BIP4COVID19: Releasing impact metrics data for articles relevant to COVID-19

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Abstract

Since the beginning of the 2019-20 coronavirus pandemic, a large number of relevant articles has been published or become available in preprint servers. These articles, along with earlier related literature, compose a valuable knowledge base affecting contemporary research studies or, even, government actions to limit the spread of the disease and treatment decisions taken by physicians. However, the number of such articles is increasing at an intense rate making the exploration of the relevant literature and the identification of useful knowledge in it challenging. In this work, we describe BIP4COVID19, an open dataset compiled to facilitate the coronavirus-related literature exploration, by providing various indicators of scientific impact for the relevant articles. Finally, we provide a publicly accessible web interface on top of our data, allowing the exploration of the publications based on the computed indicators.

Background & Summary

COVID-19 is an infectious disease, caused by the coronavirus SARS-CoV-2, which may result, for some cases, in progressing viral pneumonia and multi-organ failure. After its first outbreak in Hubei, a province in China, it subsequently spread to other Chinese provinces and many other countries. On March 11th 2020, the World Health Organisation (WHO) declared the 2019–20 coronavirus outbreak a pandemic. Until now more than 1,000,000 cases have been recorded in more than 200 countries, counting more than 80,000 fatalities.

At the time of writing, an extensive amount of coronavirus related articles have been published since the virus’ outbreak (indicatively, our collected data contain about 3,550 articles published in 2020). Moreover, the number of weekly publications is increasingly growing, numbering up to 985 publications in the week from March 30\textsuperscript{th} to April 5\textsuperscript{th} (based on the statistics presented in [2]). Taking additionally into account previous literature on coronaviruses...
and related diseases, it is evident that there is a vast literature on the subject. Consequently, effective exploration of this literature by researchers is a difficult task. However, the literature exploration can be facilitated through the use of publication impact measures. A variety of such measures have been proposed in the fields of bibliometrics and scientometrics [7, 8]. Some of them rely on network analysis of the underlying citation graph, formed by publications and the references between them. Other approaches utilise measures commonly known as “altmetrics”, which analyse data from social media and/or usage analytics in online platforms. Both approaches have their benefits and shortcomings, each better capturing a different aspect of an article’s impact. Thus, using a broad overview of different measures can help better uncover a comprehensive view of each article’s impact.

In this context, the objective of this work is to produce BIP4COVID19, an openly available dataset, which contains a variety of different impact measures calculated for COVID-19-related literature. Two citation-based impact measures (PageRank [5] and RAM [6]) were chosen to be calculated, as well as an altmetric indicator (tweet count). The selected measures were chosen so as to cover different impact aspects of the articles. Furthermore, to select a representative set of publications, we rely on two open datasets of COVID-19-related articles: the CORD-19 [3] and the LitCovid [2] datasets. BIP4COVID19 data are updated on a regular basis and are openly available on zenodo.

Methods

BIP4COVID19 is a regularly updated dataset (the current plan involves weekly updates). Data production and update is based on the semi-automatic workflow presented in Figure 1. In the following paragraphs the major processes involved are elaborated.

Article Data Collection and Cleaning. The list of COVID-19-related articles is created based on two main data sources: the CORD-19\(^1\) Open Research Dataset [3], provided by the Allen Institute for AI, and the LitCovid\(^2\) collection [2] provided by the NLM/NCBI BioNLP Research Group. CORD-19 offers a full-text corpus of more than 40,000 articles on coronavirus and COVID-19, collected based on articles that contain a set of COVID-19 related keywords from PMC, biorXiv, and medRxiv and the further addition of a set of publications on the novel coronavirus, maintained by the WHO. LitCovid, is a curated dataset which currently contains more than 2,000 papers on the novel coronavirus.

The contents of the previous datasets are integrated and cleaned. During this process, the eSummary tool\(^3\) from NCBI’s eTool suite is utilised to collect extra metadata for each publication using the corresponding PubMed or PubMed Central identifiers (pmid and pmcid, respectively), where available.

\(^{1}\)https://pages.semanticscholar.org/coronavirus-research


\(^{3}\)https://www.ncbi.nlm.nih.gov/books/NBK25500/
The collected metadata are semi-automatically processed for the removal of duplicate records. The resulting dataset contains one entry for each distinct article. Each entry contains the pmid, the DOI, the pmcid, and the publication year of the corresponding article. This information is the minimum required for the calculation of the selected impact metrics.

