### Extracting Definienda in Mathematical Scholarly Articles with Transformers

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October 27. 2023

The 2nd WIESP @ IJCNLP-AACL 2023









Proposed Approach

### Context

- Mathematical scholarly articles contain mathematical statements such as axioms, theorems, proofs, etc.
- Semantic knowledge in these articles are not captured by traditional ways of navigating the scientific literature, e.g., keyword search.
- We aim to propose a better knowledge discovery from mathematical papers, especially those with PDF versions only.

Mathematical definition in PDF  $\downarrow$ Text of the definition  $\downarrow$ *definienda* (terms defined within)

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**Definition 2.1.** Let V be a vector space over the field F. We say that the collection  $\sigma$  of subspaces is a spread if (1)  $A, B \in \sigma$ ,  $A \neq B$  then  $V = A \oplus B$ , and (2) every nonzero vector  $x \in V$  lies in a unique member of  $\sigma$ . The members of  $\sigma$  are the components of the spread.

```
\begin{definition}
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collection $\sigma$ of subspaces is a \emph{spread} if (1) $A,B
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vector $x\in V$ lies in a unique member of $\sigma$. The members
of $\sigma$ are the \emph{components} of the spread.
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spread components

High-quality (but not complete!) dataset construction:

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- Examine by hand 1024 labeled entries. Only 30 annotated texts out of 1024 were incorrectly labeled. Manually corrected, to obtain a test data set of 999 labeled texts with 1552 definienda. (The rest of the dataset, not manually checked, becomes training data.)

# Fine-tuning Pre-trained Language Models for Token Classification

- ► We experimented with an out-of-the-box and general language model Roberta-base (Liu et al., 2019) and a domain-specific model cc\_math\_roberta (Mishra et al., 2023).
- ► We experimented with 1024, 2048, and 10240 samples to see the performance of the classifiers with low resources.
- To evaluate the predictions, we used the predicted tag of the first word piece of each word and regrouped the IOB2-tagged word into definienda.

### Alternative Approach: Querying GPT

### SYSTEM

You will be provided with a block of text that might define one or multiple mathematical terms. Your task is to extract the defined term(s). For example, the phrase "An interval of \_n is called new if it cannot be obtained as the grafting of two intervals" defines "new". It is possible that the sentence does not contain the defined term. You should only return me the terms that you find, separate the terms with ###.Please do not print anything else.

USER	For a bipartite graph G(U,V) with $ V $ = k $ U ,  U $ disjoint copies of K_1,k (a star) is a k-matching from U to V.
ASSISTANT	k-matching###bipartite
gpt-3.5-turbo	graph###U###V###K_1,k###disjoint copies###star
ASSISTANT gpt-4	k-matching
Add mess	age

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Conclusion 000

### **Experimental Results**

Model	GPT-3.5	GPT-4	cc_ep01	cc_ep10	Rob.
Training data	1	1	10240	10240	10240
Precision	0.1929	0.6248	0.420	0.652	0.697
Recall	0.8312	0.8821	0.473	0.743	0.794
$F_1$	0.3131	0.7315	0.442	0.692	0.742

### Conclusion

- Fine-tuned classifiers have more balanced precision and recall and much smaller cost
- GPT's answers have better recall but much poorer precision than fine-tuned models
- GPT-3.5 tends to over predict formulas and mathematical expressions, while GPT-4 shows an impressive capacity to understand mathematical texts with only one example in the prompt

### Future Work

- ► Test on a broader, more diverse, dataset of PDF papers (but if no \u00e9TEX source available, no automatic construction of a labeled dataset)
- Extract terms elsewhere in the paper to link them back to their original definition
- Improve the robustness of domain-specific language models over different NLP tasks beyond extraction of definienda

Proposed Approach

# Thank you for your attention!

### Related Work

Name	Dataset	Remarks
ArGot	mathematical	Use static word embeddings and
(Berlioz,	arXiv papers	hand-codes features.
2021)		Mathematical expressions and formu-
-		las are masked out.
Scholarphi	papers in	Use transformer-based architecture
(Head et al.,	general	syntactic features & heuristic rules.
2021)	domain	Only processes papers with LATEX
		sources.
NaturalProofs	mathematical	Extract definitions with hand-crafted
(Welleck et	papers &	rules.
al., 2021)	textbook	Definienda are not annotated.

### Results of Fine-tuning PLM

Model	cc_ep01	cc_ep10	Rob.
Extracted	2093.0	1710.8	1764.2
True positive	514.9	881.2	934.2
TP+Split Term	693.8	1056.5	1127.5
Too Long	170.2	209.1	268.8
Cut Off	522.6	405.2	326.1
Precision	0.354	0.623	0.646
Recall	0.447	0.681	0.726
F <sub>1</sub>	0.383	0.647	0.679

### Results of Fine-tuning PLM with more training data

Model	cc_ep01	cc_ep10	Rob.
Extracted	1775.2	1779.2	1770.5
True positive	540.3	972.6	1082.6
TP+Split Term	733.9	1152.5	1232
Too Long	143.5	201.3	233.7
Cut Off	509.6	438.2	274.1
Precision	0.420	0.652	0.697
Recall	0.473	0.743	0.794
F <sub>1</sub>	0.442	0.692	0.742

### Evaluation of GPT's answers

Model	GPT-3.5	GPT-4
Extracted	6867	2245
True Positive	1072	942
TP+Split Term	1315	1383
Too Long	379	595
Cut Off	656	138
Precision	0.1929	0.6248
Recall	0.8312	0.8821
$F_1$	0.3131	0.7315

#### Appendix 0000●

### References

- Luis Berlioz. ArGoT: A Glossary of Terms extracted from the arXiv. Electronic Proceedings in Theoretical Computer Science, 342:14–21, 2021.
- Andrew Head et al. Augmenting scientific papers with just-in-time, position-sensitive definitions of terms and symbols. In Proceedings of the 2021 CHI Conference on Human Factors in Computing Systems, pages 1–18, 2021.
- Sean Welleck et al. NaturalProofs: Mathematical Theorem Proving in Natural Language. In The 35th NeurIPS, Datasets and Benchmarks Track (Round 1), 2021.
- Yinhan Liu et al. *Roberta: A robustly optimized BERT pretraining approach. arXiv preprint arXiv:1907.11692*, 2019.
- Shrey Mishra, Antoine Gauquier, and Pierre Senellart. Multimodal Machine Learning for Extraction of Theorems and Proofs in the Scientific Literature. arXiv preprint arXiv:2307.09047, 2023.