in agreement with other studies in the

590 Paraná Basin (Bueno et al., 2016). Looking at the basin scale, we estimated an increase 591 of annual ET rates by 15 mm/year due to existence of reservoirs. The net reservoir 592 evaporation (i.e., reservoir evaporation minus the evapotranspiration that would occur 593 without the lake, which is equal to the difference between blue and red lines in **Figure** 594 **4**b) for the assessed lakes varied between 21 ± 12 mm/month (mean \pm SD) for Itaipu and 595 70±41mm/month for Itumbiara (located in the north of the basin; ID 19 in Figure 1). 596 This loss can be relevant during dry periods, and thus must be accounted for in large 597 scale models. For instance, loss in energy production due to reservoir evaporation in the 598 Brazilian southeast region was estimated as 2% (over 900 MW; Zambon et al., 2018), 599 and it is also an important measure to assess regional scale reservoirs' water footprint 600 (Semertzidis et al., 2019). At Itumbiara dam, the modeled evapotranspiration was highly 601 constrained by soil moisture during austral winter, what explains the large net 602 evaporation losses. This difference led to a higher peak simulated under the scenario 603 without open water evaporation (Figure 4a). When looking at finer scales, 604 evapotranspiration rates will drastically differ. Since the lumped (offline) method is not 605 able to represent the dynamic conversion between dry and flooded soil/vegetation, the 606 representation of local scale coupled processes between surface and atmosphere will 607 perform poorly, as well as the local scale runoff estimation.



610 Figure 4. (a) Daily climatology of simulated dam outflows for the three types of storage representation 611 (Lumped, Equal bathymetry, Variable bathymetry) and for Barra Bonita (ID 3 in Figure 1), Itumbiara 612 (ID 19), Foz do Areia (ID 15) and Itaipu (ID 18) dams. Results adopt the operation R. Bold colors refer to 613 default scenarios with two different flow width values (which converge to very similar values), while 614 dashed lines with light colors are scenarios not computing reservoir open water evaporation. (b) Monthly 615 climatology of open water evaporation (ET-Res; Penman equation) and evapotranspiration without 616 reservoir effects (ET-NoRes; i.e., Penman-Monteith equation not considering reservoir surface area) at 617 the location of the dam sites. The operation scheme REG is adopted for all plotted results.

(a) Assessment at basin scale



(b) Longitudinal profile along Itaipu reservoir



619

Figure 5. (a) Spatial assessment of RMSD regarding simulated water level, for scenarios lumped x variable bathymetry (left column) and equal x variable bathymetry (right column). Higher RMSD values indicate higher discrepancy between storage representation types to estimate backwater effects. Green circles refer to the simulated dams. (b) Validation of the simulated longitudinal water level profile along Itaipu reservoir with ICESat altimetry data for three different dates. The Itaipu reservoir lake area is highlighted in the panel a. The operation scheme REG is adopted for all plotted results.

626

627

629 **3.2** Effects of reservoir operation

630

This section compares the different reservoir operation (H06, 3PT, REG, in order of increasing data requirement) and storage/bathymetry representation schemes (Lum, Eq, Var), addressed in terms of dam outflow and reservoir hydrodynamics. The differences among the simulated operation schemes are larger than among the reservoir bathymetry types for discharge and active storage estimation (**Figure 6** and **Figure 7**).

636 Results for the same four dams analyzed in the previous section show that the 637 REG operation scheme led to far better outflow estimation for Itumbiara and Itaipu 638 dams. In these cases, operations H06 and 3PT also outperformed the natural flows 639 scenario (i.e., without reservoir effects). The overall model performance in representing 640 basin-wide hydrologic regime alteration (as depicted by Itaipu dam) shows that the best 641 performance was obtained for REG (NSE 0.69), followed by H06 (0.47) and 3PT 642 (0.19), and that all of them outperformed the scenario without dams (-0.26). For 643 Itumbiara, the better performance of REG for outflow compared to the other scenarios 644 can be seen in the better depicted seasonality, also reflected on the storage simulation. 645 For Foz do Areia dam, located in a river with low precipitation seasonality, all model 646 versions led to similar estimates as the natural flow scenario, i.e., the inclusion of 647 reservoirs did not lead to improvements. The simulation performance for active storage 648 (NRMSE) was similarly satisfactory for the four dams and all scenarios, except for 649 Itaipu under operation REG, which outperformed the others by significant margins (9% 650 for REG, against 23% and 27% for H06 and 3PT, respectively).

