



1 HESS Opinions: Are soils overrated in hydrology?

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12 **Abstract**

13 Traditional "physically-based" hydrological models are based on the assumption that soil is key in
14 determining water's fate. According to these models, soil properties determine water movement in
15 both saturated and unsaturated zones, described by matrix flow formulas known as the Darcy-
16 Richards equations. Soil properties would also determine plant available moisture and thereby
17 control transpiration. These models are data demanding, computationally intensive, parameter rich
18 and, as we shall show, founded on a wrong assumption. Instead, we argue the reverse: it is the
19 movement of the water through a porous medium, creating preferential patterns, that determines
20 soil properties; while water movement is primarily controlled by the ecosystem's reaction to the
21 climatic drivers. According to this assumption, soil properties are a "consequence", rather than a
22 "cause" of water movement. It is not the soil that is in control of hydrology, it is the ecosystem. An
23 important and favourable consequence of this climate and ecosystem-driven approach is that
24 models developed with this approach do not require soil information, are computationally cheap,
25 and parsimonious. Our assumption is motivated by several arguments. Firstly, in well-developed
26 soils the dominant flow mechanism is preferential, which is not particularly related to soil
27 properties, such as pF curves. Secondly, we observe that it is the ecosystem, rather than the soil,



28 ~~that~~ determines the land-surface water balance and hydrological processes. Top-down analysis by
29 large-sample datasets reveal that soil properties are often a poor predictor of hydrological
30 signatures. Bottom-up hydrological models usually do not directly use field measurements of the
31 soil, but "rebalance" the observed soil texture and translate it to soil structure by vegetation
32 indices. Thus, soil-based models may be appropriate at small spatio-temporal scale in non-
33 vegetated and agricultural environments, but introduce unnecessary complexity and sub-optimal
34 results in catchments with permanent vegetation. Progress in hydrology largely relies on
35 abandoning the compartmentalized approach, putting ecosystem at the centre of hydrology, and
36 moving to a landscape-based modelling approach. This change in perspective is needed to build
37 more realistic and simpler hydrological models that go beyond current stationarity assumptions,
38 but instead regard catchments as the result of ecosystems' coevolution with atmosphere, biosphere,
39 hydrosphere, pedosphere, and lithosphere.

40

41 1 Introduction

42 Soil is important. It forms the substrate of the terrestrial ecosystem and hence is a crucial element
43 of the critical zone of life on Earth (Banwart et al., 2017). ~~Recently it has been demonstrated that~~
44 ~~the soil~~ forms an ecosystem in itself, full of micro-biotic and macrobiotic life (Ponge, 2015).
45 Fungi forming dense underground networks live in symbiosis with vegetation, exchanging
46 nutrients for carbon, ~~which makes~~ them responsible for the larger part of subterranean carbon
47 storage (Domeignoz-Horta et al., 2021). ~~Soils are full of life.~~ Above ground life cannot survive
48 without sub-surface life; they are both part of the same ecosystem. How this ecosystem functions,
49 evolves and survives depends on the climatic and geological boundary conditions. Through
50 evolution and natural selection, the ecosystem has found ways to make best use of the climatic and
51 geological resources. In doing so, it manipulates the substrate on which it grows. It manipulates
52 key hydrological characteristics such as: ~~interception capacity,~~ infiltration capacity, moisture
53 storage capacity, preferential pathways to replenish moisture ~~stocks and~~ recharge, and subsurface
54 drainage. ~~So,~~ soil is regarded as the key to hydrological processes (Lin et al., 2006). Yet, in (most



55 established) hydrological models, soils are simplified to soil classifications (Freeze and Harlan,
56 1969; Refsgaard et al., 2022), such as sandy, clayey, or loamy soils with their related soil
57 characteristics, such as porosity, moisture retention capacity, wilting point, plant available
58 moisture, etc., which are subsequently combined by the rooting depth of the dominant vegetation
59 (e.g. Drewniak, 2019, Lu et al., 2019). ~~But this is the wrong way round,~~

60 ~~Traditionally,~~ hydrologists assume that the root zone storage is a function of plant available
61 moisture and rooting depth (Yang et al., 2016). This approach considers plant available moisture
62 and rooting depth as the independent variables, and root zone storage as the dependent variable.
63 But it is the opposite. Plant available moisture and rooting depth are a function of root zone
64 storage, given some environmental constraints such as soil type and soil depth. Obtaining the
65 detailed spatio-temporal root and soil information at a global scale is virtually impossible (Or,
66 2019), but fortunately not necessary.

67 It is not the soil that determines the hydrology, it is the interaction between ecosystem and climate.
68 The climate provides the driver to the survival strategy of the vegetation. As a result, the
69 vegetation develops a rooting strategy, which results in a certain moisture buffering capacity (Gao
70 et al., 2014; Wang-Erlandsson et al., 2016). Depending on the soil, this buffer translates in a root
71 depth, extent, density, and possibly access to deeper groundwater (McCormick et al., 2021). So, for
72 hydrological modelling ~~it is not useful to measure soil characteristics and rooting depth.~~ We do not
73 need to know these dependent variables if we can determine root zone storage directly.

74 2 The soil-centred hydrological perspective

75 It is a deeply rooted perception in hydrology that soil is key in controlling rainfall-runoff
76 processes (Vereecken et al., 2022). Over a century ago, soil physicists developed the laboratory-
77 scale matrix flow theory, which is still widely used in so-called "physically-based" hydrological
78 and land surface models (LSMs) (Brutsaert et al., 2015; Xie et al., 2020). This theory describes
79 flow in porous media based on equations that depend on soil hydraulic properties (e.g. porosity,
80 hydraulic conductivity), which are typically associated to soil texture classes (e.g. clay, loam,
81 sand). Darcy's law describes matrix flow under saturated conditions through a porous medium



82 under a head gradient. Richards' equation regards matrix flow under unsaturated conditions in the
83 vadose zone, ~~determining water flow direction and velocity.~~ The matrix flow in most cases is
84 regarded as a well-established theory, much like classical mechanics. To simulate catchment
85 hydrology, matrix flow equations have been extended to distributed physically-based models,
86 cutting catchments into rectangles or triangle fragments, and using topography to drive soil water
87 movement vertically and laterally (Duffy, 1998; Refsgaard et al., 2022). ~~The water~~ infiltration and
88 vertical redistribution in most land surface models are also based on simplified Richards
89 equations, for example in diffusivity form (Vereecken et al., 2019). These models generally
90 require many space and soil-dependent parameters that require substantial calibration and
91 generally lead to **equifinality**.

92 However, ~~by~~ tracer field experiments, such as dye and isotope studies, matrix flow ~~has been~~
93 ~~scarcely~~ observed, simply because a well-prepared homogenous soil under lab conditions does not
94 exist in nature. Most soils contain crevices, ~~preferential~~ channels, and openings that transmit ~~free~~
95 water quite rapidly to the sub-surface, ~~which is~~ termed as preferential flow (Beven and Germann,
96 1982; McDonnell et al., 2007; Beven, 2018; Zehe et al., 2021).

97 To allow for preferential flows, modelers proposed **more complicated models**, which involved
98 even more ~~free~~ parameters to fit with observations. Unlike matrix flow, ~~which has clear~~
99 ~~description of water movement~~ in homogenous porous media, it is much more challenging to
100 describe preferential flow by succinct mathematic equations. Although there have been dual-
101 continuum, dual-porosity, or dual-permeability modifications (Jarvis et al., 2016), most models are
102 still based on matrix flow theory (Weiler, 2017). Because of the extreme complexity of soil
103 preferential flow in nature, it is extremely hard, if not impossible, to develop an accurate model to
104 describe and predict even microscale water movement, ~~the~~ challenge is exponentially greater
105 when we upscale preferential flow from spot-scale to hillslope or catchment scales (Davies et al.,
106 2013; Germann, 2014; Or, 2019). **However, this increase of model complexity, with more free**
107 **parameters to calibrate, did not show** a clear improvement in the simulation of soil moisture,
108 **groundwater head, and discharge (Glaser et al., 2019).**

109 From the beginning, preferential flow theory was regarded as the main culprit challenging the



110 foundation of physically-based hydrological models, but it did not help the model builder. The
111 focus on preferential flow appeared not to shed light on a solution, but rather made the fog even
112 denser.

