

# SCIENTIFIC DATA

**CONFIDENTIAL**

COPY OF SUBMISSION FOR PEER REVIEW ONLY

**Tracking no:** SDATA-18-00628

***A novel replicability survey tool to measure and promote reproducibility in hydrology***

**Authors:** James Stagge (The Ohio State University), David Rosenberg (Utah State University), Adel Abdallah (Utah State University), Hadia Akbar (Utah State University), Nour Attallah (Utah State University), and Ryan James (Utah State University)

**Abstract:**

There is broad interest to improve the reproducibility of published research. We developed a simple survey tool to assess the 1) availability of data, models, code, directions, and other digital artifacts, and 2) replicability of results published in peer-reviewed journal articles. We used the tool to assess 360 random sampled articles of the 1,989 articles published in 2017 in six well-regarded hydrology and water resources journals. Like several prior studies, we replicated results for just a small fraction of articles (1.6%; estimated between 0.6% and 6.8% for all 1,989 articles with 95% confidence). Unlike prior studies, the tool helped identify key bottlenecks to make work more reproducible. These bottlenecks include: only some digital artifacts available (44% of articles), no directions (89%), or all artifacts available but results not replicable (5%). These results suggest the tool (or extensions) can help authors, journals, funders, and institutions to self-assess the availability and replicability of manuscripts, provide feedback to authors to improve manuscript reproducibility, and recognize and reward reproducible articles as examples to follow.

**Datasets:**

Repository Name	Dataset Title	Dataset Accession Number	URL	Reviewer Passcode
	jstagge/reproduc_hyd: Source code accompanying		<a href="https://github.com/jstagge/reproduc_hyd">https://github.com/jstagge/reproduc_hyd</a>	

# A novel replicability survey tool to measure and promote reproducibility in hydrology

James H. Stagge<sup>1,2\*</sup>, David E. Rosenberg<sup>1</sup>, Adel M. Abdallah<sup>1</sup>,  
Hadia Akbar<sup>1</sup>, Nour A. Attallah<sup>1</sup>, Ryan James<sup>1</sup>

October 20, 2018

1. Utah State University, Department of Civil and Environmental Engineering, Logan, UT 84321

2. The Ohio State University, Department of Civil, Environmental and Geodetic Engineering, Columbus, OH 43210

\*corresponding author(s): James H. Stagge (stagge.11@osu.edu)

## Abstract

There is broad interest to improve the reproducibility of published research. We developed a simple survey tool to assess the 1) availability of data, models, code, directions, and other digital artifacts, and 2) replicability of results published in peer-reviewed journal articles. We used the tool to assess 360 random sampled articles of the 1,989 articles published in 2017 in six well-regarded hydrology and water resources journals. Like several prior studies, we replicated results for just a small fraction of articles (1.6%; estimated between 0.6% and 6.8% for all 1,989 articles with 95% confidence). Unlike prior studies, the tool helped identify key bottlenecks to make work more reproducible. These bottlenecks include: only some digital artifacts available (44% of articles), no directions (89%), or all artifacts available but results not replicable (5%). These results suggest the tool (or extensions) can help authors, journals, funders, and institutions to self-assess the availability and replicability of manuscripts, provide feedback to authors to improve manuscript reproducibility, and recognize and reward reproducible articles as examples for others to follow.

## 1 Introduction

The science community is broadly interested to improve the reproducibility of research [1, 2, 3, 4, 5]. While exact definitions of reproducibility vary [6, 7, 1, 8, 9, 10, 11, 12, 13], one overarching theme is that reproducibility requires multiple, progressive components such as (i) all data, models, code, directions, and other digital artifacts used in the research are available for others to reuse (hereafter, “availability”; [14, 15, 16], (ii) the artifacts can be used to exactly replicate published results (replicability, sometimes called bit or computational

9 reproducibility; [17, 18], and (iii) existing and new datasets can be processed  
10 using the artifacts to reproduce published conclusions (reproducibility).

11 Replicable and reproducible scientific work is currently uncommon because  
12 of misaligned incentives and poor coordination among authors, journals, in-  
13 stitutions, and funding agencies that conduct, publish, and support scientific  
14 research [19, 20, 8]. For example, making artifacts available requires authors  
15 to document additional materials [21, 22] and learn new skills and technologies  
16 [23]. Authors may worry that shared materials will never be used [24, 9] or that  
17 other scientists will scoop them on follow-up studies [25]. Further, universities  
18 typically reward peer-reviewed journal publications, rather than data reposi-  
19 tories or documentation, while current scientific culture rewards novelty rather  
20 than replicating prior efforts [17, 2].

21 Several efforts are underway to encourage more reproducible science [8]. Au-  
22 thors can share research materials in a growing number of online repositories  
23 such as Github, Figshare, Harvard Dataverse, Dryad, and HydroShare. Insti-  
24 tutional libraries are transitioning to offer online repositories to house digital  
25 research artifacts [26, 27, 28, 29]. Within our field of hydrology and water re-  
26 sources, recent tools provide environments to repository data and allow software  
27 to operate on the data as well as create virtual environments that package code,  
28 data, and a working operating system to reduce problems of incompatibility  
29 (e.g., [30, 31]). Authors can assign digital object identifiers (DOIs) to research  
30 packages to create persistent links and use umbrella research licenses to describe  
31 the manner in which these digital artifacts and their associated paper may be  
32 legally used by others [19]. Additionally, authors can specify the level of repro-  
33 ducibility that readers and reviewers can expect from each publication. And  
34 yet, despite these powerful tools, few authors are making their work available  
35 for others to replicate or reproduce.

36 To quantify the current state of reproducible science in hydrology and to  
37 understand the factors preventing more replicable or reproducible publications,  
38 we present here a simple 15-question survey tool (Fig. 1) designed to assess  
39 the availability of digital artifacts and replicability of results in peer-reviewed  
40 journal articles (see Methods). We use this survey tool to assess 360 stratified  
41 random-sampled articles from the 1,989 articles published in 2017 across six  
42 reputable hydrology and water resources journals. Sampling was stratified by  
43 journal and reproducibility keywords of interest. Results identify bottlenecks  
44 to making digital artifacts available and replicating output. We also use results  
45 to generalize replicability for the entire sample of articles, test a hypothesis  
46 about use of keywords to identify replicable articles, compare the effectiveness  
47 of different journal data availability policies, and highlight how authors, jour-  
48 nals, funders, and institutions can use the enclosed survey tool to encourage,  
49 recognize, and reward the publication of more reproducible articles.

## Results

Applying our survey tool to 360 random-sampled hydrology articles published in 2017 shows a decreasing number of articles are able to satisfy the progressively stricter reproducibility requirements of artifact availability and ultimately replication of published results (Fig. 2). For example, 70.3% of the 360 sampled articles stated some materials were available, but we could only access 48.6% of those materials online (Fig. 3). Only 5.6% of sampled articles made data, model/code, and directions publicly available while just 1.1% of sampled articles made artifacts available and were fully replicated. We partially replicated an additional 0.6% of articles.

### Artifact Availability

Across all journals, the most common artifact provided was input data, followed by code/model/software, and then directions to run (Fig. 4). Other artifacts, such as hardware/software requirements, common file formats, unique and persistent digital identifiers, and metadata, were made available at much lower rates than the primary artifacts. Articles published in Environmental Modeling & Software (EM&S) made artifacts available online at a high rate and these articles also provided data/model/code, directions, hardware/software requirements, common file formats, and metadata at rates 2 to 3 times the next most frequent journal.

