

1 **Seasonal hydrologic buffer on continents: patterns, drivers and ecological**
2 **benefits**

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15 **Keywords**

16 seasonal water storage, hydrologic buffer, lateral water transfer, delayed evaporation, climatic
17 water imbalance, snow storage

18

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19 **Abstract**

20 Continental precipitation returns to the atmosphere and the ocean with a delay that is critical in
21 regulating seasonal water supply to ecosystems and societies. We quantify the magnitude and
22 spatial patterns of this seasonal hydrologic buffer, its climatic and terrain drivers, and its apparent
23 benefits to ecosystems using observed precipitation, climate reanalysis evaporation, GRACE
24 seasonal water storage change, and MODIS vegetation index for a $1^\circ \times 1^\circ$ global grid. We found
25 that (1) seasonal hydrologic buffering is widespread and averages 241 mm.yr^{-1} on land (a quarter
26 of continental precipitation); it supports evaporation 3-to-9 months of the year over all regions
27 except the per-humid tropics and energy limited high latitudes, (2) the seasonal climatic water
28 imbalance, with surplus in some months and deficit in others, drives hydrologic buffering in lower
29 latitudes, while it is controlled by snow/ice storage in high latitudes, (3) the main terrain effect at
30 our scale of analysis is grid-to-grid water transfer via large rivers providing lateral subsidy to
31 lowland basins, and (4) buffering is manifested in global patterns of plant water use, as shown by
32 high evaporation levels in water deficit conditions, particularly under tropical monsoonal climate.
33 Our results highlight the paramount role of seasonal land water storage and redistribution in
34 supporting ecosystem productivity, and provide a reference to understanding likely impacts of
35 global change on the water cycle and ecosystem dynamics in the future.

36

37 **1. Introduction**

38

39 Precipitation (P) falling on land is temporarily stored in snow packs, soils, groundwater,
40 rivers, lakes and wetlands. This storage and its delayed release regulate the timing of the global
41 water cycle, so that water availability to land ecosystems and societies is partially decoupled from
42 precipitation, as evidenced by flowing rivers and thriving vegetation in the absence of P. This
43 storage and release occur over a range of time scales (hours to millennia) and regulates a range of
44 processes (flooding [1] to long-term carbon cycle through weathering [2]). In this study we focus
45 on the seasonal time scale which is critical to understanding vegetation activity and water supply
46 to human needs and has a growing body of available global data. In the remainder of this paper we
47 will refer to this seasonal storage and release as the “seasonal hydrologic buffer” (SHB). Knowing
48 where, how much, for how long, and why precipitation is held on land through the course of a
49 growing season will give clues to where on Earth the service (both for nature and humans) of

50 seasonal hydrological buffering is most significant, how it is controlled by regional environmental
51 conditions, and to what extent it may be affected by global environmental change.

52

53 At a given location and in a given month, how much of the water storage is recharged or
54 depleted depends on several fluxes, broadly grouped here into climatically-driven vertical fluxes
55 and terrain driven lateral fluxes. Of the former, an example is found in strongly seasonal climates
56 where the wet-season surplus ($P > E$, E =evaporation) is carried over to fill the dry-season deficit
57 ($P < E$) in a tropical monsoonal or Mediterranean climate, as emphasized at least since
58 Thornthwaite's seasonal diagrams of climatic water imbalance [3]. Another climatic factor is snow
59 and ice in cold regions where winter P is held until the melt season. Of the terrain-driven fluxes,
60 those operating at hourly to daily scales, such as rapid runoff to local streams, are less important
61 for seasonal hydrological buffering, but regional-scale lateral transport through large river
62 networks towards floodplains can delay the outflow, holding water that later will be exported to
63 downstream areas or evaporated in-situ. Soil storage capacity is another terrain driver, which can
64 either restrict buffering where the bedrock is close to the surface or enhance it where soils and
65 plant roots are deep and/or groundwater is an important source – through phreatophytic vegetation
66 or groundwater-fed irrigation. Lastly, lakes and reservoirs offer an additional land storage
67 mechanism.

68

69 We ask the following questions (1) What is the magnitude and geographic distribution of
70 seasonal water buffering? (2) What are the drivers of such storage and release? Where on Earth
71 are climate or the terrain the primary drivers of seasonal water buffering? (3) How does this
72 seasonal storage correlates with geographic and seasonal patterns of plant productivity? These
73 questions will shed light on where and how seasonal hydrologic buffering allows terrestrial life to
74 escape strict dependence on current and local precipitation, how human activities may alter natural
75 buffering, and how they could be steered to maximize ecological and social benefits. We will
76 address these questions through a set of global data analyses over $1^\circ \times 1^\circ$ grids and at monthly steps
77 over the 2003-2010 period.

78

79 **2. Materials and methods**

80

81 Over a given grid, the monthly water budget can be written as:

$$\frac{\Delta S}{\Delta t} = P - E - Q, \quad (1)$$

82 where ΔS is the change in terrestrial water storage (hereafter S) over the monthly timespan Δt , P
83 is monthly precipitation and E monthly evaporation from the grid, and Q is the lateral, grid to grid
84 transport (run-on or runoff).

85

86 **2.1. Data sources**

87

88 The storage change is estimated from GRACE (the Gravity Recovery And Climate
89 Experiment) satellites [4], which have enabled independent constraints on the storage term over
90 global grids [5,6] from April 2002 onwards. Starting from 2003 (first full calendar year covered),
91 here we combined the three RL05 solutions releases of the $1 \times 1^\circ$ near-monthly product provided
92 by the Center for Space Research at the University of Texas at Austin (CSR), the Jet Propulsion
93 Laboratory (JPL), and the German Research Centre for Geosciences (GFZ). After applying the
94 scaling factor (provided along with the data sets) to compensate for signal alteration from sampling
95 and post-processing of the original data [6], we averaged the three data sources to derive a mean
96 TWS record, as this approach allows reducing the noise within the available gravity field solutions
97 [7]. A quality criterion for the S record is defined as $q_S > 1$, with

$$q_S = \frac{A_S}{2 \cdot e_S}, \quad (2)$$

98 where A_S is the mean seasonal amplitude of S over one year, and e_S is the time-invariant error
99 associated to the GRACE record, which bears two uncorrelated components:

$$e_S = \sqrt{e_{S,\text{meas}}^2 + e_{S,\text{leak}}^2}, \quad (3)$$

100 where $e_{S,\text{meas}}$ is the measurement error and $e_{S,\text{leak}}$ is the leakage error arising from data filtering
101 and rescaling from the original surface mass variations measured by GRACE [6]. Like all seasonal
102 metrics used in this study, A_S is calculated for each 12-month moving window of the study period
103 (85 overlapping annual cycles over 2003-2010), and then time-averaged, to best use the seasonal
104 content in the data.

