## **Archive Scholar Reference Dataset**

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### **Abstract**

As part of its scholarly data efforts, the Internet Archive releases a citation graph dataset (ASREF) derived from scholarly publications and additional data sources. It is composed of data gathered by the fatcat cataloging project<sup>1</sup> and related web-scale crawls targeting primary and secondary scholarly outputs. In addition, relations are worked out between scholarly publications, web pages and their archived copies, books from the Open Library project as well as Wikipedia articles. This first version of the graph consists of over X nodes and over Y edges. We release this dataset under a Z open license under the collection as an archive item<sup>2</sup>. All code used in the derivation process is releases under an MIT license<sup>3</sup>.

Index terms— Citation Graph, Web Archiving

### 1 Introduction

The Internet Archive releases a first version of a citation graph dataset derived from a raw corpus of about 2.5B references gathered from metadata and from data obtained by PDF extraction tools such as GROBID[5]. The goal of this report is to describe briefly the current contents and the derivation of the Archive Scholar Citations Dataset (ASC). We expect this dataset to be iterated upon, with changes both in content and processing.

Modern citation indexes can be traced back to the early computing age, when projects like the Science Citation Index (1955)[3] were first devised, living on in existing commercial knowledge bases today. Open alternatives were started such as the Open Citations Corpus (OCC) in 2010 - the first version of which contained 6,325,178 individual

references[6]. Other notable sources from that time include CiteSeerX[9] and CitEc[1]. The last decade has seen an increase of more openly available reference dataset and citation projects, like Microsoft Academic[8] and Initiative for Open Citations[2][7]. In 2021, according to [4] over 1B citations are publicly available, marking a tipping point for open citations.

#### 2 Related Work

There are a few large scale citation dataset available today. COCI, the "OpenCitations Index of Crossref open DOI-to-DOI citations" was first released 2018-07-29. As of its most recent release on 2021-07-29, it contains 1,094,394,688 citations across 65,835,422 bibliographic resources.

The WikiCite<sup>4</sup> project, "a Wikimedia initiative to develop open citations and linked bibliographic data to serve free knowledge" continuously adds citations to its data base and as of 2021-06-28 tracks 253,719,394 citations across 39,994,937 publications<sup>5</sup>.

Microsoft Academic Graph<sup>6</sup> is comprised of a number of entities<sup>7</sup> with PaperReferences being one relation among many others. As of 2021-06-07 the PaperReferences relation contains 1,832,226,781 edges across YYY bibliographic entities.

Numerous other projects have been or are concerned with various aspects of citation discovery and curation, among them Semantic Scholar, CiteSeerX or Aminer.

As mentioned in [4], the number of openly available citations is not expected to shrink in the future.

<sup>1</sup>https://fatcat.wiki

<sup>2</sup>https://archive.org/details/fatcat-asref-todo

<sup>3</sup>https://gitlab.com/internetarchive/cgraph

<sup>4</sup>https://meta.wikimedia.org/wiki/WikiCite

<sup>&</sup>lt;sup>5</sup>http://wikicite.org/statistics.html

 $<sup>^6</sup>A$  recent copy has been preserved at https://archive.org/details/mag-2021-06-07

<sup>&</sup>lt;sup>7</sup>https://docs.microsoft.com/en-us/ academic-services/graph/reference-data-schema

### **3** Citation Dataset

We release the first version of the ASREF dataset in an format used internally for storage and display (and which we call *biblioref*). The format contains source and target fatcat release and work identifiers, as well as few attributes from the metadata (such as year or release stage) as well as information about the match provenance (like match status or reason). For ease of use, we include DOI as well, if available.

The dataset currently contains X unique bibliographic entities and Y citations.

TODO: how matches are established and a short note on overlap with COCI DOI.

# 4 System Design

The constraints for the systems design are informed by the volume and the variety of the data. In total, the raw inputs amount to a few TB of textual content, mostly newline delimited JSON. More importantly, while the number of data fields is low, certain schemas are very partial with hundreds of different combinations of available field values found in the raw reference data. This is most likely caused by aggregators passing on reference data coming from hundreds of sources, each of which not necessarily agreeing on a common granularity for citation data and from artifacts of machine learning based structured data extraction tools.

Each combination of fields may require a slightly different processing path. For example, references with an Arxiv identifier can be processed differently from references with only a title. Over 50% of the raw reference data comes from a set of eight field manifestations, as listed in Table 1.

