

rOpenSci tools for accessing research literature for text mining

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Abstract

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9 Background. xxxx.

10 Methods. xxxx.

11 Results. xxxx.

Discussion. xxxx.

12 Introduction

13 There's more than 100 million articles published (source: Crossref API), representing an enormous
14 amount of knowledge. In addition to simply reading these articles, they contain a vast trove of
15 information of interest to researchers for machine aided questions.

16 For example, many researchers are interested in statistical outcomes of articles: questions about P-values,
17 about effect sizes, and more. With regard to effect sizes, these are of particular interest, as they are
18 often combined in meta-analyses to draw broad conclusions about a particular question.

19 Text-mining is the broad term associated with pulling information out of articles. Given the importance
20 of text-mining, good text-mining tools are needed to make it easier for researchers. In particular, the R
21 programming language is used widely throughout many academic fields and thus tools in R for text
22 mining are of particular importance.

23 Here, we present an overview of text-mining tools in the R programming language. We do not cover
24 analysis tools per se, but rather those tools for searching for, acquiring, and “mashing up” text.

25 Digital articles: technical aspects

26 Those articles that are digital can be split into two groups: easily machine readable and non-machine
27 readable.

28 The machine readable articles are those in XML, JSON, or plain text format. The former two, XML
29 and JSON, are ideal for the machine readable types because they are structured data, whereas plain
30 text has no structure - it's simply a set of characters with line breaks and spaces in between.

31 Of the non-machine readable kind, there's PDFs. These can be broken out into two groups: text based
32 PDFs and scanned PDFs. The former are converted from digital versions of various kinds (MS Word,
33 OpenOffice, markdown, etc.), while the latter are PDFs created by scanning in print articles for which
34 there is no digital version.

35 Digital articles: the access landscape

36 Acces to full-text is the holy grail in text-mining. Some use cases can get by with article metadata
37 (authors, title, etc.), some with abstracts, but many use cases need full-text.

38 The landscape of access to full-text is extremely hetergeous, with the majority of variation along the
39 publisher axis. The major hurdle is paywalls. The majority of articles are published by the big three
40 publishers - Wiley, Springer, Elsevier - and the majority of their articles are behind paywalls.

41 A promising sign is an increasing number of open access publishers, yet these represent a very small
42 portion of the total articles (XXXXX) (ref.).

43 With respect to paywalled articles, access varies by institution, depending on what each institution
44 decided to pay for. In addition, some users may not realize access varies with IP address so that access
45 from campus vs. from home (if not on a VPN) will drastically differ.

46 We can not end this section without mentioning SciHub. This is a last resort option for many probably,
47 providing access to full text of articles that are normally paywalled. No tools in this manuscript provide
48 access to SciHub.

49 **The discovery problem**

50 XXX

51

52 XXX

53 **Data sources**

54 There is increasing open access scientific literature content available online. However, only a small
55 proportion of scientific journals provide access to their full content; whereas, most publishers provide
56 open access to their metadata only (most often through Crossref; Table 1). The following is a synopsis
57 of the major data sources and associated R tools.

58 Table 1. Sources of scientific literature, their content type provided via web services, whether rOpenSci
59 has an R packages for the service, and where to find the API documentation.

Data Provider	Content Type	rOpenSci Package	Documentation
Crossref	Metadata only	rcrossref/crminer	1
DataCite	Metadata only	rdatacite	2
Biodiversity Heritage Library	Full content/Metadata	rbhl	3
Public Library of Science (PLOS)	Full text/altmetrics	rplos	4
Scopus (Elsevier)	Full content/Metadata	fulltext	5
arXiv	Full content/Metadata	aRxiv	6
Biomed Central (via Springer)	Full content/Metadata	fulltext	7
bioRxiv	Full content/Metadata	fulltext	8
PMC/Pubmed (via Entrez)	Full content/Metadata	rentrez	9
Europe PMC	Full content/Metadata	europepmc	10
Microsoft Academic Search	Metadata	fulltext/microdemic	11
Directory of Open Access Journals	Metadata	jaod	12
JSTOR Data for Research	Full content	jstor	13
ORCID	Metadata	rorcid	14
Wikimedia's Citoid	Citations	rcitoid	15
Open Citation Corpus	Citations	citecorp	16

