

Data Mining for Entity Relationship Associations

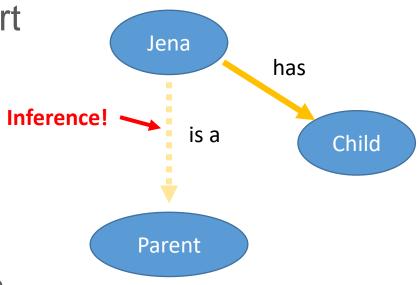
School of Engineering and Applied Science
Department of Computer Science CSCI 6443— Data Mining

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Problem Definition

- Many chatbots are a combination of expert systems and machine learning.
- A knowledge base is often used as the "brain" of the chatbot due to its ability to perform inference.
- Traditionally knowledge bases perform inference based on inference rules, which are brittle and don't scale well.



IF <subject> has Child THEN <subject> is a Parent



Problem Definition Continued

- Unsupervised learning of entity relationships is difficult and supervised learning datasets are costly to create.
- Performance is subjective and language dependent.
- State-of-the-art NLP algorithms struggle to perform Relationship Extraction (RE) with the precision and recall of a person.

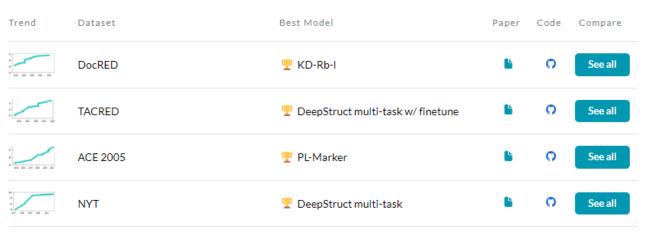
Java was originally developed by James Gosling at Sun Microsystems

https://corenlp.run/

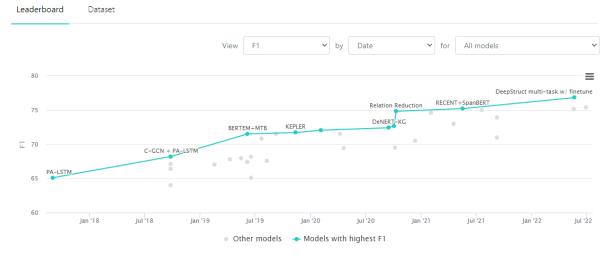


Related Work

- Datasets (not all are free):
 - <u>https://paperswithcode.com/datasets?task=</u> relation-extraction
- Papers:
 - https://paperswithcode.com/task/relation-extraction#papers-list
- Notable Algorithms Types:
 - OpenIE
 - Convolutional Neural Network (CNN)
 - Sequence based networks (RNN, LSTM, GRU, Transformers)



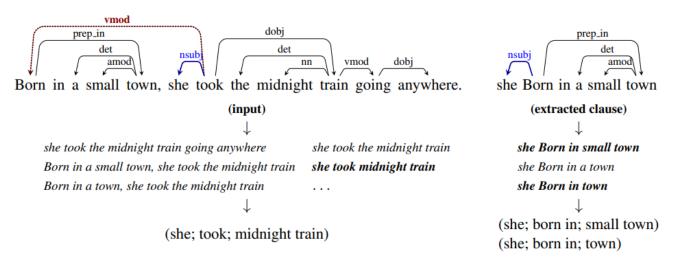
Relation Extraction on TACRED



https://paperswithcode.com/task/relation-extraction



- Methods of doing RE :
 - OpenIE
 - Greedy search on dependency tree
 - Goal is to reduce sentence, to utterance, to triple



https://nlp.stanford.edu/pubs/2015angeli-openie.pdf

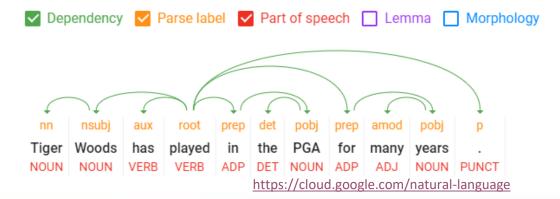


- Methods of doing RE:
 - Named Entity Extraction (NER) + Classification of relations between entities using neural networks
 - Text and/or Part of Speech (POS) tags are input features along with entities
 - Graph Convolutional Neural Networks (GCN)
 - Create graph from dependencies
 - GCN learns features and structures in graph that are associated to training data

 $\langle Tiger Woods \rangle_1$ has played in the $\langle PGA \rangle_2$ for many years.

