

Data Mining and Information Extraction Methods for Large-Scale High Quality Representations of Scientific Publications

Dissertation Defense

Tarek Saier | 22. April 2024



Data Mining and Information Extraction Methods
for Large-Scale High Quality
Representations of Scientific Publications

**Data Mining and Information Extraction Methods
for Large-Scale High Quality
Scholarly Data**

Scholarly Data

■ Usage

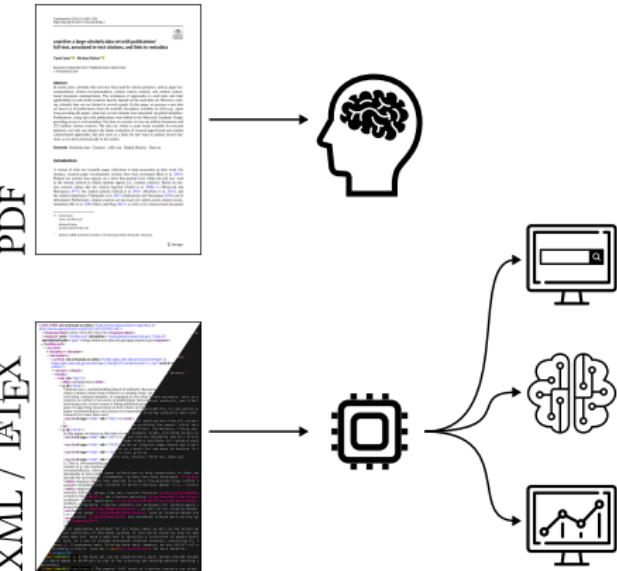
- Services (Search, Recommendation, Stats)
 - ML models (LLMs, summarization, recommender systems)
 - Analyses (temporal, geographic, institutional)

■ Flavors

- Metadata (MAG, OpenAlex, ORKG, crossref)
 - Documents (Core, arXMLiv, PMC)
 - Linked Documents (unarXive, S2ORC)

■ Data Sources

- PDF (Core, ACL Anthology)
 - XML (PubMed, PLOS, publisher internal)
 - LaTeX (arXiv)



Motivation

Background

Outline

Corpus

Citations & Non-English ooo

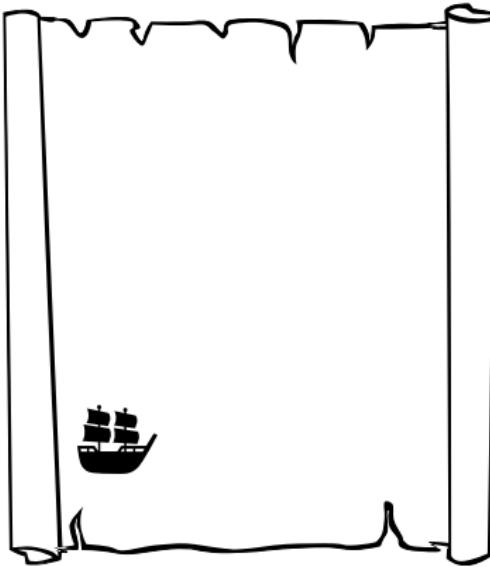
Artifact Parameters

Conclusion

References

Analogy

Maps of the Sea



Motivation



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Citations & Non-English ooo

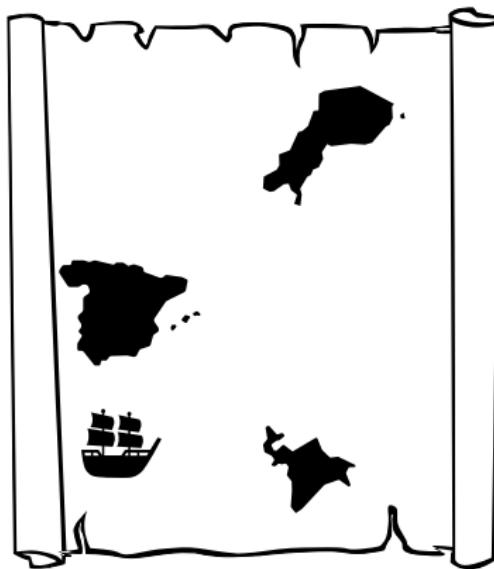
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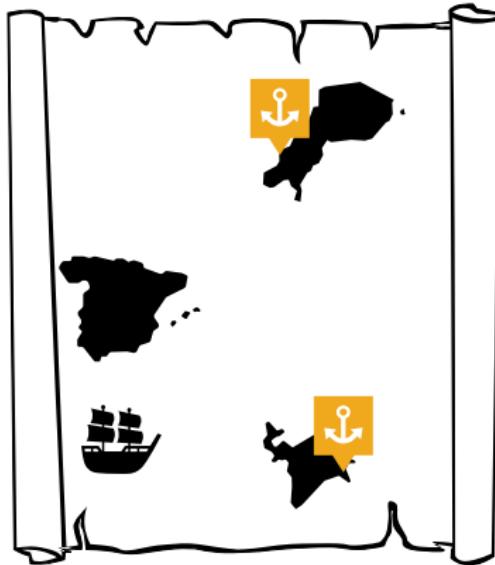
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Maps of the Sea



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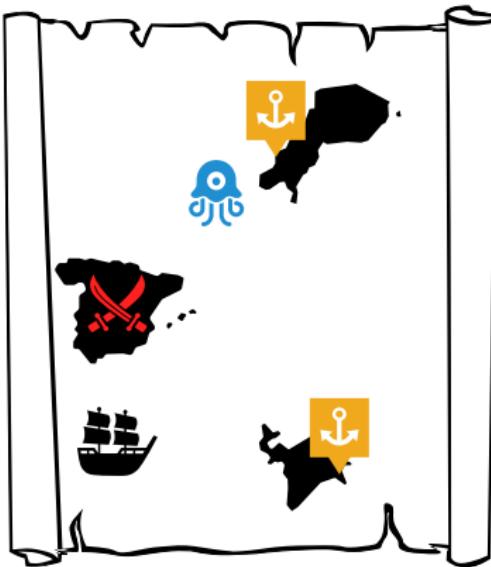
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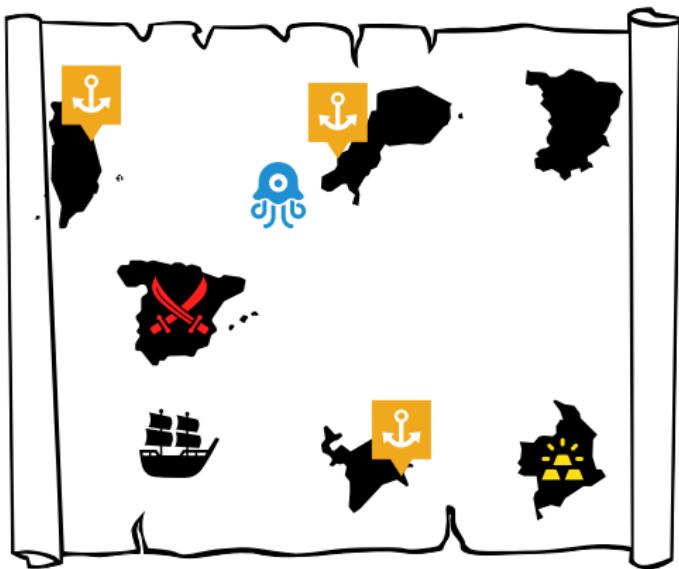
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Maps of the Sea



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Maps of the Sea / Maps of Science

The Sailor looks for

- Port to trade
- Island to explore

The Scientist looks for

- Paper to read
- Venue to publish at
- Research idea to explore

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Maps of the Sea / Maps of Science

The Trade Company looks for

- Routes to expand
- Ports to build
- Sailor to hire

The University/Funding Body looks for

- Research to fund
- Researcher to hire
- Policy to establish

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Maps of the Sea / Maps of Science

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- Research to fund
- Researcher to hire
- Policy to establish

better maps¹ ⇒ better decisions

¹Abstract representations of the real world

Maps of the Sea / Maps of Science

The Trade Company looks for

- Routes to expand
- Ports to build
- Sailor to hire

The University/Funding Body looks for

- Research to fund
- Researcher to hire
- Policy to establish

false maps ⇒ misleading/false analyses, models, etc.

Maps of the Sea / Maps of Science

The Trade Company looks for

- Routes to expand
- Ports to build
- Sailor to hire

The University/Funding Body looks for

- Research to fund
- Researcher to hire
- Policy to establish

our maps of science are insufficient¹

¹Used data sets are small, incomplete, noisy, etc.

Maps of the Sea / Maps of Science

The Trade Company looks for

- Routes to expand
- Ports to build
- Sailor to hire

The University/Funding Body looks for

- Research to fund
- Researcher to hire
- Policy to establish



Research Objective

Develop methods for generating high-quality scholarly data.

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Data Mining and Information Extraction Methods
for **Large-Scale High Quality**
Representations of Scientific Publications

Maps of Science



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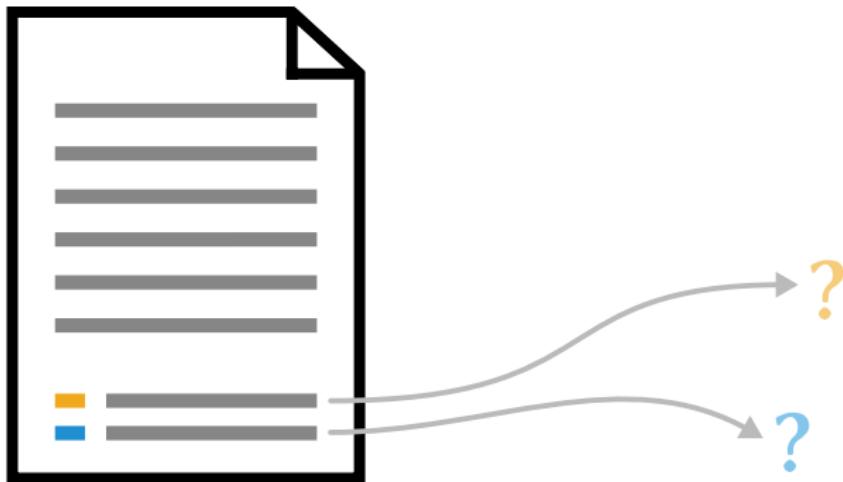
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Maps of Science



