

1 **Past, present, and future of the Hurst-Kolmogorov dynamics in**
2 **Stochastics: A bibliometric analysis of the last 50 years in water**
3 **resources**

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19 **Abstract**

20 Hurst's paper on the Nile's flow variability marked a pivotal moment in
21 hydrology and beyond by introducing what was called the Hurst phenom-
22 enon. Independently, Kolmogorov developed a mathematical model de-
23 scribing this behaviour a decade earlier. The Hurst-Kolmogorov dynamics
24 (HKd) is used to express this phenomenon physically and mathematically,
25 which is characterised by high uncertainty and persistence (across spatial
26 and temporal scales) and challenges traditional analytical frameworks,
27 particularly in water resources-related topics and implications in engineer-
28 ing designs. Given the importance of HKd, a bibliometric analysis of it in
29 water resources is helpful to trace its historical development, current
30 state, and (possible) future trajectories. The latter intends to offer a com-
31 prehensive perspective on HKd, serving as a guide for new readers seek-
32 ing an entry point into this field. Using the Web of Science database, 617
33 publications from 1974 to 2023 are analysed, revealing a consistent
34 growth trend in research outputs up to 2018. Collaborative efforts among
35 researchers worldwide have been prominent, with the USA and China
36 leading in international collaborations. High-impact journals on topics re-
37 lated to water resources and geosciences are primary outlets for research
38 related to the HKd. Interestingly, only two journals published the 20 most
39 cited papers on this topic. A clear pattern from "groundwater" to "stream-
40 flow" to "soil moisture" to "precipitation" was observed from the past to

41 the present. Overall, this analysis provides a comprehensive overview of
42 the past, present, and future trends in HKd research, and highlights its
43 contribution to the scientific literature of water resources.

44

45 **Keywords:** Bibliometric analysis; Hurst-Kolmogorov dynamics; Sto-
46 chastics; Water resources.

47

48 **1. INTRODUCTION**

49 Hydrology underwent a transformative shift with the work of Hurst
50 (1951), who published his paper on the variability of natural processes,
51 including the Nile flow records. Hurst's analysis revealed physical statisti-
52 cal behaviours that challenged prevailing analytical frameworks, laying
53 the groundwork for what would become known as long-term persistence
54 (LTP), long-range dependence (LRD), or Hurst-Kolmogorov dynamics
55 (HKd). The latter name emerged to give credit to the completely inde-
56 pendent contributions of Hurst (1951) and Kolmogorov (1940), who es-
57 tablished the grounds for its theoretical analysis (Koutsoyiannis, 2010).

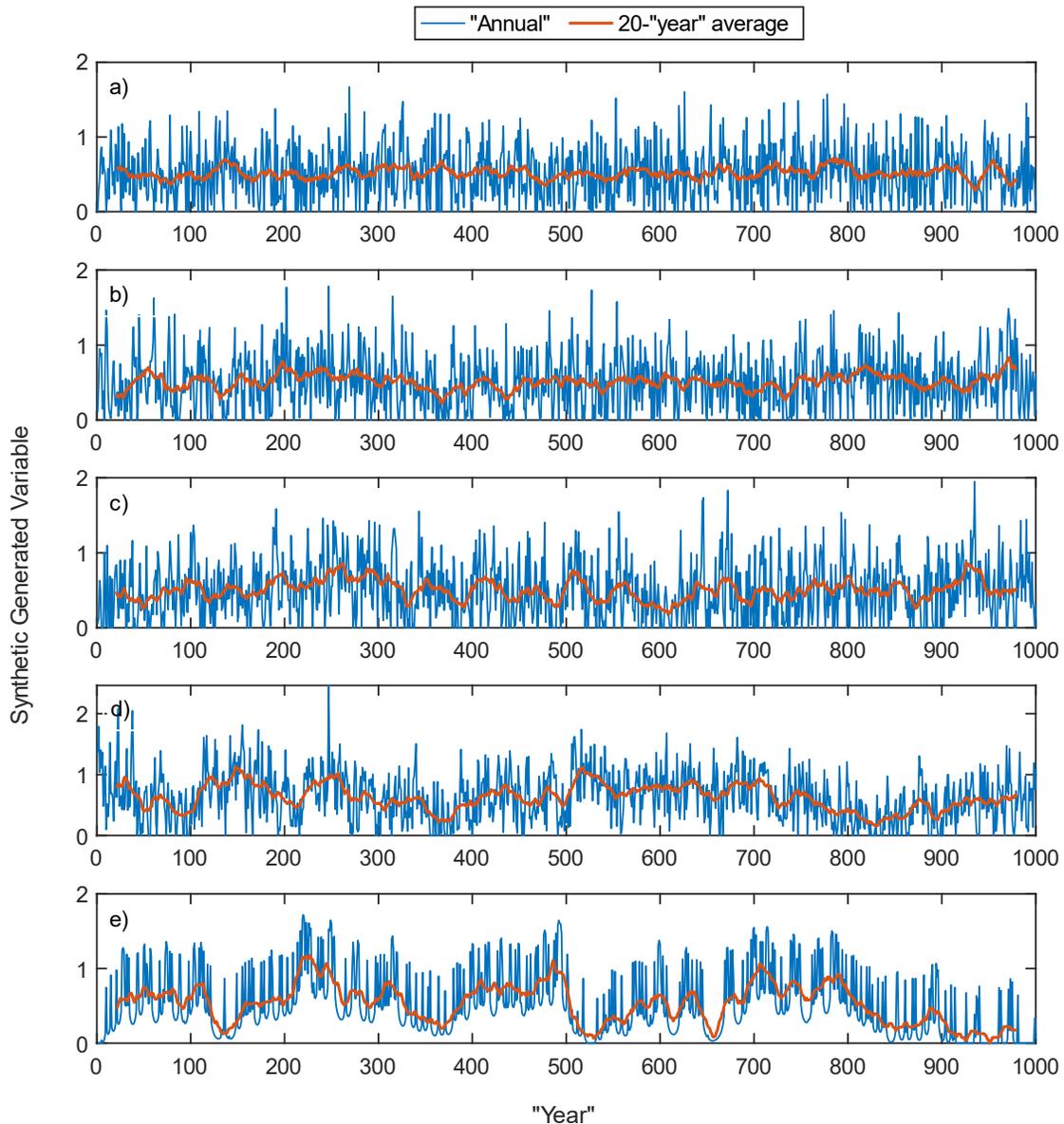
58 Among the many complex dynamics of different phenomena in water re-
59 sources (and related topics), HKd is directly related to high uncertainty
60 and persistence (McMillan et al., 2018; Westerberg et al., 2022;
61 Westerberg & McMillan, 2015). Even if this topic originated in hydrology-