Calculation of Citation-based Measures. A prerequisite for calculating the citation-based impact measures of the collected articles, is the compilation of their citation network, i.e., the network which has articles as nodes and citations between them as directed edges. The citations of the articles, required to construct this network, are gathered using NCBI’s eLink tool. The tool returns, per query pmid/pmcid, other pmids/pmcids that cite, or are cited by, the query article. Two citation-based impact measures are calculated on the constructed network: the PageRank [5] and the RAM scores [6]. These two measures were selected based on the results of a recent experimental study [7], which found them to perform best in capturing the overall and the current impact of an article, respectively. Both measures are calculated by performing citation analysis. PageRank evaluates the overall impact of articles by differentiating their citations based on the importance of the articles making them. However, it is biased against recent articles that haven’t accumulated many citations yet, but may be the current focus of the research community. RAM alleviates this issue by considering recent citations as being more important.
Calculation of Tweet-based measure. In addition to the citation-based measures the number of tweet posts that mention each article is calculated as a measure of social media attention. The COVID-19-TweetIDs dataset [4] is used for the collection of COVID-19-relevant tweets. Currently (v1.2), this dataset contains a collection of 72,403, 796 tweet IDs, each of them published by one of 9 predetermined Twitter accounts (e.g., @WHO) in the period between January 21st and April 3rd and containing at least one out of 68 predefined coronavirus-related keywords (e.g., “Coronavirus”, “covid19”, etc).

At the time of writing, a subset of this dataset containing tweets posted from March 1st to 12th (19,285,031 unique tweet IDs) has been integrated in BIP4COVID19. This subset was extracted from the COVID-19-TweetIDs v1.1 release, which contains totally 63,616,072 tweet IDs collected from January 21st to March 12th. The corresponding fully hydrated Tweet objects were collected using the Twitter API. The result was a collection of 17,957,947 tweet objects (106,94 GB). The difference between the number of IDs and hydrated objects is due to the fact that 1,327,084 tweets have been deleted in the meantime (6.81%) and are, therefore, impossible to retrieve.

To find those tweets which are related to the articles in our database, we rely on the URLs of the articles in doi.org, PubMed, and PMC. These URLs are easily produced based on the corresponding identifiers. In addition, when possible, the corresponding page in the publisher’s website is also retrieved based on the doi.org redirection. After the collection of the URLs of all articles, the number of appearances of the URLs related to each one are produced. However, since the Twitter API returns either shortened or not fully expanded URLs, the fully expanded URLs are collected using the unshrtn5 library.

Code availability

Article data cleaning and citation network building are based on using the NCBI’s eTools suite. The URL extraction script is available on GitHub6, under a GNU/GPL license. The same holds for the tweet counting process7. For the calculation of popularity (RAM [6]) and influence (PageRank [5]) scores, the open PaperRanking [7] library8 was used. Finally, retrieving Twitter objects based on given tweet IDs was done using the twarc Python library9.

Data Records

The BIP4COVID19 dataset, produced by the previously described workflow, is openly available on Zenodo [1], under the Creative Commons Attribution 4.0 International license. At the time of publication, the third release of this

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4https://github.com/echen102/COVID-19-TweetIDs
5https://github.com/docnow/unshrtn
6https://github.com/diweis/paper-url-extractor
7https://github.com/danae-pla-karidi/social_kg
8https://github.com/diweis/PaperRanking
9https://github.com/DocNow/twarc
The BIP4COVID19 dataset is comprised of three files in tab separated (TSV) format. The files contain identical data, but each one is sorted based on a different impact measure for the articles (popularity, influence, number of tweets). The data attributes included in each file are summarised in Table 1.

Finally, it should be noted that the current version of BIP4COVID19 inte-

10Current plan involves weekly data updates.
Attribute Interpretation

**PubMed identifier** Unique identifier of the article in PubMed, as collected from the source data files. Articles missing this identifier are indicated with the value “N/A”.

**DOI** The Digital Object Identifier of the article, as collected from PubMed. Articles missing a DOI are indicated with the value “N/A”.

**PCM identifier** Unique identifier of the article in Pubmed Central (PMC), as collected from the source data files. Articles missing an identifier in PMC have the value “N/A”.

**Popularity score** The value of the corresponding citation-based measure (RAM [6]) for the respective article.

**Influence score** The value of the corresponding citation-based measure (PageRank [5]) for the respective article.

**Social media attention** The calculated tweet count for the article corresponding to the record.

Table 1: Data attributes inside the TSV files.


