

1 **Discoverability of open data is critical to Earth system science**

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13 This paper is a non-peer reviewed preprint submitted to EarthArXiv.

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## 18 **Abstract**

19 There is consensus throughout the Earth system science research community that “open data” is  
20 of critical importance. However, discoverability and accessibility are often overlooked, raising  
21 the question of how useful archived, but not easily discoverable data are. As part of evaluating  
22 databases suitable for our own research data archival, we conducted a data discovery exercise  
23 (aggregators and repositories) with search parameters to evaluate (i) feasibility of discovering  
24 data, and (ii) number of relevant results found (defined by exact matches to our search). We  
25 found that search parameters need more options (and perhaps community driven development of  
26 thematic keyword search options), repositories affiliated with funding agencies/large scale  
27 research datasets were more likely to reveal relevant results, broad aggregators with poor  
28 metadata requirements yield the most irrelevant results, and current practices may drive smaller  
29 datasets to disappear thereby promoting a non-inclusive open data world that is not truly open for  
30 all. There are encouraging signs, however, whereby commitment to open data practices is  
31 leading to datasets becoming public--with due credits--prior to analysis and associated  
32 publication. Ideally, making data meet FAIR principles means more than depositing data as a  
33 journal or funding requirement: community buy-in and consensus is needed across the spectrum  
34 of data generators, hosts, and users to agree on how to best achieve the ideal of data being  
35 findable, accessible, interoperable, and reusable.

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37 **Keywords:** Data, sharing, repository, best practices, metadata

## 38 **Introduction**

39 Multi-disciplinary Earth-system science is based on generating, synthesizing, and evaluating  
40 knowledge and data within and across environmental systems and scales. This requires

41 collaboration across disciplinary and geographical borders (Reid et al., 2009) with accessible and  
42 shareable datasets serving a critical role in the process. However, accessibility of datasets largely  
43 depends on data curation, metadata assembly, data deposition practices, and data  
44 discovery (Popkin, 2019). With various repositories available in the data repository realm [e.g.,  
45 funding agency mandated data deposition requirements like the United States Department of  
46 Energy's Environmental Systems Science Data Infrastructure for a Virtual Ecosystem (ESS-  
47 DIVE) for Earth and environmental science research data (Varadharajan et al., 2018), or general  
48 data repositories like Zenodo (Nielsen, 2017)], data generators and users have a wide variety of  
49 choices for depositing and accessing data, respectively, but no clear guidance on successful data  
50 discovery and reusability practices. Furthermore, agency-affiliated repositories like ESS-DIVE  
51 have mandatory meta-data standards and consistent data structures that are usually lacking in  
52 general repositories. These standards are critical for making data Findable, Accessible,  
53 Interoperable, and Reusable (FAIR) (Castelvecchi, 2018; McQuilton et al., 2016; Nature  
54 Editorial, 2019; Wilkinson et al., 2016) which is the current gold-standard for ensuring that  
55 public data can be used.

56

57 As researchers generating data, we conducted a data search exercise to inform ourselves about  
58 the data deposition options available to use for an ongoing research project and evaluate which  
59 repositories yield data that is both discoverable and useful. Furthermore, our goal was also to  
60 evaluate how small datasets not affiliated with programmatic requirements of funding agencies  
61 fare in terms of data deposition and discoverability. For example, ESS-DIVE is a great  
62 repository, but is necessarily limited to DOE funded datasets. To be beneficial to users, data  
63 contributors like us need to ascribe to all aspects of FAIR data practices in addition to data  
64 deposition. The first requirement is that data are findable. This is arguably the most critical

65 element of the FAIR principles; if data cannot be found, they cannot be reused. We therefore  
66 evaluate the degree to which openly available data can be found (sometimes referred to as  
67 ‘discovered’), assess whether discovered data aligns with user needs, and discuss the need for  
68 community accepted and enacted data-stewardship practices that support small but valuable  
69 datasets that may not be affiliated with large research programs to promote open data discovery.

