1 A comparative analysis reveals irreproducibility in searches of scientific

2 literature

3 Short title: Search location and the reproducibility of systematic reviews

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37 Abstract

38 Repeatability is the cornerstone of science and it is particularly important for systematic reviews. However, little is known on how database and search engine choices influence replicability. Here, 39 40 we present a comparative analysis of time-synchronized searches at different locations in the world, revealing a large variation among the hits obtained within each of the several search terms using 41 42 different search engines. We found that PubMed and Scopus returned geographically consistent results to identical search strings, Google Scholar and Web of Science varied substantially both in 43 44 the number of returned hits and in the list of individual articles depending on the search location and computing environment. To maintain scientific integrity and consistency, especially in 45 46 systematic reviews, action is needed from both the scientific community and scientific search 47 platforms to increase search consistency. Researchers are encouraged to report the search location, 48 and database providers should make search algorithms transparent and revise access rules to titles 49 behind paywalls.

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51 Key words: Database, search engine, search location, repeatability

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53 **Introduction**

54 Since the 17th century and Newton's strict approach to scientific inquiry[1], research has increasingly 55 relied on rigorous methodological constrains. One of the cornerstones of the scientific method is 56 reproducibility. However, a recent study shows that most scientists believe that a substantial proportion of methods published in peer-reviewed papers are not reproducible, creating a 'reproducibility crisis'[2]. 57 58 Following similar arguments, narrative reviews are increasingly being replaced by systematic reviews, 59 also called "evidence-based synthesis"[3]. Transparency and repeatability are also cornerstones of this 60 method of knowledge synthesis. However, the repeatability of systematic reviews remains rarely 61 examined. Though repeatability in such studies is of utmost importance, and detailed protocols are 62 available[4,5], the technical aspects of these underpinning databases and search engines have not been 63 systematically tested and, at present, there is no recommendation on these technical aspects.

64 As primary scientific literature is rapidly expanding[6], scientists are unable to keep track of new discoveries by focusing only on the primary literature [7,8], so systematic reviews have become 65 66 increasingly important[9]. Recognized weaknesses of the traditional, narrative reviews include the non-67 transparency of the literature selection process, evaluation criteria, and eventual level of detail devoted to individual studies[10]. With the advent and rapid development of Internet-based databases and search 68 69 engines, the role of narrative reviews is now being overtaken by new, quantitative methods of evidence 70 synthesis[11,12]. A core requirement in these activities, repeatability, crucially depends on reliable 71 databases[13]. Large scientific databases/search engines, such as PubMed, Web of Science and Scopus, 72 are essential in this process. They have been primary electronic search engines for scientists since 1997 73 with the inauguration of PubMed[14]. Today, nearly all scientists working on various forms of evidence-74 based synthesis use these databases/search engines to find relevant papers as the basis for further analysis.

An important condition in the whole process is that the evidence base must be solid: a given search string
in a database should generate identical results, independent of search locations, provided the searches are

running at the same time. If this assumption were violated, it would have serious consequences for the reliability and repeatability of the data and papers selected for a specific systematic review. Therefore, there is a need to know what variables and/or parameters should be included in the methodology of any search to ensure its repeatability. One of the most crucial steps is to define which database and engine search is going to be used for obtaining the data to be synthesized.

82 Differences among the most commonly used scientific search engines and databases are well 83 documented [13,15,16] but knowledge of the consistency within databases in relation to geographical 84 location where the search is requested from (but see Gusenbauer and Haddaway[13]), software 85 environment, or computer configuration remain surprisingly limited. Since the search histories of users may be stored in the browsers' cache, and considered by the scientific search engines, repeated and 86 87 identical searches may result in different outcomes. During a recent systematic review in ecology, we 88 accidentally discovered that a multi-locus search performed on 1 February 2018, using an identical search 89 string in Web of Science, produced radically different number of hits at different institutions at Hangzhou and Fuzhou, in China, and in Denmark (2,394, 1,571, and 7,447, respectively). 90

Since there is no known study comparing the consistency of returned papers over successive identical searches using several databases in one machine, we examined the way search engines deliver results and decided to systematically explore the inconsistencies found. Our study aimed to evaluate the consistency of search engines by comparing the outcomes from identical search strings ran on different computers from a wide range of localities across the world, with various software backgrounds, and using different search engines.