62 related works, its importance has been recognised across many fields out-
63 side hydrology, such as economy, climatology, and statistics (see, e.g.,
64 O’Connell et al. (2016) for a recent review paper and Dimitriadis (2017)
65 for a variety of analyses of different hydroclimatic variables). HKd means
66 that the probability of an event occurring again depends strongly on the
67 history of the events before it. As a result, its autocorrelation function
68 decays as a power law. The latter implies that the correlation between
69 observations persists over longer lags than expected (under usually ap-
70 plied modelling frameworks, e.g., Beran et al., 2013; Mandelbrot & Wallis,
71 1968). HKd implications are significant at research and operational levels
72 since they indicate that traditional modelling approaches (based on the
73 independence principle among observations) may not be appropriate for
74 describing some natural phenomena (see, e.g., Dimitriadis &
75 Koutsoyiannis, 2018; Koutsoyiannis, 2016, 2023, for the stochastic sim-
76 ulation of HKd with an arbitrary marginal distribution and with focus on
77 the extremes of hydroclimatic processes, respectively).

78 Figure 1 shows different values of the Hurst parameter (H) to help readers
79 visualise how H affects the behaviour of time series data. Such a figure
80 presents synthetic data generated using the Filtered Hurst-Kolmogorov
81 Cauchy-Dagum (FHK-CD) model, with parameter values set as follows:
82 $a_1 = a_2 = b_1 = b_2 = M = 0.5$, and H ranging from 0.5 to 0.9. The synthetic
83 data follows an exponential distribution with $\lambda = 2$, and non-positive

84 values were truncated to zero. Synthetic data was generated at an
85 "annual" scale (in blue) along with the 20-"year" average (in red). The
86 length of the generated time series was 1000 "years". Some remarks as
87 follows:

- 88 • All generated data follows the same probability distribution;
- 89 • $H = 0.5$ represents white noise with no long-term persistence (i.e.,
90 past values have no significant influence on future values);
- 91 • $H > 0.5$: Increasing H values indicate a higher degree of
92 persistence, where past values increasingly influence future values.
93 Additionally, variability at higher time scales (in this case, 20
94 "years" in red) increases with H .



95
 96 **Figure 1.** Synthetically generated data (blue colour, "annual" scale) and 20-"year" av-
 97 erage (red colour) with different values of the Hurst parameter (H): a) $H = 0.5$; b) $H =$
 98 0.6 ; c) $H = 0.7$; d) $H = 0.8$; and, e) $H = 0.9$.

99 Since the HK behaviour was discovered, several works have been pub-
 100 lished in hydroclimatic processes evidencing, quantifying, and simulating
 101 it. Table 1 summarises some examples in terms of the research topic/var-
 102 iable of interest and its respective reference.

103 **Table 1.** Recent works evidencing, quantifying, and simulating HKd in terms of research
 104 topic or variable of interest.

Research Topic or Variable	References
Hydroclimatic processes	Dimitriadis & Koutsoyiannis (2018)
Temperature	(Glynis et al., 2021; Koutsoyiannis, 2003; Pelletier, 1998)
Rainfall	(Fraedrich & Larnder, 1993; Pelletier & Turcotte, 1997; Iliopoulou et al. 2018; Iliopoulou and Koutsoyiannis, 2019)
Wind power	(Haslett & Raftery, 1989; Katikas et al., 2021; Koutsoyiannis et al., 2018)
North-Atlantic Oscillation index	(Stephenson et al., 2000)
Streamflow	(Aguilar et al., 2017; Graves et al., 2017, pp. 1951–1980; Koutsoyiannis, 2011; Montanari, 2012; O’Connell et al., 2016; Pizarro et al., 2022; Vavoulogiannis et al., 2021)
Rock formation	(Dimitriadis et al., 2019)
Groundwater	(Varouchakis et al., 2012)
Solar radiation	(Koudouris et al., 2018)
Evapotranspiration	(Dimitriadis, Tegos, et al., 2021)
Water-energy nexus	(Mamassis et al., 2021; Sargentis, 2022)
revised tools and methodologies for key water-cycle hydrological processes	(Dimitriadis, Koutsoyiannis, et al., 2021; Koutsoyiannis, 2020a)
History and relevance of HK behaviour	O’Connell et al. (2016)
Quantification of HK behaviour through autocovariance, power-spectrum, and climacogram	Dimitriadis & Koutsoyiannis (2015)
Simulation of HKd	Dimitriadis, Iliopoulou, et al. (2021); Koutsoyiannis (2020b)
Stochastic investigation of hydrological extremes	(Iliopoulou, 2020; Koutsoyiannis, 2023)

105 Recognising the dissemination of Hurst's physical evidence and Kolmogo-
 106 rov’s mathematical description in the subsequent developments in the last
 107 decades, this bibliometric analysis aims to uncover the multifaceted as-
 108 pects of HKd in water resources, shedding light on its historical develop-
 109 ment, current state, and possible future trajectories. This study offers a

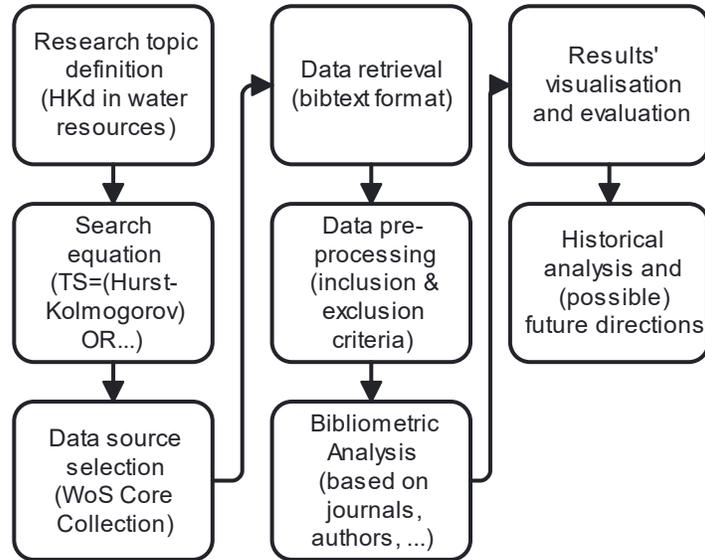
110 panoramic view of the research status by examining key metrics such as
111 publication growth, authorship patterns, international collaboration, jour-
112 nal impacts, and the frequency of used keywords. Furthermore, it seeks
113 to identify authors with significant contributions to HKd research, influen-
114 tial publications, and emerging trends within this field. Through this bib-
115 liometric analysis, we aim to offer new readers a detailed and accessible
116 introduction to this subject.