113 Both matrix flow and preferential flow models are based on the *a priori* assumption that: soil
114 properties are key in describing hydrological processes. In this study, we question the general
115 validity of this assumption, and ask whether thinking outside of the box of the soil-based
116 modelling framework could promote alternative, perhaps simpler theories, that help explain the
117 observed patterns of hydrological variability.

118 3 Does soil variability matter in catchment hydrology?

119 3.1 Does soil influence the long-term water budget

120 Globally, based on ERA-5 data (the fifth generation of ECMWF atmospheric reanalysis), annual
121 average terrestrial precipitation (P) is 745 mm/yr (111 km³/yr), of which 440 mm/yr (65.5 km³/yr,
122 60% of P) returns to the atmosphere as evaporation (E), and 305 mm/yr (45.5 km³/yr, 40% of P)
123 generates runoff (Q) (Figure 1). It is clear that evaporation rather than runoff is the dominant flux
124 after precipitation.

125 For terrestrial evaporation, the total global land area is 145 million km², of which 104 million km²
126 (72%) is vegetated. Of total evaporation (E_{total}), soil evaporation only happens in the topsoil, with
127 a limited amount, only 6% of E_{total} (Good et al., 2015). As evaporation from vegetation and ground
128 interception accounts for 27%, transpiration is the largest flux at about 60-70% of E_{total} (Coenders-
129 Gerrits et al., 2014; Lian et al., 2018). An increasing number of studies confirmed that landuse
130 change, e.g. deforestation and afforestation, significantly alter runoff generation (Fenicia et al.,
131 2009; Nijzink et al., 2016; Sun et al., 2020). Thus, vegetation, and more correctly the ecosystem,
132 rather than soil, is the determining factor of the global and catchment water balance.

133 3.2 Do soil-centred models reproduce observed patterns of 134 hydrological variability?

135 In reductionist thinking the lead paradigm is that “the whole is the sum of the parts” (Sivapalan et
136 al., 2003). Bottom-up models are developed on the basis of small laboratory-scale laws, where the



137 assumption is that microscale laws remain valid after upscaling. Due to the difficult-to-measure
138 soil hydraulic properties (SHPs), such as the water retention characteristics, unsaturated and
139 saturated hydraulic conductivity, modelers have used soil pedotransfer functions (PTFs) to
140 ~~correlate~~ readily available soil properties, such as soil texture (i.e. sand-, silt-, clay- content),
141 organic matter, and bulk density ~~with SHPs~~ (van Looy et al., 2017; Or, 2019).

142 There are several critical issues with this approach, which result in inevitable biases: 1) most soil
143 ~~property~~ parameters are measured ~~by~~ pedologic surveys, at great expense and efforts (Van Looy et
144 al., 2017). Due to limited field measurement, most soil property maps rely on uncertain
145 interpolations and upscaling, which are based on soil-forming factors such as climate, parent
146 material, topography, biota, and time (van Looy, 2017); 2) PTFs are usually obtained by using
147 measurements from ~~uniform~~ soil samples, and performed in laboratory-scale experiments, which
148 merely reflect disturbed and therefore unnatural conditions; 3) ~~the parameters obtained at the~~
149 ~~laboratory scale are not necessarily the same at the modelled scale, which requires upscaling~~
150 ~~assumptions, which are difficult to verify, or recalibration, which is hampered by equifinality;~~ 4)
151 as emerging properties are scale-dependent, the processes that are important at lab or plot scale are
152 not necessarily important at catchment scale (Blöschl and Sivapalan, 1995).

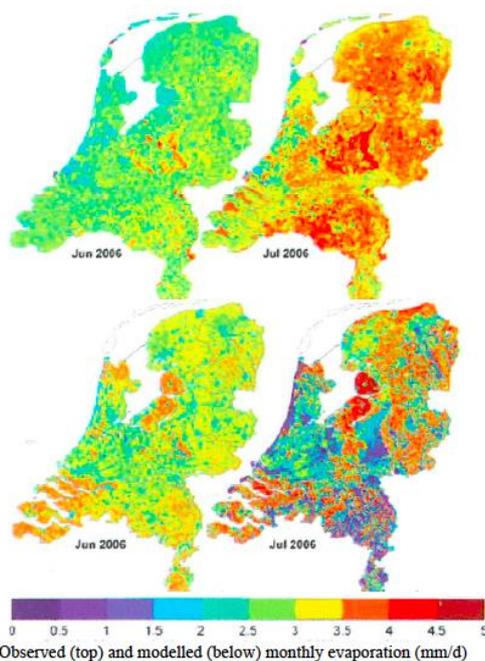
153 Figure 1. is a nice illustration of how a "physically based" model, making use of detailed soil
154 information, can make completely wrong predictions under extreme circumstances (Beekman et
155 al., 2014). On average, these models may function adequately, as almost all hydrological models
156 do under average conditions, but ~~the example of Figure 1. shows how~~ during a relatively extreme
157 drought ~~in The Netherlands~~ the modelled evaporation diverts considerably from ~~observations~~.

158 ~~The observed evaporation was obtained by~~ interpolation of eddy-covariance and lysimeter
159 observations using ETLook, ~~an energy balance-based evaporation product~~ (Bastiaanssen et al.,
160 2012). The modelled evaporation was generated by the ~~dominant~~ Dutch hydrological model (NHI)
161 which heavily relies on detailed soil maps and soil observations. The June 2006 picture in the
162 lower part, in fact, ~~mimics the soil map~~. Red (high evaporation) ~~is seen~~ on clay soils and purple
163 (almost no evaporation) on sand. ~~It is funny to see that in~~ reality (top right) the forested sandy part
164 at the centre of The Netherlands was ~~evaporating lushly~~, whereas in the model (bottom right), this



165 ecosystem appeared to be dead. In fact, all ecosystems continued to evaporate well during this dry
166 month. Apparently, the ecosystems had prepared for this eventuality and had created enough root
167 zone buffer to overcome this period of drought.

168 Although such mismatch between distributed model outputs and observed patterns are not
169 infrequent, they are typically not regarded as a challenge to the basic model assumptions, but
170 rather, as a problem associated to the uncertainty in model inputs. Hence, such soil-centred
171 hydrological models remain vivid under the hope that “novel, highly resolved soil information at
172 higher resolutions than the grid scale of LSMs may help in better quantifying sub-grid variability
173 of key infiltration parameters” (Vereecken et al., 2022).



174 Observed (top) and modelled (below) monthly evaporation (mm/d)
175 Figure 1. Evaporation during June (left) and July (right) of 2006 in The Netherlands. Modelled
176 below and observed on top (from Beekman et al., 2014)

177 3.3 Is soil a good predictor for hydrological signatures?

178 The top-down approach, as a genuine way of learning from data, is amenable to systematic
179 learning and hypothesis testing (Sivapalan et al., 2003), which is more suitable to understand the