97 To investigate the repeatability of scientific searches in four of the major databases and search engines, 98 Web of Science, Scopus, PubMed, and Google Scholar, we generated search strings with two complexity 99 levels in ecology and medicine and ran standardized searches from various locations in the world, within 100 a limited timeframe. According to our null hypothesis, every search engine should give the exact same

101 number of results to the same search (after the search term has been adjusted to match the specific 102 requirements for each of these search engines), and therefore, a metric, showing the proportional deviance 103 of the search hits, should always be zero. We, therefore, first tested if summarized average absolute 104 deviation proportions (AADPs) for each search engine were significantly different from the ideal value 105 (zero) by using robust non-parametric tests. AADPs of search engines were compared to each other and 106 factors driving the differences were investigated. Similarly, the publications found by any given search 107 engine from identical searches should be also identical, thus, the mean similarities between search runs 108 should be 100%, and the scatter of the ordinated points should be zero. In order to test whether these 109 requirements were met, Jaccard distances[17] of the first twenty hits were used for within and between 110 group ordinations and multivariate analysis.

111 **Results**

Our time-synchronized, cross-institution and multi-location search exercise resulted in a large variation among the hits obtained using any of the search terms. Google Scholar generally yielded a greater number of hits than any other databases for all the locations (Table 1). As expected, less complex and medical search terms tended to result in greater hit numbers than complex ecological ones.

116Table 1. Comparison of the mean numbers of hits (SD) resulting from simple vs. complex search strings in the fields of ecology117and medicine using different search engines, different browsers and cache handling

			Number of hits of search strings in thousands			
			<u>Ecolo</u>	<u>egv</u>	Me	<u>dicine</u>
Search Engine	Browser	Cache	Simple	Complex	Simple	Complex
Google Scholar	Chrome	Full	1157.188± 991.840	2.069± 1.663	1165.170± 1167.252	28.117±25.262
Scholar		Cleaned	871.186±	1.595±	1013.800±	22.718±25.643
	Internet	Full	1065.303 1077.496±	1.699 1.945±	1178.801 1263.791±	28.140± 25.266
	Explorer		1018.818	1.685	1154.650	
		Cleaned	862.614± 1054.802	1.595± 1.699	1012.371± 1177.043	22.689±25.608
	Firefox	Full	905.849± 1026.956	1.945± 1.684	1263.791± 1154.650	28.113±25.266
		-	1020.330	1.004	1124.020	

		Cleaned	985.978±	1.816±	1169.975±	26.100±25.602
			1036.853	1.693	1179.213	
PubMed	Chrome	Full	2.881± 0.001	0.006± 0	147.726±	0.233± 0
					0.030	
		Cleaned	2.881± 0.001	0.006± 0	147.727±	0.233±0
					0.030	
	Internet	Full	2.881± 0.001	0.006± 0	147.729±	0.233±0
	Explorer				0.030	
		Cleaned	2.881± 0.001	0.006± 0	147.734±	0.233±0
	- 6		2 004 1 0 004	0.0001.0	0.030	0 0 0 0 1 0
	Firefox	Full	2.881± 0.001	0.006± 0	147.728± 0.030	0.233±0
		Cleaned	2.881± 0.001	0.006± 0	0.030 147.731±	0.233± 0
		Cleaned	2.881± 0.001	0.000±0	147.751± 0.030	0.255±0
Scopus	Chrome	Full	19.912±0	0.078± 0	545.558± 0	0.711± 0
Scopus	emente	Cleaned	19.912±0	0.078± 0		
					545.558±0	0.711±0
	IE	Full	19.912±0	0.078± 0	545.558± 0	0.711± 0
		Cleaned	19.912 ± 0	0.078± 0	545.558± 0	0.711± 0
	Mozilla	Full	19.912 ± 0	0.078± 0	545.558± 0	0.711± 0
		Cleaned	19.912±0	0.078± 0	545.558± 0	0.711± 0
Web of	Chrome	Full	17.295±1.214	15± 0	190.899±	0.357± 0.041
Science					24.163	
		Cleaned	17.561± 0.798	15± 0	195.432±	0.367± 0.026
					22.271	
	Internet	Full	17.642±0.740	15± 0	200.904±	0.373± 0.018
	Explorer				15.646	
		Cleaned	17.587±0.832	15± 0	199.665±	0.372± 0.020
					17.580	
	Mozilla	Full	17.492±0.967	14.9±0.49	192.108±	0.364± 0.031
		<u>.</u>	47 270 0 070	44.01.0 ==	24.784	0.001.0.005
		Cleaned	17.370±0.978	14.8±0.55	203.694±	0.36± 0.035
					38.988	