**Technical Validation**

To ensure the proper integration and cleaning of the CORD-19 and LitCovid datasets, we rely on NCBI’s eTool suite. In particular, we collect pmids and pmcids from the aforementioned datasets and use them as queries to gather each article’s metadata. After cleaning the article title (e.g., removing special characters) we automatically identify duplicates by comparing each record’s content and eliminate them. Finally, manual inspection is performed to produce the correct metadata for a limited number of duplicates that remain (e.g., duplicate records containing the title of the same pub in two different languages).

Further, to guarantee the correctness of the compiled citation graph we apply the following procedures. After gathering all citing - cited records using NCBI’s eTools, those that include identifiers not found in the source data are removed. Since many citing - cited pairs may have been found both with pmids and pmcids, the resulting data may still contain duplicate records. These records are removed, after mapping all pmids/pmcids to custom identifiers, with pmid-pmcid pairs that refer to the same article being mapped to the same identifier. The final resulting citation graph is based on these mapped identifiers. As an extra cleaning step, any links in the graph that denote citations to articles published at a later time than the citing article are removed.

To guarantee a complete representation of the links to each article required
for our tweet-based analysis, we aimed to integrate as many different links as possible. Hence, apart from URLs in doi.org, Pubmed, and PMC, we also aimed to include URLs to each article from its publisher’s website, since these are the most commonly used, where possible. To avoid incorrect tweet counts due to duplicate tweets, we used a simple deduplication process after the Tweet object retrieval. Moreover, the use of the unshrttn library to expand the short URLs from tweet texts ensures that our measurements derive from all available URL instances of each publication record, no matter how they were shortened by users or Twitter.

Usage Notes

Our data are available in files following TSV format, allowing easy import to various database management systems and can be conveniently opened and edited by any text editor, or spreadsheet software. We plan to update the data on a weekly basis, incorporating any additions and changes from our source datasets, as well as to expand the tweet counts based on all available data for 2020. Additionally, we plan to incorporate any further sources on coronavirus related literature that may be released and which will index the literature based on pmids and/or pmcids.

The contents of the BIP4COVID19 dataset may be used to support multiple interesting applications. For instance, the calculated scores for each impact measure could be used to rank articles based on their impact to help researchers prioritise their reading. In fact, we used our data to implement such a demo service described in more detail in the next section. Additionally the rank scores may be useful for monitoring the research output impact of particular sub-topics or as features in machine learning applications that apply data mining on publications related to coronavirus.

Finally, the following limitations should be taken into consideration with respect to the data: while we take effort to include as many articles as possible, there are many cases where our source data do not provide any pmids or pmcids. As a consequence, no data for these articles are collected and they are not included in the BIP4COVID19 dataset. Furthermore, with respect to the calculated impact scores, it should be noted that the citation analysis we conduct is applied on the citation graph formed by citations from and to collected publications only, i.e., our analyses are not based on pubmed’s complete citation graph, but on a COVID-19-related subgraph. Consequently, the relative scores of publications may differ from those calculated on the complete PubMed data. Finally, regarding the tweet-based analysis, since our data come from the COVID-19-TweetIDs dataset which only tracks tweets from a predefined set of accounts and which is based on a particular set of COVID-19-related keywords, the measured number of tweets is only based on a subset of the complete COVID-19-related tweets.
Web Interface

A Web interface has been developed on top of the BIP4COVID19 data. Its aim is to facilitate the exploration of COVID-19-related literature. The option to order articles according to different impact measures is provided. This is expected to be useful since users can better prioritise their reading based on their needs. For example, a user that wants to delve into the background knowledge about a particular COVID-19-related sub-topic could select to order the articles based on their influence. On the other hand, another user that needs to get an overview of the latest trends in the same topic, could select to order the articles based on their popularity.

The information shown to users, per publication, includes its title, venue, year, and the source dataset where it was found. Moreover, each result is accompanied by color coded icons that denote the publication’s importance based on each calculated impact measure. In this way, the users can easily get a quick insight about the different impact aspects of each article. The tooltips of these icons provide the exact scores for each measure. Each publication title functions as a link to the corresponding article’s entry in its publisher’s website, or to Pubmed. Finally, a page containing interesting statistics is provided. This page contains various charts that visualise, for example, the number of articles per year, or the number of articles that have substantial impact based on each of the provided impact measures, per year.

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Author contributions

T.V., I.K., and S.C designed the database and the Web user interface. T.V., I.K, S.C, and D.P.K. implemented the data collection, cleaning, and processing workflows. T.V, I.K, and D.P.K. wrote the manuscript and T.D. provided guidance throughout the process. All authors read and approved the final version of the manuscript.

Competing interests

The authors declare no competing interests.

11https://bip.covid19.athenarc.gr/
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