70

71 Data search is inherently subjective to the knowledge domain of a scientist and therefore non-  
72 uniform amongst users. A general approach is to begin with a thorough literature review for  
73 published results and the available data. Keyword-defined searches on platforms such as Google  
74 Scholar usually yield literature results that eventually guide readers to databases/repositories.  
75 Lack of standardized keyword-based searches means searches are user-defined and highly  
76 variable, which may differentially shape search output, and therefore influence downstream  
77 access and reuse of relevant data. A second approach is to search known databases including  
78 archives/repositories/aggregators with relevant keywords. However, the data hosting platforms  
79 may be constrained by available keywords that one can apply. These non-uniform search options  
80 therefore present a possibility that searches may yield biased results yielding long-term studies  
81 with significant data presence and/or datasets affiliated with programmatic requirements that are  
82 more commonly known than small datasets/studies. This has the potential to skew data reuse,  
83 with large numbers of short-term and/or non-programmatic datasets being effectively lost. In  
84 turn, if individual investigators perceive that their small/non-programmatic datasets will not be  
85 reusable, there is little incentive for investing resources into making these data FAIR, leading to  
86 a negative feedback loop.

87

88 Federal agency mandated data centers like the National Aeronautics and Space Administration  
89 (NASA) supported Distributed Active Archive Centers, the National Oceanic and Atmospheric  
90 Administration (NOAA) supported National Center for Environmental Information, US DOE  
91 supported ESS-DIVE, and the US Geological Survey's (USGS) Earth Resources Observation  
92 Systems (EROS) Data Center (Downs et al., 2015) have defined data archiving and search  
93 guidelines whereas stand-alone data-repositories like Zenodo and figshare (Scientific Data,  
94 2019) are less structured in terms of data types and metadata requirements with limited directions  
95 on data searches. While all archives provide data-hosting and archival of diverse data-types (e.g.,  
96 geophysical and ecological data) (National Research Council, 1995; Scientific Data, 2019) , the  
97 extent to which open data is discoverable and usable is unclear.

98 We conducted a search exercise across multiple repositories and aggregator databases to evaluate  
99 variation in data search outcomes. We observed that data is increasingly being made open ahead  
100 of publications, which we believe is a positive sign. However, we discovered that public data is  
101 not necessarily findable or usable (i.e., relevant) data. The disparity in data search capabilities  
102 and results in terms of keyword usage and the relevance of the search results to our intended  
103 search suggests that data generators, data users, and repositories must collaboratively promote  
104 FAIR data practices in a coordinated way. Researchers must make efforts to upload data with  
105 standardized metadata and themselves evaluate how FAIR their data are. Users have the  
106 responsibility of ascribing to FAIR data principles and to help make improvements by engaging  
107 with repositories (e.g., help standardize keyword searches and provide feedback about their  
108 experiences). Finally, repositories need to consciously choose to operate and implement FAIR  
109 data principles by requiring standardized metadata, providing streamlined search parameters, and  
110 mandating data formats and file structures. As a community, we need to agree that making data  
111 open is not just a publication requirement but meets all principles of FAIR data practices.

112

113 **Data discovery exercise**

114 The goal of the data discovery exercise was to (i) evaluate the degree to which data are findable,  
115 and (ii) identify current challenges in making data FAIR. For the purpose of this opinion piece,  
116 we conducted searches across two data aggregators and five repositories for data  
117 released/published between 2014-2018. The search focused on soil carbon data for coastal  
118 ecosystems in Florida. This was chosen as an example since coastal areas provide significant C  
119 storage but are vulnerable to C loss in a warming climate (Osland et al., 2018) and therefore need  
120 to be accurately mapped and monitored (Holmquist et al., 2018).

