

A review of open data for studying global groundwater in social-ecological systems

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72 **Abstract**

73 Global data have served an integral role in characterizing large-scale groundwater systems,
74 identifying their sustainability challenges, and informing on socioeconomic and ecological
75 dimensions of groundwater. These insights have revealed groundwater as a dynamic component
76 of both the water cycle and social-ecological systems, leading to an expansion in groundwater
77 science that increasingly focuses on interactions between groundwater with ecological,
78 socioeconomic, and Earth systems. This shift presents many opportunities that are conditional on
79 broader, more interdisciplinary system conceptualizations, models, and methods that require the
80 integration of a greater diversity of data in contrast to conventional hydrogeological investigations.
81 Here, we catalogue 144 global open access datasets and dataset collections relevant to
82 groundwater science that span elements of the hydrosphere, biosphere, atmosphere, lithosphere,
83 food systems, governance, management, and other socioeconomic system dimensions. The
84 assembled catalogue offers a reference of existing data for use in interdisciplinary assessments,
85 and we summarize these data across their primary system, spatial resolution, temporal range,
86 data type, generation method, level of groundwater representation, and institutional location of
87 lead authorship. The catalogue includes 15 groundwater datasets, 23 datasets explicitly linked
88 with groundwater, and 106 datasets with implicit or potential groundwater connections. We find
89 the majority of datasets are temporally static and that temporally dynamic data availability
90 currently peaks during the 2000-2010 decade. Only a small fraction of temporally dynamic data
91 are explicitly linked to groundwater, representing a significant opportunity for future work to
92 address. We find that most groundwater datasets are generated by a small number of countries,
93 including the USA, Germany, the Netherlands, and Canada. We raise three themes of possible
94 priorities for future global groundwater data initiatives, which include: data improvements through
95 more explicit integration of groundwater and prioritizing observed and temporally dynamic data;
96 elevating regional and local scale data and perspectives to address challenges relating to equity
97 and bias; and advancing and promoting data sharing initiatives founded on reciprocal benefits
98 between global initiatives and data providers.

99 **1. Introduction**

100 Groundwater, a critical resource for drinking water, agriculture, and ecosystems, is under
101 increasing pressure from human activities and climate change (Famiglietti 2014, Abbott *et al*
102 2019, Gleeson *et al* 2020, Kuang *et al* 2024, Scanlon *et al* 2023, Jasechko *et al* 2024, Reinecke
103 *et al* 2024, Taylor *et al* 2013, Bierkens and Wada 2019). Recognizing groundwater connections
104 in social-ecological systems has been proposed as a more holistic approach to address the
105 complexities of groundwater's evolving role in the Anthropocene and support equitable
106 groundwater management (Huggins *et al* 2023, Kuang *et al* 2024). Applying this understanding
107 requires a broader conceptualization where hydrogeological systems are not understood as
108 stand-alone resource systems but rather as systems embedded within a network of
109 socioeconomic, ecological, and Earth systems (Gleeson and Cardiff 2013, Huggins *et al* 2023).
110 This conceptual extension of groundwater systems to a more holistic social-ecological model is
111 supported by both groundwater and social-ecological systems theory (Berkes *et al* 1998, Zellner

112 2008) and implicitly by a wide and yet unquantified variety and volume of data. Understanding the
113 scope of data available to study groundwater in social-ecological systems is necessary to inform
114 data generation and sharing priorities, and more broadly support theory development of
115 groundwater in the Anthropocene.

116 There is a long history of scholarship on groundwater as a common-pool resource (Ostrom 1990,
117 Blomquist *et al* 1994), and there is a rich literature detailing the benefits of applying social-
118 ecological system framings to understand groundwater within social, economic, human, and
119 ecological contexts (Rica *et al* 2017, Barreteau *et al* 2016, Bouchet *et al* 2019). The recent
120 groundwater-connected systems framing (Huggins *et al* 2023), which identifies and details
121 groundwater systems and processes across elements of the Social-Ecological Systems
122 Framework (McGinnis and Ostrom 2014), provides the thematic scope of this review. Whereas
123 conventional groundwater investigations typically integrate hydrogeological, climatic, and
124 topographic data (e.g., hydraulic conductivity, precipitation, and land surface elevation),
125 assessments of social-ecological systems require an integrated consideration of biophysical and
126 social systems that extend beyond this scope to include data on ecosystems, governance,
127 economic activity, and the broader socioeconomic context (e.g., environmental flows,
128 groundwater institutions, groundwater irrigation, and human development). This social-ecological
129 framing is centered on groundwater interactions with connected systems and thus foregrounds
130 groundwater's diverse roles in systems such as the Earth system (Gleeson *et al* 2020), human
131 health (Wang *et al* 2023), food systems (Siebert *et al* 2010, Dalin *et al* 2019) and cultures (Re
132 2015, Zwarteveen *et al* 2021). We aim to capture existing open data availability across all of these
133 systems as they relate to groundwater in this review.

134 Groundwater-connected systems operate across a range of scales from the local to the global:
135 stream-aquifer interactions at the reach scale (Brunner 2017, Yang *et al* 2025), basin scale
136 management agencies and actions (e.g., Groundwater Sustainability Agencies in California,
137 USA), groundwater-dependent ecosystems sustained by regional groundwater flow (Aldous and
138 Gannett 2021, Yao *et al* 2018), transboundary aquifer governance (Shaminder and Villholth
139 2017), international virtual water trade networks (Dalin *et al* 2017), and climate change drivers of
140 regional groundwater storage (Wu *et al* 2020). In recognition of the interconnectedness between
141 these processes and the widespread nature of groundwater challenges, a global research agenda
142 on groundwater has emerged over recent decades (Konikow 2005, Foster *et al* 2013, Giordano
143 2009, Gleeson *et al* 2020; Kuang *et al* 2024) that is rooted in the development of global
144 sustainability frameworks and enabled by growth in data and computational methods and abilities.
145 To support this agenda, we focus this review on data with global spatial coverage. This focus on
146 data with global coverage is tailored to support analyses on systematic comparison between
147 regions, Earth system processes and interactions with groundwater (Gleeson *et al* 2020),
148 evaluation or validation of continental and global models (Gleeson *et al* 2021, Gnann *et al* 2023),
149 and to mirror the extent of the broader global groundwater sustainability discourse (Famiglietti,
150 2014, Gleeson *et al* 2020; Scanlon *et al* 2023, Mukherjee 2024). Furthermore, these global
151 datasets can be used as place-holder data in regions where localized data are either unavailable
152 or of insufficient quality.

153 Here, we assemble and review a large catalogue of open-access datasets to support this growing
154 global research agenda on groundwater in social-ecological systems. Open science is “perhaps
155 the most important paradigm shift in the recent history of scholarly publishing” (Clark *et al* 2021)
156 through its democratization of information access and its promotion of transparency and
157 reproducibility, among other benefits. Thus, we limit our review to open access datasets to align
158 this work with the open science movement and to encourage the open sharing of global datasets
159 in future work. While we focus on global data, we note that these data can represent processes
160 that operate across a variety of spatial and temporal scales. To review the catalogue, we derive
161 a wide variety of metadata including the primary system to which the dataset relates, the dataset’s
162 spatial resolution, temporal range, level of groundwater representation, data generation method
163 and data format.

164 This review of global data doubles as an opportunity to assess geographic trends in institutional
165 data authorship. Clear biases towards institutions in the global North have been identified in
166 climate, environmental, and conservation sciences (Karlsson *et al* 2007; Maas *et al* 2021, Hazlett
167 *et al* 2020), and for study areas of hydrological climate hazards (Stein *et al* 2024). Similar biases
168 have been recently identified for groundwater modelling regarding model extents and with respect
169 to non-local model development (Zamrsky *et al* 2025), however these forms of bias have yet to
170 be explored and discussed with respect to global groundwater data. Understanding which regions
171 are driving global groundwater data development, including the coordination of global data
172 sharing initiatives, can be instructive to evaluate representation of regional values and needs in
173 these processes. Further, this analysis can be used as a basis to explore potential opportunities
174 and tensions between global groundwater data ambitions and local to regional datasets, priorities,
175 and realities.