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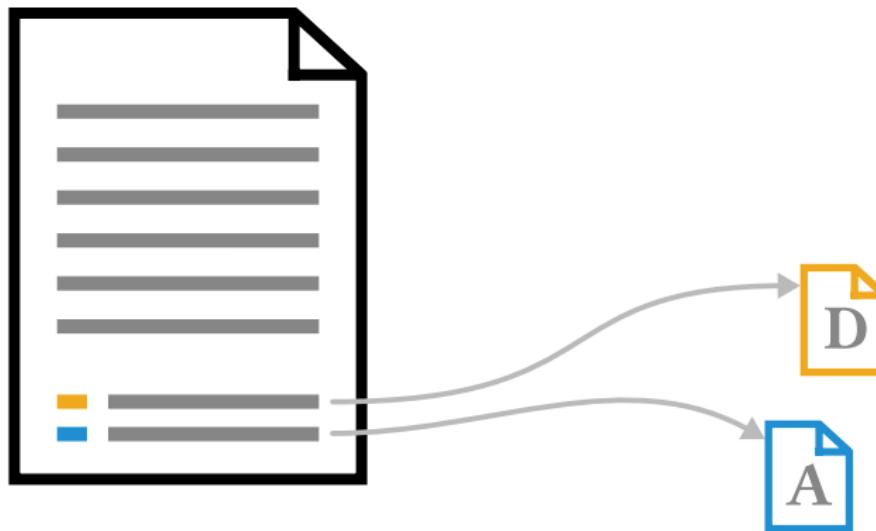
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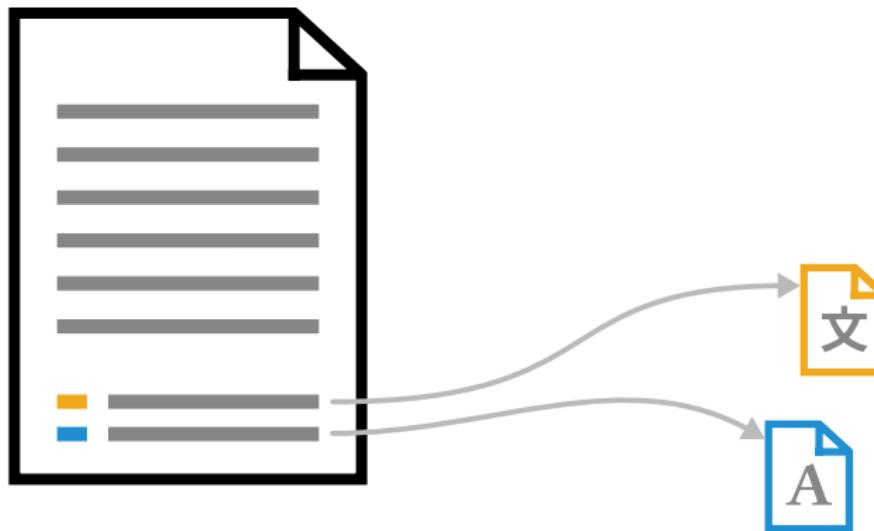
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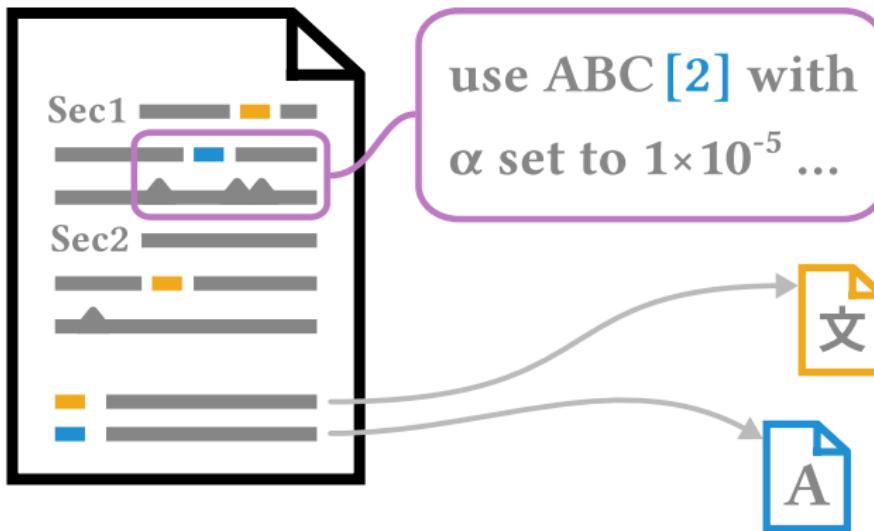
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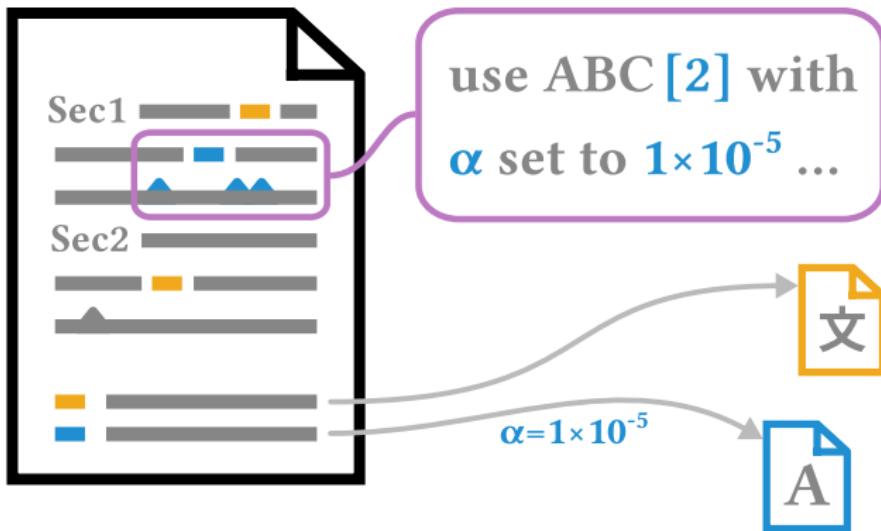
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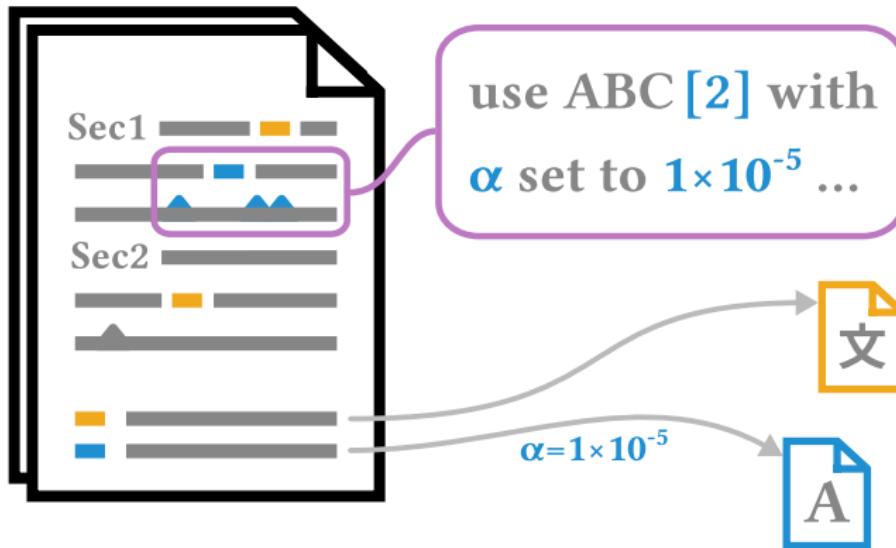
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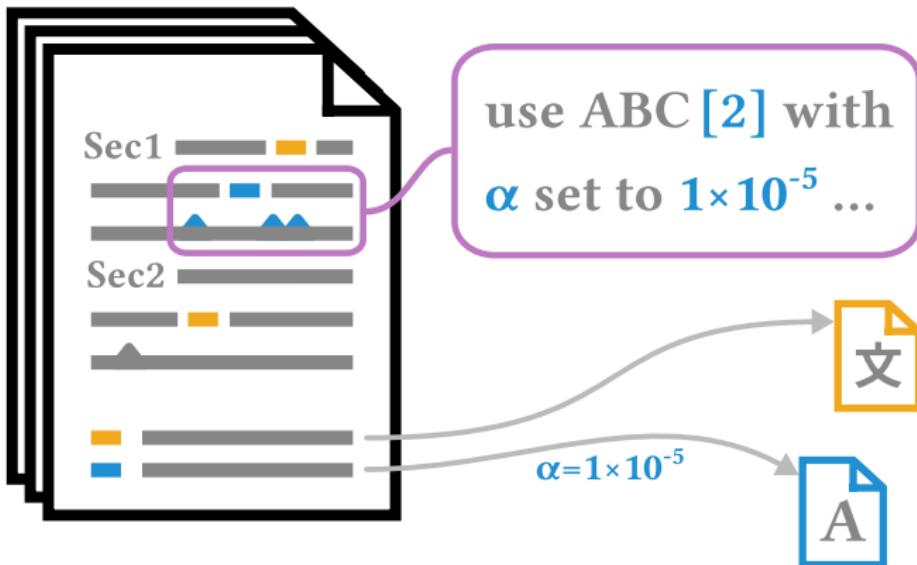
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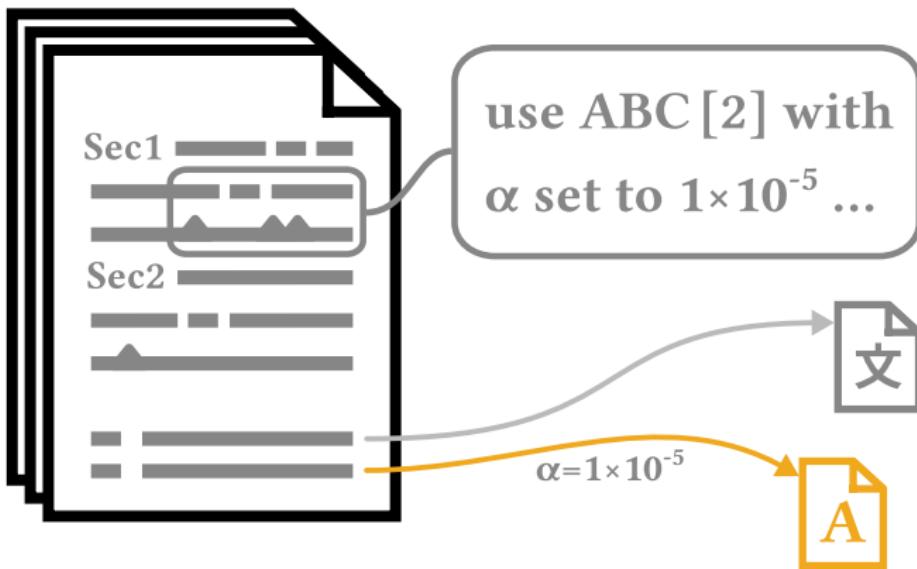
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Research Gap: Citation Network



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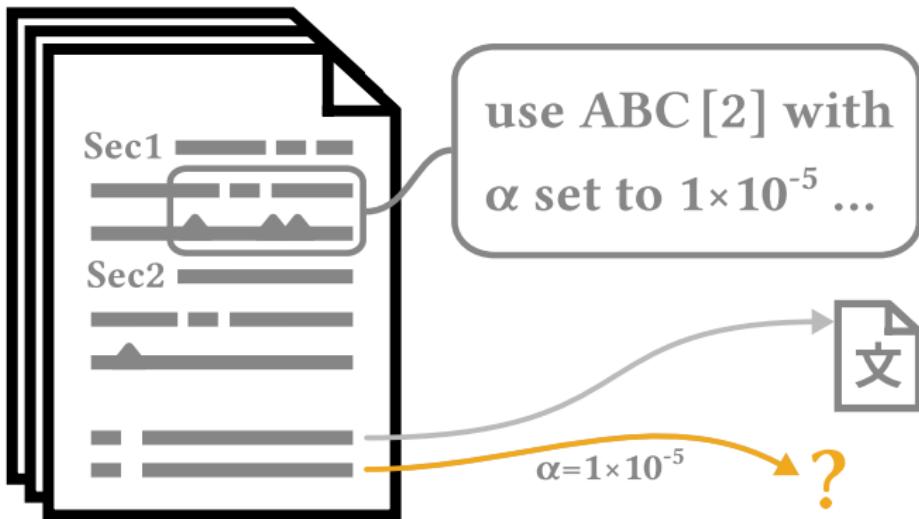
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Research Gap: Citation Network



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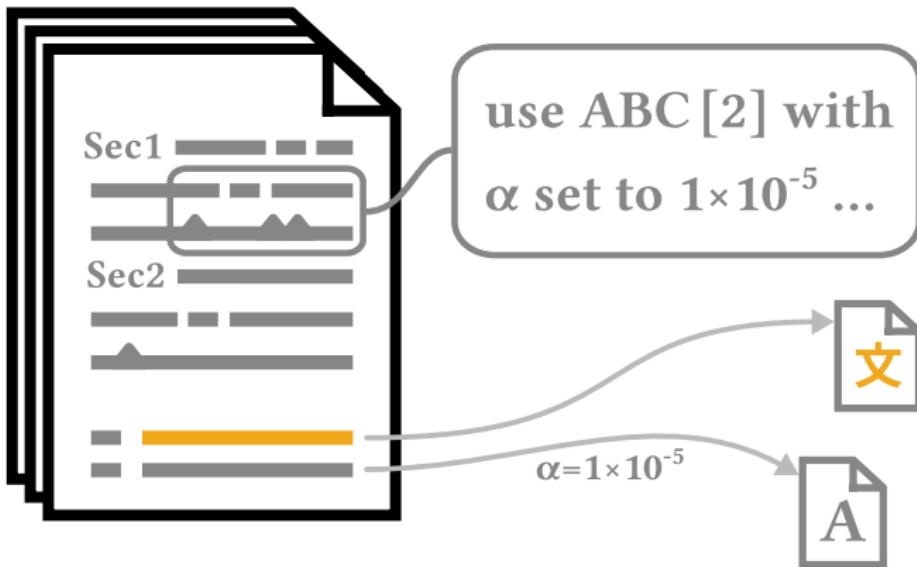
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Research Gap: Non-English Documents



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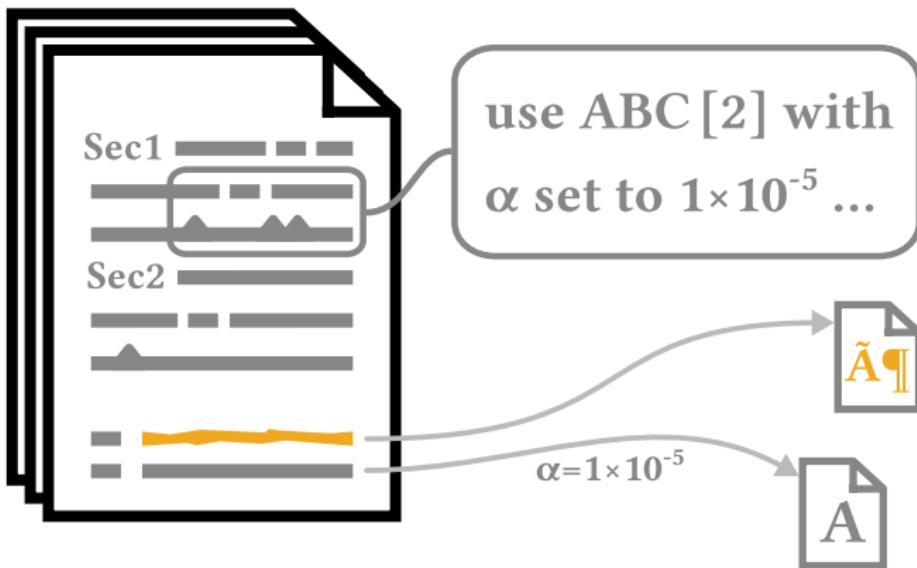
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Research Gap: Non-English Documents



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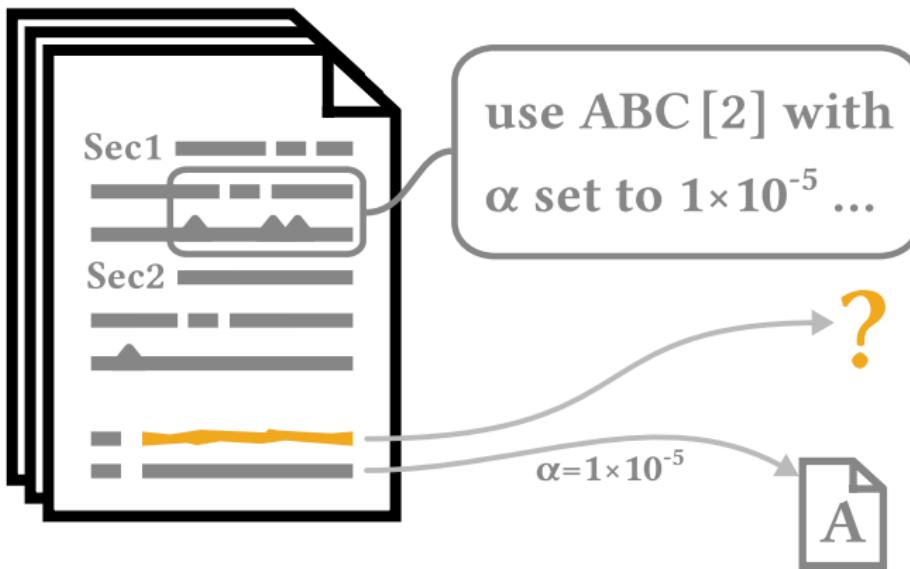
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Research Gap: Non-English Documents