A similar behavior was observed when looking at the ensemble of 31 dams (Figure
7 and Figure 8), which was supported by a basin-wide assessment for the whole drainage
network (Supplementary Material S3). The highest differences were obtained for NSE,

654 where REG had the best performance, followed by H06 and 3PT, and for active storage 655 r, for which REG was followed by 3PT and H06. Indeed, a more satisfactory 656 performance was expected for REG given its more data-intensive nature. The three-657 point rules (3PTLum, 3PTEq, 3PTVar and 3PTGlobVar) had the lowest performance 658 for discharge in terms of NSE, but this was not the case for high (Q_{10}) and low flows 659 (Q_{90}) and peak discharges. Interestingly, for low flows, all operation types were 660 outperformed by the natural flow scenario, showing that the tested operations led to 661 excessive discharge attenuation (i.e., overestimated base flows) during dry periods.

The operation scheme REG, which relies on observed data, provided the best discharge estimates with a median NSE of 0.3 and a maximum of 0.75 for the 31 reservoirs. Although the basin-wide hydrological alteration was relatively well captured, e.g., at Itaipu dam location (**Figure 6**), the non-data intensive schemes (H06 and 3PT) need further improvements if aiming at locally relevant estimates of dam operation.

667 The analysis of regional (H06, 3PT, REG) versus global-based parameterizations 668 (H06Glob, 3PTGlob, REGGlob) showed that the global ones had a relatively poorer 669 performance in relation to their regional counterparts. For instance, the regression with 670 monthly values (REGGlob), considering long term averages as outflow, presented the 671 poorest performance for peak NRMSE and high and low flows, while 3PTGlobVar 672 scenario presented the poorest representation of active storage. However, for certain 673 purposes these global approaches could already provide valuable discharge estimates, 674 e.g., for providing a general understanding of regional scale hydrological alteration. For 675 example, median NSE values were 0.1 (0.1) for scenario H06 (H06Glob), and 0.3 (0.1) 676 for REG (REGGlob), showing the just slightly better performance of the regional 677 parameterization. The global setups were also more accurate than naturalized flows for 678 all metrics except for low flows.

679 There was an overall satisfactory model performance to estimate peak attenuation, 680 with Pearson correlation between 0.72 and 0.91, and NRMSE between 10% and 22% 681 (Figure 8). The different types of storage representation led to very similar NRMSE 682 values between simulation and observation, and the same occurred for the reservoir 683 operation, although H06 was slightly better than 3PT, which in turn was marginally 684 better than REG. This is interesting given the low degree of data requirement in the H06 685 scheme. The global-based parameterization led to less accurate results for scenarios 686 H06Glob and REGGlob, but not for 3PTGlob, in relation to their counterparts H06, 687 REG and 3PT. Among all assessed metrics in Figure 7, the only one for which a 688 noticeable difference was obtained regarding storage representation was the correlation 689 of peak attenuation, for which the variably distributed storage (Var) yielded better 690 values than the other ones.

691 The capability of the dams' regulation capacity (total active storage divided by 692 long term average discharge; red to blue colors in Figure 8) to predict peak attenuation 693 was also investigated. A positive trend between regulation capacity and peak attenuation 694 was clearer for REG (i.e., lower attenuation values with red color and higher ones with 695 blue). The lack of a clear relation resulted from the behavior of the three dams with 696 largest regulation capacity (Serra do Facão, Nova Ponte and Emborcação dams; ID's 30, 697 24 and 14 in Figure 1, respectively), which were associated to a relatively small peak 698 attenuation (around 10%).