176 Summarizing metadata across the compiled data catalogue enables the research questions listed
177 below to be investigated. After reviewing these outcomes, we discuss a suite of needs and
178 priorities for future groundwater data efforts that address identified limitations and opportunities.

- 179 ● How many global datasets and dataset collections are openly accessible for studying
180 groundwater in social-ecological systems?
- 181 ● What is the distribution of datasets across social-ecological system elements, and is this
182 distribution balanced or biased toward certain elements?
- 183 ● How many datasets are temporally static, and how many are temporally dynamic?
- 184 ● What are the spatial (e.g., grid size or zonal unit) and temporal (e.g., time step) resolutions of
185 these datasets?
- 186 ● How explicitly is groundwater represented or integrated in dataset generation?
- 187 ● What is the national distribution of institutional authorship of these datasets, and how does this
188 distribution compare and relate to regional trends in groundwater challenges?

189 **2. Review methodology**

190 **2.1 Review intentions**

191 We seek to develop and review a large and representative catalogue of global datasets that are
192 available for the study of groundwater in social-ecological systems. The general approach and
193 structure of our data review is illustrated in Figure 1.

194 To ensure coverage across social-ecological system elements, we base our review in the
195 groundwater-connected systems framing that conceptualizes groundwater connections with
196 social-ecological systems. However, this framing is nascent and studying groundwater in social-
197 ecological contexts lacks an established self-identifying language and dedicated data
198 repositories. These realities introduce specific complications to our review including decentralized
199 source locations of potential datasets, and ambiguous boundaries.

200 Regarding dataset identification, we developed a multifaceted approach that considered diverse
201 sources and wide search criteria to locate relevant datasets (described in **2.2 Dataset**
202 **identification**). Regarding review scope, we were interested in not only identifying data that have
203 already been used to study global groundwater in hydrological and social-ecological system
204 contexts, but also in identifying data that have the potential for such applications. In total, we
205 catalogue datasets with thematic coverage across eight system types, including the hydrosphere,
206 lithosphere, biosphere, atmosphere, food systems, governance, other socioeconomic systems,
207 and an integrative category for datasets that span multiple systems (described in **2.3 Metadata**
208 **categories**). To fulfill this scope, a considerable amount of data with potential or implicit
209 connections and relevance to groundwater is included in the catalogue. Interpreting these
210 potential and implicit connections to groundwater requires subjective judgements, yet which are
211 unavoidable when engaging with a social-ecological framing (e.g., Andrachuk and Armitage,
212 2015; Lazurko *et al* 2024). For instance, data on language diversity may offer implicit insights into
213 the local complexity of groundwater governance and management or may provide a proxy
214 representation of the diversity of value systems relating to groundwater. Others, however, may
215 consider such data to hold little relevance for groundwater science.

216 Our large authorship team of global groundwater, Earth system, and social-ecological system
217 scientists hold a variety of perspectives that are reflected in our assembled catalogue. We thus
218 characterize the catalogue as a large and representative, but not exhaustive, resource. We
219 anticipate that the primary benefit of this initiative will be to serve as a resource for scholars,
220 practitioners, and others to identify data for use in future studies of groundwater in social-
221 ecological systems.

222 **2.2 Dataset identification**

223 We considered multiple sources when identifying datasets for inclusion in this review. We sought
224 to incorporate a wide range of sources to reflect the diverse locations where global geospatial
225 data are hosted online. Thus, we not only searched for datasets generated in publications, but

226 also screened input datasets used in thematically aligned global social-ecological system
227 assessments, leading global geospatial data platforms, compatible global data reviews, and
228 through crowd-sourcing additional inputs from this study's co-authorship. The full list of consulted
229 data sources and screening procedures are reported in Table 1.

230 Only open-access datasets are included in this review. We take this step to ensure the reviewed
231 datasets are accessible for use in future studies, to encourage data sharing practices, and to
232 broadly align this review with open science principles. In taking this step, this review implicitly
233 evaluates the scope of global groundwater data following *findable* and *accessible* data principles
234 from the FAIR initiative (Wilkinson *et al* 2016). Screening for this criterion biases toward data
235 generated within recent years (ca. 2015 and later) due to the relatively recent rise of open
236 publishing and data deposition practices (Clark *et al* 2021, Hall *et al* 2022). The data sharing
237 agreements for datasets that enable inclusion in this review include Creative Commons licenses,
238 dataset-specific user agreements, or an explicit statement encouraging the use of data where a
239 license or agreement was not readily identifiable.

240 **2.3 Metadata categories**

241 We developed several classification schemes to organize and evaluate datasets. These
242 classifications include: (1) the primary system to which the described dataset relates, (2) how
243 explicitly groundwater is represented in the dataset, (3) dataset type and (4) format, (5) spatial
244 resolution, (6) temporal range and time step for temporally dynamic data, (7) data generation
245 method, and (8) institutional country of lead authorship (Figure 1c).

246 To classify the primary *system* to which the described variable of each dataset relates, we
247 developed a composite classification scheme that combined elements from social-ecological and
248 Earth system frameworks. This composite scheme was developed to address disciplinary biases
249 in social-ecological and Earth system classifications. Our specification of individual Earth system
250 elements served to counteract the biophysical simplifications in social-ecological system schemes
251 (i.e., atmosphere, biosphere, hydrosphere, and lithosphere data would otherwise be classified
252 under the broad term of biophysical systems). Conversely, elements from the Social-Ecological
253 Systems Framework (SESF; McGinnis and Ostrom, 2014) helped to balance overgeneralized
254 human system representation in Earth system schemes (i.e., governance, food systems, and
255 other socioeconomic dimensions would otherwise be classified under the broad 'Anthroposphere'
256 term). This composite scheme thus consists of eight categories: hydrosphere, lithosphere,
257 biosphere, atmosphere, food systems, governance, other human and socioeconomic systems,
258 and an integrative category for datasets that span multiple systems (Figure 1a). We isolated food
259 systems to reflect the significance of groundwater-agriculture interactions at the global scale,
260 including agriculture representing the dominant driver of groundwater consumption globally, the
261 importance of groundwater for irrigation water supporting crop production and food security, the
262 magnitude of groundwater embedded in international food trade (Siebert *et al* 2010, Wada *et al*
263 2012, Dalin *et al* 2017), and the large volume of data on food systems in relation to other
264 socioeconomic sectors and human dimensions of groundwater.

265 To characterize the explicitness of groundwater *representation* in each dataset, we classified all
266 data into one of three orders of representation: direct, explicit, and implicit (Figure 2a). The ‘direct’
267 class was assigned to actual groundwater data (e.g., water table depth, groundwater storage, or
268 groundwater temperature), the ‘explicit’ class was assigned to data that incorporate groundwater
269 in the data generation process (e.g., groundwater-driven wetlands, water table ratio, and
270 groundwater management indicators), and the ‘implicit’ class was assigned to data that have
271 implied or potential connections to groundwater (e.g., cropland area, freshwater ecoregions, and
272 gross domestic product). We applied a literal approach when assigning explicit versus implicit
273 classes, where data were classified as implicit unless the data generation process included
274 explicit consideration of groundwater. Thus, this procedure to assign orders of groundwater
275 representation is not based on the strength of the underlying theory connecting a variable with
276 groundwater but rather on the dataset generation process itself. For instance, global cropland
277 datasets that do not indicate specific sources of irrigation water were classified as ‘implicit’ despite
278 agriculture being the dominant consumer of groundwater globally. However, should datasets
279 explicitly consider groundwater, such as identifying areas equipped for groundwater irrigation,
280 these data would be classified as ‘explicit’. Similarly, global wetland maps that do not specify
281 wetland type were identified as ‘implicit’ whereas those that do specify groundwater-dependent
282 wetlands are identified as ‘explicit’.