Sampled articles use different methods to make artifacts available and these methods differ markedly across journals (Fig. 4). EM&S had the largest proportion of sampled articles that made at least some artifacts available online (61.9%). By contrast, Hydrology and Earth Systems Sciences (HESS) and Water Resources Research (WRR) had high percentages of articles where materials were only available by first author request (38.5-40.2%). Otherwise, the Journal of Hydrology (JoH), Journal of the American Water Resources Association (JAWRA), and Journal of Water Resources Planning and Management (JWRPM) had large proportions of articles where data was not available or was available within the article and supplemental material.

### Replication of Results

Twenty sampled articles (5.6% of total sampled articles) made the required input data, software/model/code, and directions available, allowing an attempt at replicating published results. We were able to fully replicate results for four articles [32, 33, 34, 35] and partially replicated results for two additional articles [36, 37]. We were unable to replicate results for four articles [38, 39, 40, 41], which nonetheless appeared to provide the necessary materials. During the process to replicate results, we found 10 of the 20 articles did not have all the required artifacts, despite being initially considered for replication testing. Reasons we only partially replicated results for two articles and did not replicate results for four articles included unclear directions (4 articles), did not generate

91 results (3 articles), hardware/software error (2 articles), or results differed from  
 92 the publication (1 article; some articles had multiple reasons). The survey per-  
 93 mitted multiple selections for this question. A common issue across cases where  
 94 we did not generate results was that folder and file locations were hard-coded  
 95 to work on the author's computer. If these issues were obvious, we tried, with  
 96 limited success, to fix them. Other cases pointed to general data gateways, like  
 97 the USGS streamgauge network, with no further details or required expensive or  
 98 proprietary software. Of the 10 articles that had all artifacts available, five were  
 99 published in EM&S, two articles were published in HESS, and the remaining  
 100 journals each had one article.

## 101 **Estimated Replicability for Population**

102 Because the stratified sampling method oversampled articles with certain repro-  
 103 ducibility keywords, we used bootstrap resampling (see Methods) to estimate  
 104 that 0.6% to 6.8% of all 1,989 articles published in 2017 in the six journals  
 105 tested here would be replicable (95% confidence interval). We estimated 28.0%  
 106 (23.1-32.6% confidence interval) of all articles published in these journals during  
 107 2017 provided at least some of the artifacts necessary for replicability (Fig. 5,  
 108 black). EM&S differed from other journals by having a large proportion of ar-  
 109 ticles with some or all data available (31.8-64.0%) and relatively high estimates  
 110 of replicability (Fig. 5).

## 111 **Using Keywords to Identify Replicable Articles**

112 We found that five of the six articles with some or complete replicability had  
 113 certain reproducibility-related keywords of interest in their abstracts (full list  
 114 in Methods). This positive hit rate (4.2%) for articles with reproducibility  
 115 keywords is significantly greater than articles without (0.4%; 2-sample Chi-  
 116 Squared test with Yate's continuity correction ( $p = 0.014$ )). These findings  
 117 confirm the value of reproducibility keywords to identify replicable articles and  
 118 reaffirm the difficulty to know at the outset whether the results presented in an  
 119 article are replicable.

## 120 **Time Required to Determine Availability and Replicability**

121 We surveyed and analyzed the time required to complete the survey to show  
 122 the incremental effort required to determine the availability of article arti-  
 123 facts and replicability of results (Fig. 6). For example, it took us as little  
 124 as 5 to 14 minutes (25-75% range) to determine the availability of input data,  
 125 model/software/code, and directions and upwards of 25 to 86 minutes (25-75%  
 126 range) to fully replicate results with outliers of 25 and 200 minutes (minimum  
 127 and maximum). There were no statistically detectable differences in the time  
 128 between journals to determine availability of digital artifacts or replicate results.

## 129 Reproducibility and Journal Policies

130 Among the six hydrology and water resources journals we studied, the HESS and  
131 WRR policies effective during the 2017 review period require articles to state  
132 how data, models, and code can be accessed. In contrast, the 2017 policies by  
133 EM&S, WRRM, JoH, and JAWRA simply encouraged this practice. EM&S fur-  
134 ther recommends articles include an explicit "Software and/or data availability"  
135 section within the article and requires authors to make software essential to the  
136 paper available to reviewers during the review process (Supplemental Material  
137 Table S1).

138 These journal data availability policies partially but not fully explain dif-  
139 ferences in artifact availability and result replicability by journal. For exam-  
140 ple, EM&S, which only encourages authors to make artifacts available, had the  
141 highest rate of articles that made artifacts available (Fig. 3) and this high rate  
142 persisted across nearly every artifact category (Fig. 4). Although EM&S used  
143 "should" instead of "must" statements, their policy was by far the most specific  
144 for papers with a software component (Supplemental S1). This may explain  
145 their high participation rate. EM&S is also a software-focused journal, which  
146 may attract papers and authors that are more conscious of reproducible soft-  
147 ware. In contrast, HESS and WRR, which require data availability statements,  
148 had lower percentages of articles that made artifacts available and more papers  
149 that direct readers to the authors or third parties for data, models, or code (Fig.  
150 3). These directional statements tend to appear in the Data Availability section  
151 of HESS articles and the Acknowledgements of WRR articles. The final group,  
152 JoH, JAWRA, and JWRP&M, that also encouraged authors to make artifacts  
153 available, had high proportions of articles without available digital artifacts (Fig.  
154 3). The HESS and WRR policies that require data availability statements ap-  
155 pear to encourage authors to select options like contact the author rather than  
156 work to provide a research article and supporting materials that are available,  
157 replicable, and reproducible. In July 2018, JWRP&M switched to start requir-  
158 ing authors to state the availability of data, models, and code, similar to HESS  
159 and WRR [42].

## 160 Discussion

161 Our findings of low replicability of research published in six hydrology and water  
162 resources journals in 2017 mirrors low rates of replicability previously reported  
163 in psychology (100 experiments; [2]), computer systems research (613 articles;  
164 [43]), and across articles published in Science (204 articles; [5]). Unlike those  
165 studies, our survey tool additionally identified bottlenecks to making all digital  
166 artifacts available and replicating results. Here, we discuss how results for our  
167 study in hydrology and water resources can inform broader use of the survey tool  
168 by authors, journals, funders, and institutions to improve the reproducibility of  
169 published scientific research.

170 **Authors.** Authors can use the survey tool as a checklist to self-assess

the availability of data, models, code, and directions and replicability of their work before submitting work for publication. The tool can help identify missing components that, if provided, will improve reproducibility. For example, our results showed that, for all journals, the number of sampled articles with code/data/software was consistently 2 to 3 times higher than the number of articles that additionally provided directions (Fig. 4). If authors used the tool and subsequently provided directions to use their materials, the tool could potentially double the number of articles which could reasonably be tested for replicability. Another bottleneck was a large fraction of authors who chose not to make artifacts available or only made artifacts available by author or third party request. Authors can look to the 10 articles we found that made all digital artifacts available to see easy-to-access platforms to provide access. These platforms included GitHub (6 articles), HydroShare (1 article), journal supplementary material (1 article), a custom website (1 article), or Figshare (1 article). Authors who bundled their code and data together in a single GitHub repository further allowed us to download the entire project, with a higher likelihood that code pointed to valid file directories.