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The AADP (see Materials and Methods) of every search engine and database, except Scopus, significantly deviated from the desirable zero (Table 2). However, we have noticed that both PubMed and Web of Science were updated during the search process, at 17:00 GMT and 19:00 GMT, respectively. When the results from PubMed and Web of Science were split into two groups, before and after the time of the daily update, none of the AADPs from PubMed searches significantly differed from zero. In contrast, the results from Web of Science searches consistently showed significant deviation, indicating inconsistency in the number of returned hits by search location.

126	Table 2. Mean and standard deviations of recorded average absolute deviation proportions (AADP) for each investigated
127	search engines, separated by search topic and search expression complexity. Values are shown in percentage.

Topic/Complexity	GScholar	PubMed	Scopus	WoS
Ecology/Complex	85.319±9.426	0.000±0.000	0.000±0.000	0.629±1.964
Ecology/Simple	98.107±4.063	0.035±0.000	0.000±0.000	4.009±3.459
Medicine/Complex	90.889±5.966	0.000±0.000	0.000±0.000	5.845±5.757
Medicine/Simple	94.609±2.964	0.014±0.000	0.000±0.000	7.818±9.852

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129 The WelshADF test revealed significant differences in AADPs among groups (92.45% variance explained), with search engines being the most important explanatory variable (WJ = 69265.22, df = 3, p 130 131 < 0.001). Effects of the search topic (WJ = 8.49, df = 1, p = 0.005), keyword complexity (WJ = 71.71, df 132 = 1, p < 0.001), the interaction of search topic and keyword complexity (WJ = 20.40, df = 1, p < 0.001), 133 and their combination with search engine (Search engine \times Topic: WJ = 11959.03, df = 3, p < 0.001, Search engine \times Keyword complexity: WJ = 61790.69, df = 3, p < 0.001) on the outcome were all 134 significant. The effect of browsers used was not significant, either alone (WJ = 0.06, df = 2, p = 0.941) or 135 136 as a covariant of search engine choice (WJ = 0.29, df = 6, p = 0.943). Cache, whether it was emptied or 137 not, did not have a significant effect, either in its own or as a covariant (Fig 1, Supplementary Information 138 1, Supplementary Information 2-3). In spite of not being a significant predictor in the entire dataset, both browser and cache showed a tendency to influence the outcome of the Google Scholar results. None of 139 140 these influenced the search platforms with a background database. There were no differences in search 141 results when using Web of Science, PubMed and Scopus but different machines at the same location but 142 Google Scholar sometimes produced different results.

143 Fig 1. Average absolute deviation proportions (AADP) of hit numbers

AADPs are grouped by searched platforms, and separated by keyword complexity (complex, simple), and research area (ecology, medicine).Boxes represent interquartile range (IQR), with median AADP values represented as a thick horizontal band. Whiskers extend from Q1-1.5IQR to Q3+1.5IQ. Abbreviated search platforms: GScholar – Google Scholar, WoS – Web of Science.

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149 The multivariate analysis run on the first twenty papers collected from each search revealed significant 150 differences among the search engines (p = 0.01) but did not show a significant influence on browser 151 choice or cache state. Areas of convex hulls defined by these 'paper-communities' (see Methods) of the 152 first twenty hits were zero for Scopus only, and they were the largest for Google Scholar (Table 3). When 153 PubMed and Web of Science datasets were split by their update time, hulls for both PubMed subsets 154 became zero but remained greater than zero for Web of Science. Distance measures showed an analogous 155 pattern; they were zero for Scopus, indicating no difference between the first twenty papers, and deviated 156 from zero for all other platforms (Fig 2). After correcting for the database update, only Web of Science 157 and Google Scholar hulls remained significantly greater than zero.

158 Table 3 Areas of complex hulls for each search engines, separated by terms of topic and complexity.

Topic/Complexity	GScholar	PubMed	Scopus	WoS
Ecology/Complex	491.90	0.00	0.00	0.00
Ecology/Simple	322.24	490.37	0.00	8.82
Medicine/Complex	476.45	4.99	0.00	0.02
Medicine/Simple	625.03	428.56	0.03	41.81

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Fig 2. Average similarities of the first twenty papers within each search engine-topic-keyword
 complexity group, for each search platform.

