

# rOpenSci tools for textmining open source science literature

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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 likely 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.

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 to do.

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

## 23 Digital articles: technical aspects

24 Of digital articles some of which are available digitally, and some of which are not. Those that are  
25 digital can be split into two groups: easily machine readable and non-machine readable.

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

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

## 33 Digital articles: the access landscape

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

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

39 A promising sign is that there's an increasing number of open access publishers. xxxx.

40 **The discovery problem (maybe remove section)**

41 XXX

42 **Data sources**

43 There is increasing open source scientific literature content available online. However, only a small  
44 proportion of scientific journals provide access to their full content; whereas, most publishers provide  
45 open access to their metadata only (most often through Crossref; Table 1).

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

Data Provider	Content Type	rOpenSci Pkg?	API Documentation
Crossref	Metadata only	rcrossref	<a href="#">1</a>
DataCite	Metadata only	rdatacite	<a href="#">2</a>
Biodiversity Heritage Library	Full content/Metadata	rbhl	<a href="#">3</a>
Public Library of Science (PLOS)	Full text/altmetrics	rplos	<a href="#">4</a>
Scopus (Elsevier)	Full content/Metadata	fulltext	<a href="#">5</a>
arXiv	Full content/Metadata	aRxiv	<a href="#">6</a>
Biomed Central (via Springer)	Full content/Metadata	fulltext	<a href="#">7</a>
bioRxiv	Full content/Metadata	fulltext	<a href="#">8</a>
PMC/Pubmed (via Entrez)	Full content/Metadata	rentrez	<a href="#">9</a>
Microsoft Academic Search	Metadata	fulltext/microdemic	<a href="#">10</a>

<sup>1</sup><http://api.crossref.org>

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

<sup>3</sup><http://bit.ly/KYQ1Rd>

<sup>4</sup><http://api.plos.org/solr>

<sup>5</sup><http://bit.ly/J9S616>

<sup>6</sup><https://arxiv.org/help/api/index>

<sup>7</sup><https://dev.springer.com/>

<sup>8</sup><http://www.biorxiv.org/>

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

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

48 The following is a synopsis of the major data sources and associated R tools.

#### 49 *Crossref*

50 Crossref is a non-profit that creates (or “mints”) Digital Object Identifiers (DOIs). In addition, they  
51 maintain metadata associated with each DOI. The metadata ranges from simple (including author, title,  
52 dates, DOI, type, publisher) to including number of citations to the article, as well as references in the  
53 article, and even abstracts.

54 Crossref does have a text-mining opt-in program for publishers. The result of this is that some publishers  
55 deposit URLs for full text content of their articles. The majority of these links are pay-walled, while  
56 some are open access. Using any of the various tools for working with Crossref data, you can filter your  
57 search to get only articles with full text links, and further to get only articles with full text links that  
58 are open access.

59 The main interface for Crossref in R is [rcrossref](#). Parallel interfaces are available in Ruby ([serrano](#)) and  
60 Python ([habanero](#)).

#### 61 *Pubmed*

62 Pubmed is a corpus/website of NIH funded research ...

### 63 **How to text mine from R: Three case studies**

#### 64 *Case study 1*

#### 65 *Case study 2*

#### 66 *Case study 3*

### 67 **Conclusions and future directions**

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### 69 **Acknowledgments**

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## 71 **Data Accessibility**

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

## 76 **References**