117 **2. METHODOLOGY AND METHODS**

118 **2.1 Search strategy and data extraction**

119 A comprehensive search for publications on HKd in water resources was
120 conducted in the Web of Science (WoS) database on the 12th of March
121 2024. We utilised the WoS Core Collection, known for its rigorous selec-
122 tion and evaluation process of academic information, to ensure the relia-
123 bility and accuracy of our findings. The WoS database provided content
124 coverage and detailed citation analysis information, making it an ideal
125 resource for this study. The search strategy involved the use of a carefully
126 designed equation that incorporated relevant terms. Such an equation, in
127 the WoS nomenclature, is explicitly presented as follows (TS means a
128 topic search): *(TS=(Hurst-Kolmogorov) OR TS=(Hurst parameter) OR*
129 *TS=(Hurst phenomenon) OR TS=(Hurst coefficient) OR TS=(Hurst expo-*
130 *nent) OR TS=(Hurst effect) OR TS=(long-term persistence) OR TS=(long-*

131 *range dependence*)). These terms were searched within the water re-
132 sources topic field (WoS categories), encompassing titles, abstracts, au-
133 thor keywords, and keyword-plus terms. All references indexed and pub-
134 lished from the 1st of January 1974 to the 31st of December 2023 were
135 included to provide a comprehensive analysis of the last 50 complete
136 years (i.e., a filter was applied to consider full years, and therefore, doc-
137 uments published in 2024 were not considered within the analysis). Early-
138 access papers were excluded from the bibliometric analysis. Including
139 only fully published papers allows for a more stable and comparable da-
140 taset, ensuring that the bibliometric indicators reflect finalized research
141 outputs. Additionally, early-access papers will eventually be incorporated
142 into the normal publication cycle and can be included in future analyses
143 once they are fully published. Only documents written in English were
144 considered. The WoS extraction tool was employed to extract the raw data
145 from the WoS database, generating data in BibTeX format available to
146 download. The following information was extracted from each document:
147 title, journal, article type, author names, affiliations, keywords, publica-
148 tion date, research area, abstract, cited references, language, and open
149 access information. These data formed the foundation for the bibliometric
150 analysis. Figure 2 shows the core steps of the bibliometric analysis.



151

152 **Figure 2.** Flowchart of research methodology for the bibliometric analysis applied in this study.

153 **2.2 Data analyses**

154 The bibliometric analysis of the raw data obtained from the WoS database
 155 was performed using R software version 4.4.0 (R Foundation for Statisti-
 156 cal Computing, Vienna, Austria). Specifically, the analysis was performed
 157 using the Bibliometrix R package (Aria & Cuccurullo, 2017.
 158 <https://www.bibliometrix.org/home/index.php>), which encompasses a
 159 comprehensive range of bibliometric methods. These methods allow the
 160 quantification of time trends, identification of highly cited papers, detec-
 161 tion of authors with significant contributions to HKd research, journals,
 162 institutions, and countries, as well as the calculation and ranking of re-
 163 search outputs and collaboration. Descriptive bibliometric analysis of the
 164 raw data was based on the metrics of sources (journals), authors, and
 165 documents. We also analysed two knowledge structures to complete the

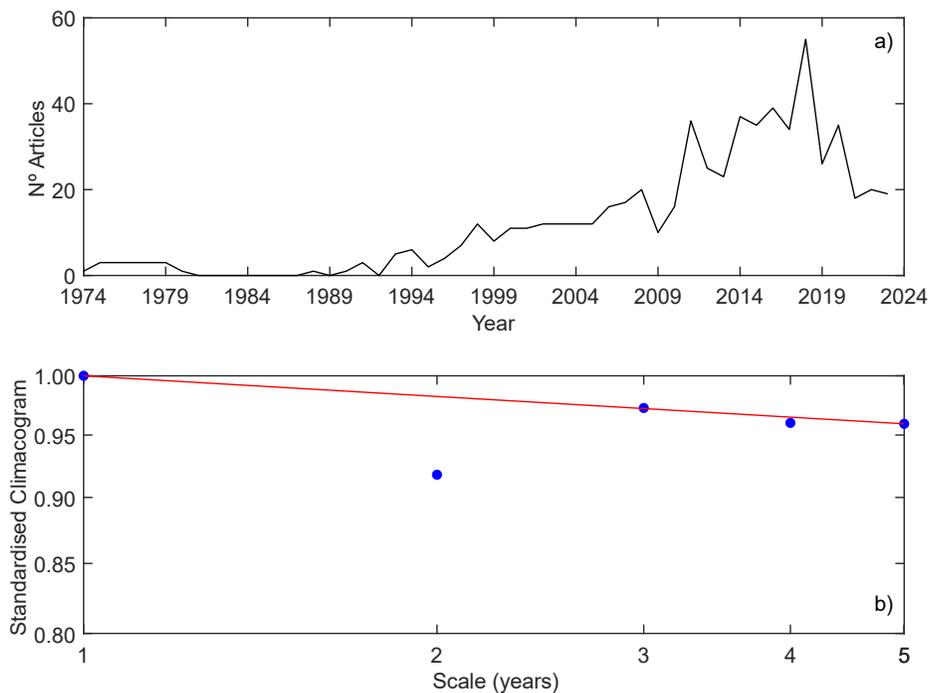
166 science mapping of HKd in water-resources research: a) conceptual: e.g.,
167 co-occurrence and thematic evolution; and, b) social: e.g., collaboration
168 network among researchers and countries. The conceptual structure was
169 measured by co-word analysis, while a collaboration network analysis
170 measured the social structure. Additionally, to enhance the bibliometric
171 results by measuring the influence of research outputs, the impact factors
172 (IF) of the journals in question were extracted from the latest Journal
173 Citation Reports (JCR, 2023) by Clarivate Analytics.

174 **3. RESULTS**

175 **3.1 Publication Analysis Based on Numbers**

176 The search strategy produced 617 publications between 1974 and 2023
177 (the last 50 years). Figure 3a shows the annual evolution of the research
178 output of HKd in water resources research. From 1974, and on average,
179 the annual growth rate was 6.19%, showing monotonically increasing
180 published papers up to 2018. From 2019 to 2023, a negative trend was
181 observed (see the Discussion section for possible explanations). However,
182 years with lack of publications are available, such as from 1981 to 1987,
183 where no documents were located from the search equation. Figure 3b
184 shows the standardised climacogram – of the time series presented in
185 Figure 3a – alongside a fitted power law to estimate the Hurst coefficient
186 H (H ranges between 0 and 1). H was estimated to have a value of 0.98,

187 which means high persistence. High persistence means that the time se-
188 ries exhibits long-term dependence, where high values are likely to be
189 followed by high values, and low values are likely to be followed by low
190 values over time. This characteristic suggests a strong "memory" effect
191 within the data, implying that past values significantly influence future
192 values. Additionally, a drop at a scale of two years was observed, provid-
193 ing hints about the existence of periodicity at this scale (see, e.g.,
194 Koutsoyiannis, 2017, page 43). It is worth mentioning that this periodicity
195 is at the border of the Nyquist frequency; therefore, it is hard to identify
196 it with a usual spectral analysis.