105

106 P, snowfall, and E products are from ECMWF-Interim reanalysis: P and snowfall have been
107 corrected for biases using GPCP gauge-satellite observations (ERA-Interim/LAND, [8]), while E
108 is taken from the ERA-Interim reanalysis [9] which assimilates atmospheric observations with
109 energy and water budget closures in the atmosphere (vs. land model estimates which are poorly
110 constrained [10]). P and snowfall are available to 2010, setting our study period as 2003-2010. We
111 also use potential evaporation (E_p) from the CRU TS3.23 monthly database, based on a modified
112 version of the Penman-Monteith equation [11].

113

114 The Enhanced Vegetation Index (EVI) is based on the level-3 MODIS composite NDVI
115 available at monthly, $0.05 \times 0.05^\circ$ resolution (MOD13C2, collection 5) [12]. Terrain slope was
116 calculated using the elevation product of the Shuttle Radar Topography Mission (SRTM) [13] at
117 3-arc-seconds (~ 90 meters) resolution, then using the mean and maximum values for the $1 \times 1^\circ$
118 resolution of this study. Finally, the land cover classification is provided by the Climate Change
119 Initiative of the European Space Agency (ESA-CCI) 300-m land cover product over the 2005-
120 2008 period [14], aggregated to $1^\circ \times 1^\circ$ (Supplementary Fig. S1).

121

122 **2.2. Quantifying seasonal hydrologic buffers**

123

124 The GRACE record does not provide absolute water storage, and all solutions releases are
125 expressed at each location as S departures from the 2004-2009 mean baseline. Storage anomalies
126 captured as variations from month to month are precisely what we require to characterize seasonal
127 storage dynamics. We define this observational seasonal hydrologic buffer (SHB) as the filling and
128 releasing of water storage in a grid cell over a season, and we quantify SHB by taking the average
129 between accumulated positive (filling) and negative (releasing) monthly S variations ($\Delta S/\Delta t$) over
130 a 12-month period:

$$SHB = \frac{1}{2} \left(\sum_{1 \text{ year}} \frac{|\Delta S|}{\Delta t} \right) . \quad (4)$$

131 In order to compare observed seasonal hydrological buffering with the levels of buffering
132 expected from climatic drivers, we consider the seasonal imbalance between atmospheric supply
133 (P) and demand (potential evaporation, E_p), which is referred to as the seasonal hydrologic buffer

134 resulting from climatic water balance (SHB_{cwb}). It can be estimated from the amplitude and timing
 135 of these two variables. Following [15], we estimate SHB_{cwb} as the cumulative deficit (sum of E_p -
 136 P when $E_p > P$) that can be met by previous water surplus (sum of $P - E_p$ when $P > E_p$):

$$SHB_{cwb} = \min \left(\int_{P > E_p} (P - E_p) dt, \int_{E_p > P} (E_p - P) dt \right)_{1 \text{ year}}, \quad (5)$$

137 and averaged over the 85 overlapping annual cycles during the study period. For this climatic
 138 approach we use E_p (not E) because E_p is a direct measure of atmospheric water vapor demand
 139 independent of vegetation and soil water characteristics.

140 Independently of the seasonal water imbalance characterized above, snow accumulation
 141 can enhance seasonal hydrological buffering. We characterized this climatic driver simply based
 142 on the continental patterns of mean annual snowfall.

143 We also account for the contribution of horizontal water transfers to total SHB based on
 144 grid-to-grid calculations of lateral transport Q via the monthly water balance equation (rearranging
 145 Eq. 1):

$$Q = P' - E' - \frac{\Delta S}{\Delta t}, \quad (6)$$

146 where P' and E' are respectively the P and E aggregated from daily values to align with GRACE's
 147 near-monthly time steps Δt , following the methodology of [16]. Separating months with net lateral
 148 inflow ($Q < 0$) and outflow ($Q > 0$), we derive an idealized, lateral water-balance-based, seasonal
 149 hydrologic buffer, or $SHB_{lateral}$:

$$SHB_{lateral} = \min \left(\int_{Q < 0} |Q| dt, \int_{Q > 0} Q dt \right)_{1 \text{ year}}, \quad (7)$$

150 where the first term in the parenthesis is the accumulated lateral inflow, and the second term
 151 outflow, averaged over the 85 overlapping annual cycles.

152

153 Lastly, in order to quantify vegetation benefits from SHB, we record the occurrence of
 154 monthly $E > P$, where the deficit is thus met by temporal and/or spatial carry-over of
 155 water surpluses. We will refer to this as subsidized E (E_s), fed by past and/or upslope surplus. We

156 examine the magnitude of the subsidized E over a year, calculated as,

$$E_s = \left(\int_{E>P} (E - P) dt \right)_{1 \text{ year}} . \quad (8)$$

157 When $E > P$, we further distinguish the relative contribution of the lateral water inflow to E_s
158 (hereafter E_{iflw}). This component is first determined at each location on a monthly basis by
159 comparing the value of runoff (Q) to the monthly subsidized E, distinguishing three cases:

- 160 • *If $Q > 0$:* there is net lateral outflow. As a result, the monthly subsidized E must come
161 exclusively from past storage ($\Delta S < 0$, Eq. (1)), so that $E_{\text{iflw}} = 0$.
- 162 • *If $Q < P - E < 0$:* there is net lateral inflow, exceeding subsidized E. All of the latter is assumed
163 to be fed by lateral inputs of this month (and the lateral surplus goes to storage, $\Delta S > 0$ from
164 Eq. (1)), so that $E_{\text{iflw}} = 1$.
- 165 • *If $P - E < Q < 0$:* there is net lateral inflow, smaller than subsidized E. All lateral inputs of this
166 month are assumed to feed $E - P$, the remainder of the latter relies on storage from past
167 months ($\Delta S < 0$, Eq. (1)), so that $E_{\text{iflw}} = |Q| / (E - P)$.

168 The results are then averaged over all the monthly time steps, providing the fraction of subsidized
169 E derived from sub-monthly lateral inflow.