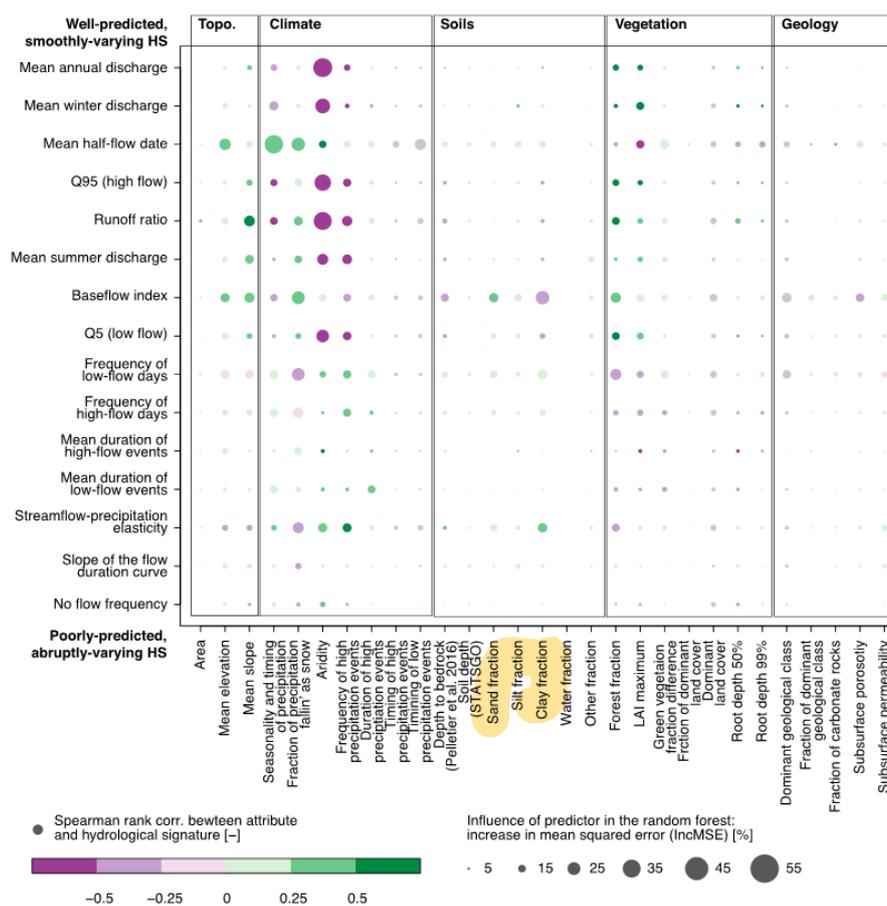
180 complex hydrological system. There are many predictors for hydrological signatures, including
181 climate, vegetation, topography, geology and soil. Interestingly, in catchment hydrology, soil
182 properties are often a poor predictor of hydrological signatures. Addor et al., (2018) used 671
183 catchments in the USA and found that, compared to soil, landscape features, i.e. vegetation and
184 topography, have stronger correlations with hydrologic signatures, not only for mean runoff, but
185 also for high-flow, low-flow, and runoff seasonality (Figure 2).

186 The soil may be important for small scale (agricultural) hydrology, but this importance is easily
187 overridden by other factors at larger scale, for example climate, topography, vegetation, and
188 geology. All these factors are intertwined, and result in preferential flow not only in the soil
189 profile, but for all hydrological processes at multi-scales (Uhlenbrook, 2006), ranging from
190 fingering, macropores, forest canopy, leaves, roots, fractures, rills and gullies, all the way to the
191 river network, and even in the groundwater system (Savenije, 2018). These preferential flow
192 patterns have great influence on all hydrological processes, including infiltration, percolation,
193 groundwater discharge, and river channel routing (Rodrigues-Iturbe and Rinaldo, 1997; Sivapalan,
194 2006; Savenije and Hrachowitz, 2017). The infiltration capacity, in the absence of vegetation, may
195 be a determining factor for infiltration excess overland runoff (Hortonian flow), but this scarcely
196 occurs in natural vegetated landscapes where the ecosystem facilitates infiltration, even in arid
197 regions (Zhao et al., 2019). In mature ecosystems, runoff is dominated by subsurface flow, which
198 is a storage-discharge relationship. The preferential flow path, as the connector between small-
199 and large-scale observations, makes the surface, rootzone and ground water systems react to
200 rainfall as an integrated fill-and-spill system (Tromp-van Meerveld and McDonnell, 2006). This
201 allows modellers to merely focus on input-output and storage change, to reproduce the complex
202 rainfall-runoff processes through the heterogeneous but fractal patterned surface and subsurface.

203 Interestingly, in practice, simple catchment hydrological models, even without explicitly
204 considering detailed matrix flow or preferential flow, can reproduce complex rainfall-runoff
205 processes, with surprisingly good performance. The reliability and robustness of the widely used
206 HBV (Seibert, 2018), GR4J (Perrin et al., 2001), Xinanjiang (Zhao, 1992), and FLEX (Fenicia et
207 al., 2011; Gao et al., 2022), have been tested and validated in a wide range of regional and global



208 hydrological studies (Bergström and Lindström, 2015; Mao and Liu, 2019; Wang et al., 2021).
 209 There are also examples of distributed models that describe the observed variability of
 210 hydrological signatures without relying on soil information (e.g. Boer-Euser et al., 2016; Fenicia
 211 et al., 2016; Gao et al., 2019; Dal Molin et al., 2020; Fenicia et al., 2022).



212
 213 Figure 2. Comparison of the influence of catchments attributes and hydrological signatures for 671
 214 U.S. watersheds (from Addor et al., 2018). Large, brightly coloured circles imply strong
 215 correlations and high influence. (Reprinted by permission of John Wiley and Sons).



216 4 Putting the ecosystem at the centre of hydrology

217 4.1 The ecosystem is the ultimate water manager

218 ~~Engineering~~ hydrologists have a long tradition to focus on runoff generation, which represents at
219 the same time the most valuable water resources and the cause of water related disasters, i.e.
220 floods and droughts. There is limited realisation that runoff is merely the ecosystem's by-product
221 of water use (Bonan, 2015), only accounting for 40% of the precipitation, while the ecosystem
222 ~~evaporates~~ 60% of precipitation back to the atmosphere in its drive to grow and survive.

223 A natural ecosystem can only survive if it ~~organizes~~ its resilience. This resilience requires not only
224 sufficient moisture storage in the root zone, to overcome critical dry spells, but also sufficient
225 infiltration capacity and subsurface drainage to maintain moisture levels between acceptable
226 boundaries: not too wet and not too dry. If given sufficient time, the ecosystem will improve the
227 infiltration capacity both to replenish water in the root zone and to prevent runoff; seldom do we
228 observe surface runoff in a natural environment. The ecosystem creates preferential flow paths
229 that facilitate infiltration in order to retain nutrients and soil particles and to direct rainfall to
230 where it is most needed. If there is too much infiltration, then sub-surface drainage by preferential
231 ~~recharge and sub-surface drainage patterns~~ on hillslopes evacuates excess water. Preferential flow
232 rapidly delivers the excess water, bypassing the rootzone, to the groundwater system (McDonnell,
233 2014). By wave propagation the groundwater head created by the recharge pushes pre-event
234 groundwater to the streams (Kirchner et al., 2003; Hrachowitz et al., 2016; Worthington, 2019)
235 also through preferential flow patterns (Savenije, 2018). In summary, prevailing preferential flow
236 paths, mostly as a result of a variety of biological activities, determine rainfall partitioning and
237 runoff generation, ~~moving~~ beyond plot-scale heterogeneity and process complexity, where
238 landscape and ecosystem are the determining factors ~~rather than the soil~~.

239 **Modelers** utilizing a bottom-up approach also noticed that the simple paradigm of using soil
240 texture as a main predictor of SHPs is problematic (Bonetti et al., 2021). Gutmann and Small
241 (2007) have shown that soil textural classes, across a range of climates and vegetation covers,
242 merely explained 5% of the variance obtained ~~from~~ the real SHPs. It was found that 95% of the
243 default soil parameters in a state-of-the-art landsurface model were significantly different from



244 region-specific observations (Kishné et al., 2017).

245 With the biased observation and the scale effect, modelers need to recalibrate numerous free
246 parameters. There are studies that "rebalance" the soil texture information and highlight the
247 importance of soil structure, originated by soil biophysical activity (Or, 2019; Fatichi et al., 2020).
248 These models do not directly use field measurements of the soil, but "rebalance" the observed soil
249 texture and translate it to soil structure by vegetation indices, e.g. the aboveground vegetation
250 biomass and leaf area index (LAI) (Bonetti et al., 2021). Involving vegetation cover and
251 productivity prominently improved the hydrologic response with respect to runoff generation,
252 infiltration and drainage (Or, 2019). Not ~~only~~ the physically-based models, but also the empirical
253 soil-based models, for example the soil conservation service (SCS) method in SWAT model (soil
254 water assessment tool), also ~~involve~~ landuse data to "rebalance" the soil-based curve number in
255 catchment simulation (Arnold et al., 2012).