197
198 **Figure 3.** a) Annual publication outputs (from 1974 to 2023) of HKd in water resources. Total
199 references within the published documents were 23,582. b) Standardised climacogram with a fitted
200 power law (red line) from data of the annual publication outputs. The Hurst coefficient was
201 estimated as $H = 0.98$ (without considering the estimation bias).

202 The 617 documents recovered from the WoS search were based on 125
203 sources, of which 1722 authors were involved. Single-authored publica-
204 tions represent less than 15% of the sample (n=77, 12.48% of all publi-
205 cations), which implies most of the publications were done by several au-
206 thors, out of which 27.71% represent international co-authorship. Accord-
207 ing to the type of document, the most relevant source of the publication
208 outputs was derived from research articles (n=524, 84.93%). The re-
209 maining documents were proceeding papers (n=64, 10.37%) and reviews
210 (n=11, 1.78%). Additionally, editorial material (n=2, 0.32%), discussion
211 paper (n=2, 0.32%), and book chapter (n=9, 1.46%) represented less
212 than 3% (all together) of the documents.

213 **3.2 Publication Analysis Based on Journals**

214 Table 2 presents the general characteristics of the 20 journals with sig-
215 nificant contributions to HKd research in water resources within the last
216 50 years. These journals published 432 articles, which accounted for
217 70.02% of all recovered publications. The journals with significant contri-
218 butions to HKd research (with an arbitrarily imposed number of 20 or
219 more published articles) were: Water Resources Research (n=96), Journal
220 of Hydrology (n=70), Water (n=35), Hydrological Processes (n=25), and
221 Hydrological Sciences Journal (n=25). These five journals published
222 40.68% (251 articles) of the total production.

Table 2. Top 20 journals with significant contributions to HKd research in water resources research within the last 50 years.

Rank	Sources	N° of Articles (%^b)	Categories (Rank)	IF (JCR)^a
1 st	Water resources research	96 (15.56%)	Water Resources; Environmental Sciences (18/127; 98/358)	4.6
2 nd	Journal of Hydrology	70 (11.35%)	Geosciences, Multidisciplinary; Water Resources (19/253; 8/127)	5.9
3 rd	Water	35 (5.67%)	Environmental Sciences; Water Resources (169/358; 40/127)	3.0
4 th	Hydrological processes	25 (4.05%)	Water Resources (45/127)	2.8
5 th	Hydrological sciences journal	25 (4.05%)	Water Resources (45/127)	2.8
6 th	Hydrology and Earth System Sciences	19 (3.08%)	Geosciences, Multidisciplinary; Water Resources (21/253; 10/127)	5.7
7 th	Stochastic environmental research and risk assessment	19 (3.08%)	Water Resources; Statistics & Probability (28/127; 8/168)	3.9
8 th	Water research	18 (2.92%)	Water Resources; Environmental Sciences (1/127; 13/358)	11.4
9 th	Advances in water resources	16 (2.59%)	Water Resources (27/127)	4.0
10 th	Journal of contaminant hydrology	16 (2.59%)	Water Resources; Environmental Sciences (32/127; 146/358)	3.5

223

224

Continued table 2

Rank	Sources	N of Articles (%)	Categories (Rank)	IF (JCR)
11 th	Water air and soil pollution	15 (2.43%)	Water Resources; Environmental Sciences (30/127; 129/358)	3.8
12 th	Environmental earth sciences	11 (1.78%)	Geosciences, Multidisciplinary; Water Resources (84/253; 45/127)	2.8
13 th	Journal of hydrologic engineering	10 (1.62%)	Water Resources; Environmental Sciences (67/127; 239/358)	2.2
14 th	Natural hazards and earth system sciences	10 (1.62%)	Geosciences, Multidisciplinary; Water Resources (34/253; 25/127)	4.2
15 th	Aquatic conservation-marine and freshwater ecosystems	9 (1.46%)	Water Resources; Environmental Sciences (54/127; 209/358)	2.5
16 th	Journal of the American water resources association	8 (1.30%)	Geosciences, Multidisciplinary; Water Resources (98/253; 51/127)	2.6
17 th	Natural hazards	8 (1.30%)	Geosciences, Multidisciplinary; Water Resources (65/253; 35/127)	3.3
18 th	Physics and chemistry of the earth	8 (1.30%)	Geosciences, Multidisciplinary; Water Resources (76/253; 40/127)	3.0
19 th	Ecohydrology	7 (1.13%)	Water Resources; Environmental Sciences (54/127; 209/358)	2.5
20 th	Environmental modelling & software	7 (1.13%)	Water Resources; Computer Science, Interdisciplinary Applications (14/127; 32/169)	4.8

^a IF (JCR), impact factor (Journal Citations Reports); Impact factor obtained from the Journal Citation Reports (2023).

^b Percentage calculated out of the retrieved 617 documents.

226 **3.3 Publication Analysis Based on Authors**

227 Between 1974 and 2023, the topic in question accumulated 617 docu-
228 ments involving 1722 authors. Most authors published multi-authored
229 documents (single authored-documents: 77; whereas authors of single-
230 authored documents: 58). The authors with significant contributions to
231 HKd research (with an arbitrarily imposed number of 10 or more published
232 articles) were: Koutsoyiannis, D. (n=41); Dimitriadis, P. (n=11); and,
233 Montanari, A. (n=10). Collaboration analysis showed an average co-au-
234 thorship of 3.45 co-authors per document and (on average) a percentual
235 international co-authorship of 27.71%.

236 **3.4 Publication Analysis Based on Countries/Regions and Institu-** 237 **tions**

238 Country publication outputs were measured using the number of author
239 appearances by country affiliation (corresponding author only). The re-
240 sults showed that the research output of HKd in water resources research
241 was geographically located in 57 countries worldwide. Table 3 shows the
242 main features of the top 20 most productive corresponding authors' coun-
243 tries in publishing research on this topic. The USA, China, Canada, and
244 Greece were the most productive countries (in terms of corresponding
245 authors), accumulating around half of the total production (n=326,
246 52.83%). The intra-country collaboration analysis (i.e., all authors from

247 the same country) showed that these countries also had the highest num-
248 ber of articles published by authors from their own countries (USA=92;
249 China=32; Canada=22; Greece=17). The USA was at the forefront of the
250 inter-country collaboration, registering 31 documents that included au-
251 thors from other countries. China (n=23) and Canada (n=15) were the
252 second and third most significant contributors. However, in relative terms,
253 South Africa (multiple country publication (MCP) ratio=0.75, total docu-
254 ments=4), Switzerland (MCP ratio=0.67, total documents=15), and Po-
255 land (MCP ratio=0.60, total documents=5) were the countries collaborat-
256 ing more actively.