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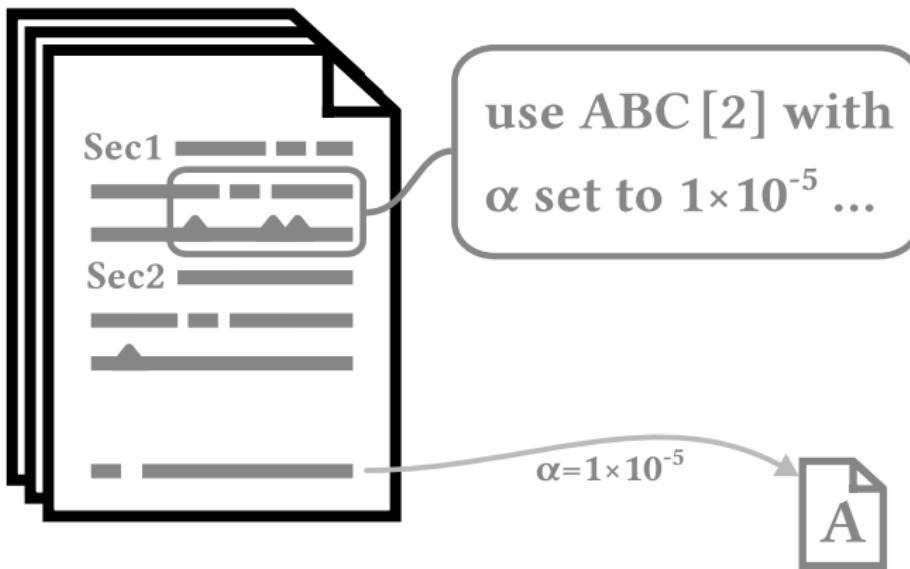
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Research Gap: Non-English Documents



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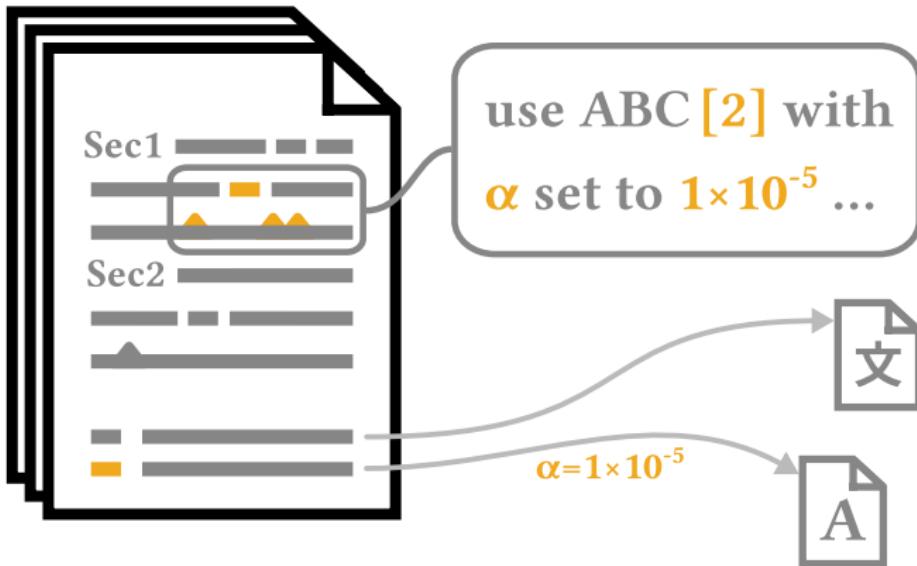
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Research Gap: Artifact Parameters



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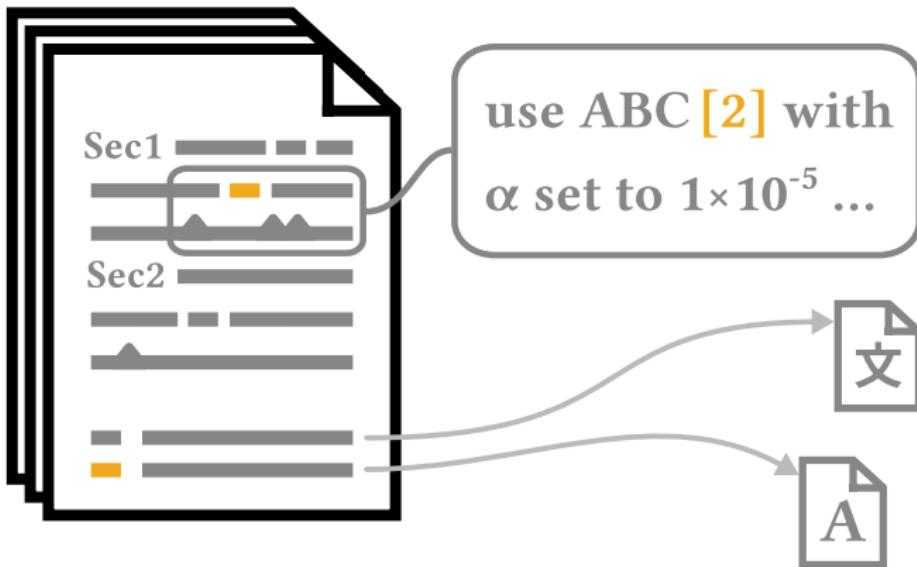
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Research Gap: Artifact Parameters



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Data Mining and Information Extraction Methods

for Large-Scale High Quality Representations of Scientific Publications



Research Objective



Research Objective

Research Gap

RT2: Citations

RT3: Non-English

RT4: Artifacts



Research Objective

Base

RT1: Corpus

Research Gap

RT2: Citations

RT3: Non-English

RT4: Artifacts



Research Objective



Data Quality

Base

RT1: Corpus

Research Gap

RT2: Citations

RT3: Non-English

RT4: Artifacts



Research Objective



Base

Research Gap

RT1: Corpus

RT2: Citations

RT3: Non-English

RT4: Artifacts

Data Quality





Research Objective



Data Quality

Base

Research Gap

RT1: Corpus

RT2: Citations

RT3: Non-English

RT4: Artifacts



✓ | X



✓ | X



✓ | X



✓ | X



✓ | X

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Research Objective



Data Quality



Base

RT1: Corpus

- ① conversion and linking pipeline
- ② extensive corpus analysis

[Scientometrics'20]

RT2: Citations

- ① improved corpus pipeline
- ② blocking method for reference linking

[ULITE'22]
[JCDL'23]

Research Gap

RT3: Non-English

- ① extraction method for cross-lingual cit.
- ② extensive citation analysis

[ICADL'20]
[IJDL'22]

RT4: Artifacts

- ① novel IE task and data set
- ② IE model development and eval.

[ECIR'24]



Research Objective



Data Quality



Base

RT1: Corpus

- ① conversion and linking pipeline
- ② extensive corpus analysis

[Scientometrics'20]

Research Gap

RT2: Citations

- ① improved corpus pipeline
- ② blocking method for reference linking

[ULITE'22]
[JCDL'23]

RT3: Non-English

- ① extraction method for cross-lingual cit.
- ② extensive citation analysis

[ICADL'20]
[IJDL'22]

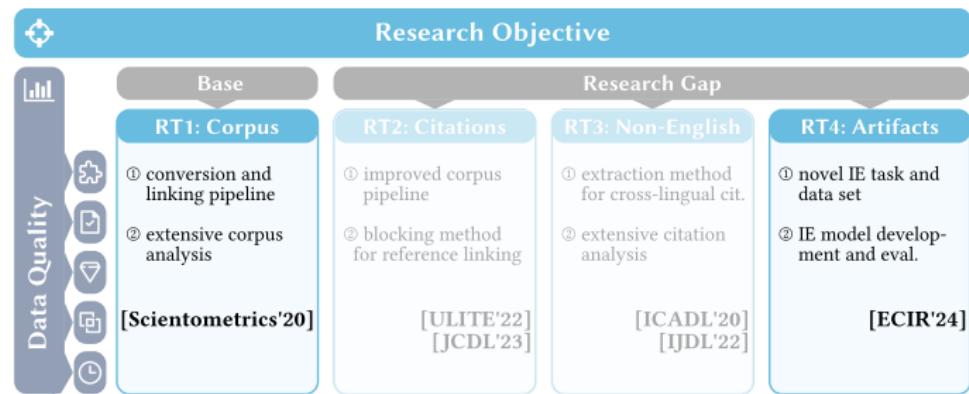
RT4: Artifacts

- ① novel IE task and data set
- ② IE model development and eval.

[ECIR'24]

Outline

- **Corpus**
 - Challenges
 - Solutions
 - Resulting Corpus
- **Citation Network**
- **Non-English Documents**
- **Artifact Parameters**
 - Task Definition
 - Methods
 - Results
- **Conclusion**
 - Contributions
 - Impact



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Corpus

unarXive

Corpus - Digest

■ Research Task

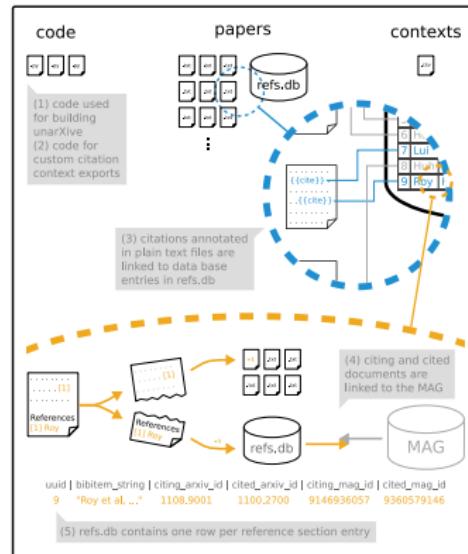
Develop a method for creating large-scale, high quality “base corpus”

■ Method

- Conversion of / IE from \LaTeX
- Joint handling of text + references
- Reference parsing + linking to metadata records

■ Results

- Corpus among 3 largest full-text corpora
- More extensive, complete, less noisy data
- State of the art reference linking



Scientometrics'20 [1]

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Corpus - Challenges

■ General

- Volume ($\sim 10^6$ docs, $\sim 10^7$ refs)
- Bridging visual medium and text information

■ Parsing

- Parser efficiency
- Typesetting info \neq semantic info
- \LaTeX is powerful and people are creative

■ Reference linking

- Choice of target set
- Parsing (bb1, not bib)
- Variance and information sparsity

```
\begin{document}
\newcommand{\nc}{\newcommand}
\nc{\be}{\begin{equation}}
\nc{\ee}{\end{equation}}
\nc{\bib}{\bibitem}
```

[10] I. Bonalde et al., Phys. Rev. Lett. **85**, 4775 (2000).

[25] Bonalde I, Yanoff B D, Salomon M B, Van Harlingen D J, Chia E M E, Mao Z Q and Maeno Y 2000
Phys. Rev. Lett. **85** 4775

[4] Jaume, S.C. and Sykes, L.R., Pure and Applied Geophysics **155**, 279-305.

Jaume, S.C. and L.R. Sykes, Evolving Towards a Critical Point: A Review of Accelerating Seismic
Moment/Energy Release Prior to Large and Great Earthquakes, Pure Appl. Geophys., 155, 279, 1999.

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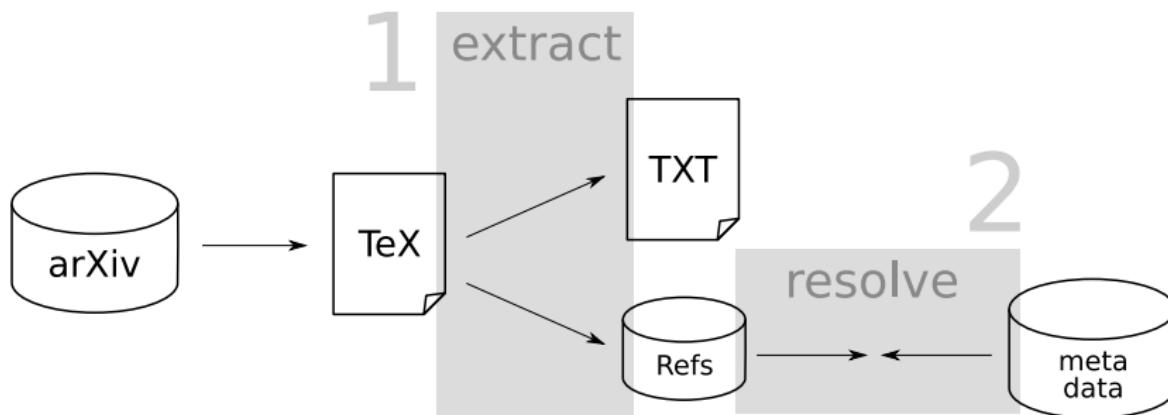
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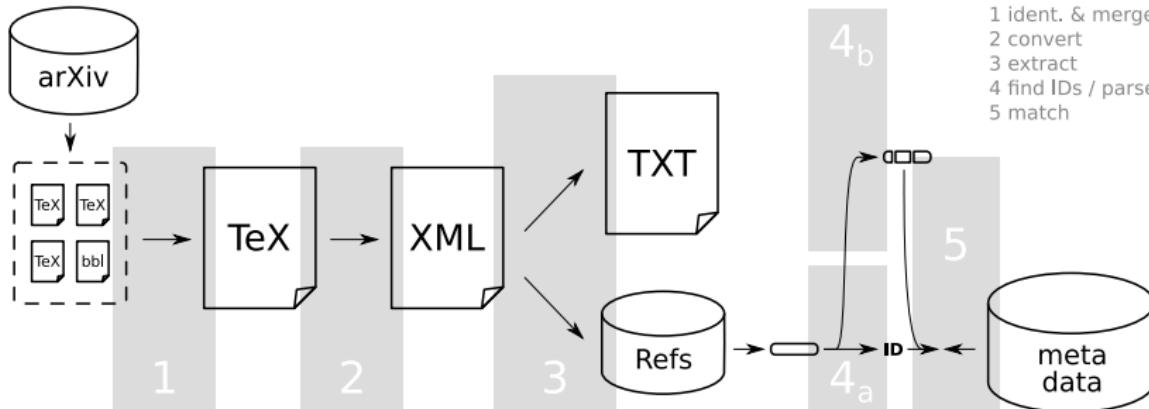
Conclusion
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References

Corpus - Solutions



Corpus - Solutions



Corpus - Result

■ Size

- 1.2 M documents (2.7 M cited)
- 16 M references
- 29 M in-text citation markers

■ Scope

- 1991–2018 (current: 2022)
- physics (63%), maths (23%), CS (11%), other (3%)

■ Reference matching

- 53% by parsing + matching
- 28% by DOI
- 19% by arXiv ID

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Corpus - Result

Data set	# Docs	Cit. markers	Disciplines	Full text	Linked
ACL-ARC [2]	11 k	no	comp. ling.	PDF	✗
ACL-AAN [3]	18 k	no	comp. ling.	PDF	✗
Scholarly Dataset 2 [4]	100 k	no	CS	PDF	✗
CiteSeerX [5] / RefSeer [6]	1 M	ambiguous	mixed	400 char excerpts	✗
PMC OAS [7]	2.3 M	exact	biomedical	XML	mixed ^a
arXiv CS [8]	90 k	exact	CS	text	✓
unarXive [1]	1.2 M	exact	phys., maths, CS	text	✓

^a No citation network due to mixed set of IDs (PubMed, MEDLINE, DOI) [9].