**Journals.** Journals can use the survey tool to assess the availability of data/model/code and directions, replicability of results of new submissions, and provide feedback to authors. Alternatively, journals can require that authors use the survey tool to self-check their work prior to submission. Feedback can be crucial as our study showed that a very low fraction of articles provided all the required artifacts. However, when artifacts were available, we fully or partially replicated results for 6 out of 10 articles. We also found that time to assess the availability of artifacts was typically less than 15 minutes, while time to replicate results was much longer. The combination of these findings suggests that promoting inclusion of digital artifacts through a relatively quick availability survey could pay significant dividends for replicability. We leave open whether responsibility for assessment should fall on a journal editor's assistant, associate editor for reproducibility, reviewers, or others. With a tool to measure reproducibility of published articles, journals could also track reproducibility over time. Tracking and publishing this information would benefit the journal as a promotional tool to show journal commitment to reproducible science. Tracking would also allow journals to acknowledge articles and authors that reach certain reproducibility standards. For example, show a bronze, silver, or gold medal icon on article webpages to recognize and reward progressively greater reproducibility corresponding to availability, replicability, and reproducibility. These badges would simultaneously communicate the expected level of reproducibility of published work. In our study, we are excited to award silver badges to the four articles whose results we fully replicated [32, 33, 34, 35] and bronze badges to six articles that made all artifacts available [36, 38, 39, 40, 41, 37]. This recognition also makes it easier for authors to find and emulate excellent reproducibility practices. Cross-journal indices could further aggregate reproducibility metrics and encourage journals and authors to improve the reproducibility of their research portfolios. To oversee these journal efforts, we envision a new role for an Associate Editor of Reproducibility to develop journal data availability and re-

217 reproducibility policy, manage reproducibility evaluations, and advocate for best  
218 reproducibility practices.

219 **Funders and Institutions.** Similar to journals, funders and institutions  
220 can use the survey tool to assess artifact availability, verify replicability of re-  
221 sults, and recognize or reward authors whose work achieves bronze, silver, and  
222 gold levels of reproducibility. Alternatively, funders and institutions could use  
223 reproducibility assessments made by journals. Funders can encourage authors  
224 to use the survey tool to self-check work prior to submitting progress or final  
225 reports or use the tool to check the reproducibility of work authors submit.  
226 Use of the tool could help verify that author submissions fulfill requirements of  
227 funder data management policies and help direct authors to improve the repro-  
228 ducibility of their work. Institutions could also use the survey tool to determine  
229 and post the expected level of reproducibility of author work deposited into  
230 institutional repositories.

231 Together, these actions by authors, journals, funders, and institutions can  
232 help nudge authors further along the reproducibility continuum to make their  
233 digital artifacts more available and to replicate published results. While these  
234 proposed policy nudges represent small shifts targeted at particular actors in  
235 the science community, this approach can produce large effects collectively [44],  
236 particularly when all parties agree that the shift will provide a net benefit, as for  
237 reproducible science. Each individual nudge is made possible by using a survey  
238 (or similar) tool to measure and quantify the availability of digital artifacts,  
239 replicability of published results, and reproducibility of findings. We welcome  
240 discussion to improve the survey tool aimed at improving the reproducibility of  
241 our science.

## 242 Methods

### 243 Online Survey Tool

244 The authors translated definitions of availability, replicability, and reproducibil-  
245 ity into a 15-question Qualtrics Research Core (Qualtrics) online survey (Fig. 1).  
246 The Qualtrics survey format has been converted into a publicly available version,  
247 provided here as an example (<https://goo.gl/forms/95S4y9BdPmVqMtm02>).  
248 The survey progressed from soliciting metadata about the article (Questions 1-  
249 4), to testing availability of artifacts (Q5-9), and ultimately testing replicability  
250 of results (Q10-14). Green or yellow shaded answers (Fig. 1) were required to  
251 progress to the next question so that survey questions followed the availability  
252 and replicability progression. Selecting a red-shaded answer stopped progression  
253 and directed the reviewer to a final question that asked how many minutes the  
254 reviewer spent to reach their stopping point (Q15). This time to complete was  
255 self-reported by reviewers rather than using the built-in Qualtrics timer so re-  
256 viewers could consider the entire time spent reading and assessing the published  
257 article, rather than the time completing the survey.

258 The authors developed the tool over four months in Fall 2017 and pre-tested



it in early 2018 on a sub-sample of five articles that spanned the availability and replicability progression. From our experience pre-testing and to improve use of the tool, we reworded some questions, altered the survey logic, discussed and addressed inter-reviewer variability. Later, after we had reviewed 23% of sampled articles, we added a final question (Q15) to ask how much time it took to complete the survey.

## Selection of Articles

360 peer-reviewed articles were random stratified sampled from the 1,989 articles found in Scopus that were published in 2017 in six well-regarded hydrology and water resources journals (measured by impact factor and reputation). The six journals were Hydrology of Earth Systems Sciences (HESS), Environmental Modeling & Software (EM&S), Journal of the American Water Resources Association (JAWRA), Journal of Hydrology (JoH), Journal of Water Resources Planning and Management (JWRPM), and Water Resources Research (WRR). Stratified random sampling was approximately proportional to the number of articles published in each journal in 2017, with extra weight placed on articles with a set of reproducibility-related keywords (Table M1).

We further adjusted the stratification so each journal had at least 30 articles (JAWRA and JWRPM were oversampled). Similarly, we oversampled articles with the keywords: analytical software, application programs, C++, cloud computing, computational reproducibility, computer modeling, computer programming, computer software, computer software usability, computer-based models, development and testing, engineering software, fortran, freely available data, freely available software, github, hardware and software, java, open code, open source, replicative validation, scientific software, code, python, cran, and http. Of the 120 articles published in the six journals in 2017 that had at least one keyword, we sampled 119 articles, principally to retain at least 15 non-keyword articles for each journal with an approximately 2:1 non-keyword to keyword ratio overall.

Table M1: Number of articles published in 2017 and number of articles sampled by journal.

	EM&S		HESS		WRR		JoH		JAWRA		JWRP&M	
	2017	Sample	2017	Sample	2017	Sample	2017	Sample	2017	Sample	2017	Sample
Keyword	49	48	9	9	23	23	24	24	7	7	8	8
Non-key word	181	15	319	43	511	59	645	79	102	23	111	22
Total	230	63	328	52	534	82	669	103	109	30	119	30

Each author was randomly assigned 60 articles stratified by journal to assess the availability of article artifacts (Q1-9). After identifying all publications that had the available artifacts, we re-assigned reviewers to assess whether the published results could be replicated (Q1-15). Reassignments matched article

software to the reviewer most familiar with those tools. The Qualtrics online format allowed us to both simultaneously and asynchronously assess journal articles and store survey responses in an accompanying Qualtrics database. After all availability and replicability assessments were complete, we exported results from the Qualtrics database to a text file which was processed in R to generate figures, tables, and results. Time spent to complete the survey (Q15) was analyzed for three key stopping points: no artifacts available (Q5), availability of artifacts (Q9), and replicability of results (Q13).

## Population Estimates

Bootstrap resampling was used to estimate the overall percentage of articles from all  $n=1,989$  articles published in 2017 in the six journals while adjusting for keyword and journal sampling. Sampled articles were sorted into six mutually exclusive categories that were stopping points in the survey: Data-less or review, Author or Third Party Only, No availability, Some Availability, Available but not reproducible, and Some or All Replicable. "Some availability" included articles with one or two data/model/code, and direction elements of the three required elements (Q7). "Available but not replicable" articles had all three required elements available on the initial review, but either could not be replicated or were found to be missing key elements when replication was attempted.

The resampling approach was based on 5,000 simulations of the population. For each simulation, the 360 sampled articles were inserted directly based on the survey results, assuming that these were known exactly. Estimates for the remaining 1,629 unsampled articles were simulated based on survey results for the sampled articles in their stratified category, i.e. journal and keyword/non-keyword. Uncertainty for the unsampled articles was calculated using the approach of Sison and Glaz [45], assuming the six potential levels of availability/replicability represent multinomial proportions of categorical data [46].