257 Figure 4 illustrates the network mapping of international collaboration
258 among the countries. The USA led the most significant number of inter-
259 national collaborations (n=156). The following most significant collabora-
260 tors were China (n=79), Greece (n=46), and Canada (n=45). Table 4
261 presents a general description of the top 20 most productive research
262 institutions/universities in publishing research on HKd in water resources.
263 These institutions produced 541 documents (87.68%) out of 617 recov-
264 ered publications. Worthy mentioning is that the first institution in Table
265 4 (National Technical University of Athens, Greece) published more than
266 three times the number of papers of the second institution in the ranking
267 in question (Hohai University, China).

Table 3. Top 20 productive corresponding authors' countries in publishing papers on HKd in water resources (1974-2023).

Countries	N of documents	%^a	SCP	MCP	MCP Ratio^b
USA	156	25.28%	128	28	0.18
China	79	12.80%	56	23	0.29
Greece	46	7.46%	37	9	0.20
Canada	45	7.29%	33	12	0.27
Australia	34	5.51%	25	9	0.27
Italy	31	5.02%	18	13	0.42
United Kingdom	23	3.73%	12	11	0.48
Germany	22	3.57%	16	6	0.27
India	18	2.92%	18	0	0.00
France	16	2.59%	11	5	0.31
Spain	14	2.27%	7	7	0.50
Iran	9	1.46%	7	2	0.22
Poland	8	1.30%	6	2	0.25
Brazil	7	1.13%	7	0	0.00
Austria	6	0.97%	5	1	0.17
Netherlands	6	0.97%	1	5	0.83
Switzerland	6	0.97%	1	5	0.83
Czech Republic	5	0.81%	4	1	0.20
Japan	5	0.81%	2	3	0.60
Mexico	5	0.81%	4	1	0.20

Abbreviations: SCP, single country publications; MCP, multiple country publications.

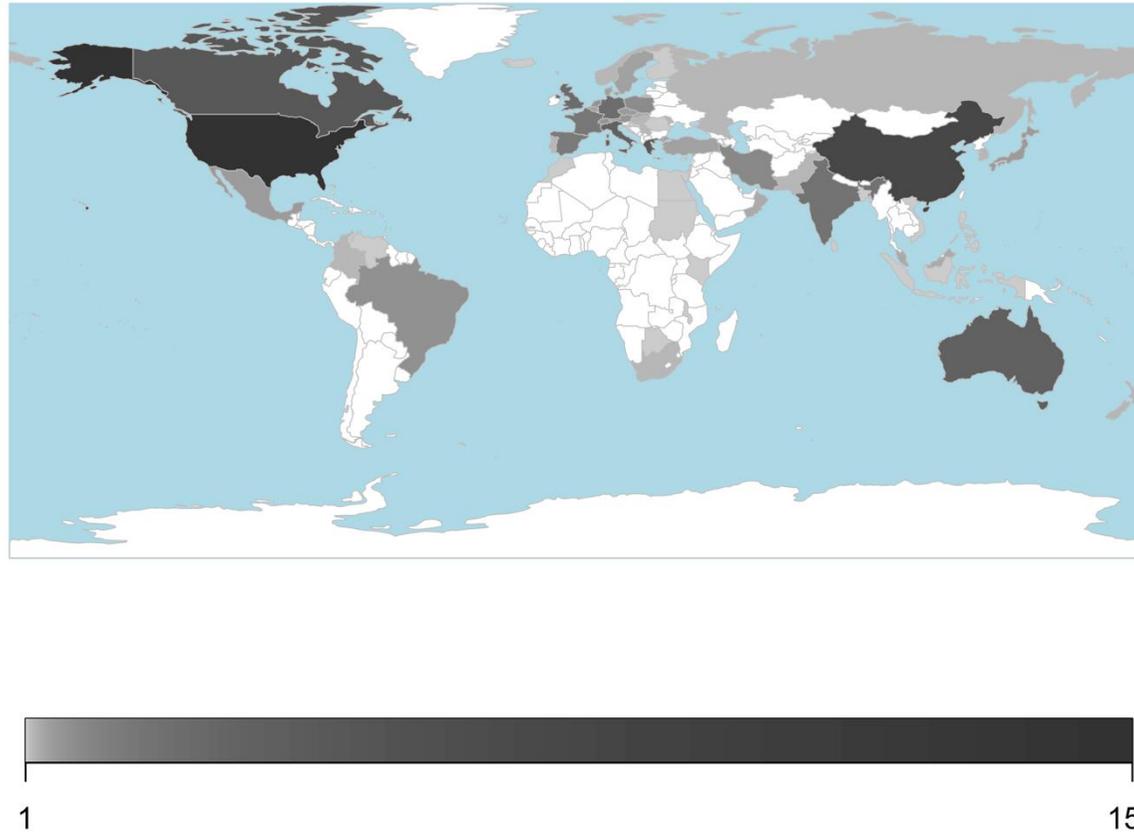
^a Percentage calculated out of the retrieved 617 documents.

^b We calculated the Multiple country publication ratio as MCP divided by the total published documents per country.

Table 4. Top 20 most productive research institutes/universities in publishing papers on HKd in water resources (1974 to 2023), sorted by the total number of articles.

University/Research Institute	Country	N of articles	%^a
National Technical University of Athens	Greece	80	12.97%
Hohai University	China	26	4.21%
University of Arizona	USA	24	3.89%
Colorado State University	USA	21	3.40%
University of Padua	Italy	20	3.24%
Beijing Normal University	China	19	3.08%
Newcastle University	UK	19	3.08%
University of Bologna	Italy	18	2.92%
Purdue University	USA	17	2.76%
University of California, Davis	USA	17	2.76%
University of Saskatchewan	Canada	17	2.76%
Vienna University of Technology	Austria	16	2.59%
University of Iowa	USA	15	2.43%
University of Newcastle	Australia	15	2.43%
Northwest A&F University	China	14	2.27%
University of Granada	Spain	14	2.27%
University of Waterloo	Canada	14	2.27%
Texas A&M University	USA	13	2.11%
University of Nevada	USA	13	2.11%
University of Tennessee	USA	12	1.94%

^a Percentage calculated out of the retrieved 617 documents.