¹<https://api.crossref.org>

²<https://support.datacite.org/docs/api>

³<http://bit.ly/KYQ1Rd>

⁴<http://api.plos.org/solr>

⁵<http://bit.ly/J9S616>

⁶<https://arxiv.org/help/api/index>

⁷<https://dev.springer.com/>

⁸<http://www.biorxiv.org/>

⁹<https://www.ncbi.nlm.nih.gov/books/NBK25500>

¹⁰<https://azure.microsoft.com/en-us/services/cognitive-services>

¹¹<https://dev.labs.cognitive.microsoft.com/docs/services/56332331778daf02acc0a50b/operations/565d9001ca73072048922d97>

¹²<https://doaj.org/api/v1/docs>

¹³<https://www.jstor.org/dfr/>

¹⁴<https://pub.orcid.org/>

¹⁵https://en.wikipedia.org/api/rest_v1/#/Citation/getCitation

¹⁶<http://opencitations.net/>

60 *Crossref/Datacite*

61 Crossref is a non-profit that creates (or “mints”) Digital Object Identifiers (DOIs). In addition, they
62 maintain metadata associated with each DOI. The metadata ranges from simple (including author, title,
63 dates, DOI, type, publisher) to including number of citations to the article, as well as references in the
64 article, and even abstracts. At the time of writing they hold 100 million DOIs.

65 One can search by DOI or search citation data to get citations. In addition, Crossref has a text-mining
66 opt-in program for publishers. The result of this is that some publishers provide URLs for full text
67 content of their articles. The majority of these links are pay-walled, while some are open access. Using
68 any of the various tools for working with Crossref data, you can filter your search to get only articles
69 with full text links, and further to get only articles with full text links that are open access.

70 The main interfaces for Crossref in R are [rcrossref](#) and [crminer](#). Similar interfaces are available in Ruby
71 ([serrano](#)) and Python ([habanero](#)).

72 Datacite is similar to Crossref, but focuses on datasets instead of articles. The main interface for
73 Datacite in R is [rdatacite](#).

74 *Biodiversity Heritage Library*

75 The Biodiversity Heritage Library (BHL) houses scans of biodiversity books, and provides web interfaces
76 and APIs to query and fetch those data. They also provide text of the scanned pages. The main R
77 interace to BHL is through [rbhl](#).

78 *Public Library of Science*

79 The Public Library of Science (PLOS) is one of the largest open access only publishers. They as of this
80 writing have published 2.1 million articles. One of the strong advantages of PLOS is that they provide
81 an API to their Solr instance, which is a very flexible way to search their articles. The main R interace
82 to PLOS is through [rplos](#).

83 *Elsevier/Scopus*

84 Elsevier is one of the largest publishers. Most of their articles are not open access. However, they have a
85 number of advantages if you have access to their articles: they are one of the few publishers to provide

86 machine readable XML (many publishers do have XML versions of articles, but do not provide it); they
87 are one of the few (two) publishers part of Crossref's text and data mining program. The packages
88 [fulltext](#) and [crminer](#) can be used to access Elsevier articles through Crossref's TDM program. There's
89 an interface to Scopus article search within [fulltext](#).

90 *arXiv/bioRxiv*

91 arXiv and bioRxiv are preprint publishers, the former in existence for many years, and the latter new
92 on the scene. You can access articles from these publishers through [fulltext](#). arXiv does provide a web
93 API that we hook into; bioRxiv does not, but we can get you articles nonetheless.

94 *Pubmed/PMC/Europe PMC*

95 Pubmed/PMC is a corpus/website of NIH funded research in the United States; while Europe PMC is
96 an equivalent for the European Union. You can access articles from Pubmed/PMC through [fulltext](#),
97 and for Europe PMC through [europepmc](#).

98 *Microsoft Academic Research*

99 Microsoft Academic Research (MAR) is a search engine for research articles. You can use their GUI
100 web interface to search, and they provide APIs for programmatic access. The R interface for MAR is
101 [microdemic](#); and [fulltext](#) hooks into [microdemic](#) as well for article search and abstract retrieval.