## Acknowledgements

This material is based upon work supported by Utah Mineral Lease Funds, the National Science Foundation, funded through OIA – 1208732, and the U.S. Fullbright Program. Additional funding was by National Science Foundation grant #1633756. Any opinions, findings, and conclusions or recommendations expressed in this material are those of the author(s) and do not necessarily reflect the views of any of the funding organizations.

The authors thank Amber S Jones for providing feedback on an early draft, Stephen Maldonado and Marcos Miranda for external review of the code repository, and Ayman Alafifi for participation in early discussions to develop the survey tool.

## 330 Author Contributions

331 Stagge, Rosenberg, James, and Abdallah conceived the idea for a survey tool  
 332 to measure the reproducibility of journal articles. Stagge and Rosenberg led  
 333 the design and refinement of the tool and Abdallah implemented the design.  
 334 All authors participated equally in the use of the survey tool to evaluate the  
 335 reproducibility of articles. Stagge led the results analysis, visualization, and  
 336 writing of the first draft. Stagge and Abdallah made all article digital artifacts  
 337 available online. Rosenberg rewrote the initial draft. All authors reviewed and  
 338 approved the final draft prior to submission.

## 339 Availability and Replicability of Research Artifacts

340 The survey tool, Qualtrics results, and all code used for analysis presented in this  
 341 article are available online through GitHub ([https://github.com/jstagge/](https://github.com/jstagge/reproduc_hyd)  
 342 [reproduc\\_hyd](https://github.com/jstagge/reproduc_hyd)) or through a permanent repository [47]. Please cite this repos-  
 343 itory for any use of the related data or code. Additionally, results can be repli-  
 344 cated using RStudio deployed in the cloud using MyBinder through the GitHub  
 345 website. An open Google Forms version of the survey tool is available for readers  
 346 to use, modify, and extend. (<https://goo.gl/forms/95S4y9BdPmVqMtm02>).

## 347 Competing financial interests

348 The author(s) declare no competing financial interests.

## 349 Figures Legends

### Paper Metadata

Q1. Assessor's name  
Q2. Journal name  
Q3. Article DOI  
Q4. Full paper citation

### Availability

Q5. How accessible to users?

Some or all applicable Not specified where Not applicable

Q6. Where available?

All online Third party Author In article

Q7. What is present?

Required Optional

Input Data Code / License Metadata Identifiers  
Directions Software Hardware / software requirements File format

Q8. Comments on availability?

Q9. Could you and readers use the available artifacts to generate results?

Yes Not sure Not familiar with resources No

Q10. Continue to replicate results?

Yes No

### Replicability

Q11. Do the outputs verify published results (in text, figures, and tables)?

Yes (explain in Q12) No (explain in Q13 and Q14)

Q12. Explain what made this replication work? Other comments?

Q13. Why did the replication fail?

Hardware / software errors Did not generate results Results differed  
Unclear directions Other

Q14. Explain why the replication failed?

### Time to Complete

Q15: How many minutes did the survey take?

Figure 1: Survey questions to assess journal article data availability and replicability. Green and grey answers continue to the next question, while red answers skip to question 15.

SciDATA\_LOGO.JPG

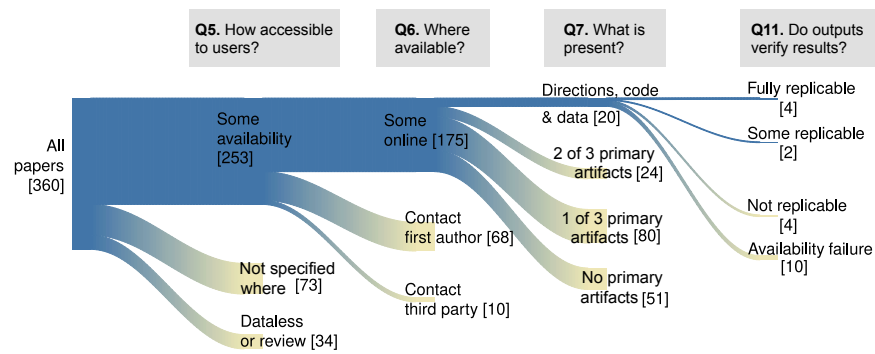


Figure 2: Number of papers progressing through the survey questions to determine availability and replicability requirements.

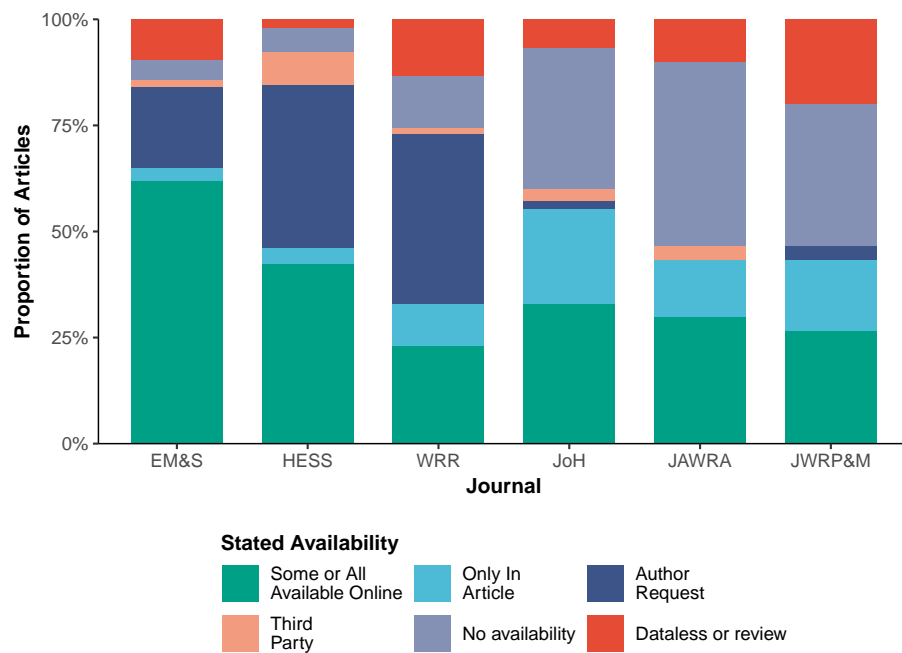


Figure 3: Data, model, code availability by journal (summary of Q4 and Q5).

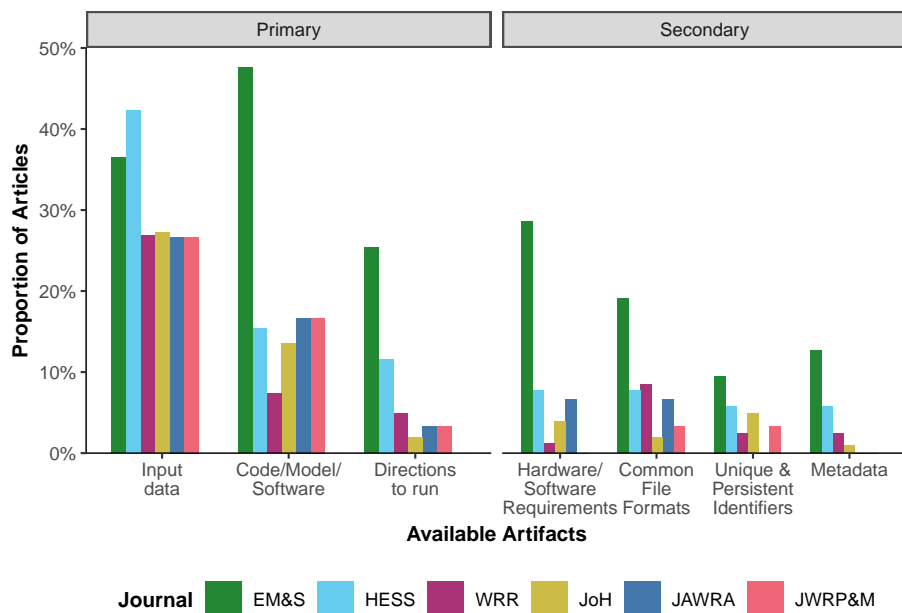
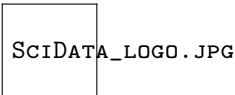


Figure 4: Availability artifacts organized by journal. All percentages are based on the total number of sampled papers for each journal.