283 Dataset *types* represent the nature of the dataset as either zonal data (e.g., climate zones), a
284 static dataset (e.g., farm field size for a given date), a time series (e.g., annual population
285 estimates), or event or process records (e.g., international water events such as freshwater
286 treaties or acts of hostility). We assigned data types as ‘zonal’ if the principal use case is as a
287 spatial unit for data summary (e.g., IPCC reference regions), and as ‘static’ if the primary use
288 case is the documentation of an underlying system property or attribute (e.g., near-surface
289 porosity), even if a secondary purpose of the data can be as a zonal layer. To differentiate
290 between ‘time series’ and ‘record’ classes, we considered time series as data that are provided
291 at regular time steps with a consistent spatial extent whereas historical record data typically have
292 irregular time steps (e.g., water-related conflicts) with potentially inconsistent temporal ranges
293 depending on individual entries within the dataset (e.g., water levels in monitoring wells). Should
294 event records be synthesized into a dataset with regular time steps, these data would be recorded
295 as a time series.

296 Dataset *formats* were assigned as raster, vector (polygon, polyline, or point), or tabular. For raster
297 data, we recorded the spatial resolution of the dataset. For vector data, we collected dataset-
298 specific spatial information such as the median size of polygons or map scale based on metadata
299 availability. For all temporally dynamic data, we recorded the start and end dates of the series
300 along with the time step if it occurs in regular intervals.

301 We additionally identified each dataset’s *generation method* as being either (1) *in situ*
302 observations, (2) remote sensing observations, or (3) modelled or simulated data, such as
303 datasets that have used statistical or process-based models to extrapolate data across larger
304 domains, historical reconstructions or future projections, or approaches that combine
305 observations with models to develop datasets for variables that are challenging to directly
306 observe.

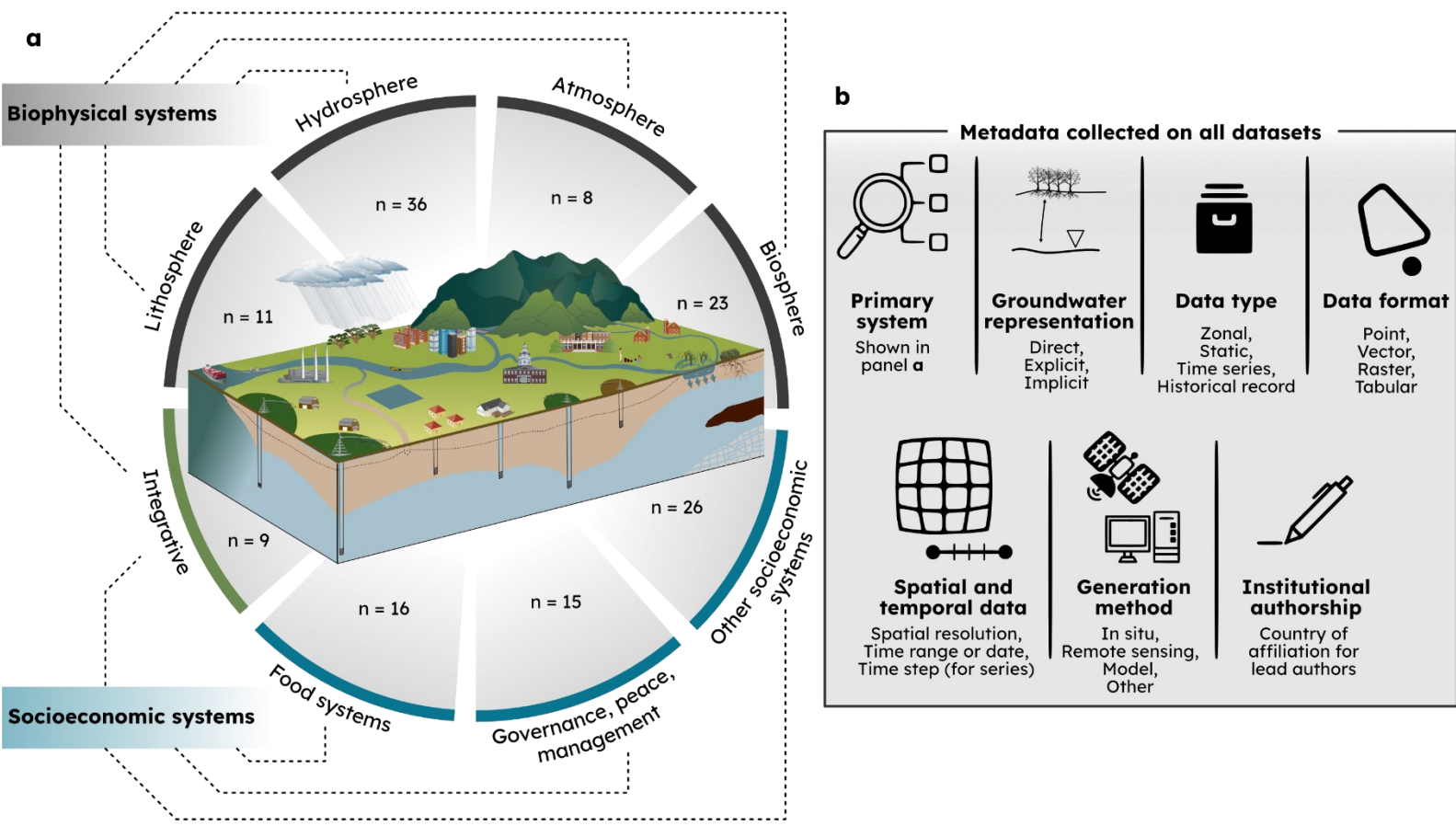
307 To assess the geographic distribution of *institutional authorship*, we recorded the country of the
308 institution affiliated with each dataset's lead author. If corresponding and lead authorship differs
309 for a dataset, we additionally included the location of the corresponding author's affiliation. For
310 data with institutions as the data provider, we used the location of the institution's headquarters.
311 When discussing results of these institutional distributions, we limited our analysis to only datasets
312 with direct and explicit groundwater representation to constrain insights and conclusions to the
313 groundwater science community.

314 **2.4 Dataset collections and nomenclature**

315 One challenging aspect of this review concerned how to best incorporate data from large,
316 coordinated research communities such as: output from global hydrological models (Reinecke *et al*
317 *et al* 2021; Schellekens *et al* 2017, Warszawski *et al* 2013), precipitation (Sun *et al* 2018), crop
318 systems (Müller *et al* 2019), and other Earth observation datasets (McCabe *et al* 2017, Jaramillo
319 *et al* 2024). These communities have respective data reviews and repositories (see preceding
320 references), and including all associated datasets risked turning our exercise into an intractable
321 'review of reviews'.

322 To maintain our focus on reviewing the variety of data available to study global groundwater in
323 social-ecological systems, we use the term 'dataset collection' to indicate when more than one
324 dataset was identified for a specific variable (e.g., precipitation). To ensure that dataset collections
325 did not skew outcomes on data accessibility when summarizing across system types, we count
326 dataset collections as a single dataset when reporting on the overall size of our catalogue (i.e.,
327 the size of our catalogue is reported as the count of unique datasets and dataset collections).

328 We also used the term 'dataset collection' to represent data initiatives that collect a wide variety
329 of variables within the same initiative. For example, the Worldwide Governance Indicators
330 initiative develops six indicators of governance dimensions, yet all indicators are available over
331 the same time range and the same spatial resolution, and are shared as a cohesive dataset.
332 Rather than listing and reporting on these indicators individually, they are included in our review
333 as a single collection.



334 **Figure 1. Scope of review. (a) System classification scheme, derived by combining**
 335 **elements from social-ecological and Earth system typologies. The number of unique**
 336 **datasets and dataset collections identified per system (n) are listed in each “slice” of the**
 337 **diagram (e.g., 36 hydrosphere datasets and dataset collections are included in the**
 338 **catalogue). (b) Summary of metadata categories and the possible values they can hold.**
 339 See Open research section for vector icon attributions.

Table 1. Description of data sources consulted to develop our open data catalogue.