271

272 **Figure 4.** Geographical distribution map of countries' research output in publishing papers on HKd in water resources from 1974 to 2023.

273 The number of articles is represented by colours in grayscale, where white and black mean the minimum and maximum number of articles,

274 respectively. Notice there are countries in white (i.e., without publications) in Latin America, Africa, and Asia.

275 **3.5 Publication Analysis Based on Citations**

276 The 617 documents included in the present bibliometric analysis had
277 23,582 citations, with an average of 38.22 citations per document (con-
278 sidering all years) and an average of 0.76 citations per document annu-
279 ally. There were 42 (6.81%) documents with no citations, 193 (31.28%)
280 were cited between one and ten times, 123 (19.94%) received between
281 11 and 20 citations, 130 (21.07%) were cited between 20 and 50 cita-
282 tions, 81 (13.13%) between 50 and 100 citations, and 50 (8.10%) were
283 cited over 100 times. The most relevant authors (in terms of number of
284 published articles) were Koutsoyiannis, D. (n=41); Dimitriadis, P. (n=11);
285 and, Montanari, A. (n=10). Additionally, the most locally cited authors
286 were Koutsoyiannis, D. (n=169); Montanari, A. (n=125); and Klemes, V.
287 (n=61). Local citations are the number of citations a document received
288 from other articles in the analysed database (n=617 articles in total).

289 The top 20 most cited articles are presented in Table 5. These articles
290 were published in only 2 scientific journals, namely: i) Water Resources
291 Research; and, ii) Journal of Hydrology. Regarding citations, Klemeš
292 (1974) was the most locally cited article, while Hamed (2008) was the
293 most globally cited one. In relative terms, Koutsoyiannis & Montanari
294 (2007) and Hamed (2008) were the most cited documents, locally and
295 globally, respectively.

Table 5. Top 20 most-cited research papers from 1974 to 2023, sorted by the number of citations (Local cited documents)

Rank	Authors	Year	Journal	DOI	LC	LCy	GC	GCy
1st	Klemeš	1974	Water Resources Research	https://doi.org/10.1029/WR010i004p00675	59	1.18	234	4.68
2nd	Koutsoyiannis & Montanari	2007	Water Resources Research	https://doi.org/10.1029/2006WR005592	58	3.41	221	13.00
3rd	Montanari et al.	1997	Water Resources Research	https://doi.org/10.1029/97WR00043	41	1.52	196	7.26
4th	Hamed	2008	Journal of Hydrology	https://doi.org/10.1016/j.jhydrol.2007.11.009	39	2.44	877	54.81
5th	Koutsoyiannis	2000	Water Resources Research	https://doi.org/10.1029/2000WR900044	26	1.08	115	4.79
6th	Kumar et al.	2009	Journal of Hydrology	https://doi.org/10.1016/j.jhydrol.2009.06.012	26	1.73	297	19.80
7th	Koutsoyiannis	2006	Journal of Hydrology	https://doi.org/10.1016/j.jhydrol.2005.09.022	24	1.33	156	8.67
8th	Koscielny-Bunde et al.	2006	Journal of Hydrology	https://doi.org/10.1016/j.jhydrol.2005.03.004	23	1.28	236	13.11
9th	Potter	1976	Water Resources Research	https://doi.org/10.1029/WR012i005p01047	22	0.46	78	1.63
10th	Mudelsee	2007	Water Resources Research	https://doi.org/10.1029/2006WR005721	22	1.29	84	4.94
11th	Thyer & Kuczera	2003	Journal of Hydrology	https://doi.org/10.1016/S0022-1694(02)00412-2	18	0.86	44	2.10
12th	Sagarika et al.	2014	Journal of Hydrology	https://doi.org/10.1016/j.jhydrol.2014.05.002	18	1.80	148	14.80
13th	Thyer & Kuczera	2000	Water Resources Research	https://doi.org/10.1029/2000WR900157	17	0.71	70	2.92
14th	Montanari et al.	2000	Water Resources Research	https://doi.org/10.1029/2000WR900012	17	0.71	134	5.58

15th	Thyer & Kuczera	2003	Journal of Hydrology	https://doi.org/10.1016/S0022-1694(02)00411-0	16	0.76	38	1.81
16th	Vogel et al.	1998	Water Resources Research	https://doi.org/10.1029/98WR02523	15	0.58	61	2.35
17th	Salas et al.	1979	Journal of Hydrology	https://doi.org/10.1016/0022-1694(79)90143-4	14	0.31	35	0.78
18th	Eltahir	1996	Water Resources Research	https://doi.org/10.1029/95WR02968	14	0.50	155	5.54
19th	Boes & Salas	1978	Water Resources Research	https://doi.org/10.1029/WR014i001p00135	13	0.28	58	1.26
20th	Khaliq et al.	2009	Journal of Hydrology	https://doi.org/10.1016/j.jhydrol.2009.02.045	13	0.87	104	6.93

Abbreviations: LC, local citations; LC_y, local citations per year; GC, global citations; GC_y, global citations per year.

297 **4. DISCUSSION AND INTERPRETATION OF RESULTS**

298 The results of this study contribute to a better understanding of the state
299 of research on HKd in water resources research. In particular, the last 50
300 years were analysed, focusing on applications in water resources. Under-
301 standing HKd provides insights not only at scientific levels but also for
302 practitioners and policymakers operating in engineering praxis, hydrolog-
303 ical modelling, and flood risk management, among others. The implica-
304 tions of HKd are substantial, suggesting that conventional modelling
305 methods, which usually rely on the assumption of independence among
306 observations, might not accurately capture certain natural phenomena.
307 This bibliometric analysis is not intended to be a usually understood re-
308 view but to extract and analyse critical information from the database
309 investigated. It aligns to give a holistic view of the HKd topic, fostering
310 and paving the path for new readers who do not know where to start (for
311 instance, reading all the papers in Table 6 and Table 6).

312 **Table 6.** Top-cited papers with reference to Hurst’s work beyond water resources
313 (adapted from O’Connell et al., 2016). Citations were retrieved from Google Scholar –
314 the 4th of June 2024. Notice that some journal papers in this table were published before
315 1974 (the starting year of the bibliometric analysis), and they were not found by the
316 search equation.