102 *Directory of Open Access Journals*

103 XXXXX

104 *JSTOR*

105 XXXXX

106 *ORCID*

107 XXXXX

108 *Citoid/Open Citation Corpus*

109 XXX

110 **fulltext: a swiss army knife for text mining in R**

111 **fulltext** is a general purpose R package for the data part of text-mining: search for articles, get links to
112 articles, get article abstracts, and fetch full text of articles. The **fulltext** package is always adding
113 additional data sources as time allows (See Table 1). Starting from searching for articles, the outputs of
114 search can be fed into a function to get links to those articles, or to get abstracts for those articles, or
115 to fetch their full text.

116 The following is a breakdown of the major distinct functional parts of **fulltext**.

117 *Search*

118 **ft_search()** provides search access to nine different data sources (PLOS, BMC, Crossref, Entrez, arXiv,
119 bioRxiv, Europe PMC, Scopus, Microsoft Academic), creating a mostly unified interface to all data
120 sources. The parts of each data source that are common are mostly factored into the parameters of the
121 **ft_search()** function, and we also allow the user to pass on data source specific options as needed.

122 XXXXX

123 *Links*

124 **ft_links()** provides two pathways to get links (URLs) for articles, with a choice of four different data
125 sources (PLOS, BMC, Crossref, Entrez). First, you can use **ft_search()**, then pass the output of that
126 function to **ft_links()**. Second, you can pass DOIs directly to **ft_links()**. Both end up at the same
127 point, links for each article, if they could be found for the user selected data source.

128 The biggest caveat with **ft_links()** is that we can't guarantee that the links will work. Link rot is one
129 way in which the links may not work: link rot is when the URL does not point to the original content
130 anymore, or fails altogether. Additionally, with Crossref, publishers can deposit URLs for articles, but
131 they make change the URLs at some later date but not update the URLs with Crossref.

132 *Abstracts*

133 **ft_abstract()** provides access to article abstracts from four different data sources (PLOS, Scopus,
134 Microsoft Academic Research, Crossref). The only way to use the function is to pass article identifiers,
135 which are for the most DOIs.

136 The advantage of abstracts over full text is that abstracts can often be retrieved even for paywalled
137 articles. That is, you can have much broader coverage of the articles you're targeting relative to full
138 text.

139 If you are after abstracts, and you are already getting or already have full text, and if the articles are in
140 XML format, then you can use [pubchunks](#) to extract out the abstracts.

141 *Fetch full text*

142 `ft_get()` fetches full text of articles from many different data sources. From the DOIs that are passed
143 in to the function, we detect the publisher, and there are specific plugins for certain publishers:

- 144 • aaas
- 145 • aip
- 146 • amersocclinoncol
- 147 • amersocmicrobiol
- 148 • arxiv
- 149 • biorxiv
- 150 • bmc
- 151 • copernicus
- 152 • crossref
- 153 • elife
- 154 • elsevier
- 155 • entrez
- 156 • frontiersin
- 157 • iee
- 158 • informa

- 159 • instinvestfil
- 160 • jama
- 161 • microbiology
- 162 • peerj
- 163 • pensoft
- 164 • plos
- 165 • pnas
- 166 • royalsocchem
- 167 • sciencedirect
- 168 • scientificsocieties
- 169 • wiley

170 If there's no built-in plugin for the publisher already, we use the FTDOI API (<https://ftdoi.org>) to try
171 to get the link for the full text of the article. If the FTDOI API doesn't bear fruit, we search Crossref
172 for a link to the full text. If Crossref doesn't have any full text links, we give up.

173 Since users can go through a lot of article requests, we cache successfully downloaded articles, and keep
174 that knowledge consistent across R sessions; all subsequent requests for the same article just use the
175 cached version. Additionally, all errors in `ft_get()` are collected in a tidy data.frame in the output of
176 the function to help the user quickly determine what went wrong.

177 **How to text mine from R: Three case studies**

178 *Case study 1: Citation mining*

179 In this example, xxxx

180 *Load libraries*

```
library("rcrossref")
library("rplos")
library("rorcid")
library("rcitoid")
library("citecorp")
```