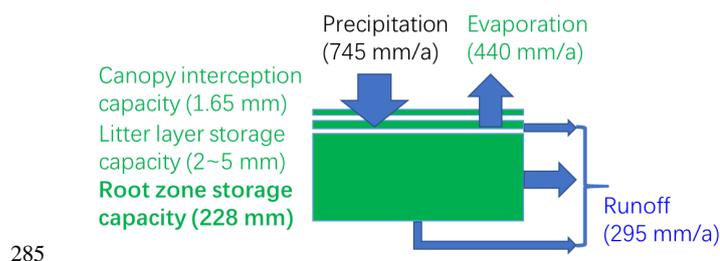
256 **4.2 The rootzone is the key element in hydrology ~~not the soil~~**

257 Vegetation as a living organism optimizes its rootzone water storage to stabilize water supply at
258 minimal carbon cost, ~~and sometimes access to the deeper groundwater,~~ to cater for periods of
259 drought (Gao et al., 2014; McCormick et al., 2021). From a hydrology perspective, the vertical
260 profile of the critical zone can be divided into different layers, i.e.: canopy, litter layer, root zone,
261 water transition zone, unconfined groundwater, and confined groundwater. The most active layers
262 in the critical zone defining rainfall partitioning and related hydrological processes include: the
263 canopy, litter layer, and root zone. Globally, the vegetation interception storage capacity of
264 terrestrial ecosystems is about 1-2 mm, estimated by remotes sensing LAI data (De Roo et al.,
265 1996). The litter layer storage capacity differs in different ecosystems, but it is likely to increase
266 the total interception storage capacity to around 2-5 mm (Shi et al., 2004). **For determining the**
267 **rootzone storage capacity of ecosystems, the mass curve technique appeared to work well locally**
268 **and globally (Gao et al., 2014; Wang-Erlandsson et al., 2016) whereby use was made of ERA-5**
269 **evaporation data to estimate, resulting in a global average root zone storage capacity of 228 mm**
270 **(Figure 3).** Hence, rootzone storage capacity is significantly larger than interception and litter
271 layer water storage capacities. Beneath the rootzone, water in the transition zone between root



272 zone and groundwater ~~and the unconfined groundwater does not have phase changes~~, and will be
273 discharged as runoff under natural condition. The deep confined groundwater has a very long
274 transit time, without much connection to surface processes, and is not considered in surface runoff
275 calculation in most cases. Hence, the root zone storage plays ~~the~~ dominant role in catchment
276 hydrology, determining how catchments respond to rainfall.

277 **It** is worthwhile noting that rootzone and soil have a strong connection, but are essentially
278 different things. The soil profile can reach over hundreds of meters' depth, e.g. the Loess Plateau
279 in China (Zhang et al., 2014), of which only the rootzone is the active ~~area, whereby the soil is~~
280 ~~merely the substrate of it~~. Rootzone storage can also be larger than soil water storage, for example
281 in karst mountainous areas where soil is thin and discontinuous, bedrock storage serves as an
282 important source of plant-available water (McCormick et al., 2021). Rootzone water storage is a
283 more accurate term ~~in hydrology~~, which cannot be measured directly, but with clear physical
284 meaning, similar to the runoff depth and runoff efficiency.



286 Figure 3. The importance of root zone storage capacity in catchment hydrology. All the numbers
287 are global average values ~~in~~ vegetated area, obtained by ERA-5 data from 1980 to 2020.

288 4.3 Landscape-based model: the giant view of hydrology

289 A soil-based model is like an ant's perspective, observing a complex world of heterogeneities and
290 randomness. Landscape, as the integration of topography and landcover, being the long-term co-
291 evolution of ecosystem, atmosphere, lithosphere, pedosphere, hydrosphere, and human activities
292 (Mücher et al., 2010; Wu, 2013; Troch et al., 2015) is rather the giant's perspective (Savenije,
293 2010). The pattern of hillslope, landscape and catchment only becomes visible when we zoom out
294 well beyond the microscale of the soil or the human scale, for that matter.



295 At landscape scale, hydrological laws can be simple: the fill and spill bucket model with
296 thresholds and associated time scales and the linear reservoir for groundwater are physically-
297 sound at hillslope and catchment scale (Savenije, 2010; Fenicia et al., 2011; Savenije and
298 Hrachowitz, 2017). Landscape-based models also use conservation of mass and energy, but
299 momentum transfer is made of conceptual relations that relate discharge to storage using
300 landscape and ecosystem characteristics, such as travel times and a probability function for runoff
301 thresholds. Geology as the substrates of hydrological processes also plays an important role on
302 slow processes. Interestingly, the underground hydrogeology system seems also self-organized
303 and structured (Savenije, 2018). For example, the recession parameter k is around 43 days in
304 worldwide catchments regardless of their climate, topography, soil, and geology (Brutsaert and
305 Sugita, 2008). Discovering these properties and related signatures benefit our understanding and
306 prediction of the dynamic adaption of ecosystems to environmental change, and the subsequent
307 impacts on hydrology (Gharari et al., 2014).

308 Explosive growth of remote sensing technology and powerful geographic information system
309 (GIS) toolkits provide unprecedented opportunities, with a giant's view, to closely read landscapes
310 and monitor their hydrological components in high spatial and temporal resolutions (Duan et al.,
311 2021; Crow et al. 2022). Various measurement techniques, e.g. eddy covariance, make it possible
312 to quantify the mass and energy budget at landscape scale (Jung et al., 2020; Lan et al., 2021). The
313 GRACE satellite provides independent estimates of moisture storage on Earth (Famiglietti et al.,
314 2015; Dong et al., 2022). These datasets form crucial complementary information to constrain
315 landscape-based model parameters, validate the internal fluxes, and eventually give more
316 physically-based and reliable simulation (Winsemius et al., 2006; Nijzink et al., 2018; Gao et al.,
317 2020).

318 **5 Why is the soil-based modelling tradition so rooted in** 319 **hydrology?**

320 Occam's razor tells us "Entities are not to be multiplied beyond necessity". This principle gives
321 precedence to simplicity: of two competing theories, the simpler explanation of an entity is to be



322 preferred. Therefore, if soil does not play a significant role neither in the water balance nor in
323 rainfall-runoff processes, and if disregarding soil does not significantly affect model performance,
324 there is sufficient and fair reason to exclude soil properties in hydrological modelling. But then:
325 why are soil-based models so rooted in hydrology?

326 **5.1 Agricultural bias**

327 Since hydrology was born from chapters of agricultural and hydraulics textbooks (Rodríguez-
328 Iturbe and Rinaldo, 2004), the misunderstanding that soils are the determining factor in hydrology
329 is probably caused by the “agricultural bias”. In agriculture, the focus is on seasonal crops. A
330 seasonal crop has limited time to develop a root zone storage that can buffer for longer term
331 variability. At best, it can buffer for average dry spells that may occur within an average year. This
332 is why modern agriculture requires water management by the farmer to buffer for natural
333 fluctuations. In agriculture, ploughing destroys preferential infiltration and sub-surface drainage. It
334 also limits the root zone storage capacity to the relatively small soil layer above the plough pan. In
335 such cases it is indeed the moisture holding capacity of the soil that determines the root zone
336 storage capacity.

337 As regards evaporation, it occurs in many ways: by transpiration, direct evaporation from leaf and
338 ground interception, bare soil evaporation, open water evaporation, sublimation, ~~and in general~~
339 ~~from all phase changes where liquid or solid water turns into vapour.~~ All these mechanisms obey
340 different constraints and need to be modelled separately to arrive at a correct representation of
341 total evaporation. In that respect, the dominant widely used Penman-Monteith equation, so
342 successful in agriculture, is likely not appropriate to describe land-atmosphere **interaction.**
343 Unfortunately, this “agricultural bias” has been dominant in most hydrological work. We argue
344 that this deeply rooted soil-based perception may limit or even mislead the further development of
345 hydrological science, especially for next generation professionals.