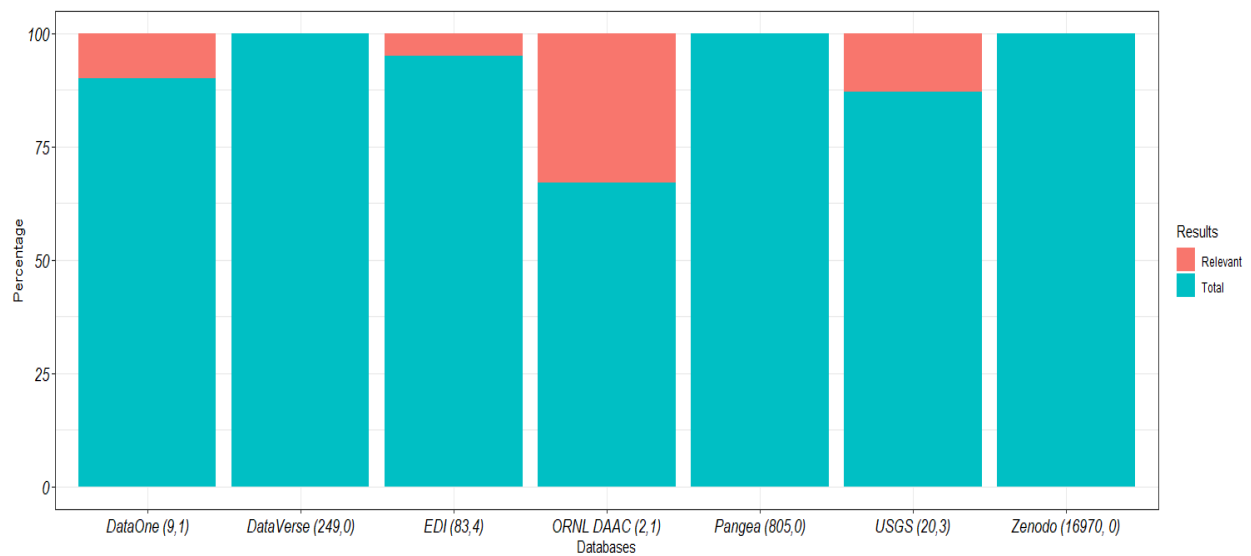
121 The repositories/aggregators ranged from programmatic [Environmental Data Initiative (EDI),  
122 USGS, Oakridge National Laboratory Distributed Active-Archive Center (ORNL DAAC)] to  
123 broad-range (Zenodo and Dataverse) archives, as well as data aggregator catalogs (DataONE and  
124 Pangaea). These repositories represent data submission options ranging from mandated data  
125 format and metadata requirements to no specific data format and metadata structure. We used  
126 keywords ‘coastal soil carbon’, attempted to place filters to restrict the search to sites in coastal  
127 Florida, United States, and limit the search to reflect data released between 2014-2018. We did  
128 not check for data type deposited (e.g., file type, data analysis codes, and raw vs processed data).  
129 A parallel search using the same keywords was conducted in Google Scholar for published  
130 research articles to serve as a benchmark of studies in the public domain. The results revealed  
131 that: (i) researchers are making their data open by hosting them on repositories, as opposed to  
132 only providing them in supplementary material of publications, (ii) irrespective of research scale,  
133 openly available data are not necessarily discoverable or usable, (ii) multiple search function  
134 options can enhance the data discovery process, (iii) improved metadata standard requirements

135 by repositories stands to improve discoverability and accessibility of data, and (iv) extra attention  
136 is needed to improve the metadata for small/short-term datasets so that these data have a high  
137 likelihood of being discovered.

138

### 139 **Results and Discussion**

140 Data were considered relevant if all individual keywords were found in the search results; exact  
141 multi-word phrases were not required. For each search, the data titles and information describing  
142 the data were assessed for matches. We consider the brief summary of data to be one type of  
143 metadata useful for discovering relevant datasets. For those searches that did not allow keyword  
144 searches (e.g., only phrases could be used), the results had to be manually curated to narrow  
145 down which datasets were considered relevant. The lack of filters made it challenging to search  
146 for relevant data, and the results showed considerable variation across repositories/aggregators in  
147 data discovery and usability (Figure 1) with some yielding results irrelevant to keywords  
148 provided.