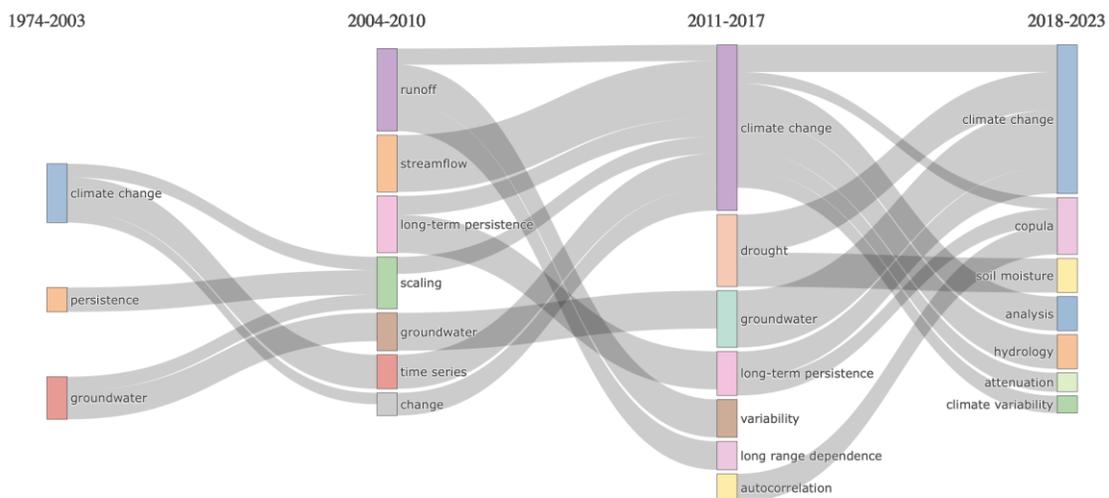
Field	Reference	DOI	Subject	Citations
Statistics and stochastic	Mandelbrot & Van Ness (1968)	https://doi.org/10.1137/1010093	Fractional Brownian motion	10158
	Hosking (1981)	http://dx.doi.org/10.1093/biomet/68.1.165	FARIMA model	4973
Geophysical sciences	Mandelbrot & Wallis (1969)	http://dx.doi.org/10.1029/WR005i002p00321	Geophysical records	1410
	Sadler (1981)	http://dx.doi.org/10.1086/628623	Sediment accumulation rate	1369
Physics	Voss & Clarke (1975)	http://dx.doi.org/10.1038/258317a0	Acoustics	839
	Borland (1998)	http://dx.doi.org/10.1103/PhysRevE.57.6634	Diffusion	413
Economics and finance	Greene & Fielitz (1977)	http://dx.doi.org/10.1016/0304-405X(77)90006-X	Stock market prices	664
	Baillie (1996)	http://dx.doi.org/10.1016/0304-4076(95)01732-1	Econometrics	2651
Medical sciences	Ellaway (1978)	http://dx.doi.org/10.1016/0013-4694(78)90017-2	Neurophysiology	528
	Kobayashi & Musha (1982)	http://dx.doi.org/10.1109/TBME.1982.324972	Heartbeat period	1026
Biology	Peng et al. (1992)	http://dx.doi.org/10.1038/356168a0	Nucleotide sequences	1809
	Arneodo et al. (1996)	http://dx.doi.org/10.1016/016	DNA sequences	223

		<u>7-2789(96)00029-2</u>		
Information sciences	Leland et al. (1994)	<u>http://dx.doi.org/10.1109/90.282603</u>	Ethernet traffic	8336
	Frost & Melamed (1994)	<u>https://doi.org/10.1109/35.267444</u>	Traffic in telecommunication networks	1025
Climate science	Koutsoyiannis (2003)	<u>http://dx.doi.org/10.1623/hysj.48.1.3.43481</u>	Climatic variability	500
	Cohn & Lins (2005)	<u>http://dx.doi.org/10.1029/2005GL024476</u>	Trends in climate	454

317
318 In terms of research outputs, this bibliometric analysis revealed an in-
319 creasing trend up to 2018. After that year, the number of published pa-
320 pers was decreasing. The latter is probably due to the COVID-19 out-
321 break's impacts on researchers and the population worldwide (see, e.g.,
322 Riccaboni & Verginer (2022), whose findings established a dropping trend
323 for research outputs in unrelated medical subject papers). Another possi-
324 ble explanation might be that in the past decades, the research focus in
325 the hydrologic literature has progressively shifted towards the identifica-
326 tion of trends under nonstationarity, i.e. adopting a deterministic perspec-
327 tive on natural variability and seeking to identify deterministic signals of
328 change (cf. Appendix I in Iliopoulou & Koutsoyiannis, 2020, and relevant
329 discussion). Additionally, machine learning (and also other) methodolo-
330 gies to capture LRD might have attracted the attention of researchers
331 recently (see, e.g., (Rozos et al., 2021)).

332

333 Thematic evolution through time can be helpful in identifying the past,
 334 present, and future of stochastics in water resources. Figure 5 shows four
 335 different periods: a) from 1974 to 2003 (i.e., the first 30 years); b) from
 336 2004 to 2010; c) from 2011 to 2017; and, d) from 2018 to 2023. Grey
 337 bands connect the research topics, expressing how those topics relate
 338 through time. Interestingly, “groundwater” has been a topic for several
 339 decades, while there has been a line of thought from “climate change” to
 340 “climate variability” (passing through “change” and “climate change”
 341 again). We hypothesise that this line of thought will be intensified in the
 342 future, changing the paradigm from climate change to climate variability.

