



1 **The ISIMIP Groundwater Sector: A Framework for Ensemble 2 Modeling of Global Change Impacts on Groundwater**

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25

26 **Abstract**

27 Groundwater serves as a crucial freshwater resource for people and ecosystems, vital in adapting to climate
28 change. Yet, its availability and dynamics are affected by climate variations, changes in land use, and excessive
29 extraction. Despite its importance, our understanding of how global change will influence groundwater in the
30 future remains limited. Multi-model ensembles are powerful tools for impact assessments; compared to single-
31 model studies, they provide a more comprehensive understanding of uncertainties and enhance the robustness of
32 projections by capturing a range of possible outcomes. However, to this point no ensemble of groundwater models
33 was available. Here, we present the new groundwater sector within ISIMIP which combines multiple global,
34 continental, and regional-scale groundwater models. We describe the rationale for the sector, present the sectoral
35 output variables, show first results of a model comparison, and outline the synergies with other existing ISIMIP
36 sectors such as the global water sector and the water quality sector. Currently, eight models are participating in
37 this sector, ranging from gradient-based groundwater models to specialized karst recharge models, each producing



38 up to 19 out of 23 modeling protocol-defined output variables. Utilizing available model outputs for a subset of
39 participating models, we find that the arithmetic mean global water table depth varies substantially between
40 models (6 - 127 m) and shows a shallower water table compared to other recent studies. Groundwater recharge
41 also differs greatly in the global mean (78 - 228 mm/y), which is consistent with recent studies on the uncertainty
42 of groundwater recharge but with different spatial patterns. Groundwater recharge changes between 2001 and
43 2006 show plausible patterns that align with droughts in Spain and Portugal during this period. The simplified
44 comparison highlights the value of a structured model intercomparison project which will help to better understand
45 the impacts of climate change on the world's largest accessible freshwater store – groundwater.

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47

48 1 Introduction

49 Groundwater is the world's largest accessible freshwater resource, vital for human and environmental well-being
50 (Huggins et al., 2023; Scanlon et al., 2023), serving as a critical buffer against water scarcity and surface water
51 pollution (Foster and Chilton, 2003; Schwartz and Ibaraki, 2011). It supports irrigated agriculture, which supports
52 17% of global cropland and 40% of food production (Döll and Siebert, 2002; Perez et al., 2024; United Nations,
53 2022; Rodella et al., 2023). However, unsustainable extraction in many regions has led to declining groundwater
54 levels, the drying of rivers, lakes and wells, land subsidence, seawater intrusion, and aquifer depletion (e.g.,
55 Bierkens and Wada (2019); de Graaf et al. (2019); Rodell et al. (2009)).

56 The pressure on groundwater systems intensifies due to the combined effects of population growth, socioeconomic
57 development, agricultural intensification, and climate change, e.g., through a change in groundwater recharge
58 (Taylor et al., 2013; Reinecke et al., 2021). Rising temperatures and altered precipitation patterns are already
59 reshaping water availability and demand, with significant implications for groundwater use. For instance,
60 changing aridity is expected to influence groundwater recharge rates (Berghuijs et al., 2024), yet the consequences
61 for groundwater levels dynamics remain limited (Moeck et al., 2024; Cuthbert et al., 2019). It is further unclear
62 how these shifts will affect groundwater's role in sustaining ecosystems, agriculture, and human water supplies.

63 Understanding the impacts of climate change and the globalized economy on groundwater systems requires a
64 large-scale perspective (Haqqiqi et al., 2023; Konar et al., 2013; Dalin et al., 2017). While groundwater
65 management traditionally occurs at local or regional scales, aquifers often span administrative boundaries, and
66 over-extraction in one area can have far-reaching effects not captured by a local model. Moreover, groundwater
67 plays a critical role in the global hydrological cycle, influencing surface energy distribution, soil moisture, and
68 evapotranspiration through processes such as capillary rise (Condon and Maxwell, 2019; Maxwell et al., 2016)
69 and supplying surface waters with baseflow (Winter, 2007; Xie et al., 2024). These interactions underscore the
70 importance of groundwater in buffering climate dynamics over extended temporal and spatial scales (Keune et
71 al., 2018) and require a global perspective on the water-climate cycle. While large-scale climate-groundwater
72 interactions are starting to become understood (Cuthbert et al., 2019), current global water and climate models
73 may not always capture these feedbacks as most either do not consider groundwater at all or only include a
74 simplified storage bucket, limiting our understanding of how climate change will affect the water cycle as a whole.



75 The inclusion of groundwater dynamics in global hydrological models remains a considerable challenge due to
76 data limitations and computational demands (Gleeson et al., 2021). Simplified representations, e.g. linear reservoir
77 (Telteu et al., 2021), often fail to capture the complexity of groundwater-surface water interactions, lateral flows
78 at local or regional scales, or the feedback between groundwater pumping and streamflow (de Graaf et al., 2017;
79 Reinecke et al., 2019). These processes are crucial for evaluating water availability, particularly in regions heavily
80 dependent on groundwater. For instance, lateral flows sustain downstream river baseflows and groundwater
81 availability, impacting water quality and ecological health (Schaller and Fan, 2009; Liu et al., 2020) and not
82 including head dynamics may lead to overestimation of groundwater depletion (Bierkens and Wada, 2019).
83 Multiple continental to global-scale groundwater models have been developed in recent years to represent these
84 critical processes (for an overview see also Condon et al. (2021); Gleeson et al. (2021)).

85 While current model ensembles of global water assessments have not yet included gradient-based groundwater
86 processes, they have already advanced our large-scale understanding of the groundwater system. The Inter-
87 Sectoral Impact Model Intercomparison Project (ISIMIP), analogous to the Coupled Model Intercomparison
88 Project (CMIP) for climate models (Eyring et al., 2016), is a well-established community project to carry out
89 model ensemble experiments for climate impact assessments (Frieler et al., 2017). The current generation of
90 models in the Global Water Sector of ISIMIP often represents groundwater as a simplified storage that receives
91 recharge, releases baseflow, and can be pumped (Telteu et al., 2021). Still, it lacks lateral connectivity and head-
92 based surface-groundwater fluxes. Nevertheless, the ISIMIP water sector provided important insights on, for
93 example, future changes and hotspots in global terrestrial water storage (Pokhrel et al., 2021), environmental flows
94 (Thompson et al., 2021), the planetary boundary for freshwater change (Porkka et al., 2024), uncertainties in the
95 calculation of groundwater recharge (Reinecke et al., 2021) and the development of methodological frameworks
96 to compare model ensembles (Gnann et al., 2023).

97 Here, we present a new sector in ISIMIP called the ISIMIP Groundwater Sector, which integrates currently
98 available groundwater models that operate at regional (at least multiple km² (Gleeson and Paszkowski, 2014)) to
99 global scales. The groundwater sector aims to provide a comprehensive understanding of the current state of
100 groundwater representation in large-scale models, identify groundwater-related uncertainties, enhance the
101 robustness of predictions regarding the impact of global change on groundwater and connected systems through
102 model ensembles, and provide insight into how to most reliably and efficiently model groundwater on regional to
103 global scales. The new groundwater sector is a separate but complementary to the existing global water sector.

104 Specifically, the ISIMIP groundwater sector will compile a model ensemble that enables us to assess the impact
105 of global change on various groundwater-related variables and quantify model and scenario-related uncertainties.
106 These insights can then be used to quantify the impacts of global change on, e.g., water availability and in relation
107 to other sectors impacted by changes in groundwater. The ISIMIP groundwater sector has natural linkages with
108 other ISIMIP sectors, such as global water, water quality, regional water, and agriculture. This paper will highlight
109 the connections between groundwater and these other sectors, providing an opportunity to improve our
110 understanding of how modeling choices affect groundwater simulation dynamics.

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113 **2 The ISIMIP framework**

114 ISIMIP aims to provide a framework for consistent climate impact data across sectors and scales. It facilitates
115 model evaluation and improvement, enables climate change impact assessments across sectors, and provides
116 robust projections of climate change impacts under different socioeconomic scenarios. ISIMIP uses a subset of
117 bias-adjusted climate models from the CMIP6 ensemble. The subset is selected to represent the broader CMIP6
118 ensemble while maintaining computational feasibility for impact studies (Lange, 2021).

119 ISIMIP has undergone multiple phases, with the current phase being ISIMIP3. The simulation rounds consist of
120 two main components: ISIMIP3a and ISIMIP3b, each serving distinct purposes. ISIMIP3a focuses on model
121 evaluation and the attribution of observed climate impacts, covering the historical period up to 2021. It utilizes
122 observational climate and socioeconomic data and includes a counterfactual "no climate change baseline" using
123 detrended climate data for impact attribution. Additionally, ISIMIP3a includes sensitivity experiments with high-
124 resolution historical climate forcing. In contrast, ISIMIP3b aims to quantify climate-related risks under various
125 future scenarios, covering pre-industrial, historical, and future projections. ISIMIP3b is divided into three groups:
126 Group I for pre-industrial and historical periods, Group II for future projections with fixed 2015 direct human
127 forcing, and Group III for future projections with changing socioeconomic conditions and representation of
128 adaptation. Despite their differences in focus, time periods, and data sources, both ISIMIP3a and ISIMIP3b require
129 the use of the same impact model version to ensure consistent interpretation of output data, thereby contributing
130 to ISIMIP's overall goal of providing a framework for consistent climate impact data across sectors and scales.

131 In the short term, the groundwater sector will focus on the historical period 1901-2019 in ISIMIP3a
132 (https://protocol.isimip.org/#/ISIMIP3a/water_global/groundwater) with the climate-related forcing based on
133 observational data (obsclim) and the direct human forcing based on historic data (histsoc). We aim to utilize these
134 simulations for an in-depth model comparison, including a comparison to observational data such as time series
135 of groundwater table depth (e.g., Jasechko et al. (2024)) and by utilizing functional relationships (Reinecke et al.,
136 2024). This will yield a new understanding of how these models differ, what the reasons for these differences are,
137 and how they could be improved. In addition, it will provide a basis for implementing impact analyses with
138 ensemble runs based on future scenarios using ISIMIP3b inputs.