346 **5.2 Unreliable intuition**

347 Hydrologists intuitively **concern** about soil. ~~It could very well~~ **by the perspective of the ant that**
348 **causes it.** As people, we are just too small to observe the larger scale processes that dominate
349 landscape hydrology. We tend to dig holes in the Earth and try to infer larger scale behaviour from



350 what we observe inside this hole. The human scale prevents us from seeing the larger picture. We
351 need the giant's perspective to recognise the patterns present in the landscape.

352 At the human scale, assuming that soil properties, such as texture and porosity, matter makes
353 intuitive sense. People tend to describe what they see (i.e. the consequence) rather than the cause,
354 and if they see water flowing or disappearing in the ground they think that it is because of such
355 soil properties. The role of the ecosystem as the driver is much more difficult to recognize,
356 because it is hidden to the eye. It requires seeing the environment as a living thing, which
357 continuously evolves and adjusts to changing circumstances. It also implies that the hydrological
358 properties are not constant over time. The root zone storage, the most critical control on rainfall-
359 runoff processes, is continuously changing in response to changing climatic and human drivers.
360 Instead of describing the 'now' as an invariant and static condition, with environmental properties
361 ~~as a given~~, one has to think of the history that determined these environmental conditions, which
362 much more difficult to realise.

363 6 Conclusion

364 ~~Hydrology is the blood stream of the ecosystem.~~ Runoff is merely the excess water after
365 precipitation has replenished ecosystem's water deficit. Traditional "physically-based" models put
366 soil physical properties at the heart of hydrology which is misleading for both process
367 understanding and model development. In contrast we need an ecosystem-centred approach.
368 ~~Nowadays there is much concern about the Critical Zone of the Earth, where virtually all life takes~~
369 ~~place.~~ The ecosystem is the ultimate manager of the critical zone, with hydrology as its blood
370 stream. We advocate a paradigm shift to put ecosystem and landscape at the heart of terrestrial
371 hydrology, and develop holistic and ~~live~~ ecosystem-based hydrological models, with better
372 representation of hydrological processes in Earth system science.

373

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377 **References**

- 378 Addor, N., Nearing, G., Prieto, C., Newman, A. J., Le Vine, N., and Clark, M. P.: A ranking of
379 hydrological signatures based on their predictability in space, *Water Resour. Res.*, 54, 8792–8812,
380 <https://doi.org/10.1029/2018WR022606>, 2018.
- 381 Arnold, J., Moriasi, D., Gassman, P., Abbaspour, K., White, M., Srinivasan, R., Santhi, C.,
382 Harmel, R., van Griensven, A., Van Liew, M., Kannan, N., and Jha, M.: SWAT: Model use,
383 calibration, and validation, *Trans. ASABE*, 55, 1491–1508, <https://doi.org/10.13031/2013.42256>,
384 2012.
- 385 Banwart, S., Bernasconi, S., Blum, W., Souza, D. M. de, Chabaux, F., Duffy, C., Kercheva, M.,
386 Krám, P., Lair, G. J., Lundin, L., Menon, M., Nikolaidis, N., Novak, M., Panagos, P.,
387 Ragnarsdóttir, K., Robinson, D., Rousseva, S., Ruiten, P., Gaans, P., and Zhang, B.: Soil functions
388 in Earth's critical zone: Key results and conclusions, *Adv. Agron.*, 142, 1–27,
389 <https://doi.org/10.1016/bs.agron.2016.11.001>, 2017.
- 390 Beekman, W., Caljé, R., Schaars, F., and Heijkers, J.: Vergelijking van enkele schattingsmethoden
391 voor de actuele verdamping (Comparison between several methods to estimate actual
392 evaporation), *Stromingen*, 20, 39–46, 2014.
- 393 Bergström, S. and Lindström, G.: Interpretation of runoff processes in hydrological modelling—
394 experience from the HBV approach, *Hydrol. Process.*, 29, 3535–3545,
395 <https://doi.org/10.1002/hyp.10510>, 2015.
- 396 Beven, K. and Germann, P.: Macropores and water flow in soils, *Water Resour. Res.*, 18, 1311–
397 1325, <https://doi.org/10.1029/WR018i005p01311>, 1982.
- 398 Beven, K.: A century of denial: Preferential and nonequilibrium water flow in soils, 1864–1984.,
399 *Vadose Zone J.*, 17, 180153, <https://doi.org/10.2136/vzj2018.08.0153>, 2018.
- 400 Blöschl, G. and Sivapalan, M.: Scale issues in hydrological modelling: A review, *Hydrol. Process.*,
401 9, 251–290, <https://doi.org/10.1002/hyp.3360090305>, 1995.
- 402 Bonan, G. B. *Ecological climatology concepts and applications*, 3rd edition, Cambridge
403 University Press, <https://doi.org/10.1017/CBO9781107339200>, 2015.
- 404 Bonetti, S., Wei, Z., and Or, D.: A framework for quantifying hydrologic effects of soil structure
405 across scales, *Commun. Earth Environ.*, 2, 107, <https://doi.org/10.1038/s43247-021-00180-0>,
406 2021.



-
- 407 Bouma, J., Bonfante, A., Basile, A., van Tol, J., Hack-ten Broeke, M. J. D., Mulder, M., Heinen,
408 M., Rossiter, D. G., Poggio, L., and Hirmas, D. R.: How can pedology and soil classification
409 contribute towards sustainable development as a data source and information carrier?, *Geoderma*,
410 424, 115988, <https://doi.org/10.1016/j.geoderma.2022.115988>, 2022.
- 411 Brutsaert, W. and Hiyama, T.: The determination of permafrost thawing trends from long-term
412 streamflow measurements with an application in eastern Siberia, *J. Geophys. Res.-Atmos.*, 117, 1–
413 10, <https://doi.org/10.1029/2012JD018344>, 2012.
- 414 Brutsaert, W. *Hydrology: An Introduction*, Cambridge University Press, [https://doi.org/](https://doi.org/10.1017/CBO9780511808470)
415 [10.1017/CBO9780511808470](https://doi.org/10.1017/CBO9780511808470), 2005.
- 416 Coenders-Gerrits, M., Van der Ent, R., Bogaard, T., Wang-Erlandsson, L., Hrachowitz, M., and
417 Savenije, H.: Uncertainties in transpiration estimates, *Nature*, 506, E1-2,
418 <https://doi.org/10.1038/nature12925>, 2014.
- 419 Crow, W. T., Dong, J., and Reichle, R. H.: Leveraging pre-storm soil moisture estimates for
420 enhanced land surface model calibration in ungauged hydrologic basins, *Water Resour. Res.*, 58,
421 e2021WR031565, <https://doi.org/10.1029/2021WR031565>, 2022.
- 422 Dal Molin, M., Schirmer, M., Zappa, M., and Fenicia, F.: Understanding dominant controls on
423 streamflow spatial variability to set up a semi-distributed hydrological model: the case study of the
424 Thur catchment, *Hydrol. Earth Syst. Sci.*, 24, 1319–1345, [https://doi.org/10.5194/hess-24-1319-](https://doi.org/10.5194/hess-24-1319-2020)
425 [2020](https://doi.org/10.5194/hess-24-1319-2020), 2020.
- 426 Davies, J., Beven, K., Rodhe, A., Nyberg, L., and Bishop, K.: Integrated modeling of flow and
427 residence times at the catchment scale with multiple interacting pathways, *Water Resour. Res.*, 49,
428 4738–4750, <https://doi.org/10.1002/wrcr.20377>, 2013.
- 429 de Boer-Euser, T., McMillan, H. K., Hrachowitz, M., Winsemius, H. C., and Savenije, H. H. G.:
430 Influence of soil and climate on root zone storage capacity, *Water Resour. Res.*, 52, 2009–2024,
431 <https://doi.org/10.1002/2015WR018115>, 2016.
- 432 De Roo, A. P. J., Wesseling, C. G., and Ritsema, C. J: LISEM: A Single-event Physically Based
433 Hydrological and Soil Erosion Model for Drainage Basins. I: Theory, Input and Output, *Hydrol.*
434 *Process.*, 10, 1107–1117, [https://doi.org/10.1002/\(SICI\)1099-1085\(199608\)10:8<1107::AID-](https://doi.org/10.1002/(SICI)1099-1085(199608)10:8<1107::AID-HYP415>3.0.CO;2-4)
435 [HYP415>3.0.CO;2-4](https://doi.org/10.1002/(SICI)1099-1085(199608)10:8<1107::AID-HYP415>3.0.CO;2-4), 1996.
- 436 Domeignoz-Horta, L. A., Shinfuku, M., Junier, P., Poirier, S., Verrecchia, E., Sebag, D., and
437 DeAngelis, K. M.: Direct evidence for the role of microbial community composition in the
438 formation of soil organic matter composition and persistence, *ISME Commun.*, 1, 64,
439 <https://doi.org/10.1038/s43705-021-00071-7>, 2021.