170

171 **3. Results and Discussion**

172 **3.1. Magnitude and distribution of the seasonal water buffer**

173

174 To address where on Earth significant seasonal water storage and release takes place
175 (Question 1) we derived the total seasonal hydrologic buffer (SHB, see Eq. (4)) directly from
176 GRACE observations. The global patterns of SHB are shown in Fig. 1, excluding grid cells with
177 low signal-to-noise ratio (i.e., $q_s < 1$) and Antarctica and Greenland. The values vary from < 10
178 mm.yr^{-1} in the Taklimakan desert to $> 4,700 \text{ mm.yr}^{-1}$ (doubling annual P) in the main valley of the
179 Amazon, with 17% of the land surface displaying values greater than 300 mm.yr^{-1} . A broad
180 latitudinal distribution is visible, with a narrow band of low SHB values at the equator fringed by
181 higher values in the $5\text{-}25^\circ$ latitude range. In mid latitudes SHB decreases except in snow-capped
182 mountains (e.g. Alaskan-Canadian coastal range, Japan, and Kamchatka peninsula) and large river

183 floodplains (e.g. lower Mississippi, Yangzi and Paraná). Integrated globally, this seasonal
184 hydrologic storage is $\sim 22 \times 10^3 \text{ km}^3 \cdot \text{yr}^{-1}$ (or $241 \text{ mm} \cdot \text{yr}^{-1}$ as the area-weighted mean water depth),

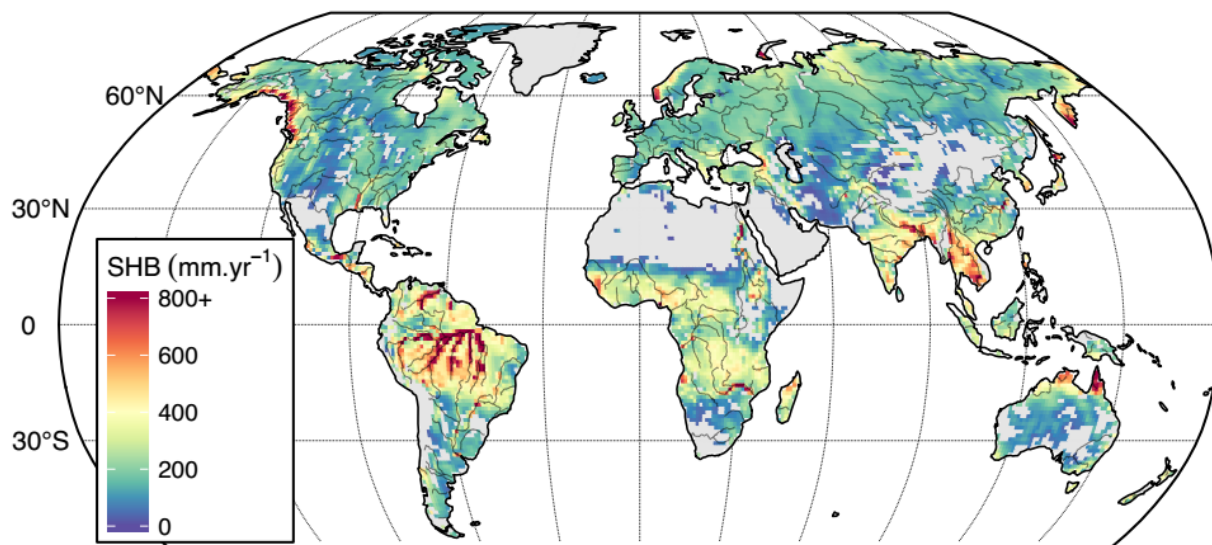


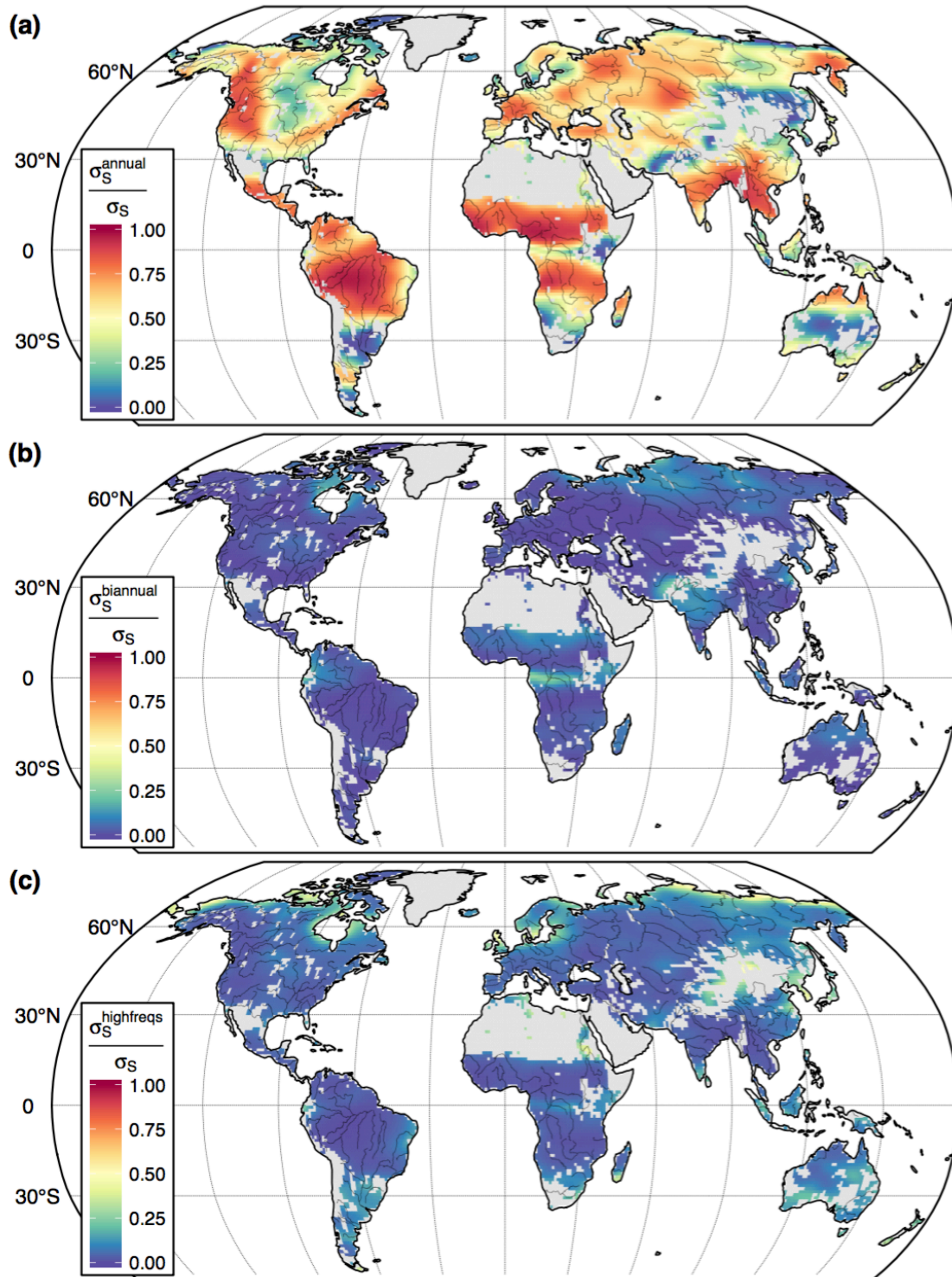
Figure 1. Vertically-integrated seasonal hydrologic buffer (SHB) on land, excluding locations with low signal-to-noise ratio (light grey).