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140 **3 The current generation of groundwater models in the sector**

141 Many large-scale groundwater models are already participating in the sector (Table 1), and we expect to expand
142 further. The current models are mainly global-scale, with some having a particular regional focus, and primarily
143 using daily timesteps.

144 While the main modeling purpose of most models is to simulate parts of the terrestrial water cycle, they all focus
145 on different aspects (such as karst recharge or sea-water intrusion), most investigate interactions between
146 groundwater and land surface processes, and account for human water uses. Two models (V2KARST and GGR)
147 have distinct purposes in modeling groundwater recharge and do not model any head-based groundwater fluxes.
148 Conceptually, the models may be classified according to Condon et al. (2021) into five categories: lumped models
149 with static groundwater configurations of long-term mass balance (a), saturated groundwater flow with recharge,



150 and surface water exchange fluxes as upper boundary conditions without later fluxes (b), quasi 3D models with
151 variably saturated flow in the soil column and a dynamic water table as a lower boundary condition (c), saturated
152 flow models solving mainly the Darcy equation (d), and variably saturated flow which is calculated as three-
153 dimensional flow throughout the entire subsurface below and above the water table (e). See Condon et al. (2021)
154 and also Gleeson et al. (2021) for a more detailed overview and discussion of approaches. Half of the models
155 (Table 1) simulate a saturated subsurface flux (d), V2KARST and GGR mainly use a 1D vertical approach (b),
156 and others simulate a combination of multiple approaches (ParFlow, Table 1) or can switch between different
157 approaches (CWatM, Table 1).

158 The sector protocol is defined at <https://protocol.isimip.org/#/ISIMIP3a/groundwater> and will be updated over
159 time. We have defined multiple joint outputs for this sector (23 variables in total), but not all models can yet
160 provide all outputs (Table 2). Models can provide 1-19 outputs (11 on average), and multiple models have further
161 outputs that are under development. The global water sector also contains groundwater-related variables (Table
162 A2), enabling groundwater-related analysis. We list them here to show their close connection to the global water
163 sector and facilitate an overview of future groundwater-related studies.

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165 **Table 1:** Summary of all models participating in the ISIMIP groundwater sector. This table lists only models that
166 add new variables to the ISIMIP protocol. Models already part of the global water sector and providing other
167 groundwater-related variables are not listed here. (GMD discussion formatting requires a portrait instead of a
168 landscape table)

Model name	Main model purpose	Coupling with other models	Spatial domain and resolution	Temporal resolution	Hydrogeological configuration, e.g. number of layers	Conceptual model according to Condon et al.	Calibrated	Representation of groundwater use	Main Reference
Water Balance Model (WBM)	Representation of the terrestrial hydrological cycle, including human interactions.	-	Global and regional. Spatial resolution defined by the input river network.	Sub-daily, Daily, Multi-day	1 soil layer, 2 groundwater layers	d.	Globally: no, regional: yes (NE, US)	Through calculated abstractions from groundwater.	Grogan et al. (2022) With groundwater methods based on de Graaf et al. (2015); de Graaf et al. (2017).
Community Land Model (CLM)	To simulate surface and subsurface hydrological processes, including crop growth, irrigation, and groundwater withdrawal.	Community Earth System Model (CESM)	Global and regional (0.05 (regional), 0.1, 0.25, and 0.5 degree (global))	Sub-Daily	20 soil layers extending up to 8.5 m; 1 aquifer layer, unconfined	c.	No	Yes	Felfelani et al. (2021) Lawrence et al. (2019)
Community Water Model (CWatM)	To reproduce main hydrological processes, including water	MODFLOW (optional)	Global, regional, subbasin (30 arcsecond s, 1 km, 1 arc-min, 5 arc-min,	Daily	Standard: 1 with MODFLOW W: variable	Standard: a./b. With MODFLOW W: d.	Globally: yes (with discharge), regional: tailored	Yes	Guillaut et al. (2022); Burek et al. (2020)



	management on regional to global scales.		30 arc-min)						
Global Gradient-based Groundwater Model (G ³ M)	Understanding of surface water, coastal, and ecosystem interaction with groundwater.	WaterGA P (Müller Schmied et al., 2016)	Global (5 arc minutes)	Daily, monthly, or yearly	2 layers, second layer with a reduced hydraulic conductivity	d.	No	Through calculated net abstractions from groundwater of WaterGA P	Reinecke et al. (2019); Kretschmer et al. (2025)
VIC-WUR-MODFLO W (VIC-wur)	Grid-based macro-scale hydrological model that solves both the surface energy balance and water balance equations.	WOFOST (WOrld FOod STudies) (Droppers et al 2021)	Regionally and globally: 5 arcminutes	Sub-daily to monthly	3 soil layers (variable thickness), 2 groundwater layers (variable thickness, confined/unconfined systems.	d.	Globally: no, regional: yes	Through calculated demands and allocation to surface water/groundwater.	Liu et al in prep.; Droppers et al. 2020.; Liang et al. (1994)
V2KART	A grid-based vegetation-recharge model for the global karst areas.	-	Globally: 0.25 arc degree	Daily	three soil layers and one epikarst layer	b.	Yes, based on global karst landscapes	no	Sarrazin et al. (2018)



GGR	A grid-based three-layer water balance model to estimate the daily global rain-fed groundwater recharge	-	180.0°W to 180.0°E longitudes and 60.0°N to 60.0°S latitudes, 0.1 degree	Daily	2 soil layers and 1 groundwater layer of variable thickness	b.	No	No	Nazari et al. (2025)
ParFlow	3D continuum simulations of variably saturated groundwater-surface water and land surface processes.	Common Land Model, CLM (Maxwell and Miller, 2005; Kollet and Maxwell, 2008), Terrestrial Systems Modeling Platform (Gasper et al., 2014), WRF (Maxwell et al., 2011)	Regionally and globally, $10^0 - 10^1$ km	Variable	Variable	a. - c.	Yes, in engineering applications	Yes	Kuffour et al. (2020)

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Table 2: List of output variables in the ISIMIP3a global groundwater sector. The spatial resolution is five arcminutes (even if some models simulate at a higher or coarser resolution), and the temporal resolution is monthly. Most models also simulate daily timesteps, but as most groundwater movement happens across longer time scales, we unified the unit to months. A “*” indicates that a model is able to produce the necessary output. A “+” indicates that this output is currently under development. (GMD discussion formatting requires a portrait instead of a landscape table)

Groundwater sector output variables		Unit	WBM	CLM	C-WaM	GIM	VIC-nur	V2KARST	GGR	ParFlow
Name	Description									
Capillary rise	Upward flux from groundwater to soil (leaving aquifer = negative value).	$m^3 m^{-2} month^{-1}$		*	*		*			*
Diffuse groundwater recharge	Downwards flux from soil to groundwater (entering aquifer = positive value). The unit $kg m^{-2} s^{-1}$ is equal to $mm s^{-1}$. Unit is kept equal to the global water sector.	$kg m^{-2} s^{-1}$	*	*	*		*	*	*	*
Groundwater abstractions	Groundwater pumped from the aquifer.	$m^3 m^{-2} month^{-1}$	*	*	*		+		+	
Groundwater abstractions (domestic)	<i>Groundwater abstractions</i> that are intended for domestic water use.	$m^3 m^{-2} month^{-1}$	*		*		+		+	
Groundwater abstractions (industries)	<i>Groundwater abstractions</i> that are intended for industrial water use.	$m^3 m^{-2} month^{-1}$	*		*		+		+	
Groundwater abstractions (irrigation)	<i>Groundwater abstractions</i> that are intended for irrigational water use.	$m^3 m^{-2} month^{-1}$	*	*	*		+		+	
Groundwater abstractions (livestock)	<i>Groundwater abstractions</i> that are intended for livestock water use.	$m^3 m^{-2} month^{-1}$	*		*		+			
Groundwater demands	Gross water demand	$m^3 m^{-2} month^{-1}$	*	*	*		+			
Groundwater depletion	Long-term losses from groundwater storage	$m^3 m^{-2} month^{-1}$	*	*		*	+			*



Groundwater drainage/surface water capture	Exchange flux between groundwater and surface water. Groundwater leaving the aquifer = negative value; entering the aquifer = positive value	m3 m-2 month-1	*	*	*	*	*	*	*	*
Groundwater drainage/surface water capture from lakes	Exchange flux between groundwater and surface water (lakes); if available, additional to the sum of exchange fluxes (Groundwater drainage/surface water capture) also separate components can be provided/ Leaving the aquifer = negative values; entering the aquifer = positive value.	m3 m-2 month-1				*	*			*
Groundwater drainage/surface water capture from rivers	Exchange flux between groundwater and surface water (rivers); if available, additional to the sum of exchange fluxes (Groundwater drainage/surface water capture) also separate components can be provided/ Leaving the aquifer = negative values; entering the aquifer = positive value.	m3 m-2 month-1	*			*	*			*
Groundwater drainage/surface water capture from springs	Exchange flux between groundwater and surface water (springs); if available, additional to the sum of exchange fluxes (Groundwater drainage/surface water capture) also separate components can be provided/ Leaving the aquifer = negative values; entering the aquifer = positive value.	m3 m-2 month-1				*	*			*
Groundwater drainage/surface water capture from wetlands	Exchange flux between groundwater and surface water (wetlands); if available, additional to the sum of exchange fluxes (Groundwater drainage/surface water capture) also separate components can be provided/ Leaving the aquifer = negative values; entering the aquifer = positive value.	m3 m-2 month-1				*	*			*
Groundwater return flow	Return flow of abstracted groundwater (not yet separated into different sources).	m3 m-2 month-1	*	*						*
Groundwater storage	Mean monthly water storage in groundwater layer in kg m ⁻² . The spatial resolution is 0.5° grid.	m3 m-2 month-1	*	*		*	*			*