-
- 440 Dong, N., Wei, J., Yang, M., Yan, D., Yang, C., Gao, H., Arnault, J., Laux, P., Zhang, X., Liu, Y.,
441 Niu, J., Wang, H., Wang, H., Kunstmann, H., and Yu, Z.: Model estimates of China's terrestrial
442 water storage variation due to reservoir operation, *Water Resour. Res.*, 58, e2021WR031787,
443 <https://doi.org/10.1029/2021WR031787>, 2022.
- 444 Drewniak, B. A.: Simulating dynamic roots in the energy Exascale Earth system land model, *J.*
445 *Adv. Model. Earth Syst.*, 11, 338–359, <https://doi.org/10.1029/2018MS001334>, 2019.
- 446 Duan, Z., Duggan, E., Chen, C., Gao, H., Dong, J., and Liu, J.: Comparison of traditional method
447 and triple collocation analysis for evaluation of multiple gridded precipitation products across
448 Germany, *J. Hydrometeorol.*, 22, 2983–2999, <https://doi.org/10.1175/JHM-D-21-0049.1>, 2021.
- 449 Duffy, C.: A two-state integral-balance model for soil moisture and groundwater dynamics in
450 complex terrain, *Water Resour. Res.*, 32, 2421–2434, <https://doi.org/10.1029/96WR01049>, 1996.
- 451 Famiglietti, J. S., Cazenave, A., Eicker, A., Reager, J. T., Rodell, M., and Velicogna, I.: Satellites
452 provide the big picture, *Science*, 349, 684–685, <https://doi.org/10.1126/science.aac9238>, 2015.
- 453 Fatichi, S., Or, D., Walko, R., Vereecken, H., Young, M. H., Ghezzehei, T. A., Hengl, T., Kollet, S.,
454 Agam, N., and Avissar, R.: Soil structure is an important omission in Earth System Models, *Nat.*
455 *Commun.*, 11, 522, <https://doi.org/10.1038/s41467-020-14411-z>, 2020.
- 456 Fenicia, F., Kavetski, D., and Savenije, H. H. G.: Elements of a flexible approach for conceptual
457 hydrological modeling: 1. Motivation and theoretical development, *Water Resour. Res.*, 47,
458 W11510, <https://doi.org/10.1029/2010WR010174>, 2011.
- 459 Fenicia, F., Kavetski, D., Savenije, H. H. G., and Pfister, L.: From spatially variable streamflow to
460 distributed hydrological models: Analysis of key modeling decisions, *Water Resour. Res.*, 52,
461 954–989, <https://doi.org/10.1002/2015WR017398>, 2016.
- 462 Fenicia, F., Meissner, D., and McDonnell, J. J.: Modeling streamflow variability at the regional
463 scale: (2) Development of a bespoke distributed conceptual model, *J. Hydrol.*, 605, 127286,
464 <https://doi.org/10.1016/j.jhydrol.2021.127286>, 2022.
- 465 Fenicia, F., Savenije, H. H. G., and Avdeeva, Y.: Anomaly in the rainfall-runoff behaviour of the
466 Meuse catchment. Climate, land-use, or land-use management?, *Hydrol. Earth Syst. Sci.*, 13,
467 1727–1737, <https://doi.org/10.5194/hess-13-1727-2009>, 2009.
- 468 Freeze, R. A. and Harlan, R. L.: Blueprint for a physically-based, digitally-simulated hydrologic
469 response model, *J. Hydrol.*, 9, 237–258, [https://doi.org/10.1016/0022-1694\(69\)90020-1](https://doi.org/10.1016/0022-1694(69)90020-1), 1969.
- 470 Gao, H., Birkel, C., Hrachowitz, M., Tetzlaff, D., Soulsby, C., and Savenije, H. H. G.: A simple
471 topography-driven and calibration-free runoff generation module, *Hydrol. Earth Syst. Sci.*, 23,
472 787–809, <https://doi.org/10.5194/hess-23-787-2019>, 2019.



-
- 473 Gao, H., Dong, J., Chen, X., Cai, H., Liu, Z., Jin, Z., Mao, D., Yang, Z., and Duan, Z.: Stepwise
474 modeling and the importance of internal variables validation to test model realism in a data scarce
475 glacier basin, *J. Hydrol.*, 591, 125457, <https://doi.org/10.1016/j.jhydrol.2020.125457>, 2020.
- 476 Gao, H., Han, C., Chen, R., Feng, Z., Wang, K., Fenicia, F., and Savenije, H.: Frozen soil
477 hydrological modeling for a mountainous catchment northeast of the Qinghai-Tibet Plateau,
478 *Hydrol. Earth Syst. Sci.*, 26, 4187–4208, <https://doi.org/10.5194/hess-26-4187-2022>, 2022.
- 479 Gao, H., Hrachowitz, M., Schymanski, S. J., Fenicia, F., Sriwongsitanon, N., and Savenije, H. H.
480 G.: Climate controls how ecosystems size the root zone storage capacity at catchment scale,
481 *Geophys. Res. Lett.*, 41, 7916–7923, <https://doi.org/10.1002/2014GL061668>, 2014.
- 482 Gao, H., Sabo, J. L., Chen, X., Liu, Z., Yang, Z., Ren, Z., and Liu, M.: Landscape heterogeneity
483 and hydrological processes: a review of landscape-based hydrological models, *Landscape Ecol.*, 33,
484 1461–1480, <https://doi.org/10.1007/s10980-018-0690-4>, 2018.
- 485 Germann, P. F.: *Preferential flow: Stokes approach to infiltration and drainage*, Geographica
486 Bernensia, Bern, Switzerland, ISBN 978-3-905835-34-2, 2014.
- 487 Gharari, S., Hrachowitz, M., Fenicia, F., Gao, H., and Savenije, H. H. G.: Using expert knowledge
488 to increase realism in environmental system models can dramatically reduce the need for
489 calibration, *Hydrol. Earth Syst. Sci.*, 18, 4839–4859, <https://doi.org/10.5194/hess-18-4839-2014>,
490 2014.
- 491 Glaser, B., Jackisch, C., Hopp, L., and Klaus, J.: How meaningful are plot-scale observations and
492 simulations of preferential flow for catchment models?, *Vadose Zone J.*, 18, 180146,
493 <https://doi.org/10.2136/vzj2018.08.0146>, 2019.
- 494 Good, S. P., Noone, D., and Bowen, G.: Hydrologic connectivity constrains partitioning of global
495 terrestrial water fluxes, *Science*, 349, 175–177, <https://doi.org/10.1126/science.aaa5931>, 2015.
- 496 Gutmann, E. D. and Small, E. E.: A comparison of land surface model soil hydraulic properties
497 estimated by inverse modeling and pedotransfer functions, *Water Resour. Res.*, 43, W05418,
498 <https://doi.org/10.1029/2006WR005135>, 2007.
- 499 Hrachowitz, M., Benettin, P., van Breukelen, B. M., Fovet, O., Howden, N. J. K., Ruiz, L., van der
500 Velde, Y., and Wade, A. J.: Transit time: the link between hydrology and water quality at the
501 catchment scale, *Wiley Interdiscip. Rev.-Water*, 3, 629–657, <https://doi.org/10.1002/wat2.1155>,
502 2016.
- 503 Jarvis, N., Koestel, J., and Larsbo, M.: Understanding Preferential Flow in the Vadose Zone:
504 Recent Advances and Future Prospects, *Vadose Zone J.*, 15,
505 <https://doi.org/10.2136/vzj2016.09.0075>, 2016.