185 which is about 25% of the annual P on land.

186

187 We note that SHB describes a flux and is thus conceptually distinct from a holding capacity,
188 i.e., the “size of the bucket”. Taking two idealized examples, qualitatively different systems can
189 display similar SHB values either through a single ample seasonal emptying-refilling sequence or
190 through several smaller non-seasonal sequences, resulting in a comparatively “shallower” storage
191 potential in the second case. In order to evaluate the proportion of water storage variability
192 explained by its annual (i.e. 1-peak seasonal cycle), biannual (i.e. 2-peak seasonal cycle) and
193 higher frequency (i.e. random variability) components, we applied a fast Fourier transform
194 algorithm to S at each location, allowing a variance decomposition into these three harmonics (Fig.
195 2). We find that although SHB integrates all water storage variations taking place between the
196 annual and monthly scale, in most areas of the world its values are predominantly explained by
197 the annual component of S, and less by the biannual and higher frequencies. The annual component
198 is most significant in most of the tropics, as well as in regions of Mediterranean climate or snow
199 domination (map in Fig. 2a and typical time series in Supplementary Figs. S2a and S2b). The
200 biannual component is weak except near the equator (Fig. 2b and Supplementary Fig. S2c).

201 Finally, the higher frequency components are significant in regions with low climatic seasonality



202

Figure 2. Decomposition of the variance of terrestrial water storage signal (ΔS) into **a)** 1-peak seasonal cycle (annual harmonic), **b)** 2-peaks seasonal cycle (biannual harmonic) and **c)** higher

frequency fluctuations (sum of remaining harmonics). Locations with low signal-to-noise ratio are excluded (light grey).

203 or where precipitation is low and sporadic (Fig. 2c and Supplementary Fig. S2d). In addition, we
204 use the mean seasonal storage amplitude (A_S , obtained directly from GRACE) as an estimate of
205 the holding capacity and compare it with SHB, finding that A_S accounts for at least 80% of SHB
206 at more than 86% of locations considered in this study, with less than 2% of the pixels showing
207 A_S /SHB values lower than 0.6 (not shown).

208

209 **3.2. Drivers behind observed storage patterns**

210 To explore the drivers of SHB and the relative importance of climate and terrain regulating its
211 patterns (Question 2) we developed independent global SHB maps.

212

213 **3.2.1. Climate**

214 Seasonal hydrologic buffering resulting from imbalance between P and E_p (SHB_{cwb} , Fig. 3a)
215 reveal the climatic inclinations for storing wet-season surplus and releasing it in the deficit periods
216 later. As expected, regions with strongly seasonal climate stand out, corresponding to the large
217 seasonal swings in P following the annual march of the ITCZ. The highest SHB_{cwb} values (>500
218 $mm.yr^{-1}$) are found in subtropical regions with winter-dry climate (Köppen-Geiger climate
219 classification, [17]) such as the Sudanian savanna south of the Sahel, the Zambezi region and
220 western Madagascar in Africa, the Indian subcontinent (where SHB_{cwb} reaches $810 mm.yr^{-1}$),
221 Northern Australia, Eastern Brazil and Venezuela. Zones with high values ($300-500 mm.yr^{-1}$) are
222 also found in smaller sectors with Mediterranean climate such as the Pacific coastal regions of the
223 US, central Chile, Peru, and the Mediterranean Sea regions. Also as expected, SHB_{cwb} is low in
224 humid equatorial regions (e.g. the western Amazon, Congo, and the Eastern Tropics), and in arid
225 locations (e.g., the Sahara, Gobi and Kalahari deserts, Central US, Saudi Arabia): the former have
226 no demand for seasonal buffering ($P > E_p$ every month), and the latter have no means for it ($P < E_p$
227 every month). About 11% of the continental surface (excluding Greenland and Antarctica) has
228 SHB_{cwb} exceeding $300 mm.yr^{-1}$, with a global volume of $\sim 16 \times 10^3 km^3.yr^{-1}$, or expressed as area-
229 weighted average water depth, $134 mm.yr^{-1}$. Comparing this climatic buffering with GRACE
230 observations (Fig. 3a vs. Fig. 1) and their respective average values (134 vs. $341 mm.yr^{-1}$), the

231 global importance and, in some regions, dominance of the climatic water balance ($P-E_p$) as a driver
232 of seasonal hydrologic buffering on land becomes evident. The global distribution of SHB_{cwb} also