Data source category	Individual sources and how they were screened
Data used in global social-ecological system characterization studies	<p>Ellis and Ramankutty (2008), Gain <i>et al</i> (2016), Sietz <i>et al</i> (2011), Václavík <i>et al</i> (2013), Varis <i>et al</i> (2019). → All input datasets used in each study were screened.</p>
Global data platforms and compendiums	<p>WRI Aqueduct (https://www.wri.org/aqueduct), WRI Resource Watch (https://resourcewatch.org/), IWRM data portal (https://iwrmdataportal.unepdhi.org/), IGRAC GIS (https://gis.un-igrac.org/), WWF Water Risk Filter (https://riskfilter.org/water/home), Protected Planet (https://www.protectedplanet.net/en), MapX (https://unepgrid.ch/en/mapx), GRID-Geneva data platform (https://unepgrid.ch/en/platforms), EarthStat (http://www.earthstat.org/), SEDAC (https://sedac.ciesin.columbia.edu/), Global Human Settlement Layer (GHSL) (https://human-settlement.emergency.copernicus.eu/datasets.php), Global Terrestrial Network - Hydrology (GTN-H) (https://www.gtn-h.info/), Copernicus Land Monitoring Service (https://land.copernicus.eu), Google Earth Engine Data Catalogue (https://developers.google.com/earth-engine/datasets/catalog), Open Land Map compendium (https://openlandmap.github.io/book/012-compendium.html). → All datasets on each platform with global coverage were screened.</p>
Compatible global data reviews	<p>Bolognesi <i>et al</i> (2018), Lindersson <i>et al</i> (2020), Kim <i>et al</i> (2021), Wang <i>et al</i> (2022). → All datasets reviewed or summarized in each paper were screened.</p>

Web of Science search	<p>Searches across the Web of Science core database were performed for the following query strings, and filtered using the “associated data” tag:</p> <p>“biophysical” AND “global” AND “dataset” (31), “ecological” AND “global” AND “dataset” (255), “governance” AND “global” AND “dataset” (36), “groundwater” AND “global” AND “dataset” (44), “socioeconomic” AND “global” AND “dataset” (95).</p> <p>Values in parentheses indicate the number of results for each query. All queries were performed in March 2024.</p> <p>→ All results from the above queries were screened.</p>
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341 **3. Results**

342 **3.1 A variety of over a hundred datasets relevant to global groundwater**

343 In total, our catalogue identifies and classifies 144 datasets and dataset collections (Table 2). All
344 datasets, including metadata and persistent web-links are provided in this study’s data repository
345 (repository will be published on Borealis, <https://borealisdata.ca/>, following manuscript
346 acceptance). An interactive table of the catalogue is accessible on this initiative’s GitHub
347 repository: <https://github.com/XanderHuggins/groundwater-SES-data-catalogue>.

348 **Table 2. Overview of unique variables and datasets included in the catalogue. The total**
349 **counts of unique datasets and dataset collections are shown in Figure 2b.**

System	List of variables * = variable has an associated <i>dataset collection</i>
Atmosphere	<p>Zonal: Köppen-Geiger climate zones, IPCC reference regions</p> <p>Static: Aridity index</p> <p>Time series: Precipitation*, Extreme precipitation projections, Evapotranspiration*, Hydrometeorological variable collections</p> <p>Records: Isotopes in precipitation, Evapotranspiration observations</p>

<p>Biosphere</p>	<p>Zonal: Freshwater and terrestrial ecoregions</p> <p>Static: Environmental flow groundwater head limit, Groundwater-dependent ecosystem extents*, Amphibian and mammal species richness, Ecological conservation prioritization index, Global wetlands, Wetlands of International Importance, Groundwater ecosystem biodiversity, Ecohydrological classes of forest growth, Ecosystem functional groups, Root zone storage capacity and depth, Groundwater-driven wetlands, Soil organic carbon content, Ramsar wetlands</p> <p>Time series: Vegetation indices* (e.g., NDVI, EVI), Maximum rooting depth, Plant functional types, Ecological vulnerability index, Vegetation health index, Wetland classification, Plant functional types, Dominant classes of grasslands</p> <p>Records: Ecosystem fluxes, Species abundances</p>
<p>Lithosphere</p>	<p>Zonal: Karst aquifer map, Sedimentary basin map</p> <p>Static: Land subsidence, Near-surface permeability and porosity, Sedimentary deposit thickness, Depth to bedrock, Coastal aquifer thickness, Soils, Lithological map, Active faults, Thickness of soil, regolith, and sedimentary deposits, Crust model</p>
<p>Hydrosphere</p>	<p>Zonal: Watersheds, River network, Aquifers, Transboundary aquifers, Karst aquifers</p> <p>Static: River width, Streamflow indices, Groundwater response time, Modern groundwater volume, River reach fragmentation, Lakes, Lake bathymetry, Lake volumes, River and stream intermittency, Groundwater recharge*, Terrestrial water storage rate of change, Surface water extent, Groundwater vulnerability to floods and droughts, Height above nearest drainage, Coastal aquifer thickness</p> <p>Time series: Streamflow*, Soil moisture, Water table depth, Terrestrial water storage anomaly*, Groundwater storage anomaly*</p> <p>Records: River discharge, Groundwater levels*, Dam locations and metadata*, Groundwater recharge, Karst spring hydrograph, Isotopes in rivers</p>
<p>Food systems</p>	<p>Static: Crop allocation to end uses, Gridded livestock systems*, Crop harvested area*, Crop type, Crop production, Crop yield*, Field size*, Cropland area*, Planting and harvesting dates for major crops, Pasture area, Virtual water trade embedded in agriculture, Area equipped for irrigation by source</p> <p>Time series: Yield gaps, Crop water footprints, Irrigated areas*, Cropland extent*, Crop yields*, Harvested areas*, Harvesting dates*, Pesticide and fertilizer application rates</p>

<p>Governance, peace, management</p>	<p>Zonal: Administrative units, Indigenous territories, Indigenous treaties</p> <p>Static: Environment, social, and governance (ESG) risk index</p> <p>Time series: Varieties of democracy, Integrated water resources management implementation indicators, Worldwide governance indicators, Environmental performance index, Subnational corruption, World values survey</p> <p>Historical records: Water related intrastate conflict and cooperation, International river basin organizations, International water events, International freshwater treaties, Water conflicts, Water related intrastate conflict or cooperation</p>
<p>Other socioeconomic systems</p>	<p>Zonal: Indigenous languages, Protected areas and other effective area-based conservation measures (OECMs)</p> <p>Static: Access to improved drinking water, Roads*, Power plants, Accessibility to cities, Development potential indices, Terrestrial human footprint, Relative deprivation index, Travel time to healthcare</p> <p>Time series: Freshwater withdrawal by sector, Human modification of terrestrial lands, Human footprint*, Gross domestic product (GDP)* and GDP per capita, Population*, Net migration, Urban land fraction, Human development index, Electricity consumption, Nighttime lights*, Migration*, Gender development inequality, Social adaptive capacity, Gini index, GNI per capita, Human development, Gender inequality</p> <p>Records: Managed aquifer recharge schemes, Living conditions of women and well-being</p>
<p>Integrative</p>	<p>Static: Population distance to surface freshwater, Land use decision making archetypes, Land system archetypes, Forest and tree proximate people, Human appropriation of net primary productivity</p> <p>Time series: Land cover*, Land use change, River basin resilience, Anthropogenic biomes</p> <p>Records: World Bank DataBank Indicators, AQUASTAT core database</p>