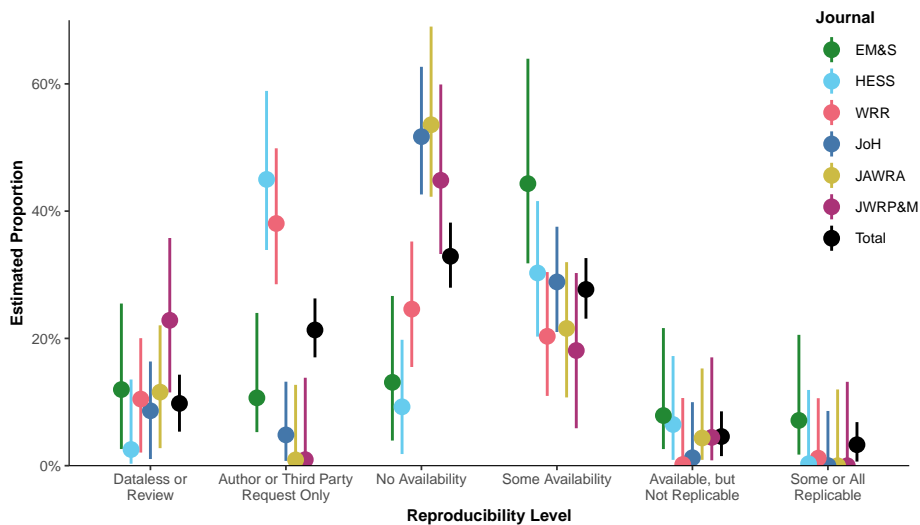


Figure 5: Population estimate of reproducibility for all papers published in 2017. Results are sorted by journal, with “Total” representing all 6 journals. Median estimate is represented by a point, vertical bars show the 95% confidence interval.

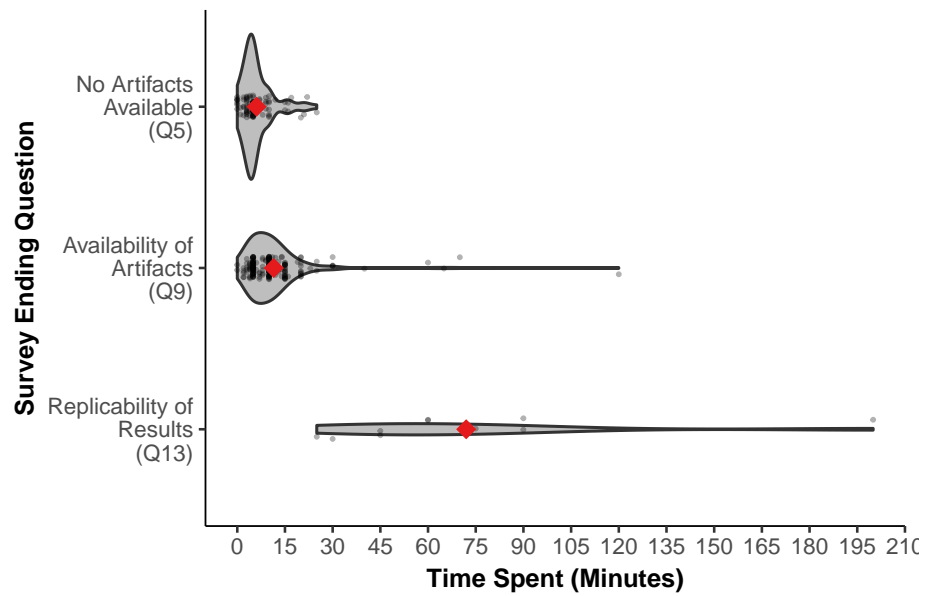
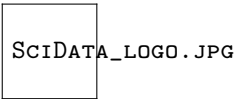


Figure 6: Self-reported time to complete survey organized by the survey’s ending question. Each reviewed paper is shown by a black dot, while the mean is represented by a red dot. Distribution density is shown by width.

## 350 Supplementary Material

Table S1: Summary of journal data policy snapshots in the selected water resources journals as of December 2017. The key word indicating the level of the requirement is highlighted in bold.

Journal	Data Policy Statement	URL
Environmental Modeling & Software (EM&S)	Where appropriate, a software/data availability section <b>should</b> be provided, containing as much of the following information as possible ...  When a software component is an essential part of the paper presentation, authors <b>should</b> be prepared to make it available to reviewers during the review process. To preserve the anonymity of reviewers, the authors <b>should</b> make the software available for a download.	<a href="https://www.elsevier.com/journals/environmental-modelling-and-software/1364-8152/guide-for-authors">https://www.elsevier.com/journals/environmental-modelling-and-software/1364-8152/guide-for-authors</a>
Hydrology and Earth Systems Science (HESS)	Each manuscript <b>must</b> have a final Data Availability section that states how the underlying research data can be accessed.	<a href="http://www.hydrology-and-earth-system-sciences.net/about/data_policy.html">http://www.hydrology-and-earth-system-sciences.net/about/data_policy.html</a>
Water Resources Research (WRR)	AGU's Data Policy states that all data necessary to understand, evaluate, replicate, and build upon the reported research <b>must</b> be made available and accessible whenever possible.	<a href="http://publications.agu.org/author-resource-center/publication-policies/data-policy/">http://publications.agu.org/author-resource-center/publication-policies/data-policy/</a>
Journal of the American Water Resources Association (JAWRA)	JAWRA <b>encourages</b> authors to share the data and other artefacts supporting the results in the paper by archiving it in an appropriate public archive.	<a href="https://onlinelibrary.wiley.com/page/journal/17521688/homepage/forauthors.html">https://onlinelibrary.wiley.com/page/journal/17521688/homepage/forauthors.html</a>
Journal of Hydrology (JoH)	To foster transparency, we <b>encourage</b> you to state the availability of your data in your submission.	<a href="https://www.elsevier.com/journals/environmental-modelling-and-software/1364-8152/guide-for-authors">https://www.elsevier.com/journals/environmental-modelling-and-software/1364-8152/guide-for-authors</a>
Journal of Water Resources Planning and Management (JWRPM)	Authors are <b>encouraged</b> to use Supplemental Data to show all necessary data	<a href="https://ascelibrary.org/doi/pdf/10.1061/9780784479018.ch03">https://ascelibrary.org/doi/pdf/10.1061/9780784479018.ch03</a>