Similarities were calculated based on binary matrices, using Jaccard distances. Median similarities are
indicated with a thick black line on the pirate plots. Abbreviated search platforms: GScholar – Google
Scholar, WoS – Web of Science.

165 **Discussion**

166 In this study, we identified a shortcoming of scientific search platforms that can decrease the transparency 167 and repeatability of the synthesis of quantitative evidence synthesis relying on database searches. Hence, 168 the creditability and reliability of the conclusions drawn from these syntheses may be compromised.

169 Our results showed significant differences in search platform consistency in terms of both the number of 170 hits (the size of the body of available evidence) and its composition when identical search terms were 171 queried at different geographic locations. We found that PubMed and Scopus had high consistencies, 172 whilst Google Scholar and Web of Science were not consistent in the number of hits they returned. 173 Google Scholar provided the greatest number of hits for every search, it also proved to be the least 174 consistent among different search runs, varying greatly in the number of hits, i.e. the total number of 175 papers. Contrarily, the composition of the evidence collected, characterized by the first twenty papers it 176 returned, was relatively consistent. Web of Science, however on a lower magnitude, showed similarly 177 poor consistency in terms of the number of hits returned from identical searches initiated from different 178 locations. Both the hit numbers and the returned list of articles from Scopus searches were consistent. 179 PubMed varied in hit numbers and had great dissimilarities among the returned sets of papers, especially 180 in those related to more general searches that necessarily had more hits. These dissimilarities were likely 181 due to a database update that happened during our search exercise. Indeed, data showed that 0, 6, 10, 25 182 papers (complex ecology, complex medicine, simple ecology, and simple medicine terms, respectively) 183 were added to the database during the course of this worldwide exercise. Since the papers listed were 184 ordered according to their time of inclusion in the dataset, the first 20 collected papers would greatly 185 differ and especially the larger values in the newly added articles can cause a disproportionally large 186 effect on the similarity of the 20 collected papers. Once the differences before and after database update 187 were accounted for, PubMed showed no deviation either in the number of returned papers or the list of the 188 first 20 listed papers. A similar change in the dataset happened with Web of Science during our search, but differences remained even after correcting for the update. This suggests that discrepancies were 189

caused by other sources, such as geographic locations. Overall, in our tests, Scopus and PubMed proved
to be the most consistent databases, and Web of Science and Google Scholar produced highly inconsistent
results.

Although we could not thoroughly decipher the influence of browser or cache on the search results, there was an indication that these factors only affected Google Scholar outcomes. Google Scholar is known to optimize search hits according to the search history of its users, thus, even the differences between browsers are likely to be the results of participants' previous browser use, and therefore different cache contents in different browsers.

198 While the disadvantages of the inconsistencies in Google Scholar search results have been repeatedly 199 illustrated[18,19], the similar behavior from Web of Science has only recently been reported[13] but in 200 neither case was the variability estimated nor were the potential solutions discussed. Given the 201 widespread use of Web of Science, neglecting this discrepancy can mislead scientists when drawing 202 conclusions from their evidence synthesis, when the body of evidence was collected by Web of Science 203 searches alone. The use of only one database is generally discouraged[5], and although some authors 204 mainly target Google Scholar-based reviews [18,20], it is clear here that relying on Web of Science alone, 205 or another single source, may lead to missing data or can make data-synthesis studies irreproducible. In 206 spite of the recommendations of the need to use multiple sources for such studies (see the PRISMA 207 statement[4]), a rapid scan of 20 recent papers in leading journals showed that recent, potentially highly 208 cited, ecology-related systematic reviews still used Web of Science as their only search engine 209 (Supplementary Information 4). In the light of the fact that using inadequate databases/search engines 210 makes systematic reviews unreliable, our findings are concerning.