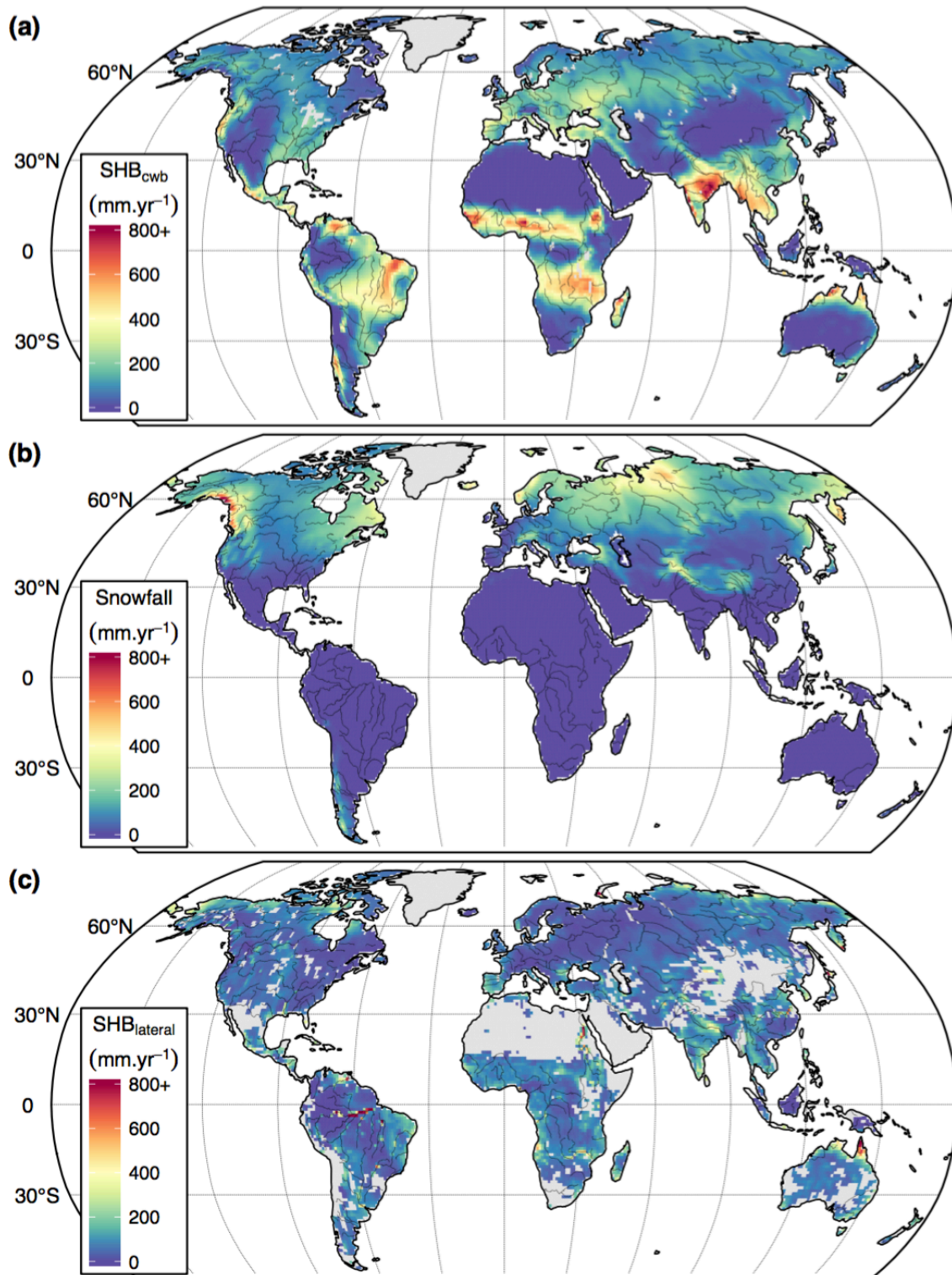


Figure 3. Continental distribution of the seasonal hydrologic buffer driven by (a) climatic water balance (precipitation - potential evaporation), (b) snow/ice storage and delayed release, and (c) lateral water transfer (inflow - outflow).

234 echoes that of the root zone storage estimates derived at the global scale by [18], both in terms of
235 patterns and magnitude. This is a striking convergence considering that their associated
236 methodologies differed on several points: 1) the quantity reported in [18] is a storage capacity
237 while SHB_{cwb} is a flux (respectively analogous to A_S and SHB), 2) in the water balance [18] uses
238 evaporation while we use potential evaporation (Eq. 4), 3) [18] takes into account water inputs
239 from irrigation in addition to P, and 4) the only comparable input, P, is obtained from independent
240 datasets (CRU TS3.22 [11] in the case of [18]).

241 The other important climatic driver that we explored was snow/ice storage and delayed melt,
242 which was shown to significantly decouple storage from runoff [19]. The magnitude and global
243 distribution of snow/ice storage is shown in Fig. 3b. Snow storage dominates the high latitudes
244 and altitudes, contributing to the total seasonal hydrologic buffering of regions where it cannot be
245 attributed to the $P-E_p$ balance discussed earlier (e.g., Alaskan-Canadian Coastal Range, and
246 Western Siberian). In regions where at least 5% of P is snow, 35% of the SHB variation across
247 grid cells is explained by annual snowfall water depth amount (Fig. 4), indicating a significant role
248 played by snow and ice storage in driving seasonal hydrologic buffering, whose area-weighted
249 average water depth is $152 \text{ mm}\cdot\text{yr}^{-1}$.

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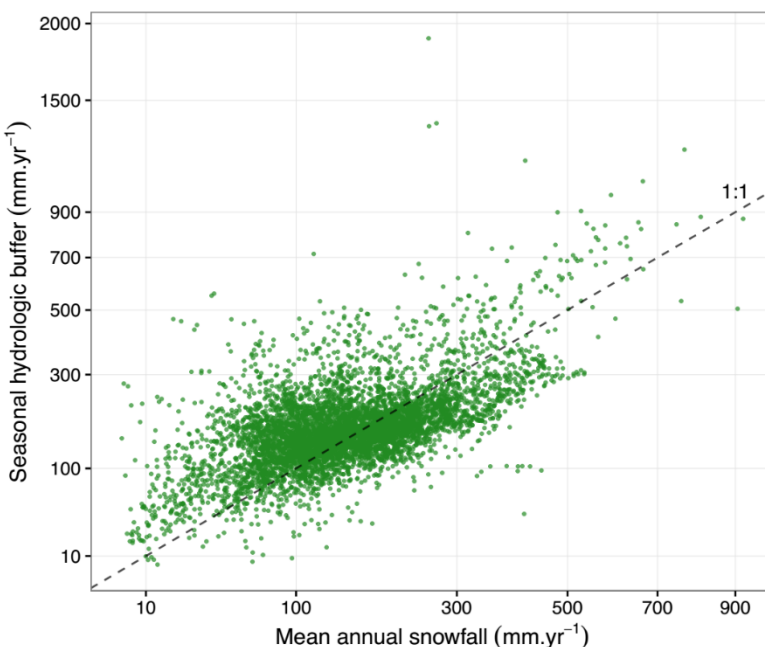


Figure 4. Vertically-integrated seasonal hydrologic buffer (SHB) versus mean annual snowfall, excluding locations with less than 5% of annual precipitation falling as snow and with low signal-to-noise ratio. Scales are non-linear for a better display of all data points.