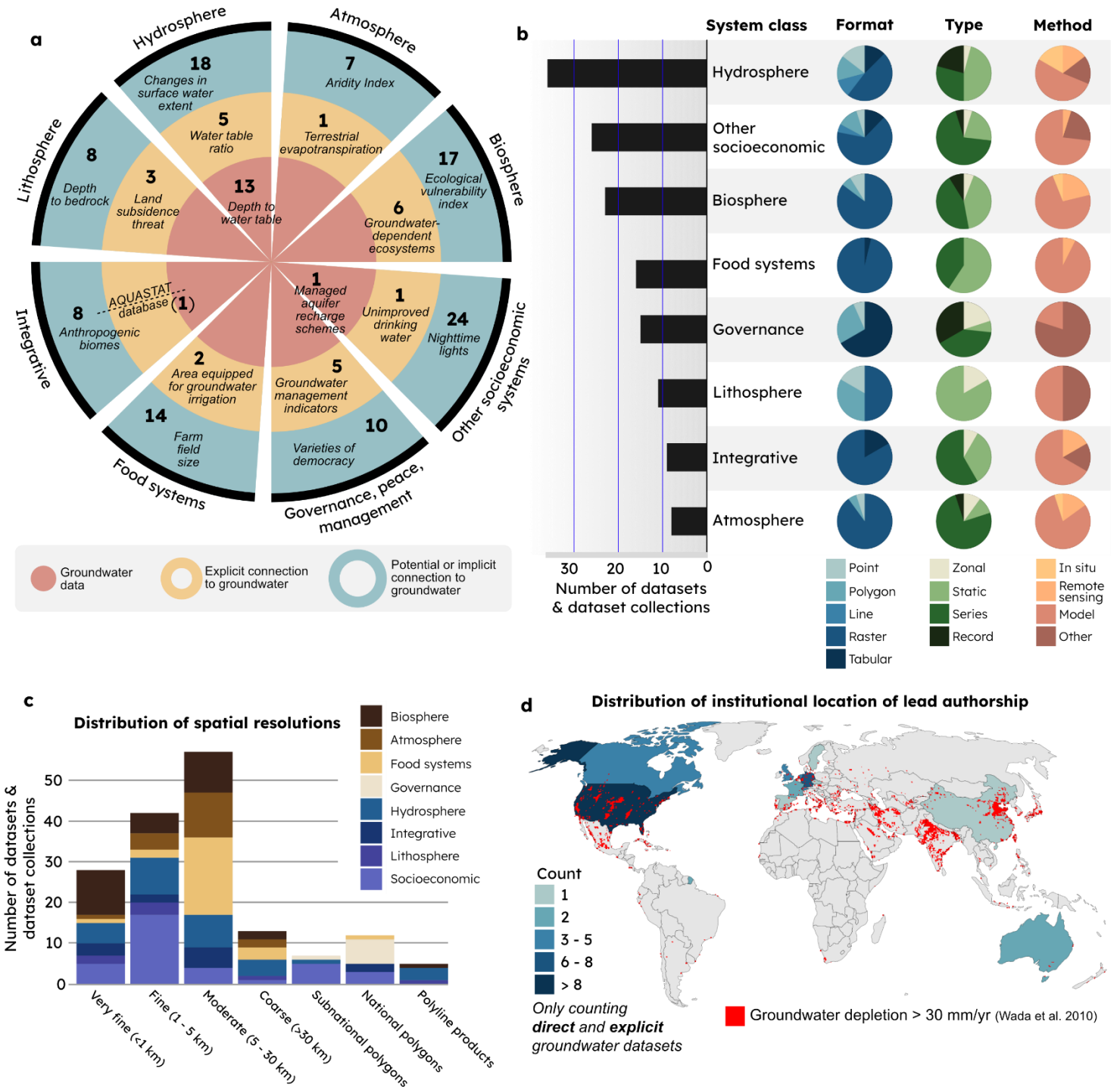
350 Of the 144 unique datasets and dataset collections, 15 were classified as direct groundwater data,
351 23 were explicitly linked to groundwater, and the remaining 106 had an implicit or potential
352 connection to groundwater (Figure 2a). All direct groundwater datasets (e.g., depth to the water
353 table, groundwater storage anomalies, groundwater temperature, etc.) are classified in the
354 hydrosphere category, whereas the 23 datasets with explicit groundwater connections are
355 distributed across all categories but are most commonly represented in the biosphere (e.g.,
356 groundwater-dependent ecosystem extents), hydrosphere (e.g., water table ratio), and
357 governance (e.g., management indicators) categories. We find most data with implicit or potential

358 connections to groundwater within the “other socioeconomic systems” category (24), which
359 includes data such as population count, gross domestic product, and gender development
360 inequalities. Trends in groundwater representation vary across the system classes. Half of all
361 hydrosphere datasets directly or explicitly consider groundwater, whereas all other system
362 categories are skewed heavily towards implicit groundwater representation.

363 Data on hydrosphere systems are the most common in the catalogue (Figure 2b), with 36 datasets
364 and dataset collections. Following the hydrosphere there are other human and socioeconomic
365 systems data (26), biosphere data (23) as these systems that have the greatest variety of
366 accessible datasets. Conversely, lithosphere (11), integrative (9), and atmosphere (8) have the
367 least representation. Governance (15) and food system (16) fall on either side of the median data
368 availability per system class. These patterns in data accessibility may reflect the overall treatment
369 and consideration of groundwater across research fields yet may also simply reflect the scope of
370 individual system categories used in this review (e.g., the class ‘other socioeconomic systems’ is
371 substantially broader than the more constrained ‘atmosphere’ or ‘lithosphere’ classes).

372 We find several patterns within dataset format, type, and generation method across the system
373 classes (Figure 2b) and which generally reflect methodological differences across disciplines that
374 relate to groundwater. For instance, the majority of governance data is tabular, consists largely of
375 event records (e.g., treaties, conflicts, etc.), and is derived through means outside of in situ
376 observations, remote sensing, and models. Conversely, biophysical data are predominantly raster
377 data, split between static and time series formats, and largely derived through models.
378 Socioeconomic data are predominantly temporally dynamic, whereas the lithosphere datasets are
379 entirely zonal or static.

380 We find these datasets are most commonly provided at moderate spatial resolutions between 5
381 and 30 km (Figure 2c). Biosphere data are most common among datasets available at very fine
382 resolutions (<1 km), while socioeconomic data are most commonly available at fine resolutions
383 (1 to 5 km), and food systems at moderate resolutions. The fine resolution is most common among
384 hydrosphere datasets, while national scale data are the most common resolution for governance
385 data.



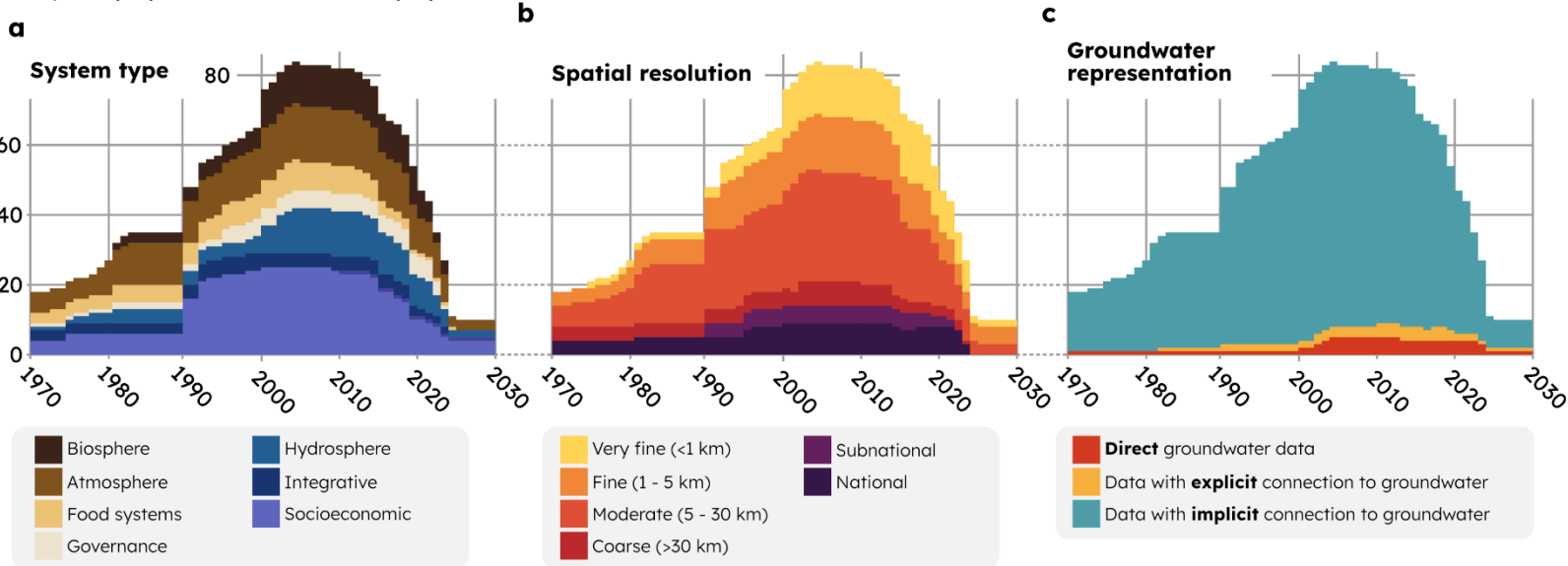
386 **Figure 2. Distribution of reviewed datasets across metadata categories. (a) Dataset count**
 387 **per system class and order of groundwater representation. Example datasets of each class**
 388 **are provided (e.g., *Nighttime lights* for data with implicit connections to groundwater in**
 389 **'other socioeconomic systems'). (b) Histogram: counts of unique datasets and dataset**
 390 **collections. Pie charts: distributions of metadata categories per system class for all**
 391 **datasets in the catalogue. (c) Distribution of spatial resolutions across system classes for**
 392 **all datasets in the catalogue. (d) Distribution of institutional authorship locations.**