-
- 506 Jung, M., Schwalm, C., Migliavacca, M., Walther, S., Camps-Valls, G., Koirala, S., Anthoni, P.,
507 Besnard, S., Bodesheim, P., Carvalhais, N., Chevallier, F., Gans, F., Goll, D. S., Haverd, V.,
508 Kohler, P., Ichii, K., Jain, A. K., Liu, J., Lombardozi, D., Nabel, J. E. M. S., Nelson, J. A.,
509 O’Sullivan, M., Pallandt, M., Papale, D., Peters, W., Pongratz, J., Roedenbeck, C., Sitch, S.,
510 Tramontana, G., Walker, A., Weber, U., and Reichstein, M.: Scaling carbon fluxes from eddy
511 covariance sites to globe: Synthesis and evaluation of the FLUXCOM approach, *Biogeosciences*,
512 17, 1343–1365, <https://doi.org/10.5194/bg-17-1343-2020>, 2020.
- 513 Kirchner, J.: A double paradox in catchment hydrology and geochemistry, *Hydrol. Process.*, 17,
514 871–874, <https://doi.org/10.1002/hyp.5108>, 2003.
- 515 Kishne, A. Sz., Yimam, Y. T., Morgan, C. L. S., and Dornblaser, B. C.: Evaluation and
516 improvement of the default soil hydraulic parameters for the Noah Land Surface Model,
517 *Geoderma*, 285, 247–259, <https://doi.org/10.1016/j.geoderma.2016.09.022>, 2017.
- 518 Lan, X., Li, Y., Shao, R., Chen, X., Lin, K., Cheng, L., Gao, H., and Liu, Z.: Vegetation controls
519 on surface energy partitioning and water budget over China, *J. Hydrol.*, 600, 125646,
520 <https://doi.org/10.1016/j.jhydrol.2020.125646>, 2021.
- 521 Lian, X., Piao, S., Huntingford, C., Li, Y., Zeng, Z., Wang, X., Ciais, P., McVicar, T. R., Peng, S.,
522 Ottle, C., Yang, H., Yang, Y., Zhang, Y., and Wang, T.: Partitioning global land evapotranspiration
523 using CMIP5 models constrained by observations, *Nat. Clim. Chang.*, 8, 640–646,
524 <https://doi.org/10.1038/s41558-018-0207-9>, 2018.
- 525 Lin, H., Bouma, J., Pachepsky, Y., Western, A., Thompson, J., van Genuchten, R., Vogel, H.-J.,
526 and Lilly, A.: *Hydropedology: Synergistic integration of pedology and hydrology*, *Water Resour.*
527 *Res.*, 42, W05301, <https://doi.org/10.1029/2005WR004085>, 2006.
- 528 Lu, H., Yuan, W., and Chen, X.: A processes-based dynamic root growth model integrated into the
529 ecosystem model, *J. Adv. Model. Earth Syst.*, 11, 4614–4628,
530 <https://doi.org/10.1029/2019MS001846>, 2019.
- 531 Mao, G. and Liu, J.: WAYS v1: A hydrological model for root zone water storage simulation on a
532 global scale, *Geosci. Model Dev.*, 12, 5267–5289, <https://doi.org/10.5194/gmd-12-5267-2019>,
533 2019.
- 534 McCormick, E. L., Dralle, D. N., Hahm, W. J., Tune, A. K., Schmidt, L. M., Chadwick, K. D., and
535 Rempe, D. M.: Widespread woody plant use of water stored in bedrock, *Nature*, 597, 225–229,
536 <https://doi.org/10.1038/s41586-021-03761-3>, 2021.
- 537 McDonnell, J. J., Sivapalan, M., Vache, K., Dunn, S., Grant, G., Haggerty, R., Hinz, C., Hooper,
538 R., Kirchner, J., Roderick, M. L., Selker, J., and Weiler, M.: Moving beyond heterogeneity and
539 process complexity: A new vision for watershed hydrology, *Water Resour. Res.*, 43, W07301,



-
- 540 <https://doi.org/10.1029/2006WR005467>, 2007.
- 541 McDonnell, J. J.: The two water worlds hypothesis: ecohydrological separation of water between
542 streams and trees?, *Wiley Interdiscip. Rev.-Water*, 1, 323–329, <https://doi.org/10.1002/wat2.1027>,
543 2014.
- 544 Muecher, C. A., Klijn, J. A., Wascher, D. M., and Schaminee, J. H. J.: A new European Landscape
545 Classification (LANMAP): A transparent, flexible and user-oriented methodology to distinguish
546 landscapes, *Ecol. Indic.*, 10, 87–103, <https://doi.org/10.1016/j.ecolind.2009.03.018>, 2010.
- 547 Nijzink, R. C., Almeida, S., Pechlivanidis, I. G., Capell, R., Gustafssons, D., Arheimer, B.,
548 Parajka, J., Freer, J., Han, D., Wagener, T., van Nooijen, R. R. P., Savenije, H. H. G., and
549 Hrachowitz, M.: Constraining conceptual hydrological models with multiple information sources,
550 *Water Resour. Res.*, 54, 8332–8362, <https://doi.org/10.1029/2017WR021895>, 2018.
- 551 Nijzink, R., Hutton, C., Pechlivanidis, I., Capell, R., Arheimer, B., Freer, J., Han, D., Wagener, T.,
552 McGuire, K., Savenije, H., and Hrachowitz, M.: The evolution of root-zone moisture capacities
553 after deforestation: A step towards hydrological predictions under change?, *Hydrol. Earth Syst.*
554 *Sci.*, 20, 4775–4799, <https://doi.org/10.5194/hess-20-4775-2016>, 2016.
- 555 Or, D.: The tyranny of small scales—On representing soil processes in global land surface models,
556 *Water Resour. Res.*, 55, <https://doi.org/10.1029/2019WR024846>, 2020.
- 557 Perrin, C., Michel, C., and Andréassian, V.: Does a large number of parameters enhance model
558 performance? Comparative assessment of common catchment model structures on 429
559 catchments, *J. Hydrol.*, 242, 275–301, [https://doi.org/10.1016/S0022-1694\(00\)00393-0](https://doi.org/10.1016/S0022-1694(00)00393-0), 2001.
- 560 Ponge, J. F.: The soil as an ecosystem, *Biol. Fertil. Soils*, 51, 645–648,
561 <https://doi.org/10.1007/s00374-015-1016-1>, 2015.
- 562 Refsgaard, J. C., Stisen, S., and Koch, J.: Hydrological process knowledge in catchment
563 modelling—Lessons and perspectives from 60 years development, *Hydrol. Process.*, 36, e14463,
564 <https://doi.org/10.1002/hyp.14463>, 2022.
- 565 Rodríguez-Iturbe, I. and Porporato, A.: *Ecohydrology of water-controlled ecosystems: Soil*
566 *moisture and plant dynamics*, Cambridge University Press,
567 <https://doi.org/10.1017/CBO9780511535727>, 2004.
- 568 Savenije, H. H. G. and Hrachowitz, M.: HESS Opinions “Catchments as meta-organisms - a new
569 blueprint for hydrological modelling,” *Hydrol. Earth Syst. Sci.*, 21, 1107–1116,
570 <https://doi.org/10.5194/hess-21-1107-2017>, 2017.
- 571 Savenije, H. H. G.: HESS Opinions “Topography driven conceptual modelling (FLEX-Topo),”
572 *Hydrol. Earth Syst. Sci.*, 14, 2681–2692, <https://doi.org/10.5194/hess-14-2681-2010>, 2010.