Overall, a map-reduce style approach is followed, which allows for some uniformity in the overall processing. We extract (key, document) tuples (as TSV) from the raw JSON data and sort by key. Then we group documents with the same key into groups and apply a function on each group in order to generate our target schema (currently named biblioref, or bref for short) or perform addition operations (such as deduplication).

The key derivation can be exact (like an identifier like DOI, PMID, etc) or based on a normalization procedure, like a slugified title string. For identifier based matches we can generate the target biblioref schema directly. For fuzzy matching candidates, we pass possible match pairs through a verification procedure, which is implemented for release entity schema pairs. The current verification procedure is a domain dependent rule based verification, able to identify different versions of a publication, preprint-published pairs or or other kind of similar documents by calculating similarity metrics across title and authors. The fuzzy matching

Fields	Share
$CN \cdot RN \cdot P \cdot T \cdot U \cdot V \cdot Y$	14%
DOI	14%
$CN \cdot CRN \cdot IS \cdot P \cdot T \cdot U \cdot V \cdot Y$	5%
$CN \cdot CRN \cdot DOI \cdot U \cdot V \cdot Y$	4%
$PMID \cdot U$	4%
$CN \cdot CRN \cdot DOI \cdot T \cdot V \cdot Y$	4%
$CN \cdot CRN \cdot Y$	4%
$CN \cdot CRN \cdot DOI \cdot V \cdot Y$	4%

Table 1. Top 8 combinations of available fields in raw reference data accounting for about 53% of the total data (CN = container name, CRN = contrib raw name, P = pages, T = title, U = unstructured, V = volume, IS = issue, Y = year, DOI = doi, PMID = pmid). Unstructured fields may contain any value.

approach is applied on all reference documents, which only have a title, but no identifier.

With a few schema conversions, fuzzy matching can be applied to Wikipedia articles and Open Library (edition) records as well. The aspect of precision and recall are represented by the two stages: we are generous in the match candidate generation phase in order to improve recall, but we are strict during verification, in order to control precision.

# 5 Fuzzy Matching Approach

### 6 Quality Assurance

### 7 Future Work

As other dataset in this field we expect this dataset to be iterated upon.

- The fatcat catalog updates its metadata continously<sup>8</sup> and web crawls are regularly conducted. Current processing pipelines cover raw reference snapshot creation and the rederivation the graph contained within.
- Metadata extraction from PDFs depends on machine learning models, which in turn depend training sets.
  With additional crawls and metadata available we hope to improve models used for metadata extraction, reducing data extraction artifacts in the process.

 $<sup>^8</sup> A$  changelog can currenly be followed here: fatcat.wiki/changelog

 As of this version, a significant number of raw reference docs remain unmatched, which means that neither exact or fuzzy matching can recover a link to a known entity. On the one hand, this can hint at missing metadata. However, parts of the data will contain a reference to a catalogued entity, but in a specific, dense and harder to recover form.

### 8 Acknowledgements

Don't forget them or you'll have people with hurt feelings. Acknowledge anyone who contributed in any way: through discussions, feedback on drafts, implementation, etc. If in doubt about whether to include someone, include them.

#### 9 Citations

# 10 Appendix A

#### References

- [1] Citations in economics. https://citec.repec.org/. Accessed: 2021-07-30.
- [2] Initiative for open citations. https://i4oc.org/. Accessed: 2021-07-30.
- [3] E. Garfield. The evolution of the science citation index. *International microbiology*, 10(1):65, 2007.
- [4] B. I. Hutchins. A tipping point for open citation data. *Quantitative Science Studies*, pages 1–5, 2021.
- [5] P. Lopez. Grobid: Combining automatic bibliographic data recognition and term extraction for scholarship publications. In *International conference on theory and* practice of digital libraries, pages 473–474. Springer, 2009.
- [6] D. Shotton. Publishing: open citations. *Nature News*, 502(7471):295, 2013.
- [7] D. Shotton. Funders should mandate open citations. *Nature*, 553(7686):129–130, 2018.
- [8] A. Sinha, Z. Shen, Y. Song, H. Ma, D. Eide, B.-J. Hsu, and K. Wang. An overview of microsoft academic service (mas) and applications. In *Proceedings* of the 24th international conference on world wide web, pages 243–246, 2015.
- [9] J. Wu, K. Kim, and C. L. Giles. Citeseerx: 20 years of service to scholarly big data. In *Proceedings of the Conference on Artificial Intelligence for Data Discovery and Reuse*, pages 1–4, 2019.