Finally, the dry years of 2000-2001 provide a stress test for our modeling system. During this period, a major drought affected the Brazilian hydropower system, which is associated to delays in generation investment leading to a large energy crisis in the country (Jardini et al., 2002). In Jan/2000, Itaipu and Barra Bonita (ID 3 in Figure 1) dams reached their lowest levels (observations available since 1993). The same occurred for Itumbiara in Nov/2001. Among the four analyzed reservoirs in Figure 6,
only Foz do Areia, located in the Brazilian southern region, did not have an extreme
year during this period. The REG scenario was able to satisfactorily simulate some of
dams' drawdowns, but there was no clear pattern among the representation of this
extreme year: this scheme estimated a too high (small) drawdown for Itaipu (Itumbiara)
dam, but yielded satisfactory estimates for Foz do Areia and Barra Bonita dams.

710



Figure 6. Simulated dam outflow (Q) and active storage (Act Sto) for the different operation types (H06,
3PT, REG), with the variably distributed reservoir simulation method, for four dams (Itumbiara, ID 19 in

Figure 1; Barra Bonita, 3; Foz do Areia, 15; and Itaipu, 18). NSE and NRMSE performance metrics for each scenario are presented for discharges (left column) and storage (right column), respectively. Pristine simulated flows (i.e., without dams; "Nat") are also presented. The unit "hm³" stands for cubic hectometers (i.e., 10⁶ m³).





Figure 7. Model performance for discharge (NSE; NRMSE of peak discharges; and errors in high (HF) and low flows (LF)) and active storage (NRMSE and r) for the 12 scenarios of operation types and reservoir simulation methods (Table 1), as well as for the naturalized (pristine) flow scenario (Nat; for discharge analysis only). Results are presented as boxplots containing values of the 31 simulated dams, and the metric median values are presented below the scenario names. From left to right, the operation schemes are ordered in terms of increasing data requirement.



Figure 8. Comparison between observed and simulated peak attenuation for each of the 12 scenarios.
Each row refer to a reservoir operation scheme (H06, 3PT, REG, with and without global parameterization – Global par), and each column to a storage representation scheme (Lumped, Equal bathymetry, Variable bathymetry). Each point refers to a simulated dam, and colors refer to the dam regulation capacity (total active storage divided by long term average discharge, in years).

734 **3.3** The relative role of floodplains and reservoirs on flood attenuation

735

River-floodplain-reservoir hydrodynamic models have been used to understand the effects of reservoirs on downstream flooding (Fleischmann et al., 2019a; Mateo et al., 2014; Pokhrel et al., 2018; Shin et al., 2019, 2020). Here we follow the methodology proposed by Fleischmann et al. (2019a) and use the developed MGB model structure,

740 with distributed representation of reservoir bathymetry and a fully coupled river-741 floodplain-reservoir scheme, to investigate the relative role of natural floodplains and 742 reservoirs on flood attenuation along the Upper Paraná Basin. Generally, natural 743 floodplains and reservoirs have a complementary role on flood attenuation in the basin. 744 While floodplains are more important along tributaries' headwaters (e.g., Iguaçu, 745 Paranapanema, Grande and Ivinhema rivers) and in the lower reaches of the Paraná 746 mainstem, reservoir effects are more relevant along medium to lower reaches of 747 tributaries (Figure 9). Part of the reservoirs' storage is currently allocated for flood 748 control during the wet season (Oct-Apr), following the coordinated operation of the 749 Paraná dam cascades (ONS, 2019).