NER Example

https://cloud.google.com/natural-language



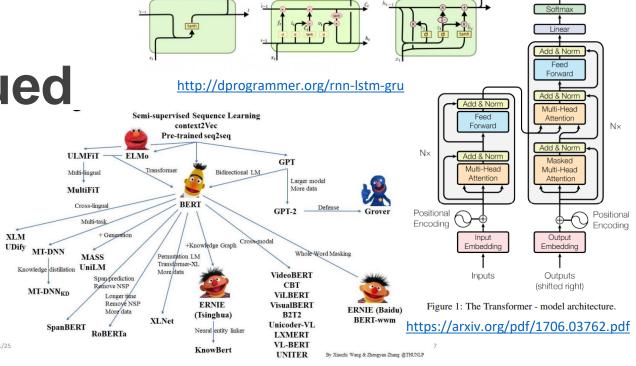


- Methods of doing RE :
 - Applying language generation/translation techniques
 - RNN, LSTM, GRU, Transformers (Seq2Seq (BERT, ERNIE, GPT, BART))



https://translate.google.com/





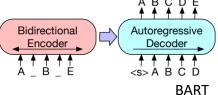
LSTM

GRU

Output Probabilities

https://www.microsoft.com/enus/research/uploads/prod/2021/06/Pre-training-Models-Xu-Tan.pdf

RNN

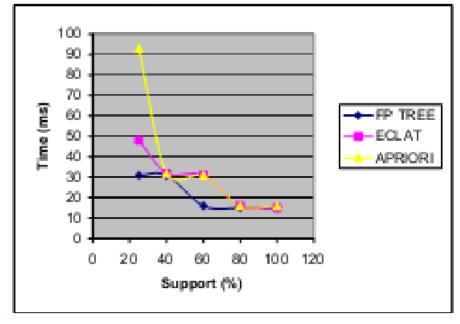


https://arxiv.org/pdf/1906.07510v8.pdf



- After RE, associations between relations occurs.
- Leading association algorithms are:
 - Apriori
 - FP Growth
 - Eclat

Figure 1. Comparison of Apriori, Eclat and FP Growth algorithm on artificial dataset.



https://research.ijcaonline.org/volume69/number25/pxc3888502.pdf



Proposed Approach

- Text already mined and transforming to JSON files
- Zipped text moved to AWS cloud environment
- This allows servers to access as needed with higher download speeds
- Download data
- Unzip
- Split into manageable data message sizes
- Publish text to queue

- High performance queue
- Allows consuming service to be parallel by ensuring only one consumer gets each message
- Perform Relation Extraction (RE) to convert unstructured data to structured data
- Write to DB

- Clean, Transform, and Filter Structured Data
- DB Scans and Queries
- Apply FP Growth







Amazon Simple Storage Service (Amazon S3)



Amazon Elastic Compute Cloud (Amazon EC2)



Amazon Simple Queue Service (Amazon SQS)



Amazon Elastic Compute Cloud (Amazon EC2)

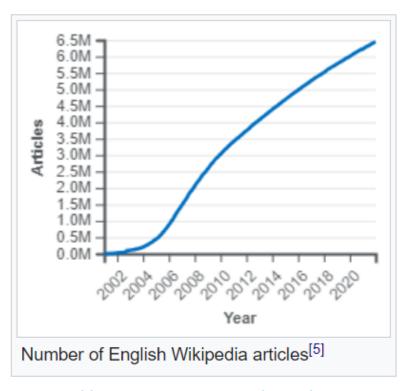


Amazon Neptune



Selected Dataset

- Wikipedia
- 6.5M+ English articles as of 2022
- 10TB of data as of 2015
 - https://dumps.wikimedia.org/enwiki/latest/
 - https://www.kaggle.com/datasets/ltcmdrdata/plain-text-wikipedia-202011
 - https://github.com/daveshap/PlainTextWikipedia



https://en.wikipedia.org/wiki/WikipediaSize of Wikipedia