Corpus - Result (2022)

Data Set	Source		Citation Network ^a			# Docs	Disciplines
	Data	Format	general	compare			
CORE [10]	multiple	PDF	0%	-	>100 M		various
S2ORC (PDF) [11]	multiple	PDF	69.4%	-	12 M		various
unarXive 2020 [1]	arXiv.org	L <small>A</small> T <small>E</small> X	42.6%	42.6%	1.2 M		phys., maths, CS
S2ORC (L <small>A</small> T <small>E</small> X) [11]	arXiv.org	L <small>A</small> T <small>E</small> X	31.1%	31.1%	1.5 M		phys., maths, CS
arXMLiv [12]	arXiv.org	L <small>A</small> T <small>E</small> X	0%	0%	1.6 M		phys., maths, CS
SciXGen [13]	arXiv.org	L <small>A</small> T <small>E</small> X	41.6%	-	205 k		CS
PMC-OAS [7]	PubMed	XML	mixed ^b	-	3.3 M		biomedical
unarXive 2022 [14]	arXiv.org	L <small>A</small> T <small>E</small> X	44.4%	44.4%	1.9 M		phys., maths, CS

^a “general”: all data; incomparable. “compare”: arXiv.org data 1991–2020; directly comparable.

^b No citation network due to mixed set of IDs (PubMed, MEDLINE, DOI) [9].

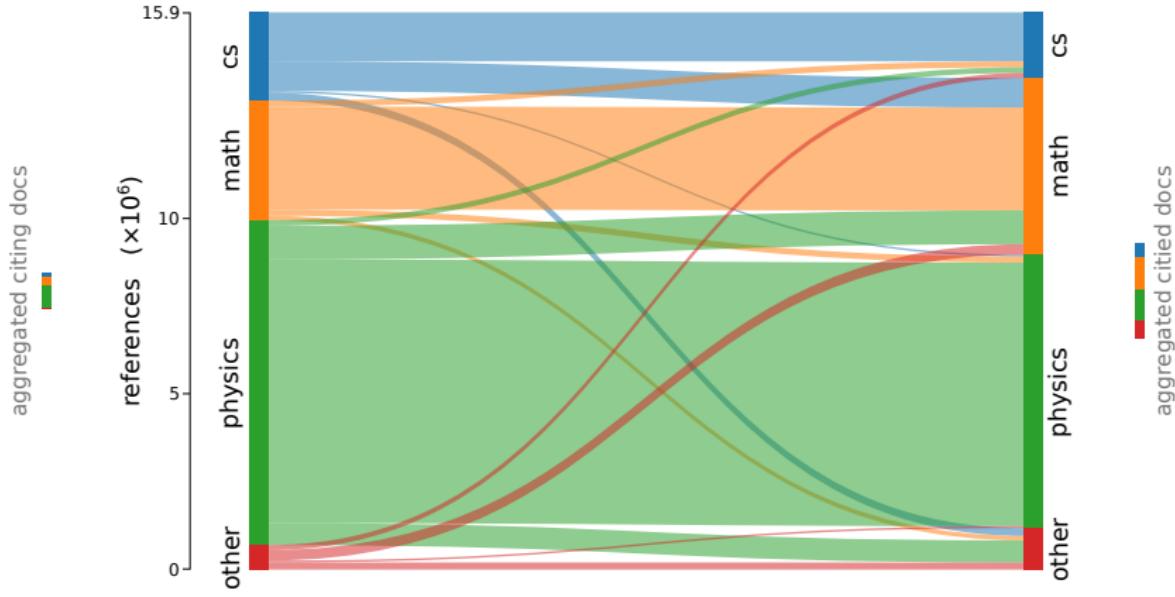
Corpus - Link Correctness

Table: Link correctness (n = 300)

Confidence level	Method ^a	Lower limit	Upper limit
0.99	Wilson	0.9613	0.9975
	Jeffreys	0.9666	0.9983
0.95	Wilson	0.9710	0.9966
	Jeffreys	0.9736	0.9972

^a Confidence interval given as Wilson score interval and Jeffreys interval [15].

Corpus - Citation Flow



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Corpus - Conclusion

■ Advancements

- More extensive
 - single domain vs multi domain
- More complete, high quality citation network
 - 13.3% increase in linked references (\leftrightarrow S2ORC_{LATEX})
 - 4% more accurate reference links (\leftrightarrow S2ORC_{LATEX})
- Less noise due to LATEX as source
- Novel types of analyses possible

■ Foundation for further studies

- → RT1 ✓

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Citations & Non-English

Citation Network - Digest

■ Research Task

Develop a method linking references more successfully without compromising accuracy

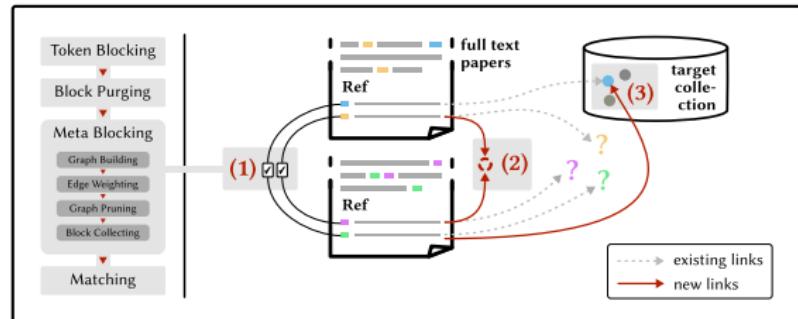
■ Method

- Use unarXive data
- Improved reference linking pipeline
(parser & target set choice, heuristics)
- Blocking & matching within set of references

■ Results

- +2% in base matching success (SOTA)
- Manifold increase in bibliographic coupling matches

■ → RT2 ✓



ULITE'22 [16]

JCDL'23 [14]

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Non-English Documents - Digest

■ Research Task

Develop an approach to include non-English publications into large-scale scholarly data

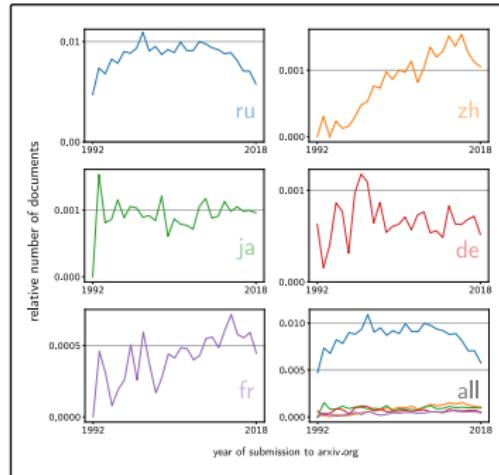
■ Method

- Use unarXive data
- Identify cross-lingual citations by reference strings
- Temporal and geographic analyses

■ Results

- Reliable method for identification
- Largest study so far ($<1k \rightarrow >1M$)
- Identification of trends and challenges

■ → RT3 ✓



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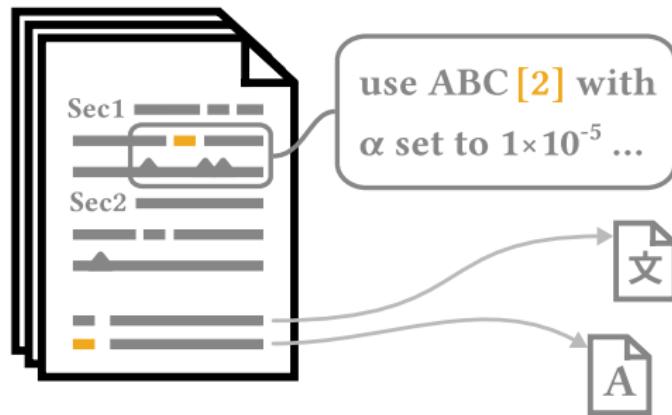
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Conclusion
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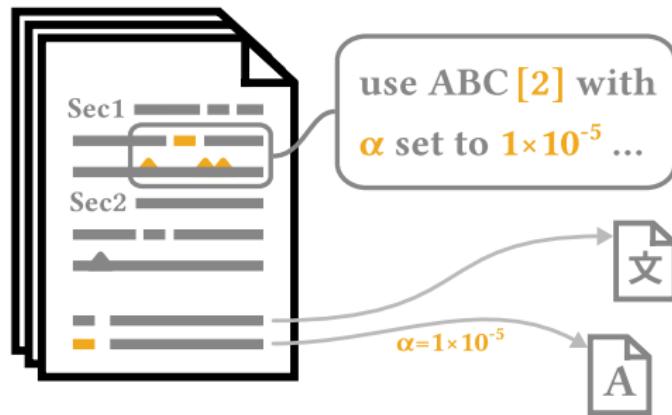
References

Artifact Parameters

Artifact Parameters

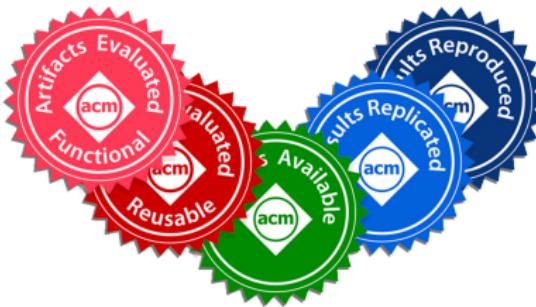


Artifact Parameters



Artifact Parameters - Motivation

The image shows two screenshots of scientific platforms. On the left, the D4Science homepage displays a "Trending Research" section featuring a paper titled "Visual Autoregressive Modeling: Scalable Image Generation via Next-Scale Prediction". The paper has 1,386 stars and was last updated on 3 Apr 2024. It includes links to the paper and code. On the right, the "Dataset Search" interface from the same platform allows users to search for datasets related to topics like "coronavirus covid-19" or "water quality site:canada.ca".



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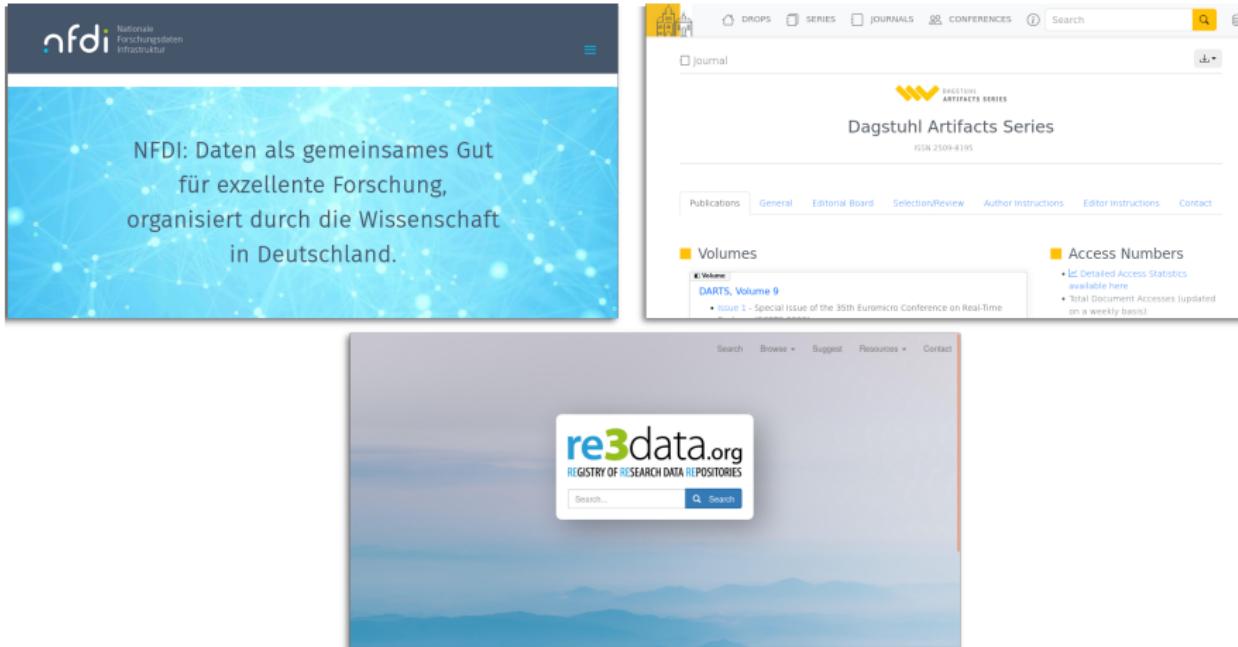
Citations & Non-English
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Artifact Parameters - Motivation



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Artifact Parameters - Motivation

NLP field increasing focus on **data** and its **algorithmic processing**

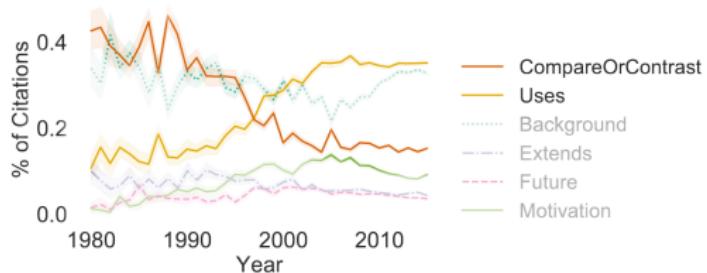


Figure 5: Changes in the average citation frame in ACL papers reveals a continued decline in the percentage of COMPARISON OR CONTRAST and increase in USES citations. The increase in BACKGROUND citations circa 2010 marks the start of the era of unlimited references in ACL conferences. Shaded regions show bootstrapped 95% confidence intervals.