393 3.2 Time series availability and groundwater underrepresentation

394 We compared temporal ranges of all temporally dynamic datasets across system type, spatial
395 resolution, and groundwater representation (Figure 3). In this comparison, we included record
396 datasets alongside time series datasets when temporal ranges are reported, even if all locations
397 within the record dataset are not available over the full reported range. We find that the greatest
398 overlap among temporally dynamic data to be available over the 2000-2010 decade, peaking with
399 over 80 available datasets. A considerable spike in time series data occurred in 1990 and a
400 consistent decline in time series availability is visible since ~2015. This creates an uncomfortable
401 reality and priority for on-going and future data initiatives: until subsequent datasets are published
402 or updated, there is a greater volume of temporally dynamic data over the 2000-2010 decade
403 than 2010-2020 despite having recently surpassed the mid-point of this century's third decade.

404 Socioeconomic data represent the most common system class with temporally dynamic data,
405 whereas zero dynamic lithosphere datasets were identified (Figure 3a). Moderate spatial
406 resolutions are the most common within temporally dynamic datasets (Figure 3b). Further, we
407 can observe the emergence of datasets with very fine spatial resolutions in the 1990s. We find
408 very few temporally dynamic datasets that directly and explicitly relate to groundwater (Figure 3c),
409 and thus these time series data largely represent systems with potential or implicit groundwater
410 connections. Only a small subset of time series datasets or dataset collections can be considered
411 actively updated through ongoing efforts.

Temporally dynamic data availability by:



412 **Figure 3: Availability of temporally dynamic datasets. (a) Dataset availability by system**
413 **type. (b) Dataset availability per spatial resolution. (c) Dataset availability by order of**
414 **groundwater representation. These plots are generated using the first and last years of**
415 **available data per dataset and do not represent the time steps of individual datasets. These**
416 **plots do not group dataset collections together and thus represent temporally dynamic**
417 **data availability across all datasets in the catalogue.**

418 3.3 Institutional leadership patterns and global North bias

419 A small set of countries lead the development of these global datasets. Of the datasets with direct
420 or explicit groundwater representation, 10 were led by institutions located in the USA, 8 by
421 institutions located in Germany, and 5 by institutions located in Canada and the Netherlands,
422 respectively (Figure 2d). Only 13 countries have led the development of all 38 datasets (and only
423 19 countries if extending to all datasets included in our collection). International agencies are
424 responsible for 8 of these direct and explicit groundwater datasets (and are thus not shown in the
425 Figure 2d map). Several countries experiencing severe groundwater sustainability challenges
426 stemming from groundwater depletion, including India, Pakistan, Iran, Mexico, and Japan (Figure
427 2d), are absent from the institutions generating these global groundwater datasets. South America
428 and Africa are absent from the lead authorship of these global data products, and Australia is the
429 only country represented from the southern hemisphere. Together, this suggests that the
430 processes of generating these global datasets have been overwhelmingly led by institutions from
431 the global North. We reflect on potential implications of this reality in section 4.3.

432 4. Discussion

433 Drawing from our review of this global open data landscape, we suggest three themes of potential
434 priorities for global groundwater data (Figure 4). These themes also structure our discussion, with
435 section 4.1 dedicated to the theme “new forms of analysis”, section 4.2 to the theme “new or
436 improved data”, and section 4.3 to the theme “more equitable processes”. Together, these themes
437 correspond with core opportunities and challenges identified in the global data landscape, and
438 are oriented in the spirit of ensuring that global groundwater science and data remain use-inspired
439 and responsive to the evolving needs of researchers, decision-makers and practitioners.

4.1 "new forms of analysis"



Synthesizing and analyzing data in new ways, including the development of groundwater or freshwater essential variables

4.2 "new or improved data"



Addressing challenges with groundwater representation, missing dimensions, and uncertainty in global data

4.3 "more equitable processes"



Elevating regional and local scale data, perspectives, and priorities to address challenges with bias and equity

440 **Figure 4. Themes of possible priorities for global groundwater data.**

441 See Open research section for vector icon attributions.

442 **4.1 Existing richness of data with significant potential for synthesis and** 443 **analysis**

444 Our assembled data catalogue paints a portrait of the open-data landscape supporting the study
445 of groundwater globally in social-ecological systems that is large and diverse. While not formally
446 evaluated by our methodology, our authorship team of global groundwater, Earth system, and
447 social-ecological system scientists shares the perspective that only a small portion of the data
448 included in our assembled catalogue has already been implemented in groundwater-focused
449 studies. In our view, this points to the significant and unrealized potential for social-ecological
450 system mental models, methods, and research objectives in large-scale groundwater science.
451 While we will focus on clear opportunities for improvement in this data landscape (section 4.2),
452 we first seek to emphasize that the research community does not need to wait for greater or
453 improved data availability: rich data are already available. Indeed, combining already existing data
454 may prove to be effective and capable of uncovering new and important relationships between
455 groundwater and social-ecological system elements and processes, and in guiding the
456 identification of which social-ecological data we most urgently lack.

457 We view the concept of essential variables, such as pursued in the fields of climate (Bojinski *et al*
458 2014), biodiversity (Pereira *et al* 2013), oceans (Miloslavich *et al* 2018), and ecosystem services
459 (Balvanera *et al* 2022), as one compelling way to integrate a coherent social-ecological framing
460 of groundwater systems with global data and analysis. Essential variables aim to identify a
461 necessary set of variables to sufficiently monitor and detect changes in the function and structure
462 of a given system. In the essential climate variables (ECV) initiative, groundwater is directly
463 included as an ECV and is implicitly represented in the terrestrial water storage variable (GCOS,
464 2024). Yet, we foresee the potential for a broader and dedicated set of essential variables for
465 groundwater in social-ecological systems, and which could include a subset of the datasets
466 available in our developed data catalogue.

467 We do not seek to establish a list of essential groundwater variables here as this would
468 necessarily involve a community-wide, iterative, and engaged process. However, we offer a
469 starting point to consider the potential for the groundwater essential variable (GEV) concept. Such
470 an initiative could lead a process of identifying a fundamental and coherent set of groundwater
471 system properties and functions in social-ecological systems that have, or could obtain,
472 observational monitoring or reporting capacities. Further, the initiative could serve the need to
473 harmonize these data into analysis-ready formats, and more broadly could act as a vehicle to
474 organize, develop, synthesize and fund global initiatives on the study of groundwater in social-
475 ecological systems.