211 There are various means of overcoming this issue:

a) Researchers conducting systematic reviews should be aware of this potential problem, and be explicitabout the methodology they use to ensure sufficient consistency and replicability. A detailed description

214 should be included on the search engines used (ideally more than one), search dates, the exact search 215 strings, as well as whether the same search was replicated by more than one person. As our study showed, 216 the location from which the search was conducted should also be reported, preferably along with the IP 217 address of the computer and the locality/institution the queries were initiated from. The exact time of the 218 search or the time window of the query are also essential. The holdings of databases, however, are not 219 constant, historical records can be added over time, and, therefore, queries even within a clearly limited 220 time period can deliver different result sets. Thus, reporting the time window of the queries can provide 221 only a partial solution.

b) The use of adequate search engines for a particular task should be an important consideration. All of the large databases have different strengths; Google Scholar searches grey literature, Web of Science has the largest (combined) dataset and, as our study confirmed, that Scopus and PubMed are the most consistent. Moreover, some databases may be more suitable for collecting information on a particular topic or have a greater historical coverage than others[14].

c) Providers of scientific search platforms should consider opening their search code and moving their paywalls to make reference lists publicly available[21], thus contributing to search consistency, and hence, scientific repeatability. Particularly Web of Science, as the most commonly used search engine, should act on making its search hits equally reachable to all users and, rather than *a priori* filtering them according to the institutions' paywall, restrict access only *after* the primary result set has been provided to the user.

d) Google Scholar, on the other hand, should open its computer code to allow researchers to understand how hit lists are generated and how results are ordered. Google Scholar has been criticized by the scientific community for the obscurity of its search algorithms[22]. Although we acknowledge that this can be against business policies for some companies, we argue that compromises must be made for the sake of research integrity and scientific rigor.

238 e) Providing well-documented, standard application programming interfaces (APIs) and generating 239 unique identifiers for searches, combining search term, result list, search time and location, and additional 240 metadata (e.g. computing environment) is required. Using an API for standardized searches would be 241 particularly beneficial for searches using Google Scholar that shows a strong dependence on the 242 computing environment. Although this solution could control for a great deal of variation derived mostly 243 from computing background and would be able to keep detailed records on the metadata of the searches, 244 it also brings up novel challenges. Firstly, APIs can be more complex to use than simple web interfaces 245 that may discourage users to use them. Moreover, collecting detailed data about search locations, or even 246 computing environment, raises both security and privacy concerns. Finally, storing individual searches 247 along with the necessary metadata may be resource heavy over a long period of time, which is likely to 248 increase maintenance costs, and therefore the subscription fees, of these services.

Should these steps towards ensuring repeatability not happen, the critical voices to web-based systematic reviews can claim unreliability of this method[11]. Given that the systematic review methodology was originally developed to handle contentious issues with various, often conflicting bodies of evidence[5], this is a critical issue. This matter can only be exacerbated by the appearance of automatic systematic reviews, relying on artificial intelligence[23].

Despite the limited number of institutions that participated in this exercise, and the overrepresentation of Europe, the lack of contribution from African, South American and other Asian countries, we found, even within the European countries, variation among the numbers of search hits. This suggests that adding more countries would have led to even greater variability in the resulting datasets. It may be interesting to test a wider range of search platforms and subjects to gain further understanding of the level of reliability of various systems and collect reliable knowledge on their strengths and weaknesses.

Since, the original set of raw data input can significantly alter/skew the output of the study and, in the age of big data, studies on already published results are becoming more common, an unbiased and timely way

of data extraction is needed. At present, updating systematic reviews using precisely repeated methodology is impossible[24]; hence a clear decision map on the advantages and disadvantages of particular databases and search engines should be drawn to ensure the integrity of publication-based studies.

266 Materials and methods

267 **Queried databases**

Three major scientific databases, PubMed, Scopus, and Web of Science, and Google Scholar, as the most used and largest scientific were used in this study. Although Google Scholar is markedly different from the other three traditionally used databases, both in business politics and search method[14,18], the increasing use of this search engine [20] justifies its inclusion in the study. The main differences between these databases have been catalogued and reviewed by Falagas et al.[14].

PubMed (https://www.ncbi.nlm.nih.gov/pubmed) is a freely available scientific database, focusing mostly on biomedical literature, which holds ca. 28 million citations covering a variety of aspects of life sciences (<u>https://www.ncbi.nlm.nih.gov/books/NBK3827/#pubmedhelp</u>. PubMed_Coverage, accessed 15/08/2018). It was developed and is being maintained by the National Center for Biotechnology Information.

Scopus, currently owned by the Elsevier group, contains bibliographic data of over 1.4 billion
publications dating back to 1970. It indexes ca. 70 million items and 22,500 journals from 5,000
publishers (<u>https://www.elsevier.com/solutions/scopus/how-scopus-works/content</u>, accessed: 17. August
2018).