## References

- [1] Sandve, G. K., Nekrutenko, A., Taylor, J. & Hovig, E. Ten Simple Rules for Reproducible Computational Research. *PLOS Computational Biology* **9**, e1003285 (2013).
- [2] Aarts, A. *et al.* Estimating the reproducibility of psychological science. *Science* **349**, 1–8 (2015).
- [3] Baker, M. 1,500 scientists lift the lid on reproducibility. *Nature* **533**, 452–454 (2016).
- [4] Brembs, B. Prestigious Science Journals Struggle to Reach Even Average Reliability. *Frontiers in Human Neuroscience* **12** (2018).
- [5] Stodden, V., Seiler, J. & Ma, Z. An empirical analysis of journal policy effectiveness for computational reproducibility. *Proceedings of the National Academy of Sciences* **115**, 2584–2589 (2018).
- [6] Kovacevic, J. How to Encourage and Publish Reproducible Research. In *2007 IEEE International Conference on Acoustics, Speech and Signal Processing - ICASSP '07*, vol. 4, IV–1273–IV–1276 (2007).
- [7] Peng, R. D. Reproducible research and Biostatistics. *Biostatistics* **10**, 405–408 (2009).
- [8] Stodden, V., Borwein, J. & Bailey, D. H. Setting the default to reproducible. *computational science research. SIAM News* **46**, 4–6 (2013).
- [9] Easterbrook, S. M. Open code for open science? *Nature Geoscience* **7**, 779–781 (2014).
- [10] Leek, J. T. & Peng, R. D. Opinion: Reproducible research can still be wrong: Adopting a prevention approach. *Proceedings of the National Academy of Sciences* **112**, 1645–1646 (2015).
- [11] Pulverer, B. Reproducibility blues. *The EMBO Journal* **34**, 2721–2724 (2015).
- [12] Goodman, S. N., Fanelli, D. & Ioannidis, J. P. A. What does research reproducibility mean? *Science Translational Medicine* **8**, 341ps12–341ps12 (2016).
- [13] Melsen, L., Torfs, P., Uijlenhoet, R. & Teuling, A. Comment on “Most computational hydrology is not reproducible, so is it really science?” by Christopher Hutton *et al.* *Water Resources Research* **53**, 2568–2569 (2017).
- [14] Akmon, D., Zimmerman, A., Daniels, M. & Hedstrom, M. The application of archival concepts to a data-intensive environment: working with scientists to understand data management and preservation needs. *Archival Science* **11**, 329–348 (2011).

- 388 [15] Hutton, C. *et al.* Most computational hydrology is not reproducible, so is  
389 it really science? *Water Resources Research* **52**, 7548–7555 (2016).
- 390 [16] Añel, J. A. Comment on “Most computational hydrology is not repro-  
391 ducible, so is it really science?” by Christopher Hutton *et al.* *Water Re-*  
392 *sources Research* **53**, 2572–2574 (2017).
- 393 [17] Casadevall, A. & Fang, F. C. Reproducible Science. *Infection and Immunity*  
394 **78**, 4972–4975 (2010).
- 395 [18] Drummond, C. Reproducible research: a minority opinion. *Journal of*  
396 *Experimental & Theoretical Artificial Intelligence* **30**, 1–11 (2018).
- 397 [19] Stodden, V. The Legal Framework for Reproducible Scientific Research:  
398 Licensing and Copyright. *Computing in Science & Engineering* **11**, 35–40  
399 (2009).
- 400 [20] Fary, M. & Owen, K. Developing an Institutional Research Data Manage-  
401 ment Plan Service (2013).
- 402 [21] Shen, Y. Research Data Sharing and Reuse Practices of Academic Faculty  
403 Researchers: A Study of the Virginia Tech Data Landscape. *International*  
404 *Journal of Digital Curation* **10**, 157–175 (2016).
- 405 [22] Shiffrin, R. M., Börner, K. & Stigler, S. M. Scientific progress despite  
406 irreproducibility: A seeming paradox. *Proceedings of the National Academy*  
407 *of Sciences* **115**, 2632–2639 (2018).
- 408 [23] Diekema, A., Wesolek, A. & Walters, C. The NSF/NIH Effect: Survey-  
409 ing the Effect of Data Management Requirements on Faculty, Sponsored  
410 Programs, and Institutional Repositories. *Data* (2014).
- 411 [24] Wallis, J. C., Rolando, E. & Borgman, C. L. If We Share Data, Will  
412 Anyone Use Them? Data Sharing and Reuse in the Long Tail of Science  
413 and Technology. *PLoS ONE* **8**, e67332 (2013).
- 414 [25] Kaufman, D. & PAGES 2k special-issue editorial team. Technical Note:  
415 Open-paleo-data implementation pilot – The PAGES 2k special issue. *Clim.*  
416 *Past Discuss.* **2017**, 1–10 (2017).
- 417 [26] Gabridge, T. The Last Mile: Liaison Roles in Curating Science and Engi-  
418 neering Research Data. *Research Library Issues* 15–21 (2009).
- 419 [27] Bracke, M. S. Emerging Data Curation Roles for Librarians: A Case Study  
420 of Agricultural Data. *Journal of Agricultural & Food Information* **12**, 65–74  
421 (2011).
- 422 [28] Pinfield, S., Cox, A. M. & Smith, J. Research Data Management and  
423 Libraries: Relationships, Activities, Drivers and Influences. *PLoS ONE* **9**,  
424 e114734 (2014).

- 425 [29] Weller, T. & Monroe-Gulick, A. Differences in the Data Practices, Chal-  
 426 lenges, and Future Needs of Graduate Students and Faculty Members.  
 427 *Journal of eScience Librarianship* **4** (2015).
- 428 [30] Horsburgh, J. S. *et al.* HydroShare: Sharing Diverse Environmental Data  
 429 Types and Models as Social Objects with Application to the Hydrology  
 430 Domain. *JAWRA Journal of the American Water Resources Association*  
 431 **52**, 873–889 (2016).
- 432 [31] Essawy, B. T. *et al.* Integrating scientific cyberinfrastructures to improve  
 433 reproducibility in computational hydrology: Example for HydroShare and  
 434 GeoTrust. *Environmental Modelling & Software* **105**, 217–229 (2018).
- 435 [32] Gillman, M. A., Lamoureux, S. F., Lafrenière, M. J. Calibration of a  
 436 modified temperature-light intensity logger for quantifying water electrical  
 437 conductivity. *Water Resources Research* **53**, 8120–8126 (2017).
- 438 [33] Horsburgh, J., Leonardo, M., Abdallah, A. & Rosenberg, D. Measuring  
 439 water use, conservation, and differences by gender using an inexpensive,  
 440 high frequency metering system. *Environmental Modelling and Software*  
 441 **96**, 83–94 (2017).
- 442 [34] Neuwirth, C. System dynamics simulations for data-intensive applications.  
 443 *Environmental Modelling and Software* **96**, 140–145 (2017).
- 444 [35] Quinn, J. *et al.* Detecting spatial patterns of rivermouth processes us-  
 445 ing a geostatistical framework for near-real-time analysis. *Environmental*  
 446 *Modelling and Software* **97**, 72–85 (2017).
- 447 [36] Buscombe, D. Shallow water benthic imaging and substrate characteriza-  
 448 tion using recreational-grade sidescan-sonar. *Environmental Modelling and*  
 449 *Software* **89**, 1–18 (2017).
- 450 [37] Yu, C.-W., Liu, F. & Hodges, B. Consistent initial conditions for the Saint-  
 451 Venant equations in river network modeling. *Hydrology and Earth System*  
 452 *Sciences* **21**, 4959–4972 (2017).
- 453 [38] Di Matteo, M., Dandy, G. & Maier, H. A multi-stakeholder portfolio  
 454 optimization framework applied to stormwater best management practice  
 455 (BMP) selection. *Environmental Modelling and Software* **97**, 16–31 (2017).
- 456 [39] Engdahl, N., Benson, D. & Bolster, D. Lagrangian simulation of mixing  
 457 and reactions in complex geochemical systems. *Water Resources Research*  
 458 **53**, 3513–3522 (2017).
- 459 [40] Güntner, A. *et al.* Landscape-scale water balance monitoring with an iGrav  
 460 superconducting gravimeter in a field enclosure. *Hydrology and Earth Sys-*  
 461 *tem Sciences* **21**, 3167–3182 (2017).