181 *rcrossref*

182 Using *rcrossref* for Crossref data:

```
x <- cr_works(query="NSF")
head(x$data)
#> # A tibble: 6 x 32
#>   alternative.id container.title created deposited published.print doi
#>   <chr>           <chr>           <chr>  <chr>      <chr>           <chr>
#> 1 S106352031630~ Applied and Co~ 2016-0~ 2019-02-- 2018-03      10.1~
#> 2 <NA>           Biogeosciences~ 2017-0~ 2017-07-- <NA>         10.5~
#> 3 <NA>           Global Biogeoc~ 2018-0~ 2019-01-- 2018-10      10.1~
#> 4 <NA>           IEEE Communica~ 2016-1~ 2017-12-- 2017         10.1~
#> 5 S002178241400~ Journal de Mat~ 2014-0~ 2018-10-- 2014-10      10.1~
#> 6 123           Light: Science~ 2019-0~ 2019-01-- 2019-12      10.1~
#> # ... with 26 more variables: indexed <chr>, issn <chr>, issue <chr>,
#> #   issued <chr>, member <chr>, page <chr>, prefix <chr>, publisher <chr>,
#> #   reference.count <chr>, score <chr>, ...
```

183 *Case study 2: Abstract mining*

184 Sometimes you just need abstracts for your research question. The benefit of only needing abstracts,
 185 and not need full text, is that there's many more articles that will have abstracts available than have
 186 their full text available.

187 As an example, let's say you xxxx

```
library("fulltext")
```

188 *xxxxx*

189 Using fulltext:

```
res <- ft_search("ecology", from = "crossref",  
  crossrefopts = list(filter = c(has_abstract = TRUE)))  
ids <- res$crossref$data$doi  
out <- ft_abstract(x = ids, from = "crossref")  
abstracts <- vapply(out$crossref, "[", "", "abstract")
```

190 Using `quanteda`, read the abstracts into a corpus

```
library("quanteda")  
corp <- corpus(abstracts)  
docvars(corp) <- ids
```

191 Get a summary of the abstracts

```
summary(corp)  
  
#> Corpus consisting of 10 documents:  
#>  
#>   Text Types Tokens Sentences          V1  
#> text1    143    262         10 10.2458/v22i1.21112  
#> text2    117    244          6 10.2458/v17i1.21696  
#> text3     75    118          4 10.2458/v25i1.23119  
#> text4      5      8          1 10.2458/v1i1.21154  
#> text5    105    171          7 10.1155/2011/868426  
#> text6    112    181          6 10.1155/2012/273413  
#> text7    117    240          8 10.5194/we-13-91-2013  
#> text8    140    245          9 10.5194/we-13-95-2013  
#> text9    107    202          7 10.1155/2014/198707  
#> text10   118    224          6 10.5402/2011/897578
```

```
#>
#> Source: /Users/sckott/github/ropensci/textmine/use-cases/* on x86_64 by sckott
#> Created: Fri Apr 5 11:36:04 2019
#> Notes:
```

192 Use the `kwic()` function to see a word in context across the abstracts

```
kwic(corp, pattern = "ecology")
#>
#> [text1, 33] knowledge production within critical political / ecology /
#> [text1, 50] in scientific articles on dryland / ecology /
#> [text1, 204] to equilibrium models in range / ecology /
#> [text1, 246] communal areas.Keywords: Critical political / ecology /
#> [text1, 255] , scientific models, rangeland / ecology /
#> [text2, 5] < jats:p> Political / ecology /
#> [text2, 23] manifestations of political economy and / ecology /
#> [text2, 45] I try to extend political / ecology /
#> [text2, 149] , in dialogue with political / ecology /
#> [text2, 177] people and resources that political / ecology /
#> [text2, 229] indigeneity scholars.Key words: political / ecology /
#> [text3, 71] an analysis from a political / ecology /
#> [text3, 114] system, supermarkets, political / ecology /
#> [text6, 134] was observed when allopatry and / ecology /
#> [text7, 167] ecosystem should be considered for / ecology /
#> [text7, 185] the" four-color issue of / ecology /
#> [text7, 201] step toward advancing knowledge in / ecology /
#> [text9, 195] or for theoretical studies integrating / ecology /
#>
#> . This article is a
#> , and investigates the functions
#> , and the fence-line photographs
#> , fence-line photography, scientific
```

```
#> , Southern Africa</
#> has expanded in multiple new
#> in the" problem"
#> to engage with ethnic studies
#> approaches to better understand the
#> focuses on cannot be adequately
#> , coloniality, Maidu,
#> standpoint allows a different interpretation
#> </ jats:p>
#> act together, leading to
#> "? Here, I
#> ", and propose that
#> and conservation biology. In
#> and biogeography.</
```