Jurgens, D. et al. Measuring the Evolution of a Scientific Field through Citation Frames. *Transactions of the Association for Computational Linguistics* 6, 391–406 (2018)

Artifact Parameters - Digest

■ Research Task

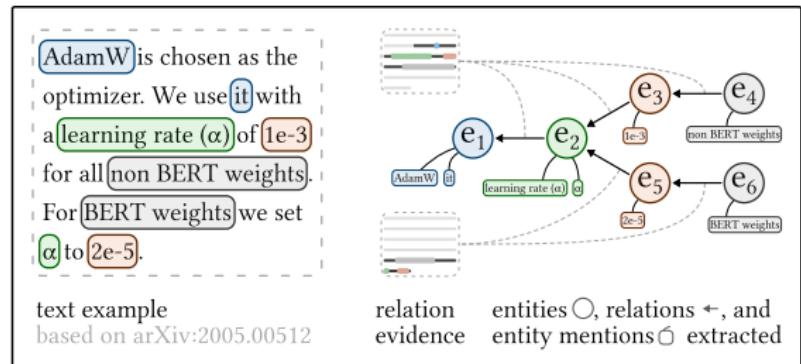
Develop a method to extract fine-grained research artifacts information

■ Method

- Task formalization
- Data annotation (data from unarXive)
- Two lines of approaches
 - BERT based model approach
 - LLM based approach

■ Results

- Novel task, novel data
- Improvements over SOTA baselines
- Methods applicable to large data sets



ECIR'24 [20]

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References

Artifact Parameters - Task

AdamW is chosen as the optimizer. We use it with a learning rate (α) of 1e-3 for all non BERT weights. For BERT weights we set α to 2e-5.

text example
based on arXiv:2005.00512

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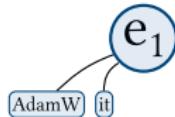
Artifact Parameters
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References

Artifact Parameters - Task

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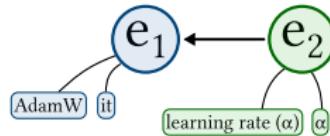


text example
based on arXiv:2005.00512

entity type
artifact

Artifact Parameters - Task

AdamW is chosen as the optimizer. We use it with a learning rate (α) of 1e-3 for all non BERT weights. For BERT weights we set α to 2e-5.



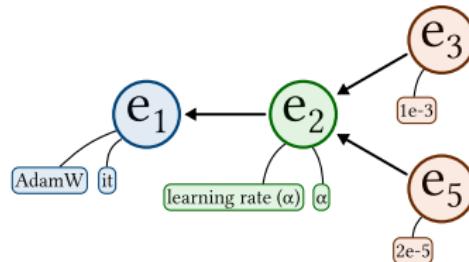
text example
based on arXiv:2005.00512

entity type
artifact parameter

Artifact Parameters - Task

AdamW is chosen as the optimizer. We use it with a learning rate (α) of $1e-3$ for all non BERT weights. For BERT weights we set α to $2e-5$.

text example
based on arXiv:2005.00512

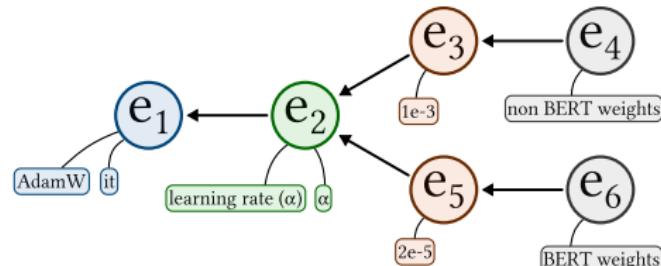


entity type
artifact parameter value

Artifact Parameters - Task

AdamW is chosen as the optimizer. We use it with a learning rate (α) of $1e-3$ for all non BERT weights. For BERT weights we set α to $2e-5$.

text example
based on arXiv:2005.00512



entity type
artifact parameter value context

Artifact Parameters - Related Work: Fine-tuned models

- SciERC dataset
 - SCIEE [21]
 - PL-Marker [22]
 - → **entity type overlap**
- [...]
- SemEval 2021 (measurements)
 - LIORI [23]
 - → **utilize mention pattern regularities**

We evaluate our model on the task of **question answering** using

Section : Dataset

SQuAD is a **machine comprehension** dataset on a large set of **Wikipedia articles** , Two metrics are used to evaluate models: **Exact Match (EM)** and a softer metric , **F1 score**

Section: Model Details .

... Each paragraph and question are tokenized by a regular - expression - based word tokenizer (**PTB Tokenizer**) and fed into the model .
....

Section : Results .

The results of our model and competing approaches on the hidden test are summarized in Table [reference]. **BiDAF (ensemble)** achieves an **EM** score of 73.3 and an **F1** score of 81.1 , outperforming all previous approaches .

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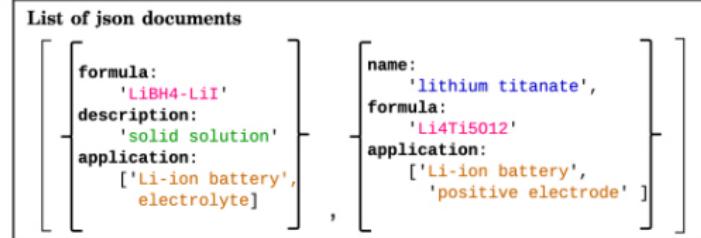
References

Artifact Parameters - Related Work: LLMs

- Medical science [24]
 - singular values
 - lists
- Material science [25–27]
 - singular values
 - lists
 - hierarchical [27] (see right)
→ **data serialization format**

Note: all of the above evaluate on GPT models only.

Document: The charge and discharge performance of an all-solid-state lithium battery with the LiBH4-LiI solid solution as an electrolyte is reported. Lithium titanate ($\text{Li}_4\text{Ti}_5\text{O}_{12}$) was used as the positive electrode and...



Dunn, A. et al. *Structured information extraction from complex scientific text with fine-tuned large language models*. Dec. 2022. doi:10.48550/arXiv.2212.05238

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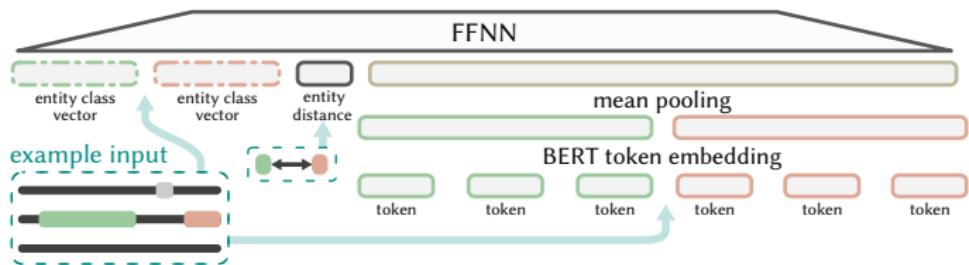
Artifact Parameters
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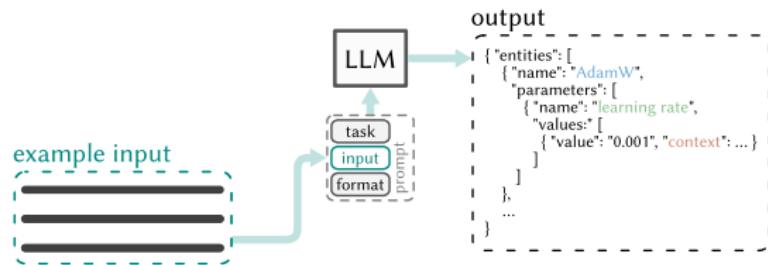
Approach: Fine-tuned models

- Based on PL-Marker (SciERC SOTA)
- (N)ER: used as is
- RE: new module, utilizing
 - Entity class embeddings
 - Entity distance



Approach: LLMs

- Data serialization format
 - JSON → YAML
- Compare 6 models
- Base prompt + tuning for each
- Zero-shot: all
- Few-shot: only Vicuna_{16k}



Approach: LLMs

- Data serialization format
 - JSON → YAML
- Compare 6 models
- Base prompt + tuning for each
- Zero-shot: all
- Few-shot: only Vicuna_{16k}

Model	Size
WizardLM [28]	13 B
Vicuna _{4k} [29]	13 B
Vicuna _{16k} [29]	13 B
Falcon [30]	40 B
GALACTICA [31]	120 B
GPT-3.5 [32]	175 B

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References

Artifact Parameters - Data

- No existing data sets
→ use unarXive ML/CV/CL/DL
 - Annotation approach
 - Paragraph level, whole papers
 - Extent
 - 444 paragraphs
 - 1,971 entities
(1,134 a, 131 p, 662 v, 44 c)
 - 614 relations
 - IAA
 - 0.867 for entities
 - 0.737 for relations