- 462 [41] Sattar, A., Jasak, H. & Skuric, V. Three dimensional modeling of free  
463 surface flow and sediment transport with bed deformation using automatic  
464 mesh motion. *Environmental Modelling and Software* **97**, 303–317 (2017).
- 465 [42] Rosenberg, D. E. & Watkins, D. W. New Policy to Specify Availability  
466 of Data, Models, and Code. *Journal of Water Resources Planning and*  
467 *Management* **144**, 01618001 (2018).
- 468 [43] Collberg, C. *et al.* Measuring reproducibility in computer systems research.  
469 *Department of Computer Science, University of Arizona, Tech. Rep* (2014).
- 470 [44] Thaler, R. H. & Sunstein, C. R. *Nudge: Improving decisions about health,*  
471 *wealth, and happiness.* Nudge: Improving decisions about health, wealth,  
472 and happiness (Yale University Press, New Haven, CT, US, 2008).
- 473 [45] Sison, C. P. & Glaz, J. Simultaneous Confidence Intervals and Sample  
474 Size Determination for Multinomial Proportions. *Journal of the American*  
475 *Statistical Association* **90**, 366–369 (1995).
- 476 [46] May, W. L. & Johnson, W. D. Constructing two-sided simultaneous con-  
477 fidence intervals for multinomial proportions for small counts in a large  
478 number of cells. *Journal of Statistical Software* **5**, 1–24 (2000).
- 479 [47] Stagge, J. H., Abdallah, A. M. & Rosenberg, D.E. jstagge/reproduc\_hyd:  
480 Source code accompanying “A novel replicability survey tool to mea-  
481 sure and promote reproducibility in hydrology”. (2018). doi: 10.5281/zen-  
482 odo.1467417

# Paper Metadata

- Q1. Assessor's name

Q2. Journal name

Q3. Article DOI

Q4. Full paper citation

# Availability

Q5. How accessible to users?

Some or all applicable

Not specified where

Not applicable

Q6. Where available?

All online

Third party

Author

In article

Q7. What is present?

Required

Optional

Input Data

Code / Software

License

Metadata

Identifiers

Directions

Hardware / software requirements

File format

Q8. Comments on availability?

Q9. Could you and readers use the available artificats to generate results?

Yes

Not sure

Not familiar with resources

No

Q10. Continue to replicate results?

Yes

No

# Replicability

Q11. Do the outputs verify published results (in text, figures, and tables)?

Yes (explain in Q12)

No (explain in Q13 and Q14)

Q12. Explain what made this replication work? Other comments?

Q13. Why did the replication fail?

Hardware / software errors

Did not generate results

Results differed

Unclear directions

Other

Q14. Explain why the replication failed?

# Time to Complete

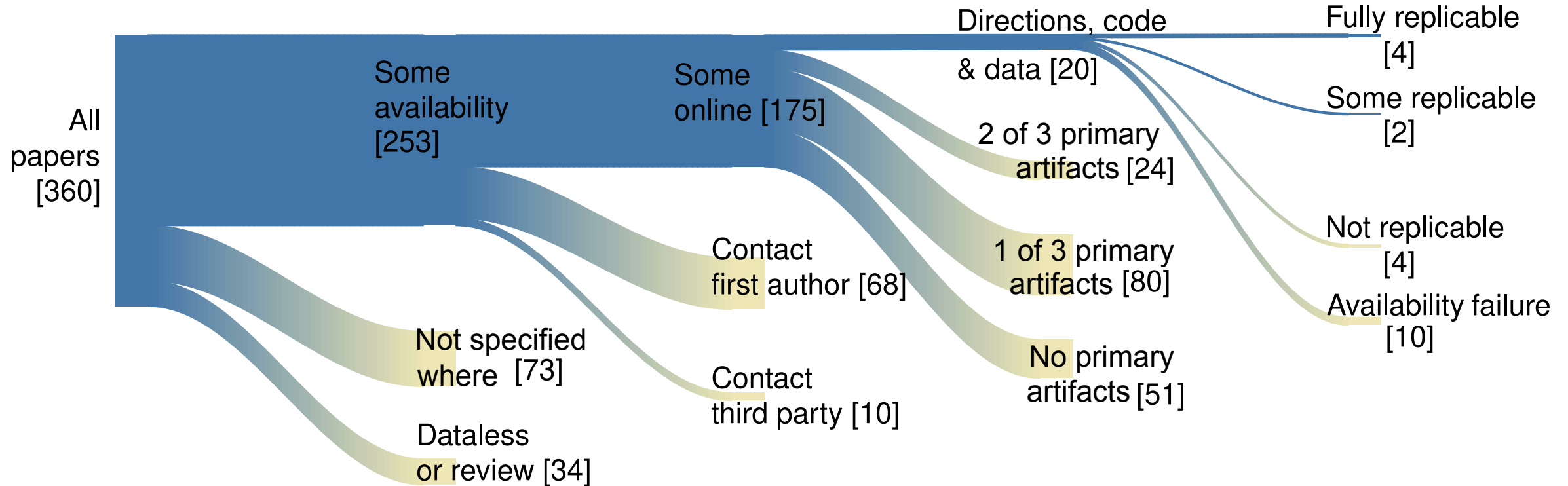
Q15: How many minutes did the survey take?