251

252 3.2.2. Terrain

253 Long distance lateral fluxes are responsible for some seasonal hydrological buffering at
254 specific locations. Water surpluses converging towards river/lake valleys and coastal lowlands,
255 where flow is retarded until its eventual release to the ocean or its evaporation, are the dominant
256 source of seasonal hydrological buffering at our scale of analysis (month to month, $1^\circ \times 1^\circ$ grid
257 cells). This lateral convergence removes water in surplus months from large areas over the source
258 region and concentrates it, typically in smaller areas, in downstream sink regions with examples
259 coming from both relatively dry as well as wet basins [20,21]. Here, lateral transfer of water
260 appears driven by the complex interplay of the myriad of terrain characteristics that shape large
261 basins (e.g., elevation, slope, soil and sediment thickness, faulting and tectonics, among others).
262 Without globally consistent sediment thickness and effective porosity data, it is difficult to assess
263 the subsurface portion of the land storage capacity in regulating total SHB. A surrogate variable is
264 the terrain slope, because steep slopes shed sediments and flat valleys accumulate sediments
265 resulting in deep sedimentary basins. However, the correlation between SHB and grid cell slope
266 (mean or maximum from a 3" DEM, see Sect. 2.1) is statistically insignificant ($|r| < 0.04$). A main
267 reason is that terrain-driven runoff is most pronounced at the hillslope scales (from ridge to valley)
268 which is too small, and the runoff time too short, to be resolved by the $1^\circ \times 1^\circ$ and monthly scales
269 considered here.

270 The continental distribution of the horizontal, seasonal hydrologic buffer defined in Eq. (7) is
271 shown in Fig. 3c. A comparison with Fig. 3a and 3b suggests that lateral transfer is far less
272 important than climatic factors in driving SHB, except for the lower reaches of large river systems
273 such as the Amazon ($>3,600 \text{ mm.yr}^{-1}$) and the Nile, funneling inflow from large headwater basins
274 into smaller storage areas. To describe the broad patterns of likely source, sink, and passing-
275 through regions engaged in this slow lateral transfer that is able to contribute to the hydrologic
276 buffering of continents, we calculated the correlation between storage change (ΔS) and the net
277 inflow ($Q < 0$), and between ΔS and net outflow ($Q > 0$). A high correlation in the former indicates
278 that the grid cell is likely a sink, a high correlation in the latter indicates a likely source, while a
279 grid cell that is both sink and source suggests a likely passing-through location. Figure 5 shows
280 cells with correlation > 0.5 . Major source areas are located in boreal regions and northern mountain
281 ranges as well as selected areas of the per-humid tropics. Passing-through areas are the most

282 abundant and include boreal regions and large river valleys (Nile, Amazon, Ganges, Indus,
283 Zambezi, Mississippi). Sink areas are rarer and include a few deserts and semiarid areas in all
284 continents.

285

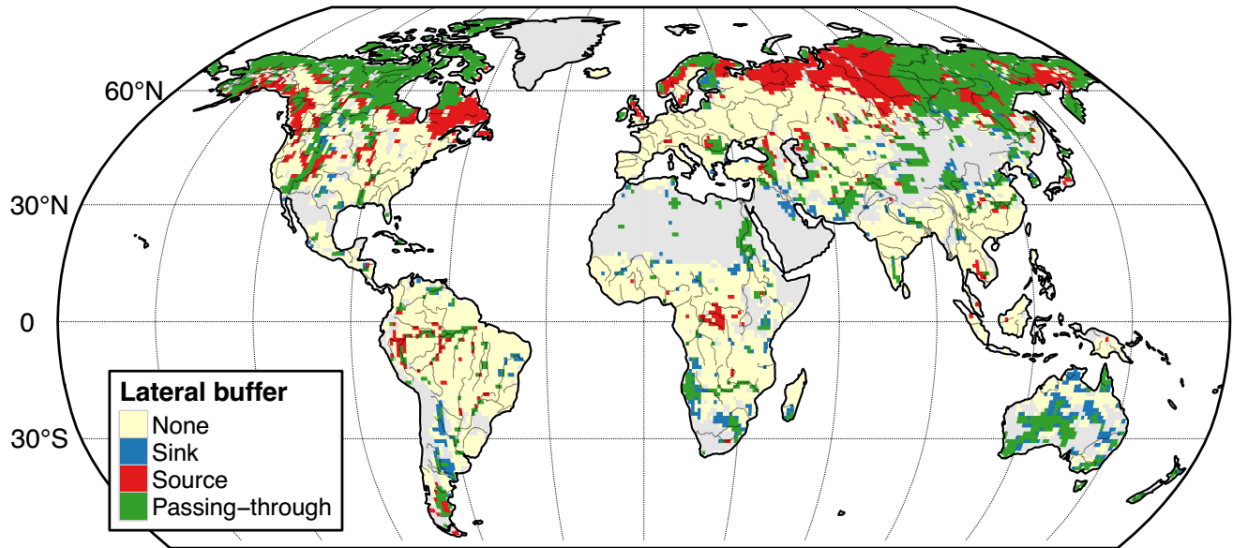


Figure 5. Locations where monthly changes in water storage are significantly correlated to monthly run-on (blue), monthly runoff (red) or both (green), excluding locations with low signal-to-noise ratio (light grey) – It highlights where seasonal hydrologic storage is significantly controlled by lateral inflow (“buffered sink”), outflow (“buffered source”), or both (“buffered passing-through”).

286

287 Finally, large man-made water reservoirs show their imprint at our coarse scale of analysis.
288 Figure 6 plots the total SHB vs. the storage capacity of 10 largest reservoirs of the world (in terms
289 of volume [22]), which shows a significant positive correlation, indicating that these reservoirs
290 have a detectable contribution to the GRACE-observed seasonal storage change.

291 These analyses suggest that climate, through seasonal water surplus/deficit imbalances and
292 snow accumulation is the first-order driver of seasonal hydrologic buffering. Terrain-driven lateral
293 transfer influences only a small portion of the land surface (yet intensely in specific regions), and
294 is only significant where/when the climatic surplus-deficit patterns align with terrain source-sink
295 drainage directions.