Annotation Guidelines	Research Article	Recommendations
General	<p>For clarity, I am providing specific examples from our own studies.</p> <ul style="list-style-type: none"> • Introduction: The introduction should be brief and clearly state the purpose of the study and why it is important. It should also provide a brief overview of the methods used. • Materials and Methods: This section should describe the materials and methods used in the study. It should include details about the participants, the interventions, and the outcome measures. • Results: The results should be presented in a clear and concise manner. They should include tables and figures where appropriate to support the findings. • Discussion: The discussion should interpret the results and draw conclusions. It should also discuss the limitations of the study and suggest areas for future research. 	<p>These are general guidelines for scientific papers.</p> <ul style="list-style-type: none"> • Introduction: The introduction should be brief and clearly state the purpose of the study and why it is important. It should also provide a brief overview of the methods used. • Materials and Methods: This section should describe the materials and methods used in the study. It should include details about the participants, the interventions, and the outcome measures. • Results: The results should be presented in a clear and concise manner. They should include tables and figures where appropriate to support the findings. • Discussion: The discussion should interpret the results and draw conclusions. It should also discuss the limitations of the study and suggest areas for future research.
Study Types	<p>Choosing the right type of study depends on the question being asked.</p> <ul style="list-style-type: none"> • Observational: These studies observe what happens to people over time without changing anything. Examples include cohort studies, case-control studies, and cross-sectional studies. • Interventional: These studies involve giving people treatments or interventions to see if they work. Examples include randomized controlled trials (RCTs) and non-randomized controlled trials (NRCTs). • Experimental: These studies involve manipulating variables to see if they cause changes. Examples include animal studies and laboratory experiments. 	<p>Choosing the right type of study depends on the question being asked.</p> <ul style="list-style-type: none"> • Observational: These studies observe what happens to people over time without changing anything. Examples include cohort studies, case-control studies, and cross-sectional studies. • Interventional: These studies involve giving people treatments or interventions to see if they work. Examples include randomized controlled trials (RCTs) and non-randomized controlled trials (NRCTs). • Experimental: These studies involve manipulating variables to see if they cause changes. Examples include animal studies and laboratory experiments.
Methodology	<p>When writing about methodology, it is important to be as detailed as possible. This includes describing the study design, the participants, the interventions, and the outcome measures.</p> <ul style="list-style-type: none"> • Study Design: The study design should be clearly described, including whether it was observational or interventional, and whether it was randomized or non-randomized. • Participants: The participants should be described in detail, including their age, gender, race, ethnicity, and other relevant characteristics. • Interventions: The interventions should be described in detail, including the type of intervention, how it was delivered, and any relevant details. • Outcome Measures: The outcome measures should be described in detail, including what was measured, how it was measured, and any relevant details. 	<p>When writing about methodology, it is important to be as detailed as possible. This includes describing the study design, the participants, the interventions, and the outcome measures.</p> <ul style="list-style-type: none"> • Study Design: The study design should be clearly described, including whether it was observational or interventional, and whether it was randomized or non-randomized. • Participants: The participants should be described in detail, including their age, gender, race, ethnicity, and other relevant characteristics. • Interventions: The interventions should be described in detail, including the type of intervention, how it was delivered, and any relevant details. • Outcome Measures: The outcome measures should be described in detail, including what was measured, how it was measured, and any relevant details.
Results	<p>The results should be presented in a clear and concise manner. They should include tables and figures where appropriate to support the findings.</p> <ul style="list-style-type: none"> • Data Analysis: The data analysis should be described in detail, including the statistical methods used and any relevant details. • Findings: The findings should be interpreted and discussed in the context of the study's objectives. 	<p>The results should be presented in a clear and concise manner. They should include tables and figures where appropriate to support the findings.</p> <ul style="list-style-type: none"> • Data Analysis: The data analysis should be described in detail, including the statistical methods used and any relevant details. • Findings: The findings should be interpreted and discussed in the context of the study's objectives.
Conclusion	<p>The conclusion should interpret the results and draw conclusions. It should also discuss the limitations of the study and suggest areas for future research.</p> <ul style="list-style-type: none"> • Limitations: The limitations of the study should be discussed, including any potential sources of bias or error. • Future Research: The future research should be suggested, including any potential directions for future studies. 	<p>The conclusion should interpret the results and draw conclusions. It should also discuss the limitations of the study and suggest areas for future research.</p> <ul style="list-style-type: none"> • Limitations: The limitations of the study should be discussed, including any potential sources of bias or error. • Future Research: The future research should be suggested, including any potential directions for future studies.
References	<p>References should be cited in the text where they are first mentioned. They should be listed at the end of the paper in alphabetical order by author.</p> <ul style="list-style-type: none"> • Journal Articles: References to journal articles should include the author(s), title of the article, name of the journal, volume number, issue number, page numbers, and year of publication. • Books: References to books should include the author(s), title of the book, publisher, location, year of publication, and page numbers. • Other Sources: References to other sources should include the author(s), title of the source, publisher, location, year of publication, and page numbers. 	<p>References should be cited in the text where they are first mentioned. They should be listed at the end of the paper in alphabetical order by author.</p> <ul style="list-style-type: none"> • Journal Articles: References to journal articles should include the author(s), title of the article, name of the journal, volume number, issue number, page numbers, and year of publication. • Books: References to books should include the author(s), title of the book, publisher, location, year of publication, and page numbers. • Other Sources: References to other sources should include the author(s), title of the source, publisher, location, year of publication, and page numbers.
Tables and Figures	<p>Tables and figures should be used to present data in a clear and concise manner. They should be labeled and explained in the text.</p> <ul style="list-style-type: none"> • Tables: Tables should be used to present data in a tabular format. They should be labeled and explained in the text. • Figures: Figures should be used to present data in a graphical format. They should be labeled and explained in the text. 	<p>Tables and figures should be used to present data in a clear and concise manner. They should be labeled and explained in the text.</p> <ul style="list-style-type: none"> • Tables: Tables should be used to present data in a tabular format. They should be labeled and explained in the text. • Figures: Figures should be used to present data in a graphical format. They should be labeled and explained in the text.
Supplements	<p>Supplements should be used to provide additional information that is not part of the main manuscript. They should be labeled and explained in the text.</p> <ul style="list-style-type: none"> • Supplemental Materials: Supplements should be used to provide additional information that is not part of the main manuscript. They should be labeled and explained in the text. 	<p>Supplements should be used to provide additional information that is not part of the main manuscript.</p> <ul style="list-style-type: none"> • Supplemental Materials: Supplements should be used to provide additional information that is not part of the main manuscript. They should be labeled and explained in the text.
Ethics	<p>Ethics should be considered throughout the entire study process, from the design to the analysis and reporting of results.</p> <ul style="list-style-type: none"> • Consent: Participants should give informed consent before participating in the study. This should be documented in writing and signed by the participant. • Confidentiality: Participants' personal information should be kept confidential and used only for the purposes of the study. • Debriefing: Participants should be debriefed after the study is completed to ensure they understand what happened and any potential risks. • Publication: Participants should be informed about the results of the study and given the opportunity to publish them. 	<p>Ethics should be considered throughout the entire study process, from the design to the analysis and reporting of results.</p> <ul style="list-style-type: none"> • Consent: Participants should give informed consent before participating in the study. This should be documented in writing and signed by the participant. • Confidentiality: Participants' personal information should be kept confidential and used only for the purposes of the study. • Debriefing: Participants should be debriefed after the study is completed to ensure they understand what happened and any potential risks. • Publication: Participants should be informed about the results of the study and given the opportunity to publish them.
Authorship	<p>Authorship should be based on significant contributions to the study. Contributions should be acknowledged in the text.</p> <ul style="list-style-type: none"> • Writing: The person who wrote the manuscript should be listed as the first author. • Conceptualization: The person who came up with the idea for the study should be listed as a co-author. • Methodology: The person who developed the methods used in the study should be listed as a co-author. • Software: The person who developed the software used in the study should be listed as a co-author. • Validation: The person who validated the results of the study should be listed as a co-author. • Resources: The person who provided resources for the study should be listed as a co-author. • Supervision: The person who supervised the study should be listed as a co-author. • Project Administration: The person who managed the project should be listed as a co-author. • Funding Acquisition: The person who secured funding for the study should be listed as a co-author. 	<p>Authorship should be based on significant contributions to the study. Contributions should be acknowledged in the text.</p> <ul style="list-style-type: none"> • Writing: The person who wrote the manuscript should be listed as the first author. • Conceptualization: The person who came up with the idea for the study should be listed as a co-author. • Methodology: The person who developed the methods used in the study should be listed as a co-author. • Software: The person who developed the software used in the study should be listed as a co-author. • Validation: The person who validated the results of the study should be listed as a co-author. • Resources: The person who provided resources for the study should be listed as a co-author. • Supervision: The person who supervised the study should be listed as a co-author. • Project Administration: The person who managed the project should be listed as a co-author. • Funding Acquisition: The person who secured funding for the study should be listed as a co-author.
Conflict of Interest	<p>Conflict of interest should be declared in the text. It should include any financial or personal relationships that could be construed as a conflict of interest.</p> <ul style="list-style-type: none"> • Financial Conflicts: Financial conflicts should be declared in the text. This should include any grants or funding received, any honoraria or consulting fees, and any other financial interests. • Personal Conflicts: Personal conflicts should be declared in the text. This should include any personal relationships that could be construed as a conflict of interest. 	<p>Conflict of interest should be declared in the text. It should include any financial or personal relationships that could be construed as a conflict of interest.</p> <ul style="list-style-type: none"> • Financial Conflicts: Financial conflicts should be declared in the text. This should include any grants or funding received, any honoraria or consulting fees, and any other financial interests. • Personal Conflicts: Personal conflicts should be declared in the text. This should include any personal relationships that could be construed as a conflict of interest.
Disclaimer	<p>A disclaimer should be included in the text to indicate that the views expressed are those of the authors and not necessarily those of the institution or funding body.</p>	<p>A disclaimer should be included in the text to indicate that the views expressed are those of the authors and not necessarily those of the institution or funding body.</p>

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Artifact Parameters

Conclusion

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Artifact Parameters - Experiments: Fine-tuned models

■ Setting

- 5-fold cross-validation
- Stratified sampling

■ Results (F_1 [%])

- ER: 78.0
- RE: 9.9 → 38.8

■ Analysis

- Parameter: low performance
- Contexts: not predicted

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References

Artifact Parameters - Experiments: Fine-tuned models

■ Setting

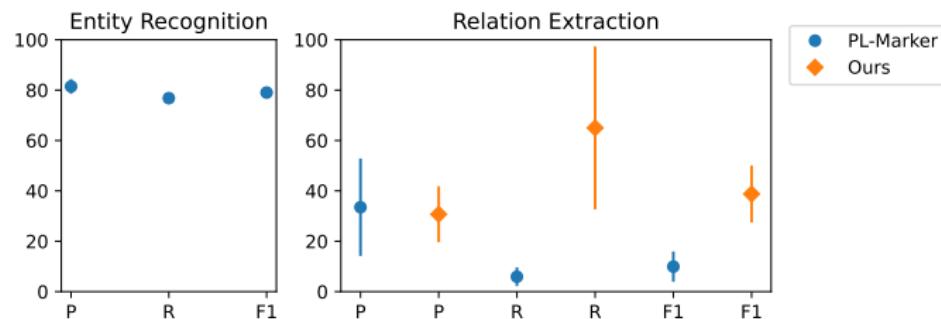
- 5-fold cross-validation
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Artifact Parameters - Experiments: Fine-tuned models

■ Setting

- 5-fold cross-validation
- Stratified sampling

■ Results (F_1 [%])

- ER: 78.0
- RE: 9.9 → 38.8

■ Analysis

- Parameter: low performance
- Contexts: not predicted

Table: Ablation study results (model inputs are: T = BERT token embeddings, C = entity class embeddings, D = entity distance)

Used	P [%]	R [%]	F_1 [%]
¬CD	15.5	8.8	11.1
T¬D	16.6	29.8	19.6
TC¬	26.5	65.0	35.5
TCD	30.7	65.0	38.8

Artifact Parameters - Experiments: LLMs

■ Setting

- Zero-/5-shot
- Compare JSON w/ YAML variant

■ Results (best ER / RE F₁ [%])

- Zero-shot: 37.4 / 7.8
- 5-shot: 44.0 / 6.1

■ Analysis

- Very low RE performance
- YAML: avg. +5% in ER
- 5-shot: +27% ER, +6% RE
- Format adherence
+ entity hallucinations

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Artifact Parameters - Experiments: LLMs

■ Setting

- Zero-/5-shot
- Compare JSON w/ YAML variant

■ Results (best ER / RE F_1 [%])

- Zero-shot: 37.4 / 7.8
- 5-shot: 44.0 / 6.1

■ Analysis

- Very low RE performance
- YAML: avg. +5% in ER
- 5-shot: +27% ER, +6% RE
- Format adherence
+ entity hallucinations

Zero-shot		ER	RE
Model	Output	F_1 [%]	F_1 [%]
WizardLM	JSON	8.6	0.1
	YAML	15.3 <small>△+6.7</small>	0.1 <small>△+0.0</small>
Vicuna _{4k}	JSON	11.5	1.2
	YAML	22.3 <small>△+10.8</small>	0.1 <small>△-1.1</small>
Falcon	JSON	10.2	0.0
	YAML	19.8 <small>△+9.6</small>	0.0 <small>△+0.0</small>
GALACTICA	JSON	19.5	0.3
	YAML	21.1 <small>△+1.6</small>	0.1 <small>△-0.2</small>
GPT-3.5	JSON	33.8	7.2
	YAML	37.4 <small>△+3.6</small>	7.8 <small>△+0.6</small>
5-shot		ER	RE
Vicuna _{16k}	JSON	39.6	1.3
	YAML	44.0 <small>△+0.4</small>	6.1 <small>△+4.8</small>

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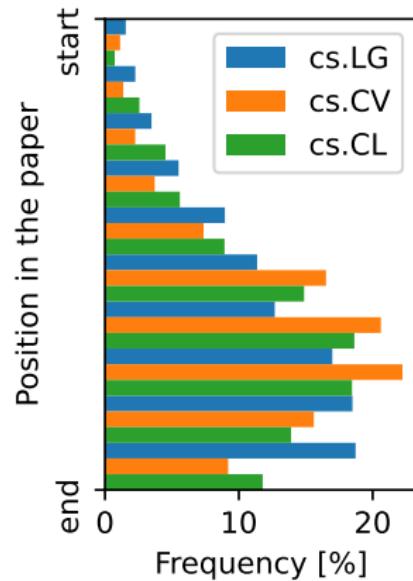
Artifact Parameters
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Artifact Parameters - Experiments: Application

- Apply best model (BERT based) on 15k paper sample
- Parameters information given in
 - 36% of ML papers
 - 42% of CV papers
 - 36% of CL papers
 - 7% of DL papers
- Distribution towards second half of paper across disciplines



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Artifact Parameters - Conclusion

■ Advancements

- Novel, relevant task
(data scheme, annotation guidelines)
- High quality manually annotated data set
- Approaches based on BERT, LLMs
 - BERT model based approach
29% F_1 increase for RE
 - LLM approach
Avg. 5.5% F_1 increase for ER
(consistent across all used LLMs)
- Trained model applicable on large scale

■ → RT4 ✓

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Overall Conclusion



Research Objective

Develop methods for generating high-quality scholarly data.

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Research Objective

Develop methods for generating high-quality scholarly data.

Research Gap

- Citation Network
- Non-English
- Research artifacts

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Research Objective



Data Quality



Base

RT1: Corpus

- ① conversion and linking pipeline
- ② extensive corpus analysis

[Scientometrics'20]

RT2: Citations

- ① improved corpus pipeline
- ② blocking method for reference linking

[ULITE'22]
[JCDL'23]

Research Gap

RT3: Non-English

- ① extraction method for cross-lingual cit.
- ② extensive citation analysis

[ICADL'20]
[IJDL'22]

RT4: Artifacts

- ① novel IE task and data set
- ② IE model development and eval.

[ECIR'24]



Research Objective

Data Quality	Base		Research Gap	
	RT1: Corpus	RT2: Citations	RT3: Non-English	RT4: Artifacts
	+	+	○	○
	+	=	○	○
	+	+	+	+
	+	+	+	+
	+	+	○	○

Overall Conclusion

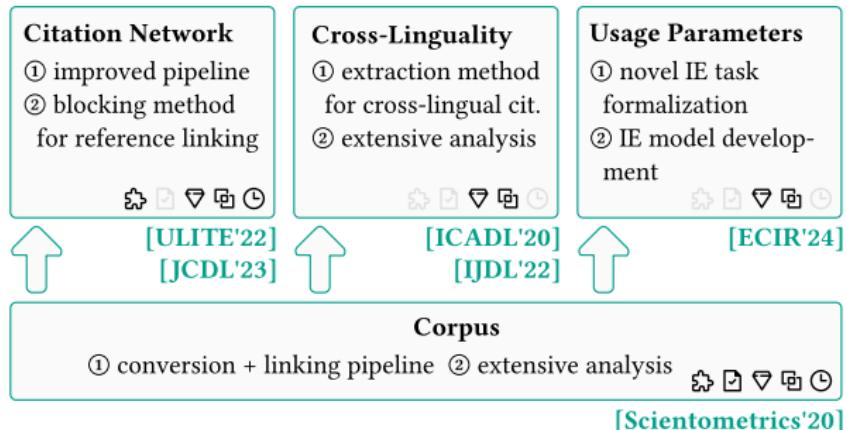
■ Big Picture

- Corpus creation method
- Advances in three focus areas
 - Citation Network
 - Non-English Documents
 - Artifact Parameters

■ Improved

- Completeness
- Accuracy
- Relevance
- Comparability
- Timeliness

■ Impact on research community



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Impact

Adoption by the research community.

■ Methodology

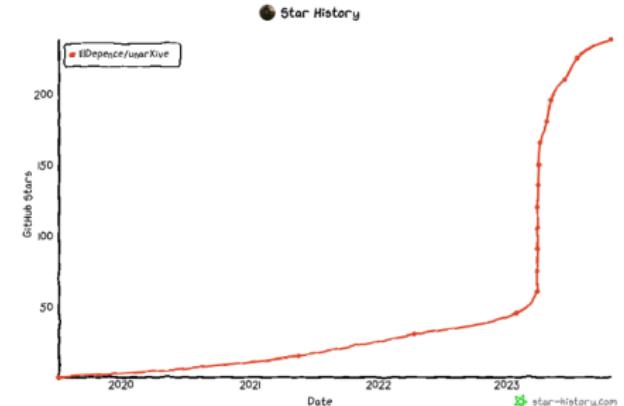
- Document Parsing Methodology [11]

■ Model dev/eval

- Citation Recommendation [33] ([34, 35])
- Citation Analysis [36, 37] ([17, 18])
- Document Retrieval [38]
- Researcher Profile Embeddings [39]
- Reference Linking ([16])

■ Dataset extension

- Link Prediction ([40])
- NER+RE ([20])



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References

Thank You ☺

Questions?