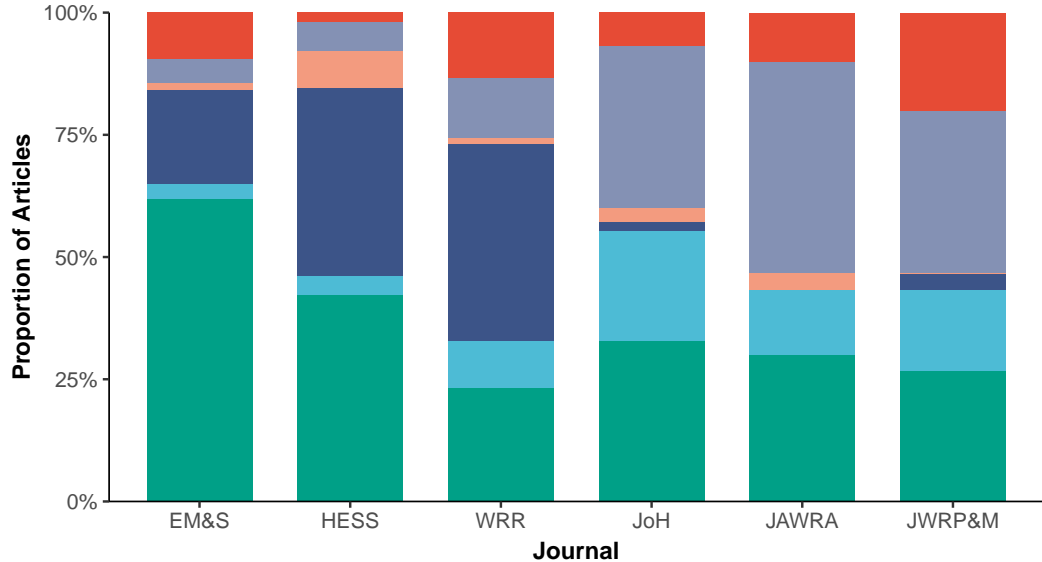
**Q5. How accessible to users?**

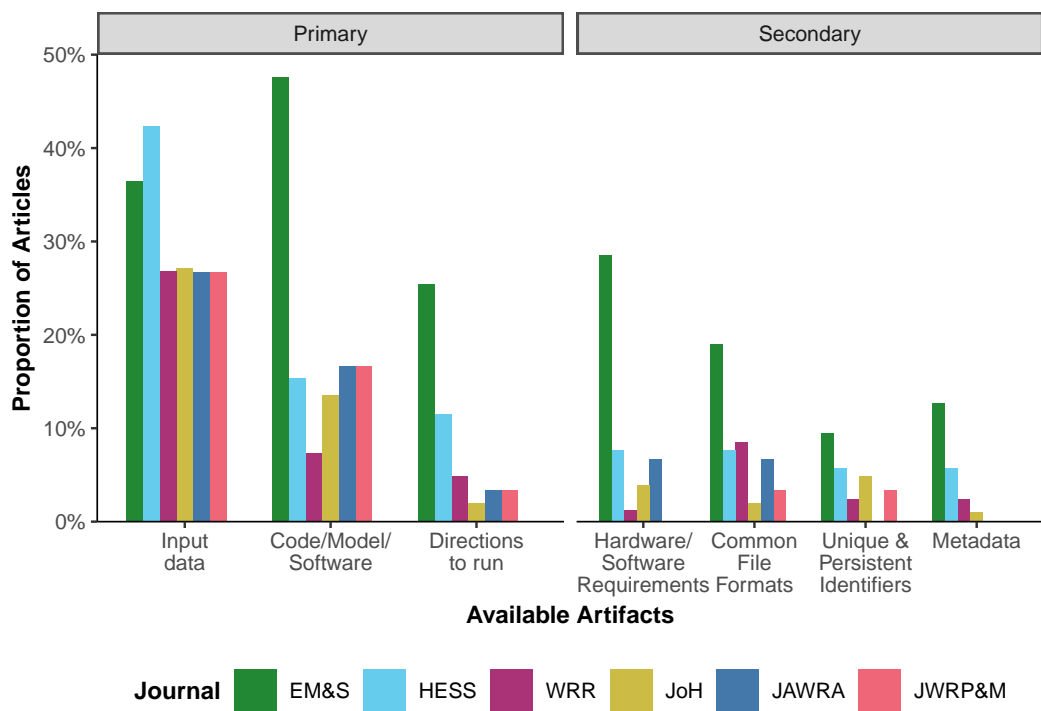
**Q6. Where available?**

**Q7. What is present?**

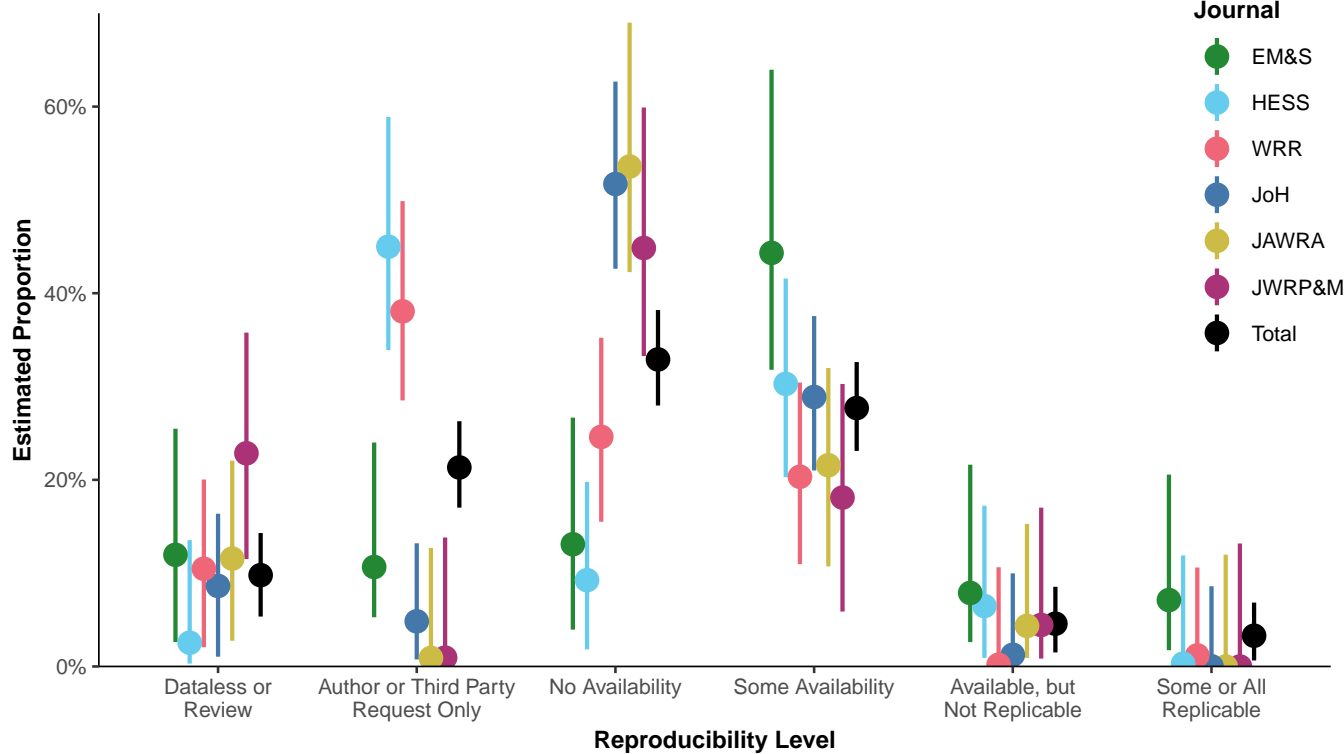
**Q11. Do outputs verify results?**











# Survey Ending Question

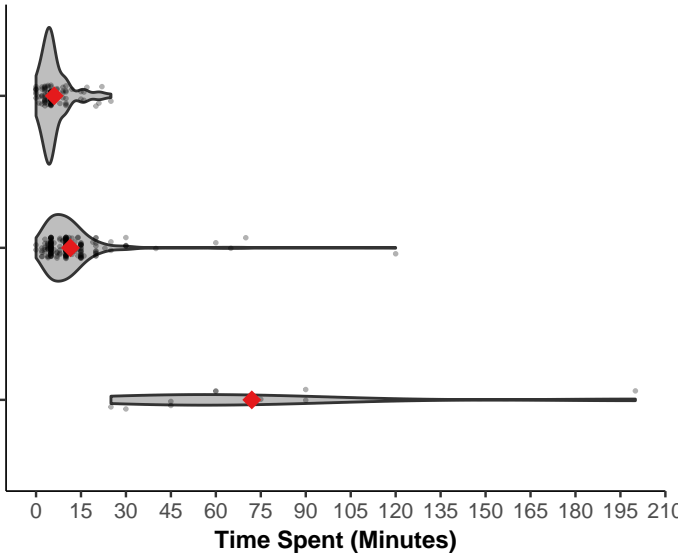
No Artifacts  
Available  
(Q5)

Availability of  
Artifacts  
(Q9)

Replicability of  
Results  
(Q13)

0 15 30 45 60 75 90 105 120 135 150 165 180 195 210

Time Spent (Minutes)



	EM&S		HESS		WRR		JoH		JAWRA		JWRP&M	
	2017	Sample	2017	Sample	2017	Sample	2017	Sample	2017	Sample	2017	Sample
<b>Keyword</b>	49	48	9	9	23	23	24	24	7	7	8	8
<b>Non-key word</b>	181	15	319	43	511	59	645	79	102	23	111	22
<b>Total</b>	230	63	328	52	534	82	669	103	109	30	119	30