Located along the Paraná mainstem, the 230 km floodplain between Porto Primavera and Itaipu dams is known as the last natural large wetland in the Upper Paraná Basin (see **Figure 1** for location), with important ecosystem processes relying on it (Agostinho et al., 2001). The flood storage along this area leads to major discharge attenuation that is propagated downstream, and it is fundamental for flood control in benefit of both Itaipu dam and riverine cities. If the reservoirs did not exist, the reaches flooded by the reservoir lakes would provide additional storage along the floodplain.

757 The comparison between scenarios with and without floodplains shows that the 758 magnitudes of maximum flows are likely to be largely overestimated if basin-wide 759 floodplain storage is not considered (Figure 9b). For instance, for the Iguaçu River at 760 Fluviópolis, ignoring this effect would lead a 10-yr flood to be estimated as 6,000 m³/s 761 (green dots in Figure 9b) instead of 3,000 m³/s (blue and red dots). The effect of 762 upstream floodplains propagate downstream (Figure 9a), although they affect the lower 763 reaches of only a few tributaries (e.g., Iguacu river, with attenuation > 20% for all 764 reaches along the river mainstem). Simulated and observed hydrographs at Água

Vermelha and Itaipu dams also stress the role of floodplains and reservoirs on discharge alteration, for both high and low flows. The large effect of floodplains relates to the difference between green and black lines in **Figure 9**b. Furthermore, the major role of reservoirs on flood attenuation along main rivers makes their representation fundamental to correctly estimate flood frequencies in the downstream reaches.

770 Finally, performing an online, fully coupled simulation of the river-floodplain-771 reservoir continuum allows a continuous representation at the regional scale of the 772 spatial-temporal variation of hydraulic variables as water levels. It is exemplified for the 773 Iguaçu river mainstem, a major southern tributary of the Paraná (Figure 10). 774 Longitudinal (maximum and minimum) water surface elevation profiles, as well as 775 maximum flooded areas, highlight the connected hydrological-hydraulic processes that 776 occur basin-wide. Along the Iguacu, major floodplains occur in the upper reaches, from 777 the most upstream parts close to Curitiba city (detail iii in Figure 10b), to União da 778 Vitória and Fluviópolis cities (see Figure 9b and detail ii in Figure 10b). A geologic 779 control creates valleys with rapids between floodplains, setting up hydraulic controls 780 and increasing upstream floodplain storage. União da Vitória is also affected by 781 backwater effects from Foz do Areia dam, located a few kilometers downstream. In this 782 study we only simulated regulation dams, while run-of-river ones were not considered 783 and thus are not represented in the simulated water surface elevation continuum. 784 Downstream of the cascade, the Iguaçu has again an incised valley with small 785 floodplains, and the river width is controlled by the hydraulic control of the large Iguaçu 786 Falls (detail i in Figure 10b). This example reinforces the model capability to represent 787 the coupled human-water system at regional scale.

(a) Basin-wide assessment



(c) Hydrographs





790 Figure 9. (a) Relative role of floodplains and reservoirs on flood attenuation, in terms of 791 average attenuation of maximum annual events. The H06Var reservoir simulation scenario is 792 adopted because of smallest peak attenuation NRMSE. (b) At-a-station assessment of flood 793 attenuation by floodplains at four locations in upstream tributaries (location in figure a), in 794 terms of simulations with and without floodplains, and observed (Obs) maximum annual 795 discharges (flood frequency analysis). Maximum simulated flood extents are presented as blue 796 areas in the left figures, together with Google Earth imagery and the location of the gauges (yellow). (c) Simulated and observed hydrographs at Água Vermelha and Itaipu dams (location 797

- in figure a) for different scenarios of floodplains (FP) and reservoirs (Res). The REGVar
- simulation scenario is used here because it led to the highest discharge NSE.



Figure 10. (a) Longitudinal profiles of maximum and minimum simulated surface water elevation (blue lines) along the Iguaçu river mainstem for the scenario REGVar. Distance is measured from the confluence between Iguaçu and Paraná rivers. The three regulation dams simulated along the Iguaçu mainstem are presented in the profile (green circles), as well as the run-of-river dams not simulated (red) and some locations of interest (yellow). (b) Maximum

simulated flood extent for the same reaches from figure a. Details (i), (ii) and (iii) showparticular areas together with Google Earth Imagery.