149

150 **Figure 1.** Data discovery and usability percentage with identifiers “coastal soil carbon, Florida,  
151 2014-2018”. Parentheses in X axis labels (a,b) indicate total number of results (a) and relevant  
152 number of results (b) for each database/repository. Results were considered relevant when all  
153 individual search words were identifiable in the title of the dataset, or the metadata brief  
154 provided by users, and/or a combination of the two.

155  
156 Detailed search results including database information, date accessed, data availability, and  
157 affiliation with programmatic requirements are provided in Supporting Table S1. The EDI,  
158 DataOne, ORNL DAAC, and USGS databases revealed five, ten, thirteen, and thirty-three  
159 percent of relevant data, respectively. The broad-range data repositories including DataVerse and  
160 Zenodo, and the aggregator catalog Pangea did not reveal any relevant datasets. Surprisingly,  
161 these repositories revealed the maximum number of results, most likely a result of a fluid data  
162 deposition requirement with no strict metadata and file requirements. For example, a dataset for  
163 soil nematode counts appeared in the search result since the word “soil” matched with our search  
164 parameters. It is likely that the search algorithm in such cases yields a match to any of the  
165 words/search phrases, without strict search parameters to tell the algorithm to "show a result only  
166 when all these words are present". Our results suggest a disconnect between data that is openly  
167 available and data that is discoverable/usable. That is, having data publicly available on a  
168 platform is only a piece of the solution, and needs to be backed up by efficient data discovery  
169 and usability.

170  
171 Making data public without making it discoverable results in data that are not very usable,  
172 thereby defeating the purpose of public data. A major hurdle in making data discoverable that we  
173 identified in our exercise was lack of user-friendly data-filtering choices. DataONE fared the best



174 in allowing all relevant filters (environment, location, measurement, year) to be used. The USGS  
175 data search included a map to focus on location of interest but the filters were broad and did not  
176 help narrow the search to soil environments and year-wise searches could not be made due to the  
177 absence of a 'time' filter. ORNL DAAC and EDI allowed phrases to be inserted in the search bar  
178 but lacked the use of filters. Pangaea only allowed broad level filters to be placed and was not  
179 able to narrow down the search to relevant locations or study systems (e.g., soils). It proved  
180 difficult to narrow down location, year, and environment in Dataverse and Zenodo. We also  
181 discovered that DataONE filtering by a location field or by a geographic map gave different  
182 results. This suggests active community engagement is needed to address the current data  
183 archival environment.

184 Data deposition practices influence the metadata labels that get recorded and ultimately matched  
185 in database searches. All repositories/archives required a manual curation step of individually  
186 checking the search results to discover relevant datasets (for example, results had to be manually  
187 evaluated to ensure data searches were only from 2014-2018), primarily because the metadata  
188 provided were inadequate for accurate identification. Therefore, metadata labels need to be  
189 standardized and informative to show up accurately in search results. As evidenced from the data  
190 search exercise, data discovery can be improved regardless of search capabilities within a given  
191 archive, etc. if the metadata is standardized and informative. For example, as pointed out earlier,  
192 if the location was not explicitly indicated in the metadata, search capabilities in DataOne would  
193 exclude the dataset. However, location-relevant records could be found by zooming in on the  
194 map feature. This ambiguity in search parameters as well as allowing broad level filters to be  
195 used as search criteria is perhaps aimed at providing choices and flexibility to data generators but  
196 ultimately results in non-uniform data deposition practices that is less likely to be helpful to data  
197 users