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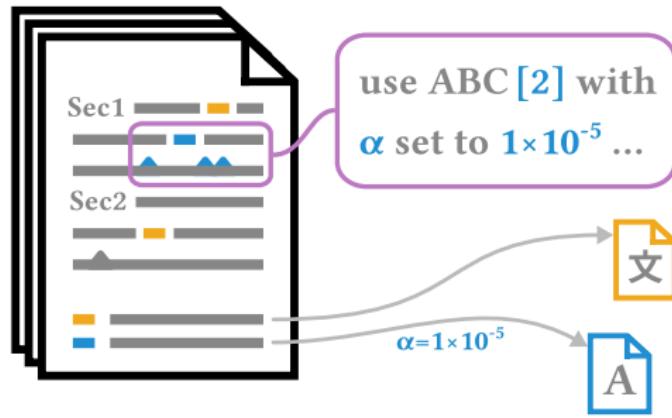
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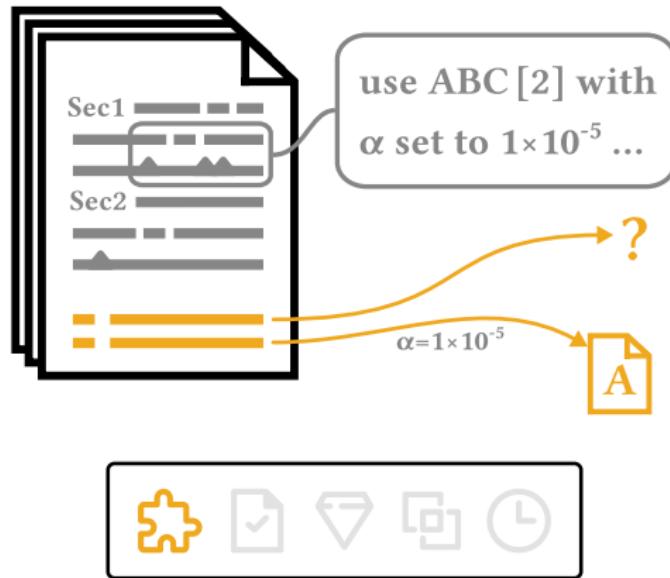
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Extra Slides

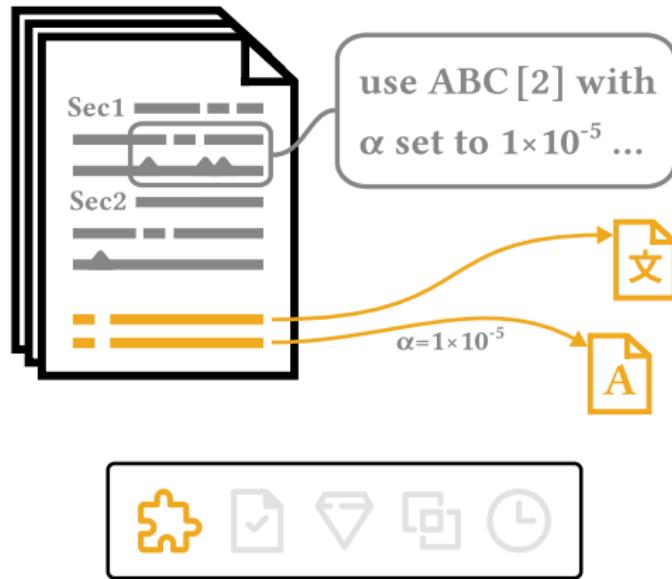
Quality Dimensions



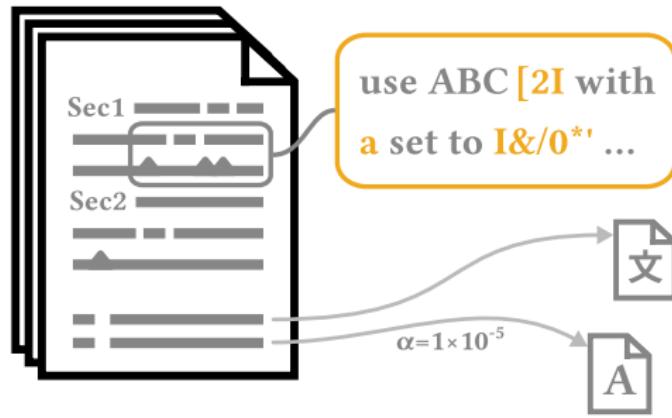
Quality Dimensions: Completeness



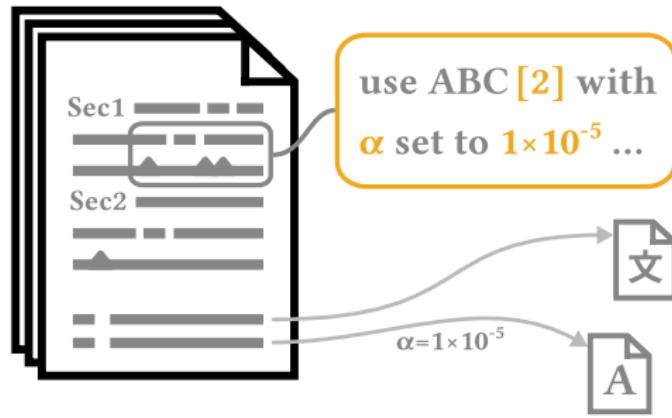
Quality Dimensions: Completeness



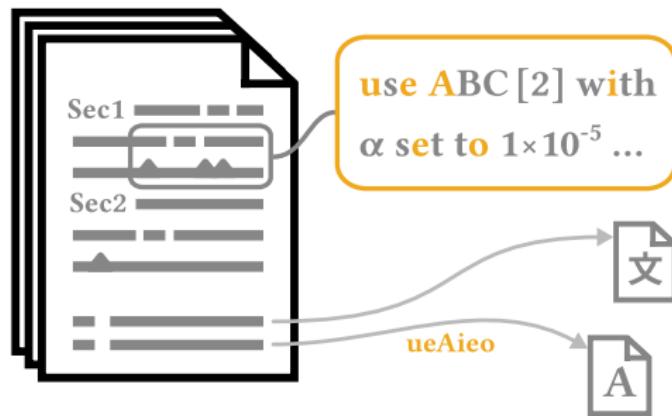
Quality Dimensions: Accuracy



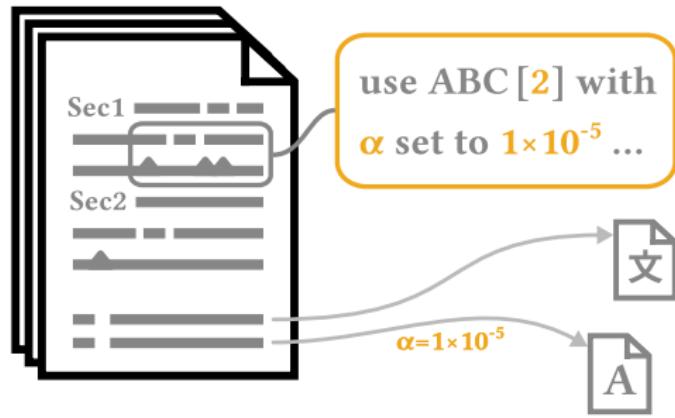
Quality Dimensions: Accuracy



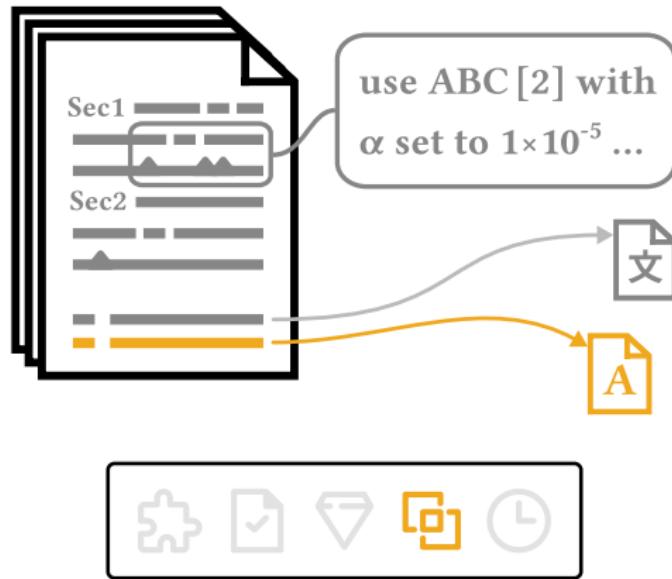
Quality Dimensions: Relevance



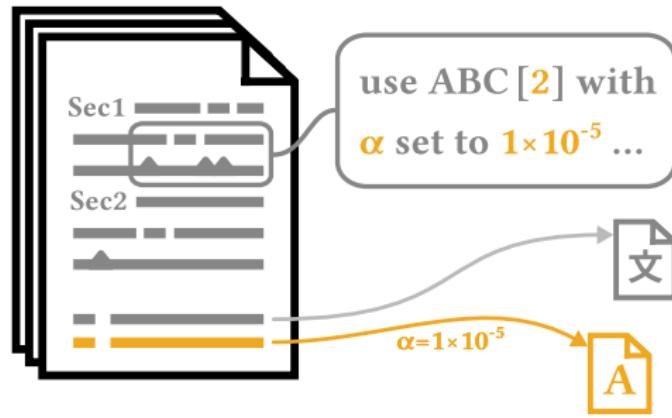
Quality Dimensions: Relevance



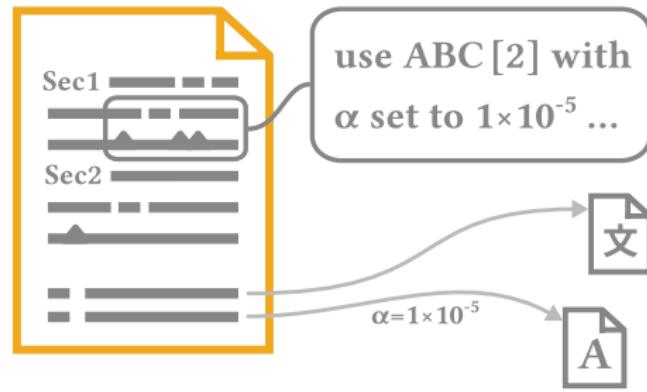
Quality Dimensions: Comparability



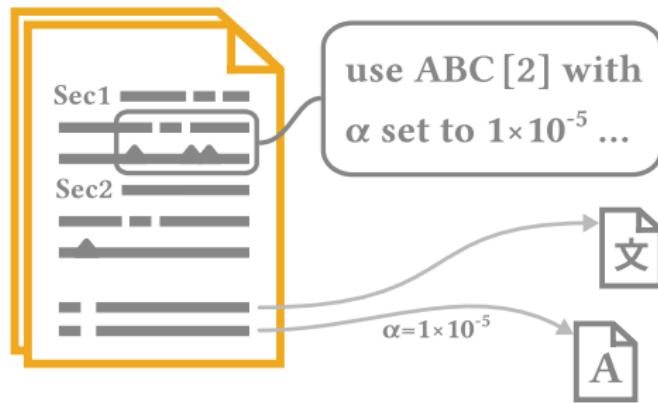
Quality Dimensions: Comparability



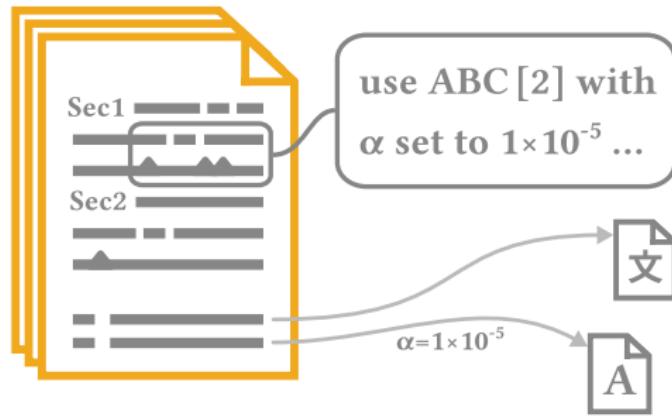
Quality Dimensions: Timeliness



Quality Dimensions: Timeliness



Quality Dimensions: Timeliness



Corpus - Stats

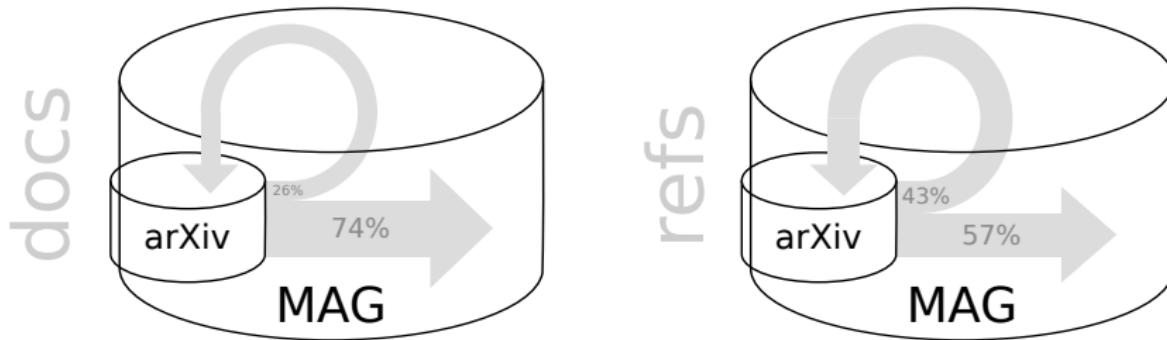
	citing documents	references		cited documents
		outgoing	incoming	
<i>full data set:</i>	1,043,126	15,954,664	15,954,664	2,746,288
full text	1,043,126	15,954,664	7,181,576	736,597
linked to MAG	994,351	15,846,351	15,954,664	2,746,288
<i>by discipline:</i>				
physics	662,894	9,300,576	7,827,072	921,852
mathematics	237,422	3,426,117	5,062,033	906,301
computer science	111,694	2,526,656	1,876,401	425,860
other	31,116	701,315	1,189,158	492,275