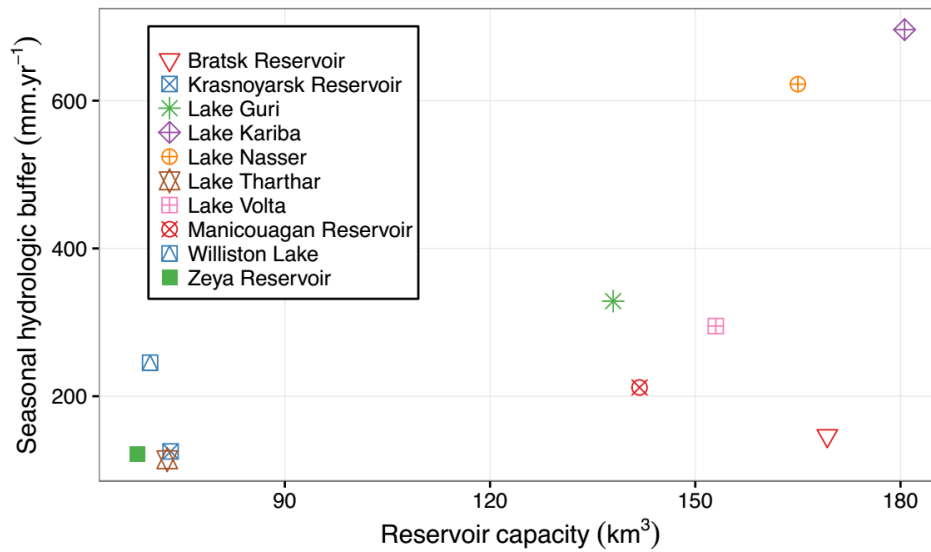


Figure 6. Vertically-integrated seasonal hydrologic buffer (SHB) against holding capacity, for the ten largest reservoir or artificial lakes.

297

298 3.3. Benefits for ecosystems

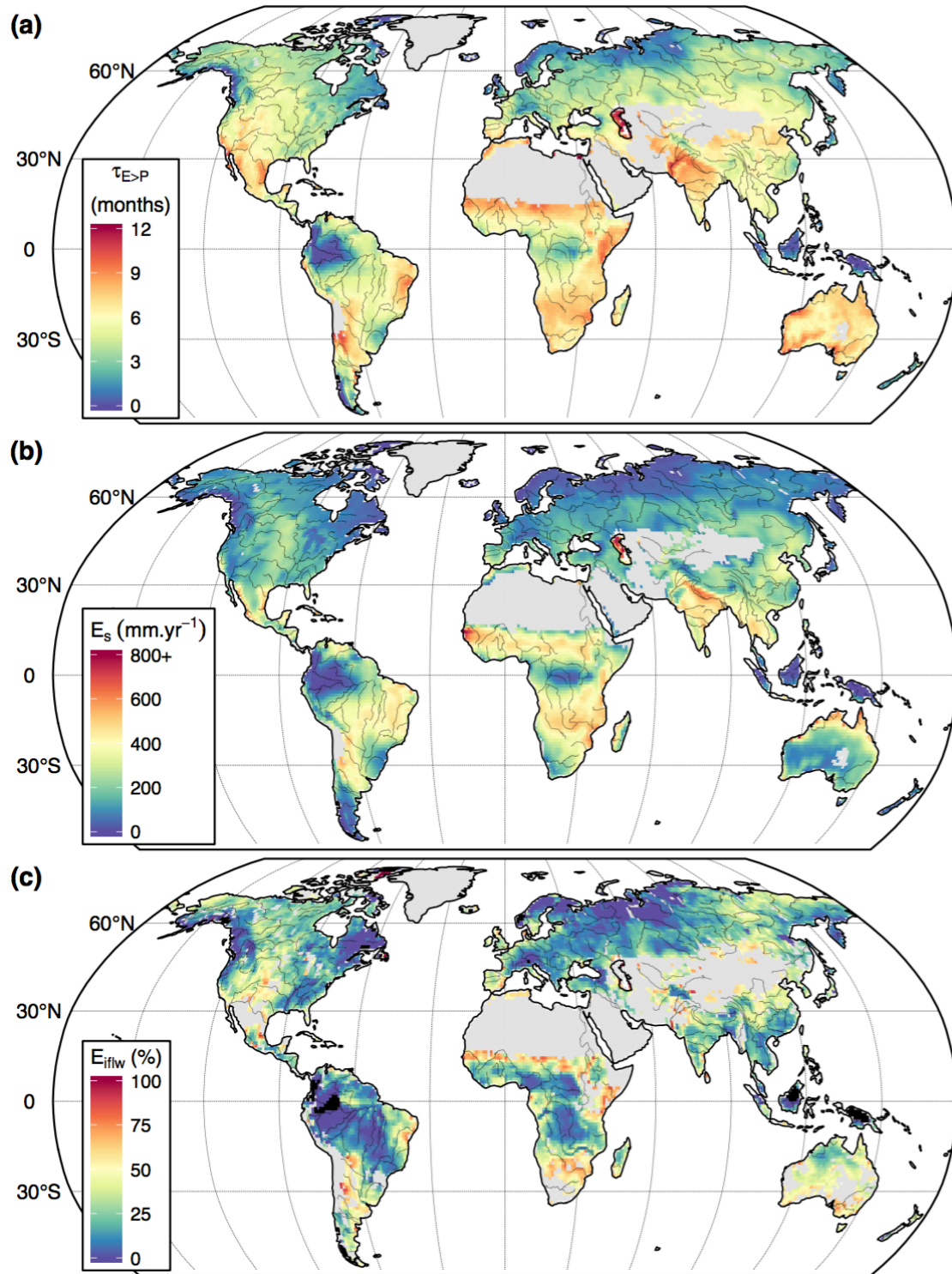
299 We identified the imprint of seasonal hydrological buffering on plant productivity (Question
 300 3) through the examination of both evaporation (E) and vegetation greenness (EVI) patterns. The
 301 most direct hydrologic connection to plant productivity is vegetation water use, or evaporation (E),
 302 since on vegetated surfaces the latter is dominated by plant transpiration [23]. A direct evidence
 303 of vegetation benefiting from SHB is the occurrence of monthly $E > \text{monthly } P$, where the deficit
 304 has to be met by temporal and/or spatial carry-over of past or neighbor's surplus, in addition to
 305 recent and local P. This was already defined as subsidized E (E_s , subsidized from past or upslope
 306 surplus, Eq. (8)).