Hydraulic head	Head above sea level in m. If more than one aquifer layer is simulated, report the heads on the top productive aquifer (confined or unconfined).	m	*				*	*			*
Lateral groundwater flux (front face)	Cell-by-cell flow (front)	m3 m-2 month-1	*	*			*	*			*
Lateral groundwater flux (right face)	Cell-by-cell flow (right)	m3 m-2 month-1	*	*			*	*			*
Lateral groundwater flux (net)	Net cell-by-cell flow	m3 m-2 month-1	*	*			*	*			*
Lateral groundwater flux (lower face)	Cell-by-cell flow (lower) when more than 1 groundwater layer is simulated.	m3 m-2 month-1	*				*	*			*
Submarine groundwater discharge	Flow of groundwater into oceans. The definition may vary by model. But in principle also models without density driven flow can submit this variable.	m3 m-2 month-1	*				*				*
Water table depth	Depth to the water table below land surface (digital elevation mode, DEM) in m.	m	*	*			*	*			*
Number of groundwater output variables in model	Counting only currently available		19	13	9	14	14	1	1	17	



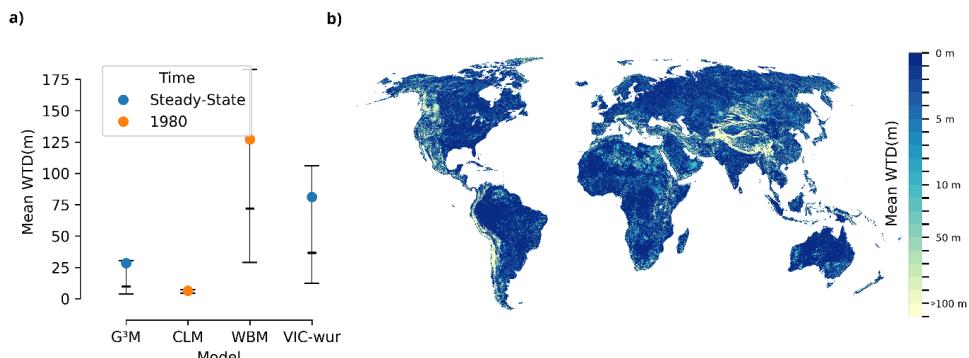
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178 **4 Unstructured experiments point out model differences that should be explored further**

179 The ISIMIP groundwater sector is in an early development stage, and we hope that an ensemble of groundwater
180 models driven by the same meteorological data will be available soon. Yet, to provide first insights into the
181 models, their outputs, and how these can be compared we collected existing outputs from the participating models
182 (see Table A1 for an overview). We opted for a straightforward initial comparison due to the various data formats,
183 model resolutions, and forcings that complicate a more thorough examination of a specific scientific inquiry. Thus,
184 this descriptive analysis serves as an introductory overview that highlights the present state of the art and identifies
185 model discrepancies warranting further investigation. In addition, relevant output data are not yet available for all
186 models. We focused on the two variables with the largest available ensemble: water table depth (G³M, CLM,
187 WBM, and VIC-wur; Table 1) and groundwater recharge (CLM, CWatM, GGR, VIC-wur, V2KARST, WBM;
188 Table 1), only on historical periods rather than future projections.

189 The arithmetic mean (not weighted by cell area) global water table depth varies substantially (6 m – 127 m)
190 between the models at the start of the simulation (1980 or steady-state) (Fig. 1a). On average, the water table of
191 G³M (28 m) and CLM (6 m) are shallower than WBM (127 m) and VIC-wur (81 m), whereas the latter two also
192 show a larger standard deviation (WBM: 133 m, VIC-wur: 105 m) than the other two models (G³M: 49 m, CLM:
193 3 m). The consistently shallower WTD of CLM impacts the ensemble mean WTD (Fig. 1b), which is shallower
194 compared to other model ensembles (5.67 m WTD as global mean here compared to 7.03 m in Reinecke et al.
195 (2024)). This difference in ensemble WTD points to conceptual differences between the models, which should be
196 investigated further, for example, by exploring spatial and temporal differences and relationships with important
197 groundwater drivers (Reinecke et al., 2024).

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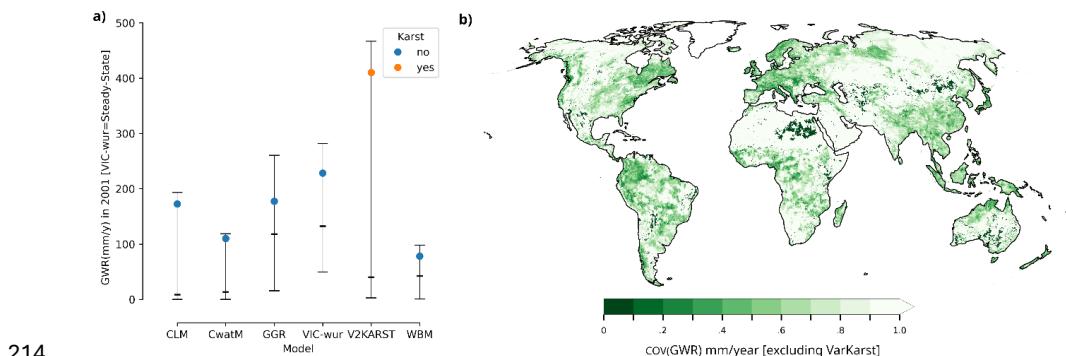
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200 **Figure 1:** Global water table depth (WTD) at simulation start (1980) or the used steady-state. The simplified
201 boxplot (a) shows the arithmetic model mean as a colored dot and the median as a black line. Whiskers indicate
202 the 25th and 75th percentiles, respectively. The global map (b) shows the arithmetic mean of the model ensemble.
203 Similarly, the global arithmetic mean groundwater recharge (not weighted by cell area) differs by 332 mm/y
204 between models (150 mm/y excluding V2KARST since it calculates recharge in karst regions only) (Fig. 2a). This



205 difference in recharge is more pronounced spatially (Fig. 2b) than differences in WTD shown before (Fig. 1b).
206 Especially in drier regions such as in the southern Africa, central Australia, and the northern latitudes show
207 coefficient of variation of 1 or greater (white areas). In extremely dry areas such as the east Sahara and southern
208 Australia, the model spread is close to 0 (dark green). While the agreement is higher in Europe and western South
209 America, the map differs slightly from other recent publications (e.g., compared to Fig. 1b in Gnann et al. (2023)).
210 In light of other publications, highlighting model uncertainty in groundwater recharge (Reinecke et al., 2021) and
211 the possible impacts of long-term aridity changes on groundwater recharge (Berghuijs et al., 2024), an extended
212 combined ensemble of the global water sector and the new groundwater sector could yield valuable insights.

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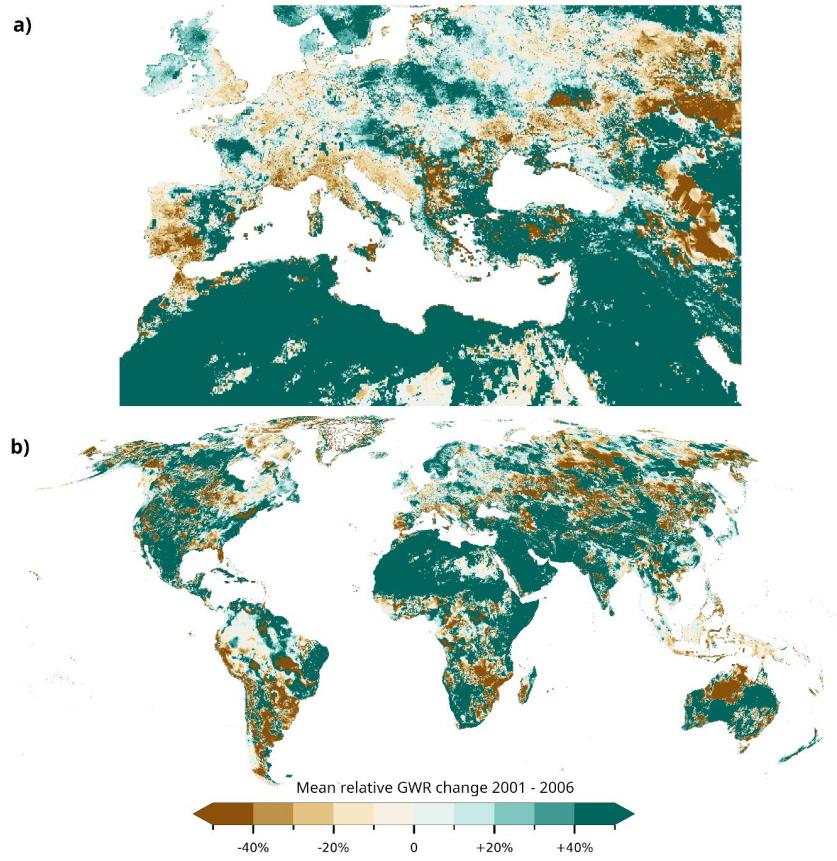


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215 **Figure 2:** Global groundwater recharge (GWR) in 2001 or at steady-state (only VIC-wur). The simplified boxplot
216 (a) shows the arithmetic model mean as a colored dot and the median as a black line. Whiskers indicate the 25th
217 and 75th percentiles, respectively. The global map (b) shows the coefficient of variation of the model ensemble
218 without V2KARST.

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220 We further calculated relative changes in groundwater recharge between 2001 and 2006 (Fig. 3) with an ensemble
221 of 7 models (CLM, CWatM, GGR, VIC-wur, V2KARST, ParFlow, and WBM). The ensemble includes two
222 models that only simulate specific regions (V2KARST: regions of karstifiable rock, ParFlow: Euro CORDEX
223 domain). This result shows a potential analysis that should be repeated within the new groundwater sector.
224 Intentionally, we do not investigate model agreement on the sign of change or compare them with observed data.
225 The ensemble still highlights plausible regions of groundwater recharge changes, such as in Spain and Portugal,
226 which aligns with droughts in the investigated period (Panque Salgado and Vargas Molina, 2015; Coll et al.,
227 2017; Trullenque-Blanco et al., 2024). Relative increases in groundwater recharge are mainly shown for arid
228 regions in the Sahara, the Middle East, Australia, and Mexico. However, it is likely that because we investigate
229 relative changes, this might be related to the already low recharge rates in these regions.