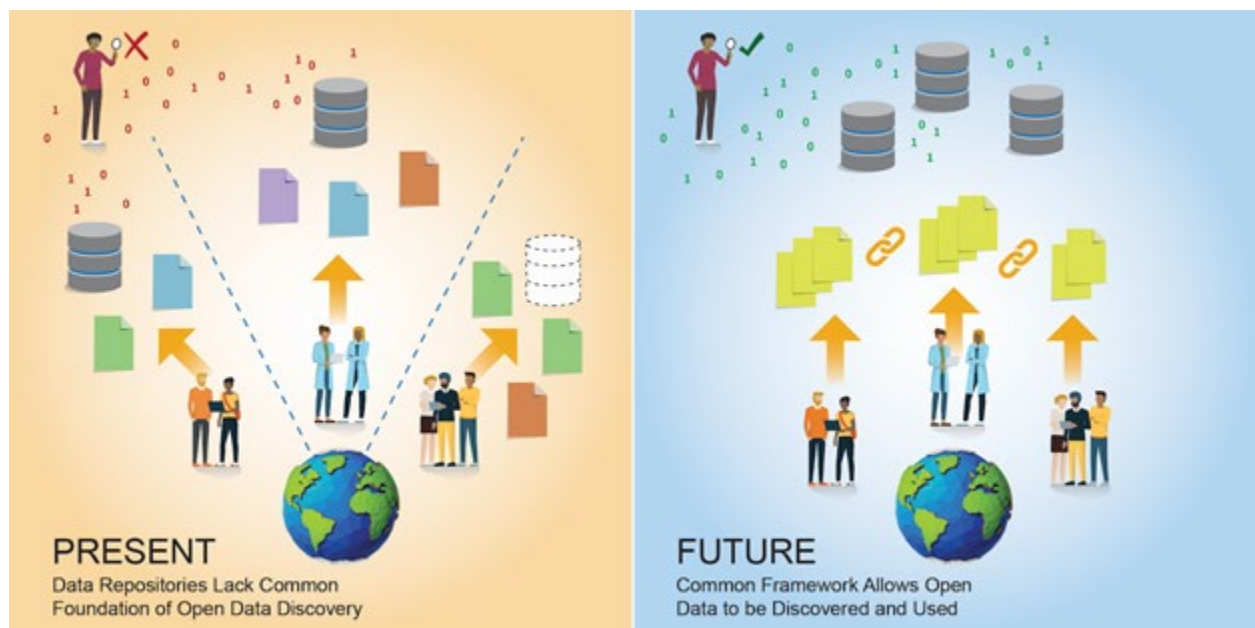
198

199 To understand how Google Scholar search compared to databases, we examined the first 100  
200 results using the same keywords as used in the database search. The search revealed twelve  
201 published articles (Supporting Table S2) as of June 14<sup>th</sup> 2019, of which only one study (Wilson et  
202 al., 2018) had the data discoverable in DataONE. As pointed out in a recent commentary  
203 (Castelvecchi, 2018) , Google Scholar searches are not yet linked to their new Dataset Search  
204 initiative, with our results showing an opportunity where linking the two will increase the  
205 chances of discovering a dataset. All but two of the Google Scholar search results had their data  
206 available in the journal article itself, with one study (Hinson et al., 2017) affiliated with a global  
207 dataset and the other (Wilson et al., 2018) affiliated with Everglades Long Term Ecological  
208 Research site data. This shows a benefit of FAIR data practices: while Google Scholar will only  
209 return results that have been published, unpublished but publicly available data can be used with  
210 proper credits for research purposes. This open data culture provides credit to data generators,  
211 recognizes the power of data-driven research, and adds value to the scientific pursuit.

212  
213 The Google Scholar and database search results highlighted a disparity between agency-affiliated  
214 large projects that can host data in programmatic databases. In contrast, small-scale studies often  
215 cannot deposit data in the programmatic databases. This disparity results in data from individual  
216 (smaller-scale) research efforts likely to be deposited in repositories with few requirements  
217 associated with making data FAIR (e.g., weak metadata standards). This leads to the associated  
218 data being difficult to discover and use. It may be that strict deposition policies in the  
219 programmatic databases may also deter scientists from uploading data, which prompts  
220 researchers to instead deposit data in a repository with few requirements due to limited time and  
221 resources.