data: <http://doi.org/10.5281/zenodo.3385851>

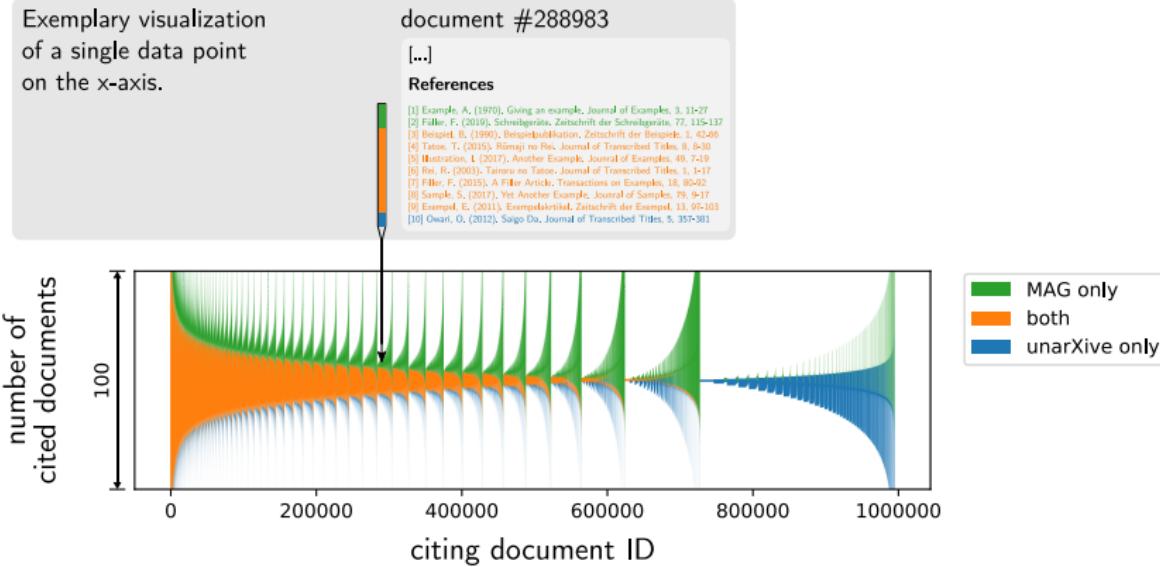
code: <https://github.com/IllDepence/unarXive>



Corpus - Citation Flow



Corpus - Reference Composition



Corpus - Target Sec. Specific Refs

	Discipline ^a	Count	Normalization factor	Normalized ratio (%)
Citing	Mathematics	298,009	4.66	8.70
	CS	9,123	6.31	0.36
	Physics	30,593	1.72	0.33
Cited	Mathematics	313,651	3.15	6.20
	CS	12,179	8.50	0.65
	Physics	31,087	2.04	0.40
Pairs	<u>Math[†]→Math[‡]</u>	200,859	5.41	6.81
	<u>Math[†]→CS</u>	5,134	92.13	2.96
	<u>Math[†]→Phys</u>	3,114	89.88	1.75
	<u>CS→Math[‡]</u>	3,456	18.82	0.41
	<u>Phys→Math[‡]</u>	3,859	16.49	0.40
	<u>CS→CS</u>	2,500	11.38	0.18
	<u>Phys→Phys</u>	10,374	2.12	0.14
	<u>CS→Phys</u>	50	307.16	0.10
	<u>Phys→CS</u>	137	101.40	0.09

^a [†]: Mathematics citing document, [‡]: Mathematics cited document, X→X: Citing and cited document are from the same discipline.



Artifact Parameters - Task

- **Task Type**

- (Named) Entity Recognition, Relation Extraction

- **4 entity classes**

- (1) research **artifact**: model, method, data set, ...
 - descriptions of how authors use the artifacts
 - (2) **parameter** (α , learning rate, k, ...)
 - (3) **value** (1e-3, five, $\frac{1}{3}$, ...)
 - (4) **context** (for fine-tuning, during grid search, ...)

- **1 relation type**

- Given by entity type pair



Artifact Parameters - Task (ext)

- **Goal**

- Automatically extract hyperparameter information from paper text

- **Motivation**

- Reproducibility indication [41], automated reproduction [42]
 - Uncover conventions and trends
 - More fine-grained paper representations (similarity measures, recommendation, search)

- **Task Type**

- (Named) Entity Recognition, Relation Extraction



Artifact Parameters - Related Work: Fine-tuned models

- SciERC dataset
 - SCIEE [21]
 - PL-Marker [22]
 - → **entity type overlap**
- SciREX dataset
 - TempGen [43] (only RE)
- SemEval 2022 (math symb. to descr.)
 - JBUU-CCLab [44]
 - AIFB [45]
- SemEval 2021 (measurements)
 - LIORI [23]
 - → **utilize mention pattern regularities**

We evaluate our model on the task of **question answering** using

Section : Dataset

SQuAD is a **machine comprehension** dataset on a large set of **Wikipedia articles** , Two metrics are used to evaluate models: **Exact Match (EM)** and a softer metric , **F1 score**

Section: Model Details .

... Each paragraph and question are tokenized by a regular - expression - based word tokenizer (**PTB Tokenizer**) and fed into the model .
....

Section : Results .

The results of our model and competing approaches on the hidden test are summarized in Table [reference]. **BiDAF { ensemble }** achieves an **EM** score of 73.3 and an **F1**-score of 81.1 , outperforming all previous approaches .



Artifact Parameters - Related Work: Fine-tuned models (other data sources)

- From code documentation [46]
- From code [47]



Artifact Parameters - Approach: LLMs

In the context of machine learning and related fields, what (if any) are the entities (datasets, models, methods, loss functions, regularization techniques) mentioned in the LaTeX Input Text below? What (if any) are their parameters and values?

```
[LaTeX Input Text start]
We use AdamW with a learning rate ($\alpha$) of 1e-3 for /* [...] */
[LaTeX Input Text end]
```

Answer in the following YAML format.

Format:

```
---
text_contains_entities: true/false
entities:
  - entity<N>:
    id: e<N>
    name: "<entity name>"
    type: dataset/model/method/loss function/regularization technique
    has_parameters: true/false
    parameters:
      - parameter<M>:
        id: p<N.M>
/* [...] */
...
```

Only include entities that are of type dataset, model, method, loss function, or regularization technique. Do not output entities that are of another type. Do not include entities of type task, metric, library, software, or API.

Only produce output in the YAML format specified above. Output no additional text.

Output:



Artifact Parameters - Approach: LLMs

In the context of machine learning and related fields, what (if any) are the entities (datasets, models, methods, loss functions, regularization techniques) mentioned in the LaTeX Input Text below? What (if any) are their parameters and values?

Task

```
[LaTeX Input Text start]  
We use AdamW with a learning rate ($\alpha$) of 1e-3 for /* [...] */  
[LaTeX Input Text end]
```

Input Text

Answer in the following YAML format.

```
Format:  
---  
text_contains_entities: true/false  
entities:  
  - entity<N>:  
    id: e<N>  
    name: "<entity name>"  
    type: dataset/model/method/loss function/regularization technique  
    has_parameters: true/false  
    parameters:  
      - parameter<M>:  
        id: p<N.M>  
/* [...] */  
...
```

Format

Only include entities that are of type dataset, model, method, loss function, or regularization technique. Do not output entities that are of another type. Do not include entities of type task, metric, library, software, or API.

Only produce output in the YAML format specified above. Output no additional text.

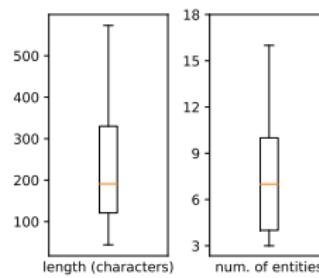
Output:

Output Prefix

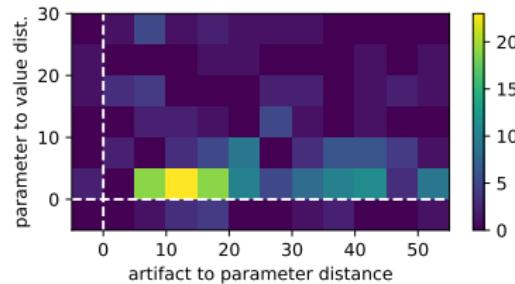


Artifact Parameters - Data

- Two annotation rounds
- Initial (pre-filtered text, exploratory, to fine-adjust scheme)
 - 151 text segments
 - 1,345 entities
 - 1,110 relations
- Main (full papers, eval data)
 - 444 paragraphs
 - 1,971 entities
(1,134 a, 131 p, 662 v, 44 c)
 - 614 relations
- IAA
 - 0.867 for entities
 - 0.737 for relations



(a) text segments



(b) relation distances (#chars)

Figure: Observations of initial annotation round

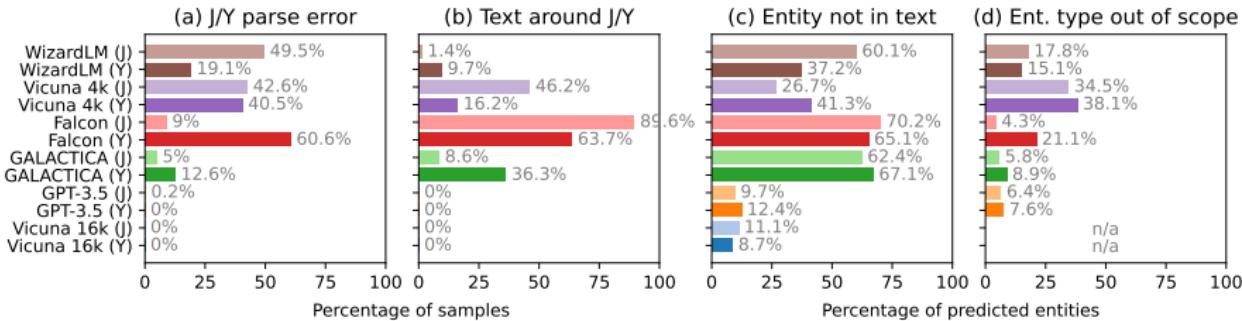


Artifact Parameters - Experiments: LLMs

Zero-shot		Entity Recognition			Relation Extraction		
Model	Output	P [%]	R [%]	F ₁ [%]	P [%]	R [%]	F ₁ [%]
WizardLM	JSON	6.9	11.3	8.6	0.1	0.8	0.1
	YAML	9.7	35.6	15.3 ^{Δ+6.7}	0.1	1.5	0.1 ^{Δ+0.0}
Vicuna _{4k}	JSON	15.1	9.3	11.5	0.7	3.8	1.2
	YAML	17.3	31.5	22.3 ^{Δ+10.8}	0.0	0.8	0.1 ^{Δ-1.1}
Falcon	JSON	37.1	5.9	10.2	0.0	0.0	0.0
	YAML	32.7	14.2	19.8 ^{Δ+9.6}	0.0	0.0	0.0 ^{Δ+0.0}
GALACTICA	JSON	25.9	15.7	19.5	0.1	2.3	0.3
	YAML	23.1	19.5	21.1 ^{Δ+1.6}	0.0	0.8	0.1 ^{Δ-0.2}
GPT-3.5	JSON	27.9	42.8	33.8	5.4	10.7	7.2
	YAML	<u>34.0</u>	<u>41.7</u>	<u>37.4^{Δ+3.6}</u>	<u>5.8</u>	<u>12.2</u>	<u>7.8^{Δ+0.6}</u>
5-shot		Entity Recognition			Relation Extraction		
Vicuna _{16k}	JSON	34.4	<u>46.7</u>	39.6	0.8	4.6	1.3
	YAML	43.9	44.1	44.0^{Δ+0.4}	4.5	9.9	6.1^{Δ+4.8}



Artifact Parameters - Experiments: LLMs



Publications - primary

Chap.	Venue	Year	Type	Length	Author Position	Venue Rating	Ref.
3	Scientometrics	2020	Journal	Full	1 of 2	SJR Q1	[1]
4	JCDL	2022	Workshop	Full	1 of 3	Core A*	[16]
	JCDL	2023	Conference	Short	1 of 3	Core A*	[14]
5	ICADL	2020	Conference	Full	1 of 2	Core A	[17]
	IJDL	2022	Journal	Full	1 of 3	SJR Q2	[18]
6	ECIR	2024	Conference	Full	1 of 4	Core A	[20]

Venue ranks from Core¹ (conferences) and SJR² (journals).³

¹See <http://portal.core.edu.au/conf-ranks/> (last accessed 2023-10-12).

²See <https://www.scimagojr.com/> (last accessed 2023-10-12).

³Ratings for publication year or, if not listed, most up-to-date ranking. Workshops ranks are that of the hosting conference.



Publications - secondary

Venue	Year	Type	Length	Author Position	Venue	Ref.
ECIR	2019	Workshop	Full	1 of 2	Core A	[34]
ECIR	2020	Conference	Full	1 of 3	Core A	[48]
NAACL	2021	Workshop	Short	3 of 4	Core A	[49]
AAAI	2022	Workshop	Full	2 of 3	Core A*	[50]
JCDL	2022	Conference	Full	3 of 3	Core A*	[51]
JCDL	2023	Conference	Short	1 of 3	Core A*	[40]

Additional publications (co-)authored leading up to and during the research period which are not a direct part of the dissertation, but nevertheless informed the overall research trajectory. Especially [34] and [49], which constitute the results of the master's thesis preceding the doctoral research period, paved the way for the dissertation.



Limitations

- **Corpus**

- L^AT_EX required (no humanities)

- **Citation Network**

- Blocking method scalability

- **Non-English Documents**

- Single “direction”
 - Dependency on author notation

- **Artifact Parameters**

- IE from text, not tables, code, etc.
 - ML specific
 - English only