Web of Science (https://webofknowledge.com) is the oldest scientific database, owned by the Clarivate
Analytics (previously Thomson Reuters). Web of Science, running under its current name since 1997, is

the successor of the first scientific citation database, the Science Citation Index, which was launched in 1964. It currently indexes 34,200 journals, books and proceedings, and, as of the last update, on 26 August 2018, it covers 151 million records altogether and over 71 million in its Core Collection (https://clarivate.libguides.com/webofscienceplatform/coverage). Currently it also includes Zoological Records, CABI Abstracts, and a number of other, formerly independent databases.

289 Google Scholar (https://scholar.google.com) is a free online tool, the sub-site of the search mogul Google 290 Inc., which is particularly designed for scholarly searches. Whilst Google Scholar has been often 291 criticized for not sharing its search algorithms, for its untraceable way of ordering search hits, and for the 292 inclusion of material from non-scholarly sources in its research hits [18,19,25], it has been playing an 293 increasing role in daily lives of scientists since its launch in 2004[20,26]. It is also estimated to include 294 160 million individual scientific publications in 2014[27] and to be the fastest growing resource for 295 scientific literature[28]. Its usefulness, however, for systematic reviews and meta-analyses has been 296 debated[16,18,19]

297 Web searches

In order to investigate the reproducibility of scientific searches in the four major search platforms, we generated keyword expressions (search strings) with two complexity levels using keywords that focused on either an ecological or a medical topic and ran standardized searches from various locations in the world (see below), all within a limited timeframe.

Simple search strings contained only one main keyword, whereas complex ones contained both inclusion and exclusion criteria for additional, related, keywords and key phrases (i.e. two-word expressions within quotation marks). Wildcards (e.g. asterisks) and logical operators were used in complex search strings. The main common keyword for ecology was "ecosystem" and "diabetes" was used for the medical topic. Search language was set to English in every case, and only titles, abstracts and keywords were searched.

- 307 Since different search engines use slightly different expressions for the same query, exact search terms
- 308 were generated for each search (Table 4).

309 Table 4 Search strings for each keyword complexity and topic, adjusted according to the search engines.

	<u>Ecology</u>		<u>Medicine</u>	
Search engine	Complex search string	Simple search string	Complex search string	Simple search string
GScholar	"ecosystem service"+"promoting"+"c rop"-"livestock"	"ecosystem services"	"diabetes"+"sugar"+"fruct ose"-"saccharose"	"diabetes mellitus"
PubMed	"ecosystem service"[Title/Abstract] AND "promoting" AND "crop"[Title/Abstract] NOT "livestock"[Title/Abstract] AND "english"[Language]	"ecosystem services"[Title/Abst ract] AND "english"[Language]	"diabetes"[Title/Abstract] AND "sugar" AND "fructose"[Title/Abstract] NOT "saccharose"[Title/Abstra ct] AND "english"[Language]	"diabetes mellitus"[Title/Abst ract] AND "english"[Language]
Scopus	TITLE-ABS-KEY ("ecosystem service" AND "promoting" AND "crop" AND NOT "livestock") AND (LIMIT- TO (LANGUAGE, "English"))	TITLE-ABS-KEY ("ecosystem services") AND (LIMIT-TO (LANGUAGE, "English"))	TITLE-ABS-KEY ("diabetes" AND "sugar" AND "fructose" AND NOT "saccharose") AND (LIMIT-TO (LANGUAGE, "English"))	TITLE-ABS-KEY ("diabetes mellitus") AND (LIMIT-TO (LANGUAGE, "English"))
WoS	TS=("ecosystem service" AND "promoting" AND "crop" NOT "livestock")	TS=("ecosystem services")	TS=("diabetes" AND "sugar" AND "fructose" NOT "saccharose")	TS=("diabetes mellitus")

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Searches were conducted on one or two machines at 12 institutions in Australia, Canada, China, Denmark, 311 312 Germany, Hungary, UK, and the USA (Supplementary Information 5), using the three main browsers 313 (Mozilla Firefox, Internet Explorer, and Google Chrome). Searches were run manually (i.e. no APIs were 314 used) according to strict protocols, which allowed to standardize search date, exact search term for every 315 run, and data recording procedure. Not all databases could have queried from every location: Google was 316 not available in China, and Scopus was not available at some institutions (Supplementary Information 5). 317 The original version of the protocol is provided in Supplementary Information 6. The first run was 318 conducted at 11:00 Australian Eastern Standard Time (01:00 GMT) on 13 April 2018 and the last search 319 run at 18:16 on 13 April 2018 Eastern Daylight Time (22:16 GMT). After each search the number of

resulted hits was recorded and the bibliographic data of the first 20 articles were extracted and saved in a file format that the website offered (.csv, .txt). Once all search combinations were run and browsers' cache had been emptied, the process was repeated. At four locations (Flakkebjerg, Denmark; Fuzhou, China; St. Catharines, Canada; Orange, Australia) the searches were also repeated on two different computers.