193 *Case study 3: Full text mining*

194 In this example, xxxx

```
library("fulltext")
# library("crminer")
```

195 *Search for articles*

196 Search for the term *ecology* in PLOS journals.

```
(res1 <- ft_search(query = 'ecology', from = 'plos'))
#> Query:
#> [ecology]
#> Found:
#> [PLoS: 47337; BMC: 0; Crossref: 0; Entrez: 0; arxiv: 0; biorxiv: 0; Europe PMC: 0; Scopus:
#> Returned:
#> [PLoS: 10; BMC: 0; Crossref: 0; Entrez: 0; arxiv: 0; biorxiv: 0; Europe PMC: 0; Scopus: 0; .
```

197 Each publisher/search-engine has a slot with metadata and data

```
res1$plos
#> Query: [ecology]
#> Records found, returned: [47337, 10]
#> License: [CC-BY]
#>
#> id
#> 1 10.1371/journal.pone.0001248
#> 2 10.1371/journal.pone.0059813
#> 3 10.1371/journal.pone.0155019
#> 4 10.1371/journal.pone.0080763
#> 5 10.1371/journal.pone.0208370
#> 6 10.1371/journal.pone.0150648
#> 7 10.1371/journal.pcbi.1003594
#> 8 10.1371/journal.pone.0102437
#> 9 10.1371/journal.pone.0175014
#> 10 10.1371/journal.pone.0166559
```

198 *Get full text*

199 Using the results from `ft_search()` we can grab full text of some articles

```
(out <- ft_get(res1))
#> <fulltext text>
#> [Docs] 10
#> [Source] ext - /Users/sckott/Library/Caches/R/fulltext
#> [IDs] 10.1371/journal.pone.0001248 10.1371/journal.pone.0059813
#> 10.1371/journal.pone.0155019 10.1371/journal.pone.0080763
#> 10.1371/journal.pone.0208370 10.1371/journal.pone.0150648
#> 10.1371/journal.pcbi.1003594 10.1371/journal.pone.0102437
#> 10.1371/journal.pone.0175014 10.1371/journal.pone.0166559 ...
```

200 *Extract text from pdfs*

201 Ideally for text mining you have access to XML or other text based formats. However, sometimes you
202 only have access to PDFs. In this case you want to extract text from PDFs. `fulltext` can help with
203 that.

204 You can extract from any pdf from a file path, like:

```
path <- system.file("examples", "example1.pdf", package = "fulltext")
ft_extract(path)
#> <document>/Library/Frameworks/R.framework/Versions/3.5/Resources/library/fulltext/examples/ex
#> Title: Suffering and mental health among older people living in nursing homes---a mixed-met
#> Producer: pdfTeX-1.40.10
#> Creation date: 2015-07-17
```

205 *Extract text chunks*

206 Requires the `pubchunks` library. Here, we'll search for some PLOS articles, then get their full text, then
207 extract various parts of each article with `pub_chunks()`.

```
library("pubchunks")
res <- ft_search(query = "ecology", from = "plos", limit = 3)
x <- ft_get(res)
x %>% ft_collect() %>% pub_chunks(c("doi", "history")) %>% pub_tabularize()
#> $plos
#> $plos$`10.1371/journal.pone.0001248`
#>
#> doi history.received history.accepted
#> 1 10.1371/journal.pone.0001248 2007-07-02 2007-11-06
#> .publisher
#> 1 plos
#>
#> $plos$`10.1371/journal.pone.0059813`
#>
#> doi history.received history.accepted
#> 1 10.1371/journal.pone.0059813 2012-09-16 2013-02-19
#> .publisher
#> 1 plos
```



```

#>
#> $plos$`10.1371/journal.pone.0155019`
#>                                doi history.received history.accepted
#> 1 10.1371/journal.pone.0155019      2015-09-22      2016-04-22
#> .publisher
#> 1      plos

```

208 Future directions

209 XXXX

210 Acknowledgments

211 XXXX

212 Data Accessibility

213 All scripts and data used in this paper can be found in the permanent data archive Zenodo under
 214 the digital object identifier (DOI). This DOI corresponds to a snapshot of the GitHub repository at
 215 <https://github.com/ropensci/textmine>. Software can be found at <https://github.com/ropensci/xxx>,
 216 xxxx, all under MIT licenses.

217 References