809

810 **4 Discussion**

811

4.1 Improving the representation of reservoir operation in large scale models

813

814 This study compared generic reservoir operation schemes (H06 scenario, based 815 on Hanasaki et al. (2006)) to data-driven ones (scenarios 3PT, related to the three-point 816 rule curve, and REG, associated to linear regressions between monthly water levels and 817 dam outflows). As expected, the data-driven approach led to more accurate discharge 818 and storage estimates. For example, while the operation H06 outperformed 3PT in the 819 hydrology metrics, approaching REG, it provided the worst results in terms of tracking 820 observed storage (Figure 7). In this case, accumulated flow errors lead to poor storage 821 estimates (Turner et al., 2020). H06 has few parameters and it is apparently too simple 822 for a complex interconnected system. In turn, the REG scheme is similar to the one by 823 Solander et al. (2016) in the way that it fits a relation between storage and outflow. It is 824 also related to Yassin et al. (2019), since it estimates the actual operational levels from 825 observed data at a monthly basis, adopting dam characteristics for levels out of the 826 observed ranges, thus emulating the actual reservoir rule curve. The satisfactory 827 performance of this rule is also associated to relatively low bias in MGB estimates (see 828 Supplementary Material S1), since inflow bias can largely affect reservoir simulation 829 schemes (Turner et al., 2020).

830 The global-based parameterization (i.e., the one that does not require 831 regionally/locally available detailed data) led to a slightly poorer performance in

comparison to the regional-based one for discharge, and it was generally more accurate
than naturalized flows, providing a reasonable approach to represent hydrological
alteration at regional scales.

835 Finally, the results from the adopted operations should be analyzed by 836 considering the context in which the real system is operated. In Brazil, the actual 837 operation of all major dams is defined considering the large-scale interconnected 838 hydrothermal power system (SIN), based on operational decisions reallocating storage 839 inter-temporally throughout the system to minimize spills and energy production costs. 840 The SIN is divided into regional interconnected subsystems (South, South-east, Central-841 west, Northeast and North) with significantly diverse hydrological characteristics. As 842 the operation of a given hydropower plant affects others units downstream, a system 843 wide operation strategy prevails over individual ones. First, an energy generation 844 solution is determined for the whole system, which is later disaggregated to individual 845 power units. Given the high contribution of hydropower in the mix (over 65%) and 846 stochasticity of inflows, operating costs depend on present and future decisions. ONS 847 defines the dispatch schedule for all generating units connected in the SIN (hydropower, 848 thermal, wind and nuclear) on a monthly basis based on a merit order (from lower to 849 higher cost), and considering current reservoir storage and flow forecasts. Hence, when 850 a group of reservoirs is low in storage in a given region, hydropower plants from 851 another region can be dispatched and the energy transferred, avoiding the use of local 852 thermal plants. The coordinated operation ranges from long term (four years) to 853 dispatch scheduling (every half hour).

The improvement of regional scale models may involve hedging operations (reducing releases to minimize the probability of more severe cutbacks in the future ; You & Cai (2008)) and coordinated operations (Marques & Tilmant, 2013; Marques et