Results were collected from each contributor, bibliographic information was stripped out from the saved files, and was stored in a standardized database, allowing unique publications to be distinguished. If unique identifiers for individual articles were missing, authors, titles, or the combination of these were searched for, and uniqueness was double checked across the entire dataset.

For the rapid scan, if authors used Web of Science as the main search platform, and if search locations were reported, we chose the first twenty papers from a Google Scholar search (7 November, 2018) with the search term "systematic review" and "ecology". Sites were restricted to sciencemag.org, nature.com, and wiley.com.

333 Statistical analysis

To investigate how consistent the number of resulting hits from each search string (i.e. the combination of the search topic and keyword expression complexity) was for each of the search engines, *average absolute deviation* (AAD, i.e. the absolute value of the difference of the actual value and the mean) was calculated and expressed as a percentage of the mean of each group (*'average absolute deviation proportion'*, AADP, i.e. search topic, search term complexity, and search engine). AADP was calculated using the equation:

$$340 \quad AADP = \frac{|e - \hat{e}_{gr}|}{\hat{e}_{gr}},$$

where *e* was the number of hits from one particular search and \hat{e}_{gr} was the mean number of hits of pooled numbers from one topic and search term complexity combination and one search engine (e.g. complex

ecological search expression queried using Scopus). This grouping was necessary because the number of
hits substantially differed depending on these three factors. Since the aim of the study was not to compare
the efficiency of different search engines, this grouping did not interfere with our analysis.

Normality of the data and homoscedasticity were tested using Kolmogorov-Smirnoff test and the Breusch Pagan test, respectively. These tests confirmed that neither the distribution of AADPs followed normal distribution, nor were the variances of residuals within each group homogenous. Indeed, the high number of zeroes resulted in a zero-inflated, an unbalanced beta distribution, as suggested by the *descdist()* function in the *fitdistrplus* R package[29], under an R programming environment[30].

AADP is expected to be zero in cases when search engines consistently give the same number of hits within groups, regardless where the search is initiated from, browser used, or whether the cache was emptied or not. Therefore, one-sided Wilcoxon signed rank tests were performed for the AADP values for each search engines within each group to test if they were significantly different from zero.

To address non-normality, unequal variances and control Type I error, non-parametric, Welch-James's statistic with Approximate Degrees of Freedom (Welch ADF) was used to investigate the differences between search engine consistencies and to select the most influential factors driving these differences. This robust estimator uses trimmed means and winsorized variances to avoid biases derived from heteroscedasticity. Bootstrapping was used to calculate empirical p-values both for between group and pairwise comparisons[31], with the help of WelchADF R package[32].

Moreover, average similarities of the first twenty papers within each of the search engine-topic-keyword complexity groups were calculated based on binary matrices, in which rows corresponded to search runs from various institutions and computers, whilst columns contained individual papers. Due to its suitability for using binary data, Jaccard distance measures were applied for similarity calculations. Distance-based redundancy analysis (dbRDA, capscale() function) was used with the same similarity matrices to ordinate the resultant article collections in each search topic-keyword complexity group. Convex hulls of the

points resulted from this ordination were then delimited for each search engine and their areas were calculated. Since similarities between article collections resulted from searches with a search engine giving consistently the same hits, regardless of search location, browser used, and cache content, should always be zero, the ideal size of these hulls would be also zero.

371 Data availability statement

- 372 All data and computer code are deposited on the Open Science Framework (OSF) website and will be
- openly available for the readers through a stable URL or DOI upon acceptance.

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380 Author contributions

Gábor Pozsgai and Geoff Gurr conceived the project. Gábor Pozsgai designed the experiment, and did the statistical analysis. Gábor Lövei, Gábor Pozsgai, Jie Zhang, and Wenwu Zhou performed the preliminary searches. All contributors were involved in running the searches and providing raw data in the given format. The first drafted version of the manuscript was prepared by Gábor Pozsgai. This draft was first edited by Gábor Lövei, Liette Vasseur, Geoff Gurr, Olivia Reynolds, and Minsheng You. All authors were included in editing the subsequent versions of the manuscript. Minsheng You funded the work.