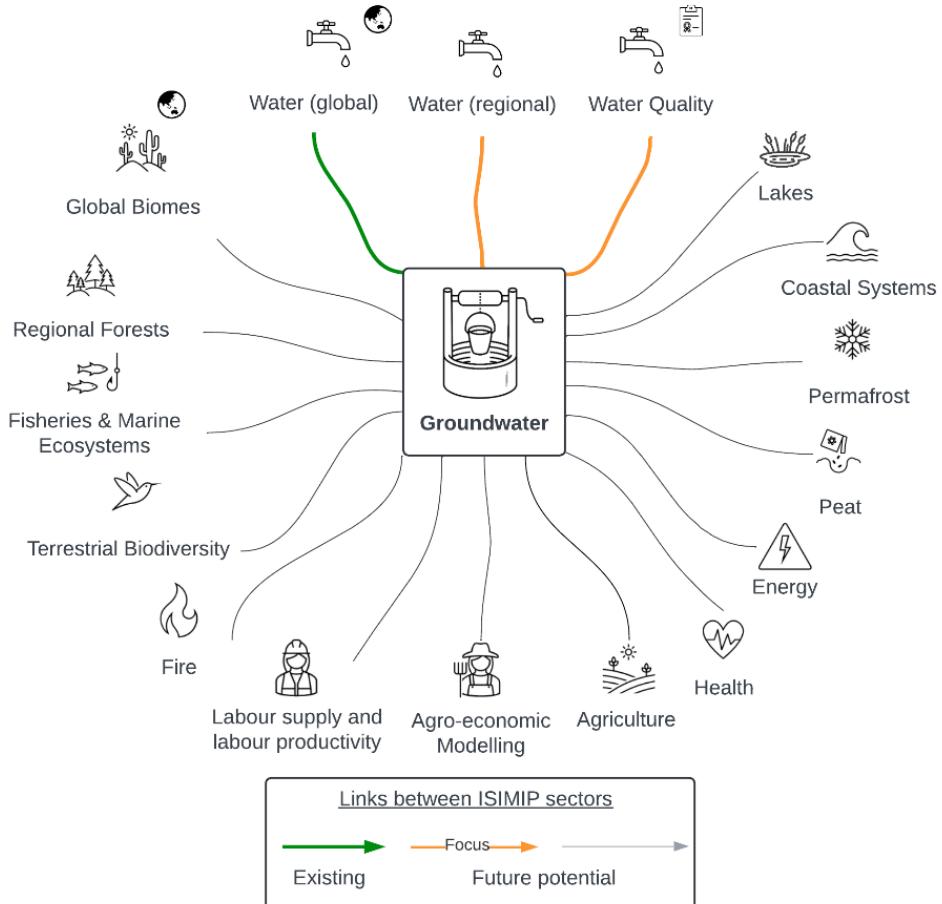
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231 **Figure 3:** Mean relative percentage change of yearly groundwater recharge between 2001 and 2006 for Europe
232 (a), and all continents except Antarctica (b). The ensemble consists of all models that provided data for the years
233 2001 and 2006 (CLM, CWatM, GGR, VIC-wur, V2KARST, WBM, and ParFlow). V2KARST (only karst) and
234 ParFlow (only Euro CORDEX domain) were only accounted for in regions where data is available.

235

236 **5 Groundwater as a linking sector in ISIMIP**

237 ISIMIP encompasses a wide variety of sectors. Currently, 18 sectors are part of the impact assessment effort. The
238 groundwater sector offers a new and unique opportunity to enhance cross-sectoral activities within ISIMIP, foster
239 interlinkages within ISIMIP, and thus deliver interdisciplinary assessments of climate change impacts.



240

241 **Figure 4:** The groundwater sector provides the potential for multiple interlinkages between different sectors
242 within ISIMIP. In the coming years, we will focus on links to three sectors (green and orange): water (global),
243 water (regional), and water quality. Other cross-sectoral linkages between non-groundwater sectors (i.e. linkages
244 between the outer circle) are not shown.

245 Some links with other sectors within ISIMIP are more evident than others with regard to existing scientific
246 community overlaps or existing scientific questions (Fig. 4). For example, the new groundwater sector will focus
247 on large-scale groundwater models, some of which are already part of global water models participating in the
248 global water sector or using outputs (such as groundwater recharge) from the global water sector (see also existing
249 groundwater variables in the global water sector Table A2). However, the groundwater sector will also feature
250 non-global representations of groundwater. Thus, collaborating with the regional water sector could provide
251 opportunities to share outputs and pursue common assessments. For example, the outputs of the groundwater
252 model ensemble, such as water table depth variations or surface water groundwater interactions, could be used as
253 input for some regional models that consider groundwater only as a lumped groundwater storage. Conversely,
254 global and continental groundwater models can learn from validated regional hydrological models, which may



255 provide insights into local runoff generation processes and impacts of water management. Furthermore, the
256 relevance of groundwater for water quality assessments is widely recognized (e.g., for phosphorous transport from
257 groundwater to surface water (Holman et al., 2008), or for salinization (Kretschmer et al., 2025)), or as a link
258 between warming groundwater and stream temperatures (Benz et al., 2024). Leveraging such connections will
259 provide valuable insights beyond groundwater itself.

260 Specifically, considering groundwater quality, a collaboration between both sectors could be achieved in multiple
261 aspects. Integrating groundwater availability with water quality helps ensure sufficient and safe drinking and
262 irrigation water. Focusing on aquifer storage levels and pollutant loads can help maintain groundwater resilience,
263 safeguard food security, and protect public health under changing climate and socioeconomic conditions. Further,
264 integrating groundwater quantity data with pollution source mapping helps prioritize remediation efforts where
265 aquifers are most vulnerable, ensuring both water availability and quality. Concerning observational data, a
266 unified approach to collecting and developing shared databases for groundwater levels and water quality
267 measurements across multiple agencies reduces bureaucratic hurdles and ensures consistent, comparable data.
268 Using standardized procedures for dealing with observational uncertainties such as data gaps, scaling issues, and
269 measurement inconsistencies would support collaborative research further.

270 Research opportunities arise in other sectors as well. Groundwater is connected to the water cycle and social,
271 economic, and ecological systems (Huggins et al., 2023). For example, health impacts (such as water- and vector-
272 borne diseases) are closely related to water quantity and quality (e.g. Smith et al. (2024)), and the roles of
273 groundwater for forest resilience (regional forest sector, (Costa et al., 2023; Esteban et al., 2021)) and forest fires
274 (fire sector) under climate change are yet to be explored (Fig. 4). To prioritize our efforts and set a research agenda
275 for the groundwater ISMIP sector, we will first focus on existing and more straightforward connections to the
276 global water sector, regional water sector, and the water quality sector and then expand to collaboration with other
277 sectors (Fig. 4).

278

279 **6 A vision for the ISMIP groundwater sector**

280 Given groundwater's importance in the Earth system and for society, it is imperative to expand our knowledge of
281 groundwater and (1) how it is impacted by global change and (2) how in turn this will affect other systems
282 connected to groundwater. This enhanced understanding is essential to equip us with the knowledge needed to
283 address future challenges effectively. The ISMIP groundwater sector serves as a foundation for examining and
284 measuring the effects of global change on groundwater systems worldwide. It facilitates cross-sector
285 investigations, such as those concerning water quality, examines the influence of various model structures on
286 groundwater dynamics simulations, and supports the collaborative creation of new datasets for model
287 parameterization and assessment.

288 Already in the short term, the creation of the groundwater sector has substantial potential to enhance large-scale
289 groundwater research by developing better modeling frameworks for reproducible research (running the multitude
290 of experiments targeted in ISMIP requires an automated modeling pipeline) and forge a community that can
291 critically examine current modeling practice. The simple model comparison presented here sparks first questions



292 on why models differ and invites us to explore model differences in more depth. Such model intercomparison
293 studies will enable us to quantify uncertainties and identify hotspots for model improvement. They will also allow
294 us to assess the impact of climate and land use change on different groundwater-related variables, such as
295 groundwater recharge and water table depth, and allow ensemble-based impact assessments on future water
296 availability.

297 In the long term, the sector will enable us to jointly reflect on processes that we currently do not model or that
298 need improvement, possibly also through new modeling approaches such as hybrid machine-learning models
299 tailored to the large-scale representation of groundwater. Since groundwater is connected to many socio-
300 ecological systems, groundwater models could also emerge as a modular coupling tool that can be integrated into
301 multiple sectors. The newly founded groundwater sector already provides a first step in that direction by
302 standardizing output names and units. If models are modular enough and define a standardized Application
303 Programming Interface (API), they could also serve as a valuable tool for other science communities.

304 In summary, the ISIMIP groundwater sector aims to enhance our understanding of the impacts of climate change
305 and direct human impacts on groundwater resources and a range of related sectors.

306

307 **Data availability**

308 The ensemble mean WTD and groundwater recharge trends are available at Reinecke (2025). For the original
309 model data publications, see Table A1.

310

311 **Author contribution**

312 RR led the writing and analysis of the manuscript. RR and IG conceived the idea. All authors reviewed the
313 manuscript and provided suggestions on text and figures.

314

315 **Competing interests**

316 None.

317

318 **Appendix**

319 **Table A1:** Original publications that describe the model outputs used in section 4.