857 al., 2006; Rougé et al., 2019), which are typically not considered. In this study, the REG 858 scheme was designed to represent an average behavior (rule curve) of the coordinated 859 system, which trades off its capability to depict anomalous years (especially dry ones). 860 Approaches focusing on a single reservoir may disregard basin-scale flood or drought 861 control that exists within a coordinated operation (Rougé et al., 2019). Furthermore, to 862 improve estimates, a detailed operation would require the representation of actual 863 hydraulic structures (spillways, outlet works, etc.) (Fleischmann et al., 2019a) which is 864 not always available to dams worldwide. As our purpose is to perform regional scale 865 simulations, simplified operations were chosen for better model applicability. Potential 866 future improvements should expand the REG operation to multiple regressions 867 including other relevant explanatory variables beyond observed levels (e.g., Solander et 868 al. (2016)). These relevant variables should be chosen based on homogeneous behavior in specific regions (i.e. not all regions would have the same explanatory variables with 869 870 the same coefficients). On the other hand, the proposed methodology could be easily 871 expanded to continental scale domains (e.g., Siqueira et al. (2018)), provided 872 information on dam characteristics as stage-area curves and observed time series of 873 storage and outflows is available. Finally, the proposed operation approaches do not 874 take into account water withdrawals and consumptive demands associated (e.g. 875 irrigation), as those are small in the context of the studied Paraná basin, and their effect 876 is localized, so that we focused on hydropower generation dams instead. In future work, 877 distributed modeling systems should explicitly simulate reservoir dynamics, as more 878 information becomes available. In Brazil, recent national scale mapping of irrigation 879 schemes (ANA, 2017) could be coupled to the MGB framework, combined with recent 880 developments in large scale modeling of reservoir operation under timely varying water

demands (Biemans et al., 2011; Haddeland et al., 2014; Hanasaki et al., 2006; Voisin etal., 2017).

883

884 4.2 On the importance of representing the river-floodplain-reservoir continuum 885 in large scale models

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887 The presented model of the river-floodplain-reservoir continuum at regional scale 888 provides a continuous depiction of the spatial-temporal variation of hydraulic variables 889 as water surface elevation and flood extent and storage. The consideration of a 890 distributed reservoir bathymetry was shown to be fundamental to estimate backwater 891 effects, as revealed by a basin-scale comparison between lumped and distributed 892 schemes (equal and variable bathymetry), and an ICESat-based validation of the Itaipu 893 reservoir longitudinal water level profile. Backwater effects are required for many 894 applications, e.g., to perform real-time monitoring of the impact of a given reservoir on 895 an upstream city (see the Iguaçu River case study in Section 3.3), or to correctly 896 estimate the dam inflow along lateral tributaries.

897 A correct representation of hydrodynamics at the basin scale was also shown to be 898 fundamental for flood frequency analysis, considering both reservoirs and floodplains' 899 effects (Fleischmann et al., 2019a; Tanaka et al., 2017; Wang et al., 2017; Zajac et al., 900 2017; Zhao et al., 2020). Project flood discharges are usually estimated with simplified 901 methods as unit hydrographs that do not consider river floodplain attenuation. An 902 interesting and open research question relates to how far upstream can these floodplain 903 storage effects go, what has major implications for water resources management. 904 Floodplains alter the celerity of flood waves at the whole basin scale, and are a major 905 driver of hydrograph shape across scales (Collischonn et al., 2017; Fleischmann et al.,

906 2016). Besides, here we have assessed the role of flood attenuation in riverine wetlands,
907 while at the very upstream reaches, upland rain-fed wetlands may also change towards
908 flood generating areas, requiring further studies (Acreman and Holden, 2013).

909 Natural floodplains provide valuable ecosystem services in terms of flood 910 attenuation and resilience, and its quantification requires new tools (Ameli & Creed, 2019; Wu et al., 2020). Building new dams (especially if designed for purposes 911 912 different than flood control), as well as new developments in floodplain areas (e.g., 913 levees), may remove the large floodplain storage effects that protect downstream 914 reaches against floods. This was shown for many rivers, as the Mississippi with 915 hundreds of kilometers of levees deactivating the river natural flood storage (Hey and 916 Philippi, 1995) and the Danube river (Schober et al., 2014). Furthermore, a benefit-cost 917 analysis of acquiring floodplain lands to avoid flood damage was performed for the 918 whole USA recently (Johnson et al., 2020), and suggested that the cumulative flood 919 damages exceeds the costs of land acquisition for a 2070 scenario. The synergic effects 920 of reservoirs and floodplains on flood attenuation have been increasingly addressed in 921 the literature with large scale models (Shin et al., 2020), and are in accordance with our 922 results. The analysis and modeling improvements provided here indicate that the 923 synergy between floodable areas and the operation of dam cascades at the whole basin 924 scale is relevant and requires further understanding, which is beyond simpler large scale 925 models relying solely on hydrodynamic simulations along downstream floodable areas. 926 In the case of the Paraná basin, inserting the proposed methodology into a proper flood 927 risk management framework will require real-time flood monitoring and forecasting.