387 Competing interests

388 The authors declare no competing interest.

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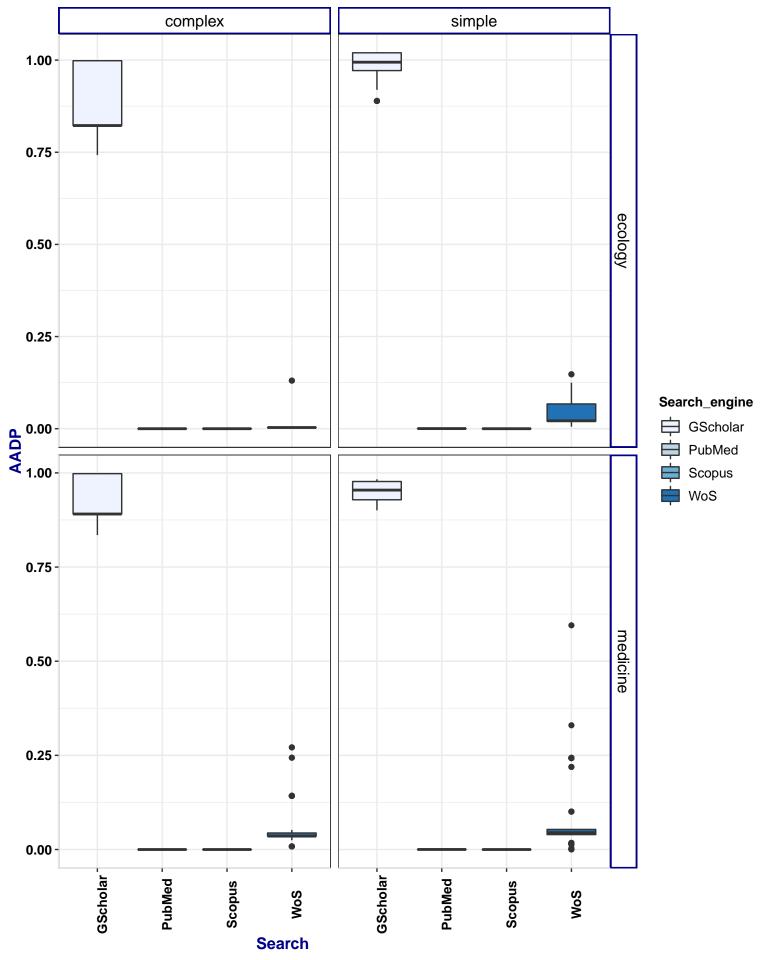
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467	Suppo	rting Information 1. The results of the Welch-James's statistic with Approximate Degrees of
468	Freedo	om. Significant ($p < 0.05$) relationships are highlighted with bold font.
469	Suppo	rting Information 2. Average absolute deviation proportions (AADP) of hit numbers, grouped by
470	search	ed platforms, and separated by grouped keyword complexity (complex, simple) – research area
471	(ecolo	gy, medicine) and cache state. Boxes represent interquartile range (IQR), with median AADP
472	values	represented as a thick horizontal band. Whiskers extend from Q1-1.5IQR to Q3+1.5IQ.
473	Abbre	viated search platforms: GScholar – Google Scholar, WoS – Web of Science.

474	Supporting Information 3. Average absolute deviation proportions (AADP) of hit numbers, grouped by
475	searched platforms, and separated by grouped keyword complexity (complex, simple) - research area
476	(ecology, medicine) and browser type. Boxes represent interquartile range (IQR), with median AADP
477	values represented as a thick horizontal band. Whiskers extend from Q1-1.5IQR to Q3+1.5IQ.
478	Abbreviated search platforms and browsers: GScholar – Google Scholar, WoS – Web of Science, Chrome
479	– Google Chrome, IE – Internet Explorer, Mozilla – Mozilla Firefox.
480	Supporting Information 4. The list of papers used in the rapid screen and the results showing how many
481	different search platforms were used and whether or not the date, search location and browser were
482	indicated.
483	Supporting Information 5. Names and affiliations of contributors and list of scientific search platforms
484	accessed during the search exercise.
485	Supporting Information 6. The exact protocol which was circulated to contributors, describing how
486	searches should be performed and how data should be saved.



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