Model	Simulation setup and used forcings	Reference
G ³ M	Steady-state model of WTD on 5 arcmin without any groundwater pumping, forced with WaterGAP 2.2d (Müller Schmied et al., 2021) groundwater recharge mean between 1901-2001.	Reinecke et al. (2019)



V2KARST	Global karst recharge model at 15 arcmin, forced with the MSWEP V2 (Beck et al., 2019) precipitation and GLDAS (Li et al., 2018) air temperature, shortwave and longwave radiation, specific humidity and wind speed for the period of 1990-2020	Sarrazin et al. (2018)
GGR	Global Groundwater Recharge model, A grid-based three-layer water balance model to estimate the daily global rain-fed groundwater recharge (2001-2020)	Nazari et al. (2025)
WBM	Time series simulation from 1980 to 2019 at 15 arc minutes, using the MERIT digital flow direction dataset (Yamazaki et al., 2019) including domestic, industrial, livestock, and irrigation water withdrawals. Forcings and key inputs: Climate: ERA5 (Prusevich et al., 2024), Reservoirs: GRanD v1.1 (Lehner et al., 2011), Inter-basin transfers (Lammers, 2022), Glaciers (Rounce et al., 2022), Impervious surfaces (Hansen and Tofteemann Thomsen, 2020), Population density (Lloyd et al., 2019), Domestic and industrial water per capita demand: FAO AQUASTAT, Livestock density and water demand (Gilbert et al., 2018), Cropland: LUH2 (Hurtt et al., 2020), Aquifer properties (de Graaf et al., 2017) aquifer depth gap-filled with terrain slope data from Yamazaki et al. 2019, Soil available water capacity: FAO soil map, Root depth (Yang et al., 2016)	Multiple, see left column.
VIC-wur	Global Hydrological model simulating the GWR and streamflow from 1970-2014 in natural condition. The mean GWR and streamflow were used to simulate the GWT in steady-state MODFLOW model in 5 arcmin. The model is forced by: GFDL-ESM4 climate model (Dunne et al., 2020), Aquifer properties (de Graaf et al., 2017).	Droppers et al. (2020)
CLM	The model was spun up for 1979 and subsequently simulated from 1979 to 2013 using	Akhter et al. (2024) (under review in WRR)



	the GSWP3 atmospheric forcing dataset at a 0.1-degree resolution. Recharge, capillary rise, drainage, irrigation pumping and cell-to-cell lateral flow were simulated within the model.	
ParFlow	The data provided here are based on Naz et al. (2023). In version 2 of the data, we provide variables including water table depth and groundwater recharge for time period of 1997-2006 at monthly time scale.	Naz et al. (2023)
CWatM	Community Water Model at 5 arcmin. Climate forcing with chelsa-W5E5v1.0 (5 arcmin) for temperature (average, maximum, minimum), precipitation, and shortwave radiation, and GSWP3-W5E5 (30 arcmin spline downscaled to 5 arcmin) for longwave radiation, wind speed, and specific humidity. Updates to Burek et al. (2020) include river network based on MERIT Hydro and upscaling with Eilander et al. (2021).	Burek et al. (2020)



Table A2: List of groundwater related output variables in the ISIMIP3a global water sector (https://protocol.isimip.org/#/ISIMIP3a/water_global). The unit of all variables is $\text{kg m}^{-2} \text{ s}^{-1}$, the spatial resolution is 0.5° grid and the temporal resolution is monthly.

Groundwater-related output variable of the Global Water Sector	Description
Groundwater runoff	Water that leaves the groundwater layer. In case seepage is simulated but no groundwater layer is present, report seepage as <i>Total groundwater recharge</i> and <i>Groundwater Runoff</i> .
Total groundwater recharge	For models that consider both diffuse and focused/localised recharge this should be the sum of both; other models should submit the groundwater recharge component that the model simulates. See also the descriptions in <i>Focused/localised groundwater recharge</i> and <i>Diffuse groundwater recharge</i> .
Focused/localised groundwater recharge	Water that directly flows from a surface water body into the groundwater layer below. Only submit if the model separates focused/localised recharge from diffuse recharge.
Potential irrigation water withdrawal (assuming unlimited water supply) from groundwater resources	Part of <i>Potential Industrial Water Withdrawal</i> that is extracted from groundwater resources.
Actual irrigation water withdrawal from groundwater resources	Part of <i>Actual Irrigation Water Withdrawal</i> that is extracted from groundwater resources.
Potential Irrigation Water Consumption from groundwater resources	Part of <i>Potential Irrigation Water Consumption</i> that is extracted from groundwater resources.
Actual Irrigation Water Consumption from groundwater resources	Part of <i>Actual Irrigation Water Consumption</i> that is extracted from groundwater resources.
Potential Domestic Water Withdrawal from groundwater resources	Part of <i>Potential Domestic Water Withdrawal</i> that is extracted from groundwater resources.
Actual Domestic Water Withdrawal from groundwater resources	Part of <i>Actual Domestic Water Withdrawal</i> that is extracted from groundwater resources
Potential Domestic Water Consumption from groundwater resources	Part of <i>Potential Domestic Water Consumption</i> that is extracted from groundwater resources.



Actual Domestic Water Consumption from groundwater resources	Part of <i>Actual Domestic Water Consumption</i> that is extracted from groundwater resources.
Potential Manufacturing Water Withdrawal from groundwater resources	Part of <i>Potential Manufacturing Water Withdrawal</i> that is extracted from groundwater resources.
Actual Manufacturing Water Withdrawal from groundwater resources	Part of <i>Actual Manufacturing Water Withdrawal</i> that is extracted from groundwater resources.
Potential manufacturing Water Consumption from groundwater resources	Part of <i>Potential manufacturing Water Consumption</i> that is extracted from groundwater resources.
Actual Manufacturing Water Consumption from groundwater resources	Part of <i>Actual Manufacturing Water Consumption</i> that is extracted from groundwater resources.
Potential electricity Water Withdrawal from groundwater resources	Part of <i>Potential electricity Water Withdrawal</i> that is extracted from groundwater resources.
Actual Electricity Water Withdrawal from groundwater resources	Part of <i>Actual Electricity Water Withdrawal</i> that is extracted from groundwater resources.
Potential electricity Water Consumption from groundwater resources	Part of <i>Potential electricity Water Consumption</i> that is extracted from groundwater resources.
Actual Electricity Water Consumption from groundwater resources	Part of <i>Actual Electricity Water Consumption</i> that is extracted from groundwater resources.
Potential Industrial Water Withdrawal from groundwater resources	Part of <i>Potential Industrial Water Withdrawal</i> that is extracted from groundwater resources.
Actual Industrial Water Withdrawal from groundwater resources	Part of <i>Actual Industrial Water Withdrawal</i> that is extracted from groundwater resources.



Potential Industrial Water Consumption from groundwater resources	Part of <i>Potential Industrial Water Consumption</i> that is extracted from groundwater resources.
Actual Industrial Water Consumption from groundwater resources	Part of <i>Actual Industrial Water Consumption</i> that is extracted from groundwater resources.
Potential livestock Water Withdrawal from groundwater resources	Part of <i>Potential livestock Water Withdrawal</i> that is extracted from groundwater resources.
Actual Livestock Water Withdrawal from groundwater resources	Part of <i>Actual Livestock Water Withdrawal</i> that is extracted from groundwater resources.
Potential livestock Water Consumption from groundwater resources	Part of <i>Potential livestock Water Consumption</i> that is extracted from groundwater resources.
Actual livestock Water Consumption from groundwater resources	Part of <i>Actual livestock Water Consumption</i> that is extracted from groundwater resources.
Total Potential Water Withdrawal (all sectors) from groundwater resources	Part of <i>Total Potential Water Withdrawal</i> that is extracted from groundwater resources.
Total Actual Water Withdrawal (all sectors) from groundwater resources	Part of <i>Total Actual Water Withdrawal</i> that is extracted from groundwater resources.

321

322 **References**

323 Beck, H. E., Wood, E. F., Pan, M., Fisher, C. K., Miralles, D. G., van Dijk, A. I. J. M., McVicar, T. R., and
324 Adler, R. F.: MSWEP V2 Global 3-Hourly 0.1° Precipitation: Methodology and Quantitative
325 Assessment, Bulletin of the American Meteorological Society, 100, 473–500,
326 <https://doi.org/10.1175/BAMS-D-17-0138.1>, 2019.

327 Benz, S. A., Irvine, D. J., Rau, G. C., Bayer, P., Menberg, K., Blum, P., Jamieson, R. C., Griebler, C., and
328 Kurylyk, B. L.: Global groundwater warming due to climate change, Nature Geosci, 17, 545–551,
329 <https://doi.org/10.1038/s41561-024-01453-x>, 2024.

330 Berghuijs, W. R., Collenteur, R. A., Jasechko, S., Jaramillo, F., Luijendijk, E., Moeck, C., van der Velde, Y.,
331 and Allen, S. T.: Groundwater recharge is sensitive to changing long-term aridity, Nature Clim
332 Change, 14, 357–363, <https://doi.org/10.1038/s41558-024-01953-z>, 2024.

333 Bierkens, M. and Wada, Y.: Non-renewable groundwater use and groundwater depletion: a review,
334 Environ. Res. Lett., 14, 63002, <https://doi.org/10.1088/1748-9326/ab1a5f>, 2019.



335 Burek, P., Satoh, Y., Kahil, T., Tang, T., Greve, P., Smilovic, M., Guillaumot, L., Zhao, F., and Wada, Y.:
336 Development of the Community Water Model (CWatM v1.04) – a high-resolution hydrological model
337 for global and regional assessment of integrated water resources management, *Geosci. Model Dev.*,
338 13, 3267–3298, <https://doi.org/10.5194/gmd-13-3267-2020>, 2020.

339 Condon, L. E. and Maxwell, R. M.: Simulating the sensitivity of evapotranspiration and streamflow to
340 large-scale groundwater depletion, *Science advances*, 5, eaav4574,
341 <https://doi.org/10.1126/sciadv.aav4574>, 2019.

342 Condon, L. E., Kollet, S., Bierkens, M. F. P., Fogg, G. E., Maxwell, R. M., Hill, M. C., Fransen, H.-J. H.,
343 Verhoef, A., van Loon, A. F., Sulis, M., and Abesser, C.: Global Groundwater Modeling and Monitoring:
344 Opportunities and Challenges, *Water Resources Research*, 57,
345 <https://doi.org/10.1029/2020WR029500>, 2021.

346 Costa, F. R. C., Schietti, J., Stark, S. C., and Smith, M. N.: The other side of tropical forest drought: do
347 shallow water table regions of Amazonia act as large-scale hydrological refugia from drought?, *The
348 New phytologist*, 237, 714–733, <https://doi.org/10.1111/nph.17914>, 2023.

349 Coll, J. R., Aguilar, E., and Ashcroft, L.: Drought variability and change across the Iberian Peninsula.
350 *Theoretical and Applied Climatology*, 130(3–4), 901–916. <https://doi.org/10.1007/s00704-016-1926-3>, 2017.