929 **4.3** Perspectives on simulating the river-floodplain-reservoir continuum at large

930 scales

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932 The development of coupled river-floodplain-reservoir modeling systems is 933 associated to the hyper-resolution global modeling agenda, aiming for example to 934 improve medium-range flood forecasts (Zajac et al., 2017) that are locally relevant 935 (Bierkens et al., 2015; Fleischmann et al., 2019c; Rajib et al., 2020; Wood et al., 2011) 936 within land surface, earth system or global hydrological models, and explicitly 937 representing reservoir dynamics within detailed grids (Shin et al., 2019; Wada et al., 938 2016; Zajac et al., 2017). From continental to global scales, these models are powerful 939 tools to assist national and world agencies on the coordinated planning of reservoir 940 expansion, as well as understanding the effects of current and future dams on water and 941 biogeochemical cycles (Bierkens et al., 2015; Wada et al., 2016), and their interaction 942 with climate change, contributing to improve global water security (Adam et al., 2007; 943 Arias et al., 2020; Dang et al., 2019; Ehsani et al., 2017; Poff et al., 2016; Williamson et 944 al., 2009). On the other hand, from local to regional scales, they can be used for actual 945 dam operation, real-time monitoring and forecasting systems, and estimation of locally 946 relevant discharges at high spatial-temporal resolution.

The necessity of spatially and temporally continuous fields of state variables as river discharges and levels has prompted the combination of remote sensing datasets and hydraulic models (Brêda et al., 2019; Gleason and Durand, 2020). ICESat altimetry data provide valuable information for lakes (Gao, 2015; O'Loughlin et al., 2016) and are very promising for validating large scale reservoir modeling systems, while new missions as ICESat-2 (threefold increase in sampling density) and SWOT will increase our capability to remotely monitor reservoirs and estimate reservoir parameters and 954 even reservoir operation (Bonnema & Hossain, 2017, 2019; Busker et al., 2019; 955 Getirana et al., 2018; Van Den Hoek et al., 2019; Yao et al., 2019; Yoon et al., 2016; 956 Yoon & Beighley, 2015). New global datasets of reservoir characteristics are also 957 promising, including new methodologies to estimate reservoir area-depth-volume 958 relationships based on remote-sensing datasets (Crétaux et al., 2016; Fassoni-Andrade 959 et al., 2020; Gao et al., 2012; Lehner et al., 2011; Li et al., 2020; Liebe et al., 2005; 960 Mulligan et al., 2020; Yigzaw et al., 2019, 2018). These advances contributed to the 961 development of reservoir representation in global hydrological models (Döll et al., 962 2009; Sutanudjaja et al., 2018; Voisin et al., 2013; Wada et al., 2016; Yassin et al., 963 2019; Zhou et al., 2017), including the data-driven operations schemes as presented here 964 and in other recent studies (Turner et al., 2020), and are shaping the new generation of 965 large scale water resources models.

966 Regarding large scale model improvement, we have adopted a 10 km river reach 967 discretization for the Paraná basin, in accordance with current practices adopted in regional to global hydrological models (i.e., 5-10 km; Shin et al., 2019; Wada et al., 968 969 2016; Zajac et al., 2017). However, higher resolution (i.e., 1 km or smaller) are required 970 to better represent relatively small dams. Our results indicate the need for better 971 representation of reservoir bathymetry distribution in order to correctly address local 972 scale hydraulic processes as backwater effects, corroborating recent studies (Adam et 973 al., 2007; Shin et al., 2019).