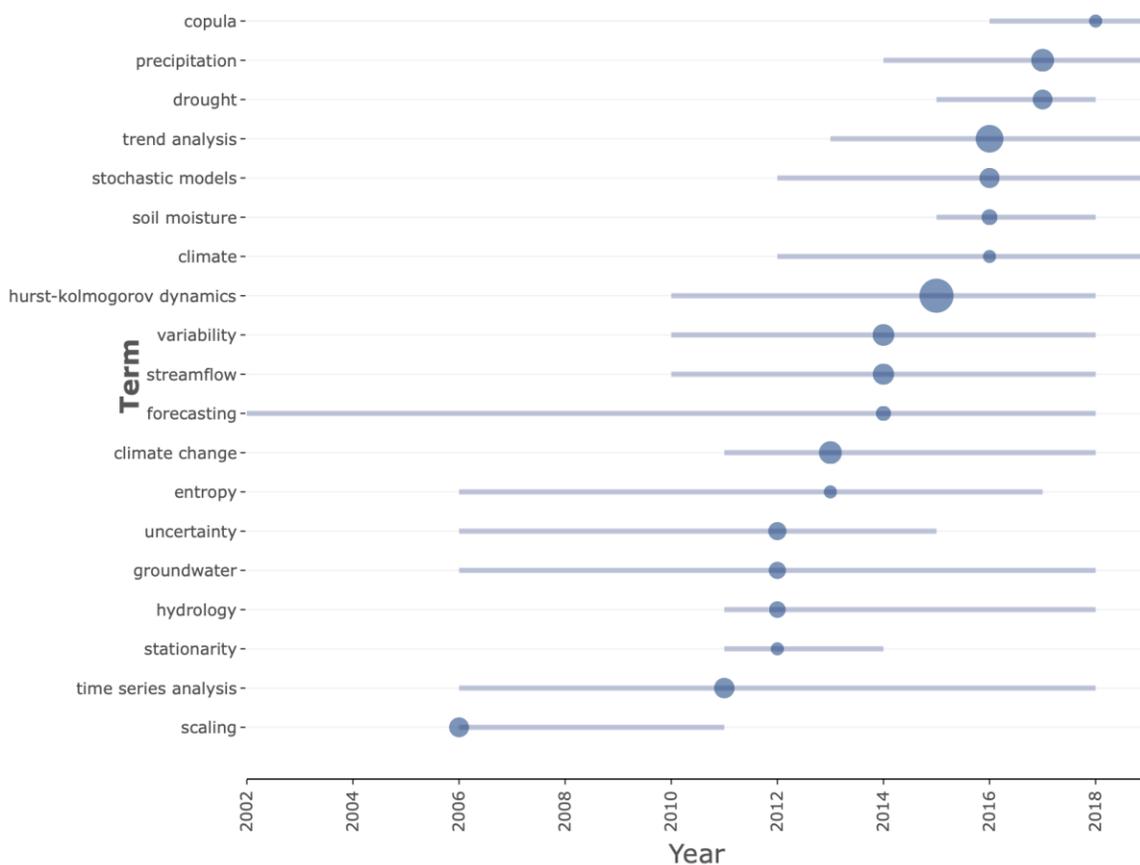


343
 344 **Figure 5.** Thematic evolution of HKd in water resources. 50 years of analysis through the author’s
 345 keywords. Connecting keywords are determined through the inclusion index weighted by word
 346 occurrences and the walktrap clustering algorithm (minimum cluster frequency per thousand docs
 347 = 14; minimum weight index = 0.1).

348 Zooming in on the last two decades and in terms of trending topics, Figure
 349 6 shows different authors' keywords with their boxplots (ranging in

350 years). The location of the dots represents the median, and the dot's size
 351 is the number of times the keyword in question has been used in different
 352 documents. Interestingly, there is a visible pattern – of research interests
 353 as a function of hydroclimatic variables – of migration from “groundwater”
 354 to “streamflow” to “soil moisture” to “precipitation”. Current research is
 355 more focused on “droughts”, “trend analysis”, and “stochastic models”.
 356 “Groundwater”, “forecasting”, and “time-series analyses” have been per-
 357 sistent through almost the entire analysed period.

358



359

360 **Figure 6.** Trending topic author's keywords for the last 20 years. The dots' location is
 361 the median of values (years), and the size represents the frequency of used terms.

362 **5. STRENGTHS AND LIMITATIONS**

363 This bibliometric analysis covers the last 50 years of HKd in water
364 resources. Even though the adopted methodology was robust, replicable,
365 and reproducible; there is room for improvement. The following list covers
366 some of them:

- 367 a) Database: expanding the source data from WoS to Scopus and
368 other databases with the intention of retrieving the maximum
369 number of articles (even though many of them will be duplicates);
- 370 b) Search equation: Sensitivity analysis of the search equation will
371 ensure a comprehensive literature capture alongside possible facets
372 of HKd research;
- 373 c) Data pre-processing: Revision of inclusion and exclusion criteria
374 (e.g., period of analysis, language of articles, among others);
- 375 **d) From bibliometric analysis to a formal systematic review and**
376 **metanalysis**: This bibliometric analysis guides new readers through
377 a first step into HKd research, analysing key aspects from literature.
378 Nevertheless, a formal systematic review will further extract critical
379 information from each article, opening avenues for an HKd research
380 metanalysis.

381 Currently, progress in HKd needs to be documented to identify the
382 strengths and limitations of methods in different contexts and transfer
383 them to other variables of interest. Establishing standardised approaches

384 represents the critical goal for future developments and collaborative net-
385 works.

386 **6. CONCLUSIONS**

387 This bibliometric analysis offers a comprehensive overview of the past,
388 present, and future of the HKd research trends in water resources. The
389 study reveals several key findings:

390 ▪ **Publications' Growth:** The field has experienced a monotonically
391 increasing growth (up to 2018) with a significant increase in
392 publications. The maximum number of published papers was 55 in
393 2018. From that year to the present, a decreasing trend has been
394 observed. We hypothesise that this trend is going to be maintained
395 for the near future due to relevant researchers in a transition phase
396 of their lives (e.g., retirement), the global impact of the COVID-19
397 pandemic, an increasing interest of researchers towards identifying
398 trends under nonstationarity, and trendy methodologies such as
399 machine learning related ones.

400 ▪ **Global Collaboration:** Researchers from diverse countries actively
401 contribute to this interdisciplinary field. In general terms, the USA,
402 China, Canada, and Greece were the most productive countries,
403 accumulating around half of the total production. Additionally, many
404 countries in Latin America, Africa, and Asia have had zero
405 publications, implying that no research on HKd in water resources is

406 available. We hypothesise that the latter is due to the lack of
407 accessibility to long time series.

408 ▪ **Journal Information:** High-impact journals in water resources and
409 geosciences are the primary outlets for HKd research. Only two
410 journals covered the 20 most relevant publications: i) Water
411 Resources Research; and, ii) Journal of Hydrology.

412 ▪ **Authors:** authors with significant contributions to HKd research from
413 various countries have contributed significantly to this research area,
414 often through multi-authored publications.

415 ▪ **Citation Impact:** Certain papers have garnered substantial
416 attention and citations, emphasising the field's relevance and impact.
417 Table 5 shows the top 20 most-cited research papers from 1974 to
418 2023. Klemeš (1974) was the most locally cited article, while Hamed
419 (2008) was the most globally cited one. Additionally, Table 6 shows
420 the top-cited papers with reference to Hurst's work beyond water
421 resources (adapted from O'Connell et al., 2016).

422 ▪ **Keyword Themes:** Analysing the last 50 years of research revealed
423 a clear pattern of migration of research topics through time. Interest
424 moved from "groundwater" to "streamflow" to "soil moisture" to
425 "precipitation" from the past to the present.

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430

431 **DATA AND RESULTS AVAILABILITY STATEMENT**

432 The data used in this study, as well as the findings, are available in the
433 following link: <http://doi.org/10.17605/OSF.IO/45V3W>

434

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