352 Cuthbert, M. O., Gleeson, T., Moosdorf, N., Befus, K. M., Schneider, A., Hartmann, J., and Lehner, B.:
353 Global patterns and dynamics of climate-groundwater interactions, *Nature Clim Change*, 9, 137–141,
354 <https://doi.org/10.1038/s41558-018-0386-4>, 2019.

355 Dalin, C., Wada, Y., Kastner, T., and Puma, M. J.: Groundwater depletion embedded in international food
356 trade, *Nature*, 543, 700–704, <https://doi.org/10.1038/nature21403>, 2017.

357 de Graaf, I. E. M., Sutanudjaja, E. H., van Beek, L. P. H., and Bierkens, M. F. P.: A high-resolution global-
358 scale groundwater model, *Hydrol. Earth Syst. Sci.*, 19, 823–837, <https://doi.org/10.5194/hess-19-823-2015>, available at: <https://hess.copernicus.org/articles/19/823/2015/>, 2015.

360 de Graaf, I., Rens L.P.H. van Beek, Tom Gleeson, Nils Moosdorf, Oliver Schmitz, Edwin H. Sutanudjaja,
361 and Marc F.P. Bierkens: A global-scale two-layer transient groundwater model: Development and
362 application to groundwater depletion, *Advances in Water Resources*, 102, 53–67,
363 <https://doi.org/10.1016/j.advwatres.2017.01.011>, 2017.

364 de Graaf, I. E. M., Gleeson, T., van Rens Beek, L. P. H., Sutanudjaja, E. H., and Bierkens, M. F. P.:
365 Environmental flow limits to global groundwater pumping, *Nature*, 574, 90–94,
366 <https://doi.org/10.1038/s41586-019-1594-4>, 2019.

367 Döll, P. and Siebert, S.: Global modeling of irrigation water requirements, *Water Resources Research*, 38,
368 <https://doi.org/10.1029/2001WR000355>, 2002.

369 Doppers, B., Franssen, W. H. P., van Vliet, M. T. H., Nijssen, B., and Ludwig, F.: Simulating human
370 impacts on global water resources using VIC-5, *Geosci. Model Dev.*, 13, 5029–5052,
371 <https://doi.org/10.5194/gmd-13-5029-2020>, 2020.



372 Dunne, J. P., Horowitz, L. W., Adcroft, A. J., Ginoux, P., Held, I. M., John, J. G., Krasting, J. P., Malyshev, S.,
373 Naik, V., Paulot, F., Shevliakova, E., Stock, C. A., Zadeh, N., Balaji, V., Blanton, C., Dunne, K. A.,
374 Dupuis, C., Durachta, J., Dussin, R., Gauthier, P. P. G., Griffies, S. M., Guo, H., Hallberg, R. W.,
375 Harrison, M., He, J., Hurlin, W., McHugh, C., Menzel, R., Milly, P. C. D., Nikonorov, S., Paynter, D. J.,
376 Ploschay, J., Radhakrishnan, A., Rand, K., Reichl, B. G., Robinson, T., Schwarzkopf, D. M., Sentman, L.
377 T., Underwood, S., Vahlenkamp, H., Winton, M., Wittenberg, A. T., Wyman, B., Zeng, Y. and Zhao, M.:
378 The GFDL Earth System Model Version 4.1 (GFDL-ESM 4.1): Overall Coupled Model Description and
379 Simulation Characteristics. *J Adv Model Earth Syst*, 12 (11), e2019MS002015.
380 <https://doi.org/10.1029/2019MS002015>, 2020. Eilander, D., van Verseveld, W., Yamazaki, D., Weerts,
381 A., Winsemius, H. C., and Ward, P. J.: A hydrography upscaling method for scale-invariant
382 parametrization of distributed hydrological models, *Hydrol. Earth Syst. Sci.*, 25, 5287–5313,
383 <https://doi.org/10.5194/hess-25-5287-2021>, 2021.
384 Esteban, E. J. L., Castilho, C. V., Melgaço, K. L., and Costa, F. R. C.: The other side of droughts: wet
385 extremes and topography as buffers of negative drought effects in an Amazonian forest, *The New
386 phytologist*, 229, 1995–2006, <https://doi.org/10.1111/nph.17005>, 2021.
387 Eyring, V., Bony, S., Meehl, G. A., Senior, C. A., Stevens, B., Stouffer, R. J., and Taylor, K. E.: Overview of
388 the Coupled Model Intercomparison Project Phase 6 (CMIP6) experimental design and organization,
389 *Geosci. Model Dev.*, 9, 1937–1958, <https://doi.org/10.5194/gmd-9-1937-2016>, 2016.
390 Felfelani, F., Lawrence, D. M., and Pokhrel, Y.: Representing Intercell Lateral Groundwater Flow and
391 Aquifer Pumping in the Community Land Model, *Water Resources Research*, 57,
392 <https://doi.org/10.1029/2020WR027531>, 2021.
393 Foster, S. S. D. and Chilton, P. J.: Groundwater: the processes and global significance of aquifer
394 degradation, *Philosophical transactions of the Royal Society of London. Series B, Biological
395 sciences*, 358, 1957–1972, <https://doi.org/10.1098/rstb.2003.1380>, 2003.
396 Frieler, K., Lange, S., Piontek, F., Reyer, C. P. O., Schewe, J., Warszawski, L., Zhao, F., Chini, L., Denvil, S.,
397 Emanuel, K., Geiger, T., Halladay, K., Hurt, G., Mengel, M., Murakami, D., Ostberg, S., Popp, A., Riva,
398 R., Stevanovic, M., Suzuki, T., Volkholz, J., Burke, E., Ciais, P., Ebi, K., Eddy, T. D., Elliott, J., Galbraith,
399 E., Gosling, S. N., Hattermann, F., Hickler, T., Hinkel, J., Hof, C., Huber, V., Jägermeyr, J., Krysanova,
400 V., Marcé, R., Müller Schmied, H., Mouratiadou, I., Pierson, D., Tittensor, D. P., Vautard, R., van Vliet,
401 M., Biber, M. F., Betts, R. A., Bodirsky, B. L., Deryng, D., Frolking, S., Jones, C. D., Lotze, H. K., Lotze-
402 Campen, H., Sahajpal, R., Thonicke, K., Tian, H., and Yamagata, Y.: Assessing the impacts of
403 1.5°C global warming - simulation protocol of the Inter-Sectoral Impact Model
404 Intercomparison Project (ISIMIP2b), *Geosci. Model Dev.*, 10, 4321–4345,
405 <https://doi.org/10.5194/gmd-10-4321-2017>, 2017.
406 Gasper, F., Goergen, K., Shrestha, P., Sulis, M., Rihani, J., Geimer, M., and Kollet, S.: Implementation and
407 scaling of the fully coupled Terrestrial Systems Modeling Platform (TerrSysMP v1.0) in a massively
408 parallel supercomputing environment – a case study on JUQUEEN (IBM Blue Gene/Q), *Geosci. Model
409 Dev.*, 7, 2531–2543, <https://doi.org/10.5194/gmd-7-2531-2014>, 2014.



410 Gilbert, M., Nicolas, G., Cinardi, G., van Boeckel, T. P., Vanwambeke, S. O., Wint, G. R. W., and Robinson,
411 T. P.: Global distribution data for cattle, buffaloes, horses, sheep, goats, pigs, chickens and ducks in
412 2010, *Scientific data*, 5, 180227, <https://doi.org/10.1038/sdata.2018.227>, 2018.

413 Gleeson, T. and Paszkowski, D.: Perceptions of scale in hydrology: what do you mean by regional scale?,
414 *Hydrological Sciences Journal*, 59, 99–107, <https://doi.org/10.1080/02626667.2013.797581>, 2014.

415 Gleeson, T., Wagener, T., Döll, P., Zipper, S. C., West, C., Wada, Y., Taylor, R., Scanlon, B., Rosolem, R.,
416 Rahman, S., Oshinlaja, N., Maxwell, R., Lo, M.-H., Kim, H., Hill, M., Hartmann, A., Fogg, G.,
417 Famiglietti, J. S., Ducharne, A., Graaf, I. de, Cuthbert, M., Condon, L., Bresciani, E., and Bierkens, M.
418 F. P.: GMD perspective: The quest to improve the evaluation of groundwater representation in
419 continental- to global-scale models, *Geosci. Model Dev.*, 14, 7545–7571,
420 <https://doi.org/10.5194/gmd-14-7545-2021>, 2021.

421 Gnann, S., Reinecke, R., Stein, L., Wada, Y., Thiery, W., Müller Schmied, H., Satoh, Y., Pokhrel, Y.,
422 Ostberg, S., Koutroulis, A., Hanasaki, N., Grillakis, M., Gosling, S., Burek, P., Bierkens, M., and
423 Wagener, T.: Functional relationships reveal differences in the water cycle representation of global
424 water models, 1079–1090, <https://doi.org/10.1038/s44221-023-00160-y>, 2023.

425 Grogan, D. S., Zuidema, S., Prusevich, A., Wollheim, W. M., Glidden, S., and Lammers, R. B.: Water
426 balance model (WBM) v.1.0.0: a scalable gridded global hydrologic model with water-tracking
427 functionality, *Geosci. Model Dev.*, 15, 7287–7323, <https://doi.org/10.5194/gmd-15-7287-2022>, 2022.

428 Guillaumot, L., Smilovic, M., Burek, P., Bruijn, J. de, Greve, P., Kahil, T., and Wada, Y.: Coupling a large-
429 scale hydrological model (CWatM v1.1) with a high-resolution groundwater flow model (MODFLOW 6)
430 to assess the impact of irrigation at regional scale, *Geosci. Model Dev.*, 15, 7099–7120,
431 <https://doi.org/10.5194/gmd-15-7099-2022>, 2022.

432 Hansen, M. and Toftemann Thomsen, C.: An integrated public information system for geology,
433 groundwater and drinking water in Denmark, *GEUS Bulletin*, 38, 69–72,
434 <https://doi.org/10.34194/geusb.v38.4423>, 2020.

435 Haqiqi, I., Bowling, L., Jame, S., Baldos, U., Liu, J., and Hertel, T.: Global drivers of local water stresses
436 and global responses to local water policies in the United States, *Environmental research letters ERL*,
437 18, 65007, <https://doi.org/10.1088/1748-9326/acd269>, 2023.