Finally, we have also discussed the role of fully coupling hydrologicalhydrodynamic processes in a two-way scheme. The MGB model considers a dynamic surface water cover, and the associated changes in evapotranspiration/runoff generation, e.g., by alternating the soil/vegetation Penman-Monteith equation with the open water Penam evaporation scheme. Considering reservoir evaporation was also implemented by other modeling systems (Adam et al., 2007; Mamede et al., 2018; Zhao et al., 2016).
This consideration is particularly important during dry periods, and even more
important for reservoirs in semi-arid regions (Bonnema et al., 2016; Celeste and Billib,
2010; Döll et al., 2009; Mamede et al., 2018). Besides a dynamic flood fraction cover,
other reservoir processes at local scale should also be included, as reservoir
sedimentation (Zhao et al., 2016) and ground seepage.

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986 5 Conclusions
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988 In this study we presented the successful development and a thorough analysis of a 989 regional scale model capable to simulate the daily river-floodplain-reservoir continuum 990 that exists along large basins. A case study was performed in the ~950,000 km² Upper 991 Paraná River Basin in South America, considering 30 regulation reservoirs and the 992 Itaipu run-of-river dam, which is the largest in world in terms of energy production. 993 Twelve simulation scenarios considering different reservoir bathymetry representation 994 and reservoir operation schemes were performed, and assessed in terms of water levels, 995 discharges, flood extent and reservoir storage. A methodology to assess the relative role 996 of floodplains and reservoirs on basin-wide flood attenuation was presented, providing a 997 powerful way to understand regional scale floods and the value of preserving natural 998 floodplains' services. We conclude that:

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• A distributed representation of reservoir bathymetry in large scale hydrological models is required for accurate predictions of backwater effects, upstream surface water elevation and flooding;

Both lumped and distributed representations of reservoir bathymetry in large
 scale hydrological-hydrodynamic models provide similar predictions of
 downstream river discharges and water levels;

A data-driven operation scheme based on historical data of reservoir storage and
 outflows adds significant value to the accuracy of reservoir storage predictions,
 if compared to more generic algorithms;

- Although the data-driven approach outperforms more generic schemes (namely,
 the H06 method) in terms of discharge estimation, the simpler generic schemes
 provide reasonable estimates, and thus can be useful to estimate regional scale
 hydrological regime alteration;
- Global-based parameterizations of operation schemes lead to only slightly
 poorer performance in comparison to more regionally-based ones, providing
 reasonable estimates of regional scale hydrological regime alteration;
- However, to properly simulate the river-floodplain-reservoir continuum at regional scale, satisfactory simulation of water levels and reservoir storages are required, and thus large-scale models should include data-driven reservoir operation approaches based on regional parameterization, and distributed reservoir bathymetry (if possible, with a variable bathymetry scheme);

In the Paraná River Basin, the floodplains are mainly located in upper parts of some tributaries and in the river mainstem, while reservoir effects are more important for flood attenuation along medium and lower reaches of tributaries.
 In this case, floodplains and reservoirs provide complementary flood attenuation at regional scale;

The existence of river floodplains across the whole basin can lead to major flood
 attenuation at the regional scale, and not only in downstream lowland reaches, as
 usually assumed in large scale models;

Major overestimation of flood design discharges can occur if the model does not
 consider upstream floodplain and reservoir storage effects, especially in the
 context of flood frequency analysis.

Finally, our results stress the importance of simulating the river-floodplain-reservoir continuum at large scales. Increasing computational capacity with intense cloud computing, and new remote sensing-based datasets and techniques, are quickly pushing the development of large to global scale models, and thus improving to a great extent our understanding and prediction capability regarding reservoir-floodplain interactions.

1036

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