-
- 573 Savenije, H. H. G.: HESS Opinions: Linking Darcy's equation to the linear reservoir, *Hydrol.*
574 *Earth Syst. Sci.*, 22, 1911–1916, <https://doi.org/10.5194/hess-22-1911-2018>, 2018.
- 575 Seibert, J., Vis, M. J. P., Lewis, E., and van Meerveld, H. J.: Upper and lower benchmarks in
576 hydrological modelling, *Hydrol. Process.*, 32, 1120–1125, <https://doi.org/10.1002/hyp.11476>,
577 2018.
- 578 Shi, P. L., Tun, B., Cheng, G. W., and Luo, J.: Water retention capacity evaluation of main forest
579 vegetation types in the Upper Yangtze Basin, *J. Nat.l Resour.*, 19, 351–360, 2004.
- 580 Sivapalan, M., Blöschl, G., Zhang, L., and Vertessy, R.: Downward approach to hydrological
581 prediction, *Hydrol. Process.*, 17, 2101–2111, <https://doi.org/10.1002/hyp.1425>, 2003.
- 582 Sivapalan, M.: Pattern, process and function: Elements of a unified theory of hydrology at the
583 catchment scale, in: *Encyclopedia of Hydrological Sciences*, edited by: Anderson, M. G., and
584 McDonnell, J. J., Wiley, 193–219, <https://doi.org/10.1002/0470848944.hsa012>, 2006.
- 585 Sun, G., Gao, H., and Hao, L.: Comments on “Large-scale afforestation significantly increases
586 permanent surface water in China's vegetation restoration regions” by Zeng, Y., Yang, X., Fang,
587 N., & Shi, Z. (2020). *Agricultural and Forest Meteorology*, 290, 108001, *Agric. For. Meteorol.*,
588 296, 108001, <https://doi.org/10.1016/j.agrformet.2020.108213>, 2021.
- 589 Troch, P. A., Lahmers, T., Meira, A., Mukherjee, R., Pedersen, J. W., Roy, T., and Valdes-Pineda,
590 R.: Catchment coevolution: A useful framework for improving predictions of hydrological
591 change?, *Water Resour. Res.*, 51, 4903–4922, <https://doi.org/10.1002/2015WR017032>, 2015.
- 592 Tromp-van Meerveld, H. and McDonnell, J.: Threshold relations in subsurface stormflow: 2. The
593 fill and spill hypothesis, *Water Resour. Res.*, 42, W02411,
594 <https://doi.org/10.1029/2004WR003800>, 2006.
- 595 Uhlenbrook, S.: Catchment hydrology—A science in which all processes are preferential, *Hydrol.*
596 *Process.*, 20, 3581–3585, <https://doi.org/10.1002/hyp.6564>, 2006.
- 597 Van Looy, K., Bouma, J., Herbst, M., Koestel, J., Minasny, B., Mishra, U., Montzka, C., Nemes,
598 A., Pachepsky, Y. A., Padarian, J., Schaap, M. G., Toth, B., Verhoef, A., Vanderborght, J., van der
599 Ploeg, M. J., Weihermueller, L., Zacharias, S., Zhang, Y., and Vereecken, H.: Pedotransfer
600 Functions in Earth System Science: Challenges and Perspectives, *Rev. Geophys.*, 55, 1199–1256,
601 <https://doi.org/10.1002/2017RG000581>, 2017.
- 602 Vereecken, H., Amelung, W., Bauke, S. L., Bogena, H., Brüeggemann, N., Montzka, C.,
603 Vanderborght, J., Bechtold, M., Blöschl, G., Carminati, A., Javaux, M., Konings, A. G., Kusche,
604 J., Neuweiler, I., Or, D., Steele-Dunne, S., Verhoef, A., Young, M., and Zhang, Y.: Soil hydrology
605 in the Earth system, *Nat. Rev. Earth Environ.*, 3, 573–587, <https://doi.org/10.1038/s43017-022->



-
- 606 00324-6, 2022.
- 607 Vereecken, H., Weihermueller, L., Assouline, S., Simunek, J., Verhoef, A., Herbst, M., Archer, N.,
608 Mohanty, B., Montzka, C., Vanderborght, J., Balsamo, G., Bechtold, M., Boone, A., Chadburn, S.,
609 Cuntz, M., Decharme, B., Ducharme, A., Ek, M., Garrigues, S., Goergen, K., Ingwersen, J., Kollet,
610 S., Lawrence, D. M., Li, Q., Or, D., Swenson, S., de Vrese, P., Walko, R., Wu, Y., and Xue, Y.:
611 Infiltration from the pedon to global grid scales: An overview and outlook for land surface
612 modeling, *Vadose Zone J.*, 18, 180191, <https://doi.org/10.2136/vzj2018.10.0191>, 2019.
- 613 Wang, J., Gao, H., Liu, M., Ding, Y., Wang, Y., Zhao, F., and Xia, J.: Parameter regionalization of
614 the FLEX-Global hydrological model, *Sci. China Earth Sci.*, 64, 571–588,
615 <https://doi.org/10.1007/s11430-020-9706-3>, 2021.
- 616 Wang-Erlandsson, L., Bastiaanssen, W. G. M., Gao, H., Jaegermeyr, J., Senay, G. B., van Dijk, A.
617 I. J. M., Guerschman, J. P., Keys, P. W., Gordon, L. J., and Savenije, H. H. G.: Global root zone
618 storage capacity from satellite-based evaporation, *Hydrol. Earth Syst. Sci.*, 20, 1459–1481,
619 <https://doi.org/10.5194/hess-20-1459-2016>, 2016.
- 620 Weiler, M.: Macropores and preferential flow—a love-hate relationship, *Hydrol. Process.*, 31, 15–
621 19, <https://doi.org/10.1002/hyp.11074>, 2017.
- 622 Winsemius, H. C., Savenije, H. H. G., van de Giesen, N. C., van den Hurk, B. J. J. M., Zapreeva,
623 E. A., and Klees, R.: Assessment of Gravity Recovery and Climate Experiment (GRACE)
624 temporal signature over the upper Zambezi, *Water Resour. Res.*, 42, W12201,
625 <https://doi.org/10.1029/2006WR005192>, 2006.
- 626 Worthington, S. R. H.: How preferential flow delivers pre-event groundwater rapidly to streams,
627 *Hydrol. Process.*, 33, 2373–2380, <https://doi.org/10.1002/hyp.13520>, 2019.
- 628 Wu, J.: Key concepts and research topics in landscape ecology revisited: 30 years after the
629 Allerton Park workshop, *Landsc. Ecol.*, 28, 1–11, <https://doi.org/10.1007/s10980-012-9836-y>,
630 2013.
- 631 Xie, Z., Wang, L., Wang, Y., Liu, B., Li, R., Xie, J., Zeng, Y., Liu, S., Gao, J., Chen, S., Jia, B., and
632 Qin, P.: Land surface model CAS-LSM: Model description and evaluation, *J. Adv. Model. Earth
633 Syst.*, 12, e2020MS002339, <https://doi.org/10.1029/2020MS002339>, 2020.
- 634 Yang, Y., R. J. Donohue, and T. R. McVicar (2016), Global estimation of effective plant rooting
635 depth: Implications for hydrological modeling, *Water Resour. Res.*, 52, 8260–8276,
636 [doi:10.1002/2016WR019392](https://doi.org/10.1002/2016WR019392).
- 637 Zehe, E., Loritz, R., Edery, Y., and Berkowitz, B.: Preferential pathways for fluid and solutes in
638 heterogeneous groundwater systems: Self-organization, entropy, work, *Hydrol. Earth Syst. Sci.*,



-
- 639 25, 5337–5353, <https://doi.org/10.5194/hess-25-5337-2021>, 2021.
- 640 Zhang, B., Wu, P., Zhao, X., Gao, X., and Shi, Y.: Assessing the spatial and temporal variation of
641 the rainwater harvesting potential (1971-2010) on the Chinese Loess Plateau using the VIC model,
642 *Hydrol. Process.*, 28, 534–544, <https://doi.org/10.1002/hyp.9608>, 2014.
- 643 Zhao, R. J.: The Xinanjiang model applied in China, *J. Hydrol.*, 135, 371–381,
644 [https://doi.org/10.1016/0022-1694\(92\)90096-e](https://doi.org/10.1016/0022-1694(92)90096-e), 1992.
- 645 Zhao, S., Hu, H., Harman, C. J., Tian, F., Tie, Q., Liu, Y., and Peng, Z.: Understanding of storm
646 runoff generation in a weathered, fractured granitoid headwater catchment in Northern China,
647 *Water*, 11, 123, <https://doi.org/10.3390/w11010123>, 2019.