476 **4.2 Addressing data limitations including groundwater representation,** 477 **uncertainty, and blind spots in global data**

478 Our findings also suggest that substantial opportunities and needs exist to improve this data
479 landscape because: all system classes outside of the hydrosphere skew towards datasets with
480 only potential or implicit connections to groundwater, and direct or explicit groundwater data only

481 comprise a small fraction of temporally dynamic datasets. Thus, while there exists substantial
482 potential to apply already available data, the question arises: how can the global-scale
483 groundwater data community make concerted efforts to address these limitations and generate a
484 more extensive and capable global groundwater data landscape?

485 Using the various classification schemes implemented in our review, we can identify a set of
486 preferences for future data efforts that include: observed over modeled data, time series over
487 static datasets, and explicit consideration or representation of groundwater systems. For instance,
488 datasets on the extents of groundwater-dependent ecosystems and areas equipped for
489 groundwater irrigation are currently available for specific time slices but could offer a myriad of
490 potential insights if both datasets were generated over consistent time ranges (i.e., improving
491 static datasets to time series). There is initial progress in this direction with the generation of a
492 dataset on the temporal evolution of irrigated areas from 1900-1980 in 10-year time steps, and
493 from 1980-2015 in 5-year time steps (Siebert *et al* 2015, Mehta *et al* 2024). However, similar
494 improvements are yet unrealized for groundwater-irrigated areas.

495 There is also a need to prioritize the development of temporally dynamic datasets. Improving
496 temporal frequency and aligning time steps across existing datasets would significantly improve
497 the scientific potential of these data and would ease the integration of multidimensional data into
498 analysis frameworks. These improvements stand alongside our finding that temporally dynamic
499 data have declined in availability since 2000-2010, and that substantial efforts are needed to
500 generate the same level of data availability for the past decade and decades ahead. We do not
501 necessarily attribute this to a decline in global time series data generation as a time lag is
502 necessary for research efforts to synthesize and publish data covering recent years, particularly
503 for variables and processes that do not benefit from near-real time observational capacities. The
504 decline in temporal data availability may also arise due to reduced incentives, perceived or real,
505 to update and extend existing datasets relative to the incentives of publishing a dataset that is a
506 'first of its kind'. Thus, it may be beneficial for groundwater-related societies, journals, and funding
507 agencies to reflect on potential initiatives that can create incentive structures to equitably reward
508 original dataset developers and dataset updaters.

509 Without sufficient temporally dynamic data, testing hypotheses on dynamic social-ecological
510 system behaviour of groundwater systems such as emergence, tipping points, context
511 dependence, and system resilience (e.g., Preiser *et al* 2018) may be limited to conceptual and
512 theoretical realms (Troy *et al* 2015; Di Baldassere *et al* 2015). These limitations not only create
513 barriers to scientific inquiry but can more problematically impede understanding of complex
514 system dynamics and contribute to erroneous decision making in applied contexts (Chávez
515 García Silva *et al* 2024).

516 A lack of globally distributed temporally dynamic datasets may point to a future of global
517 groundwater science that is more oriented towards case study and point location-based analyses.
518 These initiatives (e.g., Kreibich *et al* 2023, Tiwari *et al* 2023) may more readily be able to
519 implement existing observational capacities, and may more vividly reflect contextually rich data,
520 such as dimensions of human health or ecosystem services that may be challenging to organize
521 into globally distributed datasets that require common conceptual models and methodologies for

522 monitoring, extrapolation, or modelling over the global domain. Thus, an intermediary level of
523 analysis, consisting of globally distributed case studies integrating groundwater with social-
524 ecological data, may offer pragmatic and instructive insights on both the dynamics of groundwater
525 in social-ecological systems, and to guide future global data initiatives.

526 There is also a need for data development on currently missing or underrepresented dimensions
527 of groundwater-connected systems. For instance, datasets connecting groundwater with
528 domestic use and human health (e.g., Mukherjee *et al* 2019) such as on health outcomes linked
529 to groundwater salinization (Mueller *et al* 2024), observations on interconnections between
530 groundwater and surface water (e.g., Jasechko *et al* 2021), and that capture human factors such
531 as behavioral (e.g. Castilla-Rho *et al* 2017), economic (e.g., Bierkens *et al* 2024), infrastructural,
532 legal (e.g., Nelson and Perrone 2016, Rohde *et al* 2017), institutional, and governance (e.g.,
533 Villholth and Conti 2018) relating to groundwater are particularly rare in the literature.

534 Several other foundational groundwater datasets would benefit from continued improvement from
535 their original releases. For instance, the development of comprehensive and harmonized
536 geological datasets that combine global lithology maps with global borehole records would enable
537 a cascade of wider dataset improvements including more reliable global groundwater models,
538 improved representation of groundwater in Earth system models, and a strengthened ability to
539 convert observations in groundwater storage to changes in the water table.

540 Finally, we raise the need for more robust inclusion and reporting of uncertainty in global datasets
541 (Wagener *et al* 2021). Without a systematic practice of uncertainty reporting, dataset selection
542 can be driven by operational convenience, such as ease of integration based on spatial or
543 temporal resolutions, rather than a critical evaluation of data uncertainty and its propagation in
544 derivative analyses. Given social-ecological assessments inherently combine a wide variety of
545 data, reporting on uncertainty becomes all the more important as compounding uncertainties can
546 have important implications on study outcomes that may lead to erroneous or tenuous decision
547 making. Integrating uncertainty in social-ecological data presents additional challenges as
548 uncertainty is often reported in different ways across natural and social sciences (cf. Westerberg
549 *et al* 2017). These challenges aren't unique to the study of groundwater in social-ecological
550 systems but will be important to address to ensure this interdisciplinary research direction for
551 groundwater science is rigorous, reproducible, and relevant for applied use cases.

552 **4.3 Elevating and respecting regional and local perspectives, priorities,** 553 **and needs in global initiatives**

554 An abundance of regional and local scale data is missed in this review that is focussed on global
555 data. These range from large, nation-scale initiatives on monitoring wells, groundwater well uses
556 (Lin *et al* 2024), sub-national virtual water flows (Dang *et al* 2015), to a myriad of crowd-sourced
557 data within individual aquifers and basins. Indeed, it should be evident that the volume of global
558 data accumulated in our reviewed catalogue is but a small fraction of the total volume of data that
559 may exist to understand and manage groundwater. Groundwater data needs vary substantially,
560 mirroring the diversity of geographies, ecologies, and societies within which groundwater is
561 situated and connected. Local and regional datasets will inherently correspond better to these

562 needs than those generated through global initiatives. As we focus below on potential avenues to
563 potentially integrate these data into global initiatives, it should be emphasized that local and
564 regional data are essential, relevant, and valuable regardless of their inclusion in global initiatives.

565 Groundwater and groundwater-connected systems operate across a range of scales, and
566 integrating data generated across a variety of scales is one opportunity to better reflect multi-
567 scalar processes in global datasets. Existing data “networks of networks” such as the Global
568 Groundwater Monitoring Network (IGRAC, 2025) and FLUXNET (Delwiche *et al* 2024), and
569 community initiatives such as the Groundwater Model Portal (Zipper *et al.*, 2023) demonstrate the
570 potential for bottom-up collaboration on global dataset development yet substantial work lies
571 ahead to realize this potential across a wider set of social-ecological dimensions. Involvement of
572 non-governmental organizations, global development organizations, and intergovernmental
573 organizations could play an instrumental role in providing incentives, investment, infrastructure,
574 and/or enforcement (e.g., through strong data mandates) of engagement in global initiatives.
575 However, development of such initiatives is an inherently political process whose success will
576 depend on equitable stakeholder and rightsholder engagement, transparent data-sharing
577 frameworks, and the challenge of ensuring mobilized funding is allocated based on needs (cf.
578 Stein *et al* 2024).

579 While synthesizing existing regional datasets and initiatives is a laudable goal, it is also one
580 accompanied by substantial challenges. Our observation that global groundwater data are
581 generated and studied in institutions mostly distant from the place- and land-based realities of
582 acute groundwater sustainability challenges raises important questions about what research
583 priorities are driving global scale research and whose interests these priorities primarily serve.