438 Holman, I. P., Whelan, M. J., Howden, N. J. K., Bellamy, P. H., Willby, N. J., Rivas-Casado, M., and
439 McConvey, P.: Phosphorus in groundwater—an overlooked contributor to eutrophication?, *Hydrol.*
440 *Process.*, 22, 5121–5127, <https://doi.org/10.1002/hyp.7198>, 2008.

441 Huggins, X., Gleeson, T., Castilla-Rho, J., Holley, C., Re, V., and Famiglietti, J. S.: Groundwater
442 Connections and Sustainability in Social-Ecological Systems, *Ground water*, 61, 463–478,
443 <https://doi.org/10.1111/gwat.13305>, 2023.

444 Hurtt, G. C., Chini, L., Sahajpal, R., Frolking, S., Bodirsky, B. L., Calvin, K., Doelman, J. C., Fisk, J.,
445 Fujimori, S., Klein Goldewijk, K., Hasegawa, T., Havlik, P., Heinemann, A., Humpenöder, F., Jungclaus,
446 J., Kaplan, J. O., Kennedy, J., Krisztin, T., Lawrence, D., Lawrence, P., Ma, L., Mertz, O., Pongratz, J.,
447 Popp, A., Poulter, B., Riahi, K., Shevliakova, E., Stehfest, E., Thornton, P., Tubiello, F. N., van Vuuren,



448 D. P., and Zhang, X.: Harmonization of global land use change and management for the period 850–
449 2100 (LUH2) for CMIP6, *Geosci. Model Dev.*, 13, 5425–5464, <https://doi.org/10.5194/gmd-13-5425-2020>, 2020.

450

451 Jasechko, S., Seybold, H., Perrone, D., Fan, Y., Shamsuddoha, M., Taylor, R. G., Fallatah, O., and
452 Kirchner, J. W.: Rapid groundwater decline and some cases of recovery in aquifers globally, *Nature*,
453 625, 715–721, <https://doi.org/10.1038/s41586-023-06879-8>, 2024.

454 Keune, J., Sulis, M., Kollet, S., Siebert, S., and Wada, Y.: Human Water Use Impacts on the Strength of the
455 Continental Sink for Atmospheric Water, *Geophys. Res. Lett.*, 45, 4068–4076,
456 <https://doi.org/10.1029/2018GL077621>, 2018.

457 Kollet, S. J. and Maxwell, R. M.: Capturing the influence of groundwater dynamics on land surface
458 processes using an integrated, distributed watershed model, *Water Resources Research*, 44,
459 <https://doi.org/10.1029/2007WR006004>, 2008.

460 Konar, M., Hussein, Z., Hanasaki, N., Mauzerall, D. L., and Rodriguez-Iturbe, I.: Virtual water trade flows
461 and savings under climate change, *Hydrol. Earth Syst. Sci.*, 17, 3219–3234,
462 <https://doi.org/10.5194/hess-17-3219-2013>, 2013.

463 Kretschmer, D. V., Michael, H., Moosdorf, N., Essink, G. O., Bierkens, M. F. P., Wagener, T., and
464 Reinecke, R.: Controls on coastal saline groundwater across North America, *Environmental research
465 letters*, <https://doi.org/10.1088/1748-9326/ada973>, 2025.

466 Kuffour, B. N. O., Engdahl, N. B., Woodward, C. S., Condon, L. E., Kollet, S., and Maxwell, R. M.:
467 Simulating coupled surface–subsurface flows with ParFlow v3.5.0: capabilities, applications, and
468 ongoing development of an open-source, massively parallel, integrated hydrologic model, *Geosci.
469 Model Dev.*, 13, 1373–1397, <https://doi.org/10.5194/gmd-13-1373-2020>, 2020.

470 Lammers, R. B.: Global Inter-Basin Hydrological Transfer Database, 2022.

471 Lange, S.: Bias-correction fact sheet, <https://www.isimip.org/gettingstarted/isimip3b-bias-adjustment/>,
472 last access: 2 March 2025, 2021.

473 Lawrence, D. M., Fisher, R. A., Koven, C. D., Oleson, K. W., Swenson, S. C., Bonan, G., Collier, N.,
474 Ghimire, B., van Kampenhout, L., Kennedy, D., Kluzeck, E., Lawrence, P. J., Li, F., Li, H., Lombardozzi,
475 D., Riley, W. J., Sacks, W. J., Shi, M., Vertenstein, M., Wieder, W. R., Xu, C., Ali, A. A., Badger, A. M.,
476 Bisht, G., van den Broeke, M., Brunke, M. A., Burns, S. P., Buzan, J., Clark, M., Craig, A., Dahlin, K.,
477 Drewniak, B., Fisher, J. B., Flanner, M., Fox, A. M., Gentine, P., Hoffman, F., Keppel-Aleks, G., Knox,
478 R., Kumar, S., Lenaerts, J., Leung, L. R., Lipscomb, W. H., Lu, Y., Pandey, A., Pelletier, J. D., Perket, J.,
479 Randerson, J. T., Ricciuto, D. M., Sanderson, B. M., Slater, A., Subin, Z. M., Tang, J., Thomas, R. Q., Val
480 Martin, M., and Zeng, X.: The Community Land Model Version 5: Description of New Features,
481 Benchmarking, and Impact of Forcing Uncertainty, *J Adv Model Earth Syst.*, 11, 4245–4287,
482 <https://doi.org/10.1029/2018MS001583>, 2019.

483 Lehner, B., Liermann, C. R., Revenga, C., Vörösmarty, C., Fekete, B., Crouzet, P., Döll, P., Endejan, M.,
484 Frenken, K., Magome, J., Nilsson, C., Robertson, J. C., Rödel, R., Sindorf, N., and Wisser, D.: High-



485 resolution mapping of the world's reservoirs and dams for sustainable river-flow management,
486 *Frontiers in Ecol & Environ*, 9, 494–502, <https://doi.org/10.1890/100125>, 2011.

487 Li, B., Rodell, M., and Beaudoing, H.: GLDAS Catchment Land Surface Model L4 daily 0.25 x 0.25 degree,
488 Version 2.0, 2018.

489 Liang, X., Lettenmaier, D. P., Wood, E. F., and Burges, S. J.: A simple hydrologically based model of land
490 surface water and energy fluxes for general circulation models. *Journal of Geophysical Research: Atmospheres*, 99(D7), 14415–14428. <https://doi.org/10.1029/94JD00483>, 1994.

491 Liu, Y., Wagener, T., Beck, H. E., and Hartmann, A.: What is the hydrologically effective area of a
492 catchment?, *Environmental research letters ERL*, 15, 104024, <https://doi.org/10.1088/1748-9326/aba7e5>, 2020.

493 Lloyd, C. T., Chamberlain, H., Kerr, D., Yetman, G., Pistolesi, L., Stevens, F. R., Gaughan, A. E., Nieves, J.
494 J., Hornby, G., MacManus, K., Sinha, P., Bondarenko, M., Sorichetta, A., and Tatem, A. J.: Global
495 spatio-temporally harmonised datasets for producing high-resolution gridded population distribution
496 datasets, *Big earth data*, 3, 108–139, <https://doi.org/10.1080/20964471.2019.1625151>, 2019.

497 Maxwell, R. M. and Miller, N. L.: Development of a Coupled Land Surface and Groundwater Model,
498 *Journal of Hydrometeorology*, 6, 233–247, <https://doi.org/10.1175/JHM422.1>, 2005.

499 Maxwell, R. M., Condon, L. E., Kollet, S. J., Maher, K., Haggerty, R., and Forrester, M. M.: The imprint of
500 climate and geology on the residence times of groundwater, *Geophys. Res. Lett.*, 43, 701–708,
501 <https://doi.org/10.1002/2015GL066916>, 2016.

502 Maxwell, R. M., Lundquist, J. K., Mirocha, J. D., Smith, S. G., Woodward, C. S., and Tompson, A. F. B.:
503 Development of a Coupled Groundwater–Atmosphere Model, *Monthly Weather Review*, 139, 96–116,
504 <https://doi.org/10.1175/2010MWR3392.1>, 2011.

505 Moeck, C., Collenteur, R. A., Berghuijs, W. R., Luijendijk, E., and Gurdak, J. J.: A Global Assessment of
506 Groundwater Recharge Response to Infiltration Variability at Monthly to Decadal Timescales, *Water
507 Resources Research*, 60, <https://doi.org/10.1029/2023WR035828>, 2024.

508 Müller Schmied, H., Cáceres, D., Eisner, S., Flörke, M., Herbert, C., Niemann, C., Peiris, T. A., Popat, E.,
509 Portmann, F. T., Reinecke, R., Schumacher, M., Shadkam, S., Telteu, C.-E., Trautmann, T., and Döll,
510 P.: The global water resources and use model WaterGAP v2.2d: model description and evaluation,
511 *Geosci. Model Dev.*, 14, 1037–1079, <https://doi.org/10.5194/gmd-14-1037-2021>, 2021.

512 Müller Schmied, H., Adam, L., Eisner, S., Fink, G., Flörke, M., Kim, H., Oki, T., Portmann, F. T., Reinecke,
513 Riedel, C., Song, Q., Zhang, J., and Döll, P.: Variations of global and continental water balance
514 components as impacted by climate forcing uncertainty and human water use, *Hydrol. Earth Syst.
515 Sci.*, 20, 2877–2898, <https://doi.org/10.5194/hess-20-2877-2016>, available at:
516 <https://hess.copernicus.org/articles/20/2877/2016/>, 2016.

517 Naz, B. S., Sharples, W., Ma, Y., Goergen, K., and Kollet, S.: Continental-scale evaluation of a fully
518 distributed coupled land surface and groundwater model, ParFlow-CLM (v3.6.0), over Europe,
519 *Geosci. Model Dev.*, 16, 1617–1639, <https://doi.org/10.5194/gmd-16-1617-2023>, 2023.



522 Nazari, S., Kruse, I. L., and Moosdorf, N.: Spatiotemporal dynamics of global rain-fed groundwater
523 recharge from 2001 to 2020, *Journal of Hydrology*, 650, 132490,
524 <https://doi.org/10.1016/j.jhydrol.2024.132490>, 2025.