307 We first documented the fraction of the year in which E_s takes place. Figure 7a plots the number
 308 of months per year where/when $E > P$, excluding grid cells with no vegetation. Except for areas that
 309 are wet and/or cool throughout the year, subsidized E occurs in at least 3 months of the year over
 310 most of the continental area, and in up to 9 months in strongly seasonal climates. The magnitude
 311 of the subsidized E over a year (E_s), is shown in Fig. 7b. While the continental area-weighted
 312 average is 190 mm.yr^{-1} , significant $E_s (> 400 \text{ mm.yr}^{-1})$ is found in the seasonal tropics where large
 313 wet-season rains are followed by a long dry-season, but plants continue to transpire in the year-

314 round warm climate. Here, if it was not for the subsidy, dry-season productivity would have been
315 far more water-limited. Interestingly this large E_s was noticed in both natural forests (e.g. the
316 Sudanian savanna in Africa) as well as in highly cultivated areas (e.g. Northern India) (Fig. 7b).
317 The actual water pool supporting E_s is typically assumed to be unsaturated soil moisture and in
318 order to achieve values beyond $600 \text{ mm}\cdot\text{yr}^{-1}$ that we observed in the regions mentioned above,
319 soils and roots reaching $>3\text{m}$ of depth would be required [15]. This is likely the case of natural
320 woody ecosystems in the dry-seasonal subtropics but very unlikely in heavily cultivated areas in
321 which annual crops display much shallower roots. In these systems, however, that saturated zone
322 gets engaged into seasonal discharge and recharge cycles that provide active buffering due to
323 irrigation and/or crops directly tapping phreatic groundwater.

324 The relative contribution of recent (<1 month) lateral inflow to subsidized E (E_{iflw} , see Sect.
325 2.2) is shown in Fig. 7c. E_{iflw} is high in the plains of southern South America that are known to be
326 fed by Andean streams [21], in the inland deltas of the Okavango, the Chad and the Niger in Africa,
327 and the internal drainage basins of Central Asia including the Aral Sea, all large regions nourished
328 by rivers sourced in the more humid headwater basins. The coarse spatial resolution of the study
329 ($1^\circ \times 1^\circ$ grids) cannot articulate lateral convergence at hillslope and catchment scales, and the
330 temporal resolution (month to month, within a year) cannot capture the quicker (flash flooding) or
331 slower (groundwater transfer) lateral fluxes, yet some of the best known and most salient features
332 of lateral water subsidy of the world stand out in this simple analysis. In these regions, societies
333 and ecosystems depend on rain or snow that fall in remote regions up-gradient for a large part of
334 their water supply.

335 Finally, we correlate the subsidized E directly to the MODIS satellite observed vegetation
336 productivity, the enhanced vegetation index (EVI), over the months of $E>P$, as shown in
337 supplementary Fig. S3. High correlations are found in strongly seasonal dry or frozen climates
338 (climatic drivers) and the low basins adjacent to large mountain ranges (terrain drivers).



339

Figure 7. (a) Average number of months where evaporation (E) exceeds precipitation and (b) Mean annual subsidized E (E_s , i.e. E-P when E>P) during these deficit months, excluding locations without vegetation cover. (c) Proportion of E_s attributed to recent (<1 month) lateral inputs (E_{iflw}), excluding locations with null SET and with low signal-to-noise ratio.

340

341 4. Conclusions

342 We examined storage and delayed release of monthly P, focusing on global geographic
343 patterns, climate and terrain drivers, and likely implications to terrestrial ecosystems. We found
344 that seasonal hydrologic buffering (SHB) is primarily driven by climatic water imbalance, with
345 SHB being most significant ($>300 \text{ mm.yr}^{-1}$) in tropical monsoonal climates followed by
346 Mediterranean climates and strongly seasonal (dry or cold) climates, carrying over wet (or cold)
347 season surplus to meet the following dry (or warm) season deficit (see Fig. 7b). Globally, about
348 25% of the annual P on land is held and released in later months.

349 Seasonal snow and ice accumulation is the second dominant driver of SHB, explaining the
350 large storage and delayed release in the melt and growing seasons. Lateral water transfer through
351 the rivers is the third dominant driver, but its importance depends on the climatic water balance:
352 seasonal lateral transfer is only important where/when the higher elevation basins have large
353 surplus, so that evaporative demand can be met by the subsidy from higher –and seasonally richer,
354 e.g. [24]– neighbors. Large reservoirs, heavily regulated by humans, is the fourth detectable and
355 regionally significant driver of SHB.

356 We find a direct evidence that SHB benefits land ecosystems and society, as seen in places and
357 times where/when E exceeds local and recent P, which occurs frequently, at 3-9 months of the year
358 (Fig. 7a), and widely, in all but per-humid and energy-limited regions of the world (Fig. 7b). This
359 illustrates the paramount importance of land water storage. Without it, vegetation productivity,
360 highest in the lower latitudes, would have been far more water-limited [25].

361 We note that, in addition to this direct impact of seasonal storage on E, the positive feedbacks
362 between E and P through continental precipitation recycling (e.g., [26]) may further amplify the
363 impact of this storage, so that more E leads to a moister and cooler boundary layer (less drought
364 stress), and sometimes more P locally or downwind, benefiting the ecosystems indirectly.

365 Neglected in this study are two important factors. The first is the role of vegetation as a
366 regulator of land water storage: through high E, it reduces the surplus of the wet month and
367 accentuate the deficit. While we chose simplicity to tease out the first-order cause-effect relations,
368 future works would benefit from more emerging observation-based global databases of root zone

369 storage capacity [18]. Secondly, we neglect anthropogenic influences, except by large reservoirs,
370 in the seasonal storage and release, directly by irrigation and drainage, and indirectly by changing
371 vegetation types which alters rooting depth and E dynamics.

372 Nonetheless, these results provide insights into the future of the water cycle through hydrologic
373 buffering by continents under climate and land use changes. Although snow storage can be shrunk
374 by global warming, water balance may partially compensate it if the seasonal decoupling of E and
375 P increases, as expected if increases in dry-season length in the tropics-subtropics take place
376 [27,28]. On the other hand, effects of land use appear less important at the scale of this study,
377 perhaps due to compensating effects (e.g. shorter crop roots in high SHB_{cwb} regions are
378 compensated by seasonal irrigation/recharge cycles that do not exist under the natural deep rooted
379 vegetation). Far from illuminating future trends, our work provides a first contribution by showing
380 where and how much continental hydrologic buffering influences the seasonal timing of the most
381 relevant water fluxes for humans, i.e., those sustaining vegetation growth and river runoff.

382

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388 the Web Applications Server of the European Center for Medium Range Weather Forecasting
389 (ECMWF), respectively at <http://apps.ecmwf.int/datasets/data/interim-land/> and
390 <http://apps.ecmwf.int/datasets/data/interim-full-daily/>, while potential evaporation of CRU
391 TS3.23 can be retrieved from http://www.cru.uea.ac.uk/cru/data/hrg/cru_ts_3.23/. The SRTM data
392 can be accessed through the Long Term Archive portal of USGS (<https://lta.cr.usgs.gov/SRTM>),
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