584 Our aim in highlighting the geographical centers of global data leadership is to encourage
585 reflection on the potential implications of their concentration in the Global North. For instance,
586 recognition of the underrepresentation of scientists from tropical Africa in global climate research
587 has, over the last decade, led to more inclusive research and improved human and infrastructural
588 capacities (e.g. Lamptey *et al* 2024, Senior *et al* 2021) that are now addressing long-held gaps in
589 tropical meteorology and the representation of Africa’s monsoon-dominated climatology in climate
590 projections (WMO, 2022, Senior *et al* 2021)). Similarly, which groundwater connections and uses
591 might we, as a global groundwater community, be underrepresenting in our initiatives?

592 Thus, data sharing and equitable participation of underrepresented institutions from developing
593 regions are essential components of global data initiatives. However, engagement between local,
594 regional, and global initiatives needs to address long-standing challenges, such as research
595 imbalances and the ownership of scientific advances that are undermined by ‘helicopter’ research
596 (Gbondon and Michelsen, 2024). Developing global data sharing initiatives on the principle of
597 reciprocal benefits will be essential but alone may not be sufficient to address these challenges.
598 Potential benefits to reciprocate the sharing of data from researchers and national or regional
599 agencies may include increased visibility of data and research (such as through co-
600 authorship on derivative studies), enhanced access to technological capacities for data
601 processing (such as conversion of unprocessed data to standardized formats, data cleanup, and
602 statistical summarizing), and the inclusion and citation of data in global initiatives (e.g., WMO,

603 2024). However, such benefits will need to be established on a case-by-case basis, tailored to
604 the specific needs of individual data providers, and may require financial commitments for data
605 access to sustain groundwater monitoring programs previously reliant on data access paywalls.

606 While the above discussion addresses representational biases in global groundwater data
607 generation, it does not resolve broader ethical tensions between Big Data, open science, and
608 Indigenous Data Sovereignty principles (Walter *et al* 2020). Given that as much as 65% of land
609 area is held under Indigenous Peoples' and local community customary systems (RRI, 2015),
610 these are truly global ethical questions and priorities for land-based sciences (Meyfroidt *et al*
611 2022). In settler-Indigenous contexts and beyond, it is important to ensure that creating open
612 global datasets does not violate the privacy and security of individuals and communities by
613 sharing sensitive information (Zipper *et al* 2019).

614 Even seemingly innocuous global data such as the type and presence of groundwater-dependent
615 ecosystems may contradict Indigenous Peoples' authority to control data access that follow best
616 practices established in recent data ethics frameworks (Carroll *et al* 2021). Indeed, the Indigenous
617 territories data layer included in our review may be useful for the global groundwater community
618 to better consider how Indigenous issues, priorities, land, and knowledge systems interact with
619 one's work and may inform on how engagement with Indigenous peoples may be necessary to
620 meet ethical standards. Clarifying how the FAIR (findable, accessible, interoperable,
621 reproducible) and CARE (collective benefit, ability to control, responsibility, and Indigenous ethics)
622 data principles (Wilkinson *et al* 2016; Carroll *et al* 2020) interrelate and relate to global
623 groundwater research agendas is thus a critical priority to resolve.

624 **4.4 Review limitations**

625 Our interest in performing a wide sweep of social-ecological systems data for groundwater
626 science applications is a process that requires normative judgements on which data are
627 considered to have potential applications for groundwater studies and will reflect disciplinary
628 biases among our authorship. For instance, our approach is informed by a conceptual
629 groundwater-connected systems perceptual model (Huggins *et al* 2023) that focuses on social-
630 ecological system dynamics in relation to shallow (i.e., upper ~100 m of the subsurface) and
631 terrestrial groundwater systems. This model informs our review scope and thus underrepresents
632 offshore aquifers, deep groundwater systems, and geochemical data. These biases in our
633 catalogue can be addressed through complementary initiatives. While some geochemical
634 datasets are included in our data catalogue, they are principally oriented around human health
635 implications. Our catalogue is thus not a review of groundwater quality and geochemistry datasets
636 (Misstear *et al* 2023), nor does it consider regional datasets that serve important roles in validating
637 global models. Both of these topics warrant separate, dedicated reviews.

638 Lastly, our analysis of institutional authorship only focused on the affiliations of lead and
639 corresponding authors and did not analyze the full diversity of co-authorship lists. Many of these
640 datasets were developed in large author studies and often do include institutions from lower-
641 resource countries underrepresented in lead authorships. While we view this approach as a
642 pragmatic account of the dominant geographies and institutional contexts of these data

643 generation activities, future work would be welcomed that additionally analyzes the distribution of
644 full authorship lists. Additionally, it would be valuable to analyze the extent to which local-scale
645 data is integrated into global data products and to review the mechanisms through which data
646 providers are acknowledged and included in global data synthesis initiatives.

647 **5 Conclusion**

648 To support continental- to global-scale research on groundwater in social-ecological systems, we
649 developed and reviewed a large catalogue of open-access global datasets and dataset collections
650 (n=144) that directly, explicitly, or implicitly relate to groundwater systems and their social-
651 ecological system interactions. We reveal that a rich variety of data are available for
652 implementation in global studies, and our catalogue can serve as a reference for researchers to
653 locate sources of interdisciplinary data. We also find important limitations and biases in the
654 existing data. At the forefront of these limitations is a lack of temporally dynamic datasets that
655 explicitly represent groundwater, undermining the ability of global groundwater science to
656 generate a strong evidence base for social-ecological system dynamics in relation to
657 groundwater. We also find the institutions leading global dataset generation efforts are
658 overwhelmingly located in the global North, prompting questions about the potential mismatches
659 in needs, interests, and incentives between groundwater data generation and groundwater data
660 needs. We highlight three potential themes for global groundwater data priorities: analyzing
661 existing datasets in new ways to uncover new insights about groundwater in social-ecological
662 systems; generating new or improved datasets to address data limitations including a decline in
663 temporally dynamic data since the 2000-2010 period and a lack of temporally dynamic datasets
664 with explicit groundwater representation; and committing to better, more equitable data generation
665 processes, including providing reciprocal benefits to data providers, elevating local and regional
666 perspectives, needs, and data in global initiatives, and addressing tensions between open science
667 principles with Indigenous data ethics and data privacy. These possible priorities reach widely
668 across the data development environment and call on a variety of actors: researchers to engage
669 more deeply in interdisciplinary groundwater assessments and more inclusively with global co-
670 authors, journals and data repositories to require uncertainty estimates in geospatial data and to
671 incentivize dataset updates equally to original dataset depositions, funding bodies to support
672 research on large-scale groundwater science and data development, and international agencies
673 to providing incentives, investments, and infrastructure in support of global data sharing initiatives.

674 **Open research**

675 The associated collection of datasets, including access links and all collected metadata will be
676 deposited on Borealis (the Canadian Dataverse Repository) upon manuscript acceptance. A copy
677 of this collection is also available on the study's associated repository, available at:
678 <https://github.com/XanderHuggins/groundwater-SES-data-catalogue>.

679 All analyses were conducted using the R project for statistical computing (R Core Team, 2023),
680 using the packages *tmap* (Tennekes *et al* 2023), *ggplot2* (Wickham *et al* 2024), and *MetBrewer*
681 (Mills, 2022). Composite figures were assembled in Affinity Designer ([https://affinity.serif.com/en-](https://affinity.serif.com/en-us/designer/)
682 [us/designer/](https://affinity.serif.com/en-us/designer/)).

683 Figures 1 and 4 use the following icons from Noun Project (<https://thenounproject.com/>): “give
684 and take” by Alfredo, “collaboration” by Andika Cahya Fitriani, “classification by Sinta Maulana,
685 “groundwater” by Adrien Coquet, “plants” by Hawraa Alsalman, “satellite” and “authorship” by Lufti
686 Gani Al Achmad, “search” by Davindraaja, and the “center pivot irrigation” by Kim Kraeer and
687 Lucy Van Essen-Fishman from the Integration and Application Network (<https://ian.umces.edu>).
688 Other vector images incorporated in figures were either generated through generative AI
689 (chatGPT by OpenAI) with prompts from the author or were drawn manually in Affinity Designer.

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