

M2 Software Project

Online App for Knowledge Substantiation

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Contents

- 1 Overview
- 2 Knowledge Base Module
- 3 Cross-checking Module
- 4 Next Steps

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Overview

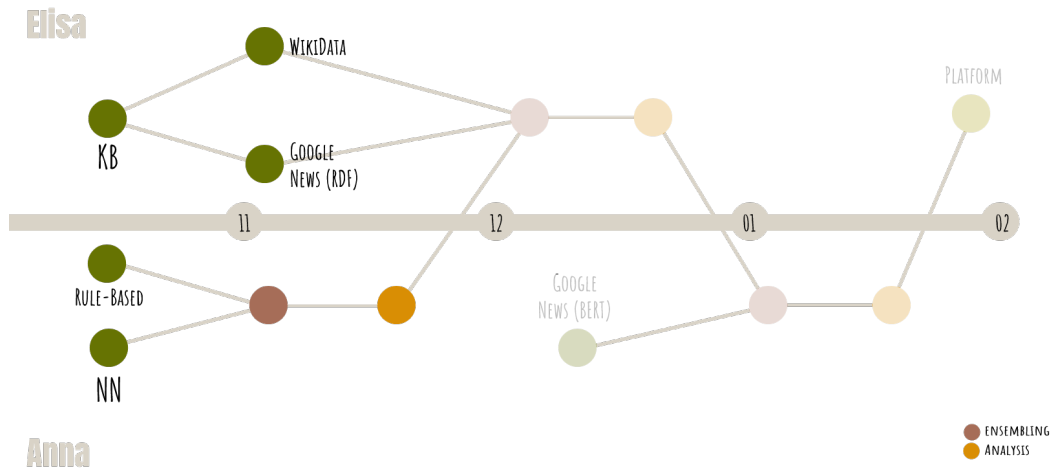


Figure: Expected Timeline

Overview

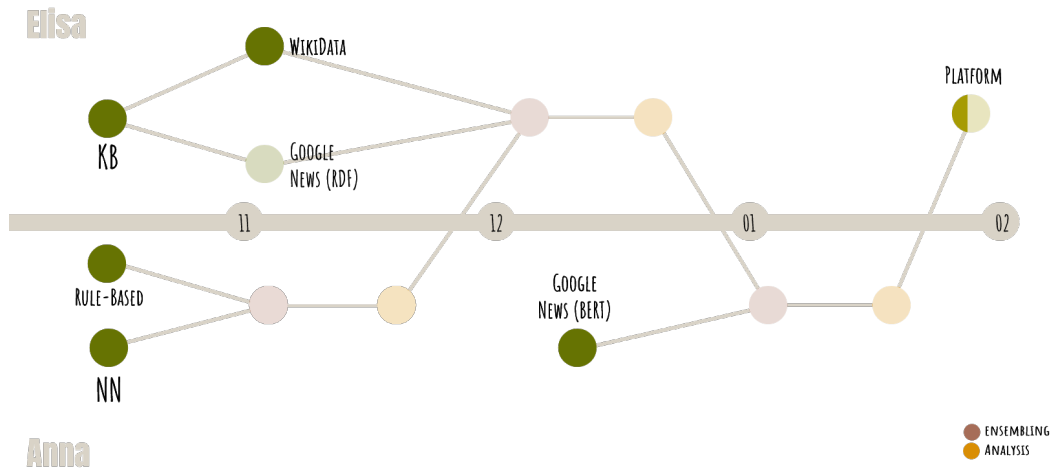


Figure: Current Timeline

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Knowledge Base

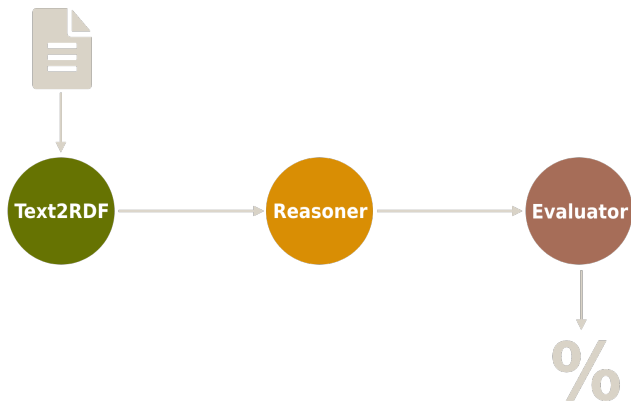


Figure: KB Pipeline

Pipeline File Structure

```
\knowledge_based_approach\  
  \data\  
  \output\  
  \src\  
    __init__.py  
    info_extractor.py  
    reasoner.py  
    util.py  
  __init__.py  
  __main__.py
```

Accessing the Module from Bash

```
$ python3 knowledge_base_approach  
--text filename.txt
```


Knowledge Base

Pipeline File Structure

\knowledge_based_approach

\data\

\output\

\src\

__init__.py


info_extractor.py

reasoner.py

util.py

__init__.py

__main__.py ----->

oaks > app > modules > knowledge_base_approach >  __main__.py > ...

```
1 def convert_to_rdf(text):
2     from src.info_extractor import extract_information
3     triples = extract_information(text)
4     return triples
5
6 def check_consistency(triples, ontology, entities):
7     from src.reasoner import get_accuracy
8     accuracy = get_accuracy()
9     return accuracy
10
11 if __name__ == '__main__':
12
13     import argparse
14
15     parser = argparse.ArgumentParser(description='knowledge-based substantiation')
16     parser.add_argument(required=True, metavar='article',
17                         help='an article to be checked')
18     args = parser.parse_args()
19
20     triples = convert_to_rdf(args)
21
22     accuracy = check_consistency(triples)
23
24     print(accuracy)
```

Text2RDF Tools

- × FRED → missing resource
- × Basic Custom → underperforming
- ✓ OpenIE [1]¹ → currently working, best performing script
- ? Neural Approaches → possible future implementation for improvement

¹OpenIE python wrapper [2]: <https://github.com/philipperemy/Stanford-OpenIE-Python>

Reasoner

- *Output*: number of inconsistent classes
- Complexity: SPARQL is PSPACE-complete
 - import .ttl entities via Virtuoso²
 - use only classes without entities

²high-performance object-relational SQL database

Evaluator

- *Output:*
 - % of inconsistent classes/entities over total
 - name of inconsistent classes
 - [other info for algo transparency]

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Cross-checking Module

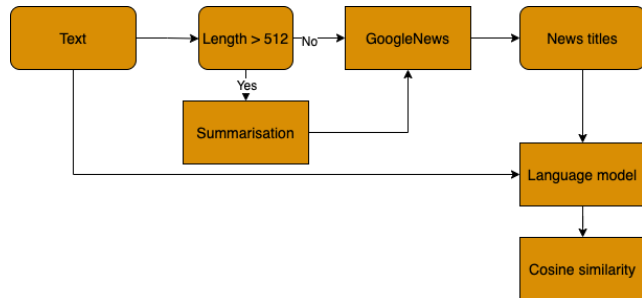


Figure: Cross-checking module pipeline

Cross-checking Module

Language model for cosine similarity

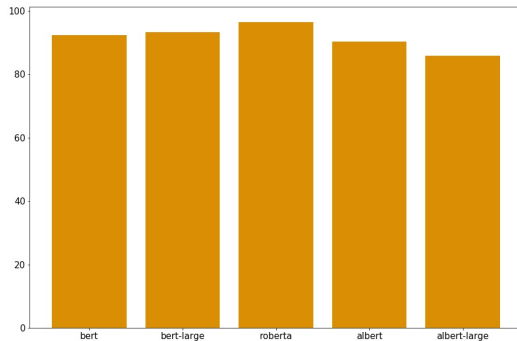


Figure: Language models comparison based on the produced similarity scores

Summarization model

- Pegasus multinews
- **Pegasus xsum**
- BigBirdPegasus large arxiv
- BART large
- BART large CNN
- **TextRank**

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Next steps

Implement and try different **ensemble methods** for feature-based, NN-based and cross-checking modules

Analyse ensembling results

Add **Google News** to the KB pipeline

Implement **neural Text2RDF approach**

Load **entities from .ttl** to a graph

Bibliography I

- [1] Lei Cui, Furu Wei, and Ming Zhou. “Neural Open Information Extraction”. In: *Proceedings of the 56th Annual Meeting of the Association for Computational Linguistics (Volume 2: Short Papers)*. Melbourne, Australia: Association for Computational Linguistics, July 2018, pp. 407–413. DOI: 10.18653/v1/P18-2065. URL: <https://aclanthology.org/P18-2065>.
- [2] Philippe Remy. *Python wrapper for Stanford OpenIE*. <https://github.com/philipperemy/Stanford-OpenIE-Python>. 2020.