525 Paneque Salgado, P., and Vargas Molina, J.: Drought, social agents and the construction of discourse in
526 Andalusia. *Environmental Hazards*, 14(3), 224–235.
527 <https://doi.org/10.1080/17477891.2015.1058739>, 2015.

528 Perez, N., Singh, V., Ringler, C., Xie, H., Zhu, T., Sutanudjaja, E. H., and Villholth, K. G.: Ending
529 groundwater overdraft without affecting food security, *Nat Sustain*, 7, 1007–1017,
530 <https://doi.org/10.1038/s41893-024-01376-w>, 2024.

531 Pokhrel, Y., Felfelani, F., Satoh, Y., Boulange, J., Burek, P., Gádeke, A., Gerten, D., Gosling, S. N.,
532 Grillakis, M., Gudmundsson, L., Hanasaki, N., Kim, H., Koutroulis, A., Liu, J., Papadimitriou, L.,
533 Schewe, J., Müller Schmied, H., Stacke, T., Telteu, C.-E., Thiery, W., Veldkamp, T., Zhao, F., and
534 Wada, Y.: Global terrestrial water storage and drought severity under climate change, *Nature Clim
535 Change*, 11, 226–233, <https://doi.org/10.1038/s41558-020-00972-w>, 2021.

536 Porkka, M., Virkki, V., Wang-Erlandsson, L., Gerten, D., Gleeson, T., Mohan, C., Fetzer, I., Jaramillo, F.,
537 Staal, A., te Wierik, S., Tobian, A., van der Ent, R., Döll, P., Flörke, M., Gosling, S. N., Hanasaki, N.,
538 Satoh, Y., Müller Schmied, H., Wanders, N., Famiglietti, J. S., Rockström, J., and Kummu, M.: Notable
539 shifts beyond pre-industrial streamflow and soil moisture conditions transgress the planetary
540 boundary for freshwater change, *Nat Water*, 2, 262–273, <https://doi.org/10.1038/s44221-024-00208-z>, 2024.

541

542 Prusevich, A. A., Lammers, R. B., and Glidden, S. J.: Delineation of endorheic drainage basins in the
543 MERIT-Plus dataset for 5 and 15 minute upscaled river networks, *Scientific data*, 11, 61,
544 <https://doi.org/10.1038/s41597-023-02875-9>, 2024.

545 Reinecke, R., Müller Schmied, H., Trautmann, T., Andersen, L. S., Burek, P., Flörke, M., Gosling, S. N.,
546 Grillakis, M., Hanasaki, N., Koutroulis, A., Pokhrel, Y., Thiery, W., Wada, Y., Yusuke, S., and Döll, P.:
547 Uncertainty of simulated groundwater recharge at different global warming levels: a global-scale
548 multi-model ensemble study, *Hydrol. Earth Syst. Sci.*, 25, 787–810, <https://doi.org/10.5194/hess-25-787-2021>, available at: <https://hess.copernicus.org/articles/25/787/2021/>, 2021.

549

550 Reinecke, R., Foglia, L., Mehl, S., Trautmann, T., Cáceres, D., and Döll, P.: Challenges in developing a
551 global gradient-based groundwater model (G³M v1.0) for the integration into a global hydrological
552 model, *Geosci. Model Dev.*, 12, 2401–2418, <https://doi.org/10.5194/gmd-12-2401-2019>, 2019.

553 Reinecke, R., Gnann, S., Stein, L., Bierkens, M., Graaf, I. de, Gleeson, T., Essink, G. O., Sutanudjaja, E. H.,
554 Ruz Vargas, C., Verkaik, J., and Wagener, T.: Uncertainty in model estimates of global groundwater
555 depth, *Environmental research letters ERL*, 19, 114066, <https://doi.org/10.1088/1748-9326/ad8587>,
556 2024.

557 Reinecke, R.: The ISIMIP Groundwater Sector: A Framework for Ensemble Modeling of Global Change
558 Impacts on Groundwater, *Zenodo*, <https://doi.org/10.5281/zenodo.14962512>, 2025.



559 Rodell, M., Velicogna, I., and Famiglietti, J. S.: Satellite-based estimates of groundwater depletion in
560 India, *Nature*, 460, 999–1002, <https://doi.org/10.1038/nature08238>, 2009.

561 Rodella, A.-S., Zaveri, E., and Bertone, F.: *The Hidden Wealth of Nations: The Economics of Groundwater*
562 in Times of Climate Change. Washington, DC: World Bank, 2023.

563 Rounce, D., Hock, R., and Maussion, F.: Global PyGEM-OGGM Glacier Projections with RCP and SSP
564 Scenarios, Version 1, 2022.

565 Sarrazin, F., Hartmann, A., Pianosi, F., Rosolem, R., and Wagener, T.: V2Karst V1.1: a parsimonious large-
566 scale integrated vegetation–recharge model to simulate the impact of climate and land cover change
567 in karst regions, *Geosci. Model Dev.*, 11, 4933–4964, <https://doi.org/10.5194/gmd-11-4933-2018>,
568 2018.

569 Scanlon, B. R., Fakhreddine, S., Rateb, A., Graaf, I. de, Famiglietti, J., Gleeson, T., Grafton, R. Q., Jobbagy,
570 E., Kebede, S., Koluś, S. R., Konikow, L. F., Di Long, Mekonnen, M., Schmid, H. M., Mukherjee, A.,
571 MacDonald, A., Reedy, R. C., Shamsudduha, M., Simmons, C. T., Sun, A., Taylor, R. G., Villholth, K.
572 G., Vörösmarty, C. J., and Zheng, C.: Global water resources and the role of groundwater in a resilient
573 water future, *Nat Rev Earth Environ*, 4, 87–101, <https://doi.org/10.1038/s43017-022-00378-6>, 2023.

574 Schaller, M. F. and Fan, Y.: River basins as groundwater exporters and importers: Implications for water
575 cycle and climate modeling, *J. Geophys. Res.*, 114, <https://doi.org/10.1029/2008JD010636>, 2009.

576 Schwartz, F. W. and Ibaraki, M.: Groundwater: A Resource in Decline, *Elements*, 7, 175–179,
577 <https://doi.org/10.2113/gselements.7.3.175>, 2011.

578 Smith, M. W., Willis, T., Mroz, E., James, W. H. M., Klaar, M. J., Gosling, S. N., and Thomas, C. J.: Future
579 malaria environmental suitability in Africa is sensitive to hydrology, *Science* (New York, N.Y.), 384,
580 697–703, <https://doi.org/10.1126/science.adk8755>, 2024.

581 Taylor, R. G., Scanlon, B., Döll, P., Rodell, M., van Beek, R., Wada, Y., Longuevergne, L., Leblanc, M.,
582 Famiglietti, J. S., Edmunds, M., Konikow, L., Green, T. R., Chen, J., Taniguchi, M., Bierkens, M. F. P.,
583 MacDonald, A., Fan, Y., Maxwell, R. M., Yechieli, Y., Gurdak, J. J., Allen, D. M., Shamsudduha, M.,
584 Hiscock, K., Yeh, P. J.-F., Holman, I., and Treidel, H.: Ground water and climate change, *Nature Clim
585 Change*, 3, 322–329, <https://doi.org/10.1038/nclimate1744>, 2013.

586 Telteu, C.-E., Müller Schmid, H., Thiery, W., Leng, G., Burek, P., Liu, X., Boulange, J. E. S., Andersen, L.
587 S., Grillakis, M., Gosling, S. N., Satoh, Y., Rakovec, O., Stacke, T., Chang, J., Wanders, N., Shah, H. L.,
588 Trautmann, T., Mao, G., Hanasaki, N., Koutoulis, A., Pokhrel, Y., Samaniego, L., Wada, Y., Mishra, V.,
589 Liu, J., Döll, P., Zhao, F., Gádeke, A., Rabin, S. S., and Herz, F.: Understanding each other's models:
590 an introduction and a standard representation of 16 global water models to support intercomparison,
591 improvement, and communication, *Geosci. Model Dev.*, 14, 3843–3878,
592 <https://doi.org/10.5194/gmd-14-3843-2021>, 2021.

593 Thompson, J. R., Gosling, S. N., Zaherpour, J., and Laizé, C. L. R.: Increasing Risk of Ecological Change to
594 Major Rivers of the World With Global Warming, *Earth's Future*, 9,
595 <https://doi.org/10.1029/2021EF002048>, 2021.



596 Trullenque-Blanco, V., Beguería, S., Vicente-Serrano, S. M., Peña-Angulo, D., and González-Hidalgo, C.:
597 Catalogue of drought events in peninsular Spanish along 1916–2020 period. *Scientific Data*, 11(1),
598 703. <https://doi.org/10.1038/s41597-024-03484-w>, 2024.

599 United Nations: The United Nations World Water Development Report 2022: groundwater: making the
600 invisible visible, UNESCO, Paris, 2022.

601 Winter, T. C.: The Role of Ground Water in Generating Streamflow in Headwater Areas and in Maintaining
602 Base Flow 1, *J American Water Resour Assoc*, 43, 15–25, <https://doi.org/10.1111/j.1752-1688.2007.00003.x>, 2007.

604 Xie, J., Liu, X., Jasechko, S., Berghuijs, W. R., Wang, K., Liu, C., Reichstein, M., Jung, M., and Koirala, S.:
605 Majority of global river flow sustained by groundwater, *Nature Geosci*, 17, 770–777,
606 <https://doi.org/10.1038/s41561-024-01483-5>, 2024.

607 Yamazaki, D., Ikeshima, D., Sosa, J., Bates, P. D., Allen, G. H., and Pavelsky, T. M.: MERIT Hydro: A High-
608 Resolution Global Hydrography Map Based on Latest Topography Dataset, *Water Resources
609 Research*, 55, 5053–5073, <https://doi.org/10.1029/2019WR024873>, 2019.

610 Yang, Y., Donohue, R. J., and McVicar, T. R.: Global estimation of effective plant rooting depth:
611 Implications for hydrological modeling, *Water Resources Research*, 52, 8260–8276,
612 <https://doi.org/10.1002/2016WR019392>, 2016.