222 The commitment to enable FAIR data is essential and promising but needs to be coupled to  
223 standardized search capabilities; data may be FAIR in principle, but if not paired with  
224 appropriate search capabilities, they cannot be discovered or used (Figure 2). For example, broad  
225 search options with inadequate filters reduces data discovery. Simultaneously, poor metadata  
226 prevents dissemination, sharing, and reuse of associated data. This problem can be overcome by  
227 agreeing as a community on integrated thematic descriptors irrespective of database affiliations,  
228 thereby improving data identifier practices and integrating the currently fragmented data  
229 ecosystem. To this end, set vocabularies such as the Global Change Master Directory keywords  
230 adopted by NASA (GCMD Keyword Governance and Community Guide, 2016), the Climate  
231 and Forecast metadata conventions adopted by the National Center for Atmospheric Research for  
232 atmospheric data (Gregory, 2003), and the Biological, Ancillary, Disturbance and Metadata  
233 protocol for the Ameriflux network (Law et al., 2008) exist but the conventions have not been  
234 translated to informative metadata identifiers across data archives.

235



236

237 **Figure 2.** Ascribing to FAIR principles is key to enabling discovery of openly available data,  
238 and the architecture of repositories is a key element; data can be FAIR in principle, but not in  
239 practice due to limitations in repository search capabilities. The ‘Present’ panel indicates a  
240 fragmented scenario where data discovery is limited due to disparate data sharing and limited  
241 search options where the left and middle groups upload their large research data with  
242 programmatic requirements (grey cylinder) in different formats while the group on the right  
243 share their small research data to stand-alone repositories. The user can discover only one of the  
244 datasets. The ‘Future’ panel indicates a common framework of open data discovery, where  
245 irrespective of research scale or repository requirements, openly available data is rich in  
246 metadata information, is searchable with dense search parameters, and therefore yields optimum  
247 user defined search results.

248  
249 It is important to consider that the cost of making data FAIR is much higher per bit of data for  
250 small stand-alone efforts, relative to large research campaigns that generate significant quantities  
251 of data. The preferential use of repositories with few or no standards is not because researchers  
252 are lazy. Instead, most research labs do not have the resources to make every dataset FAIR  
253 through strict adherence to data and metadata standards. A major need is finding ways to  
254 decrease the cost and effort of making ‘small data’ FAIR. This will require coordinated efforts to  
255 standardize metadata deposition requirements and search filters, as well as new  
256 cyberinfrastructure that streamlines the logistics of making data FAIR. Furthermore, there is a  
257 need for cultural change whereby the scientific norm is data deposition with the intent of  
258 maximizing discoverability, rather than merely ‘checking a box’ to fulfill a publication  
259 requirement. Greater multi-way communication is also needed among data generators,

260 repositories, and data users to develop streamlined practices that are useful for the involved  
261 parties and that lead to discoverable and usable data.

262

## 263 **Conclusion**

264 Inclusive accessibility, discovery, and reusability of data is necessary to ensure data stewardship  
265 and the progression of science. The data archival guidance provided by FAIR data principles is  
266 aimed at strengthening data stewardship by involving multiple stakeholders. Strides have been  
267 made by the Earth-science community to begin implementing FAIR data practices over the past  
268 two years (Stall et al., 2019), with researchers, funding agencies, the research community, and  
269 publishers adhering to FAIR data practices. However, our results show that there is a significant  
270 need to develop mechanisms/tools that minimize the effort and cost needed to make any dataset  
271 FAIR. If data are hard to find and when found are hard to use, they are public but are not truly  
272 open or FAIR. It is important to invest the time and effort to ensure that data are discoverable  
273 through standardized keywords and identifiers. Without such tools, the vast number of ‘small’  
274 datasets generated by individual research groups will effectively die despite being public. The  
275 responsibility of researchers go beyond simply depositing data in a public repository. A much  
276 broader commitment from the community is needed to promote uniform data deposition, search,  
277 recoverability, and usability, irrespective of data deposition in programmatic databases or  
278 standalone repositories.

279

## 280 **Acknowledgement**

281 This work is part of the PREMIS Initiative at Pacific Northwest National Laboratory (PNNL). It  
282 was conducted under the Laboratory Directed Research and Development Program at PNNL, a

283 multi-program national laboratory operated by Battelle for the U.S. Department of Energy under  
284 Contract DE-AC05-76RL01830. Data can be accessed in the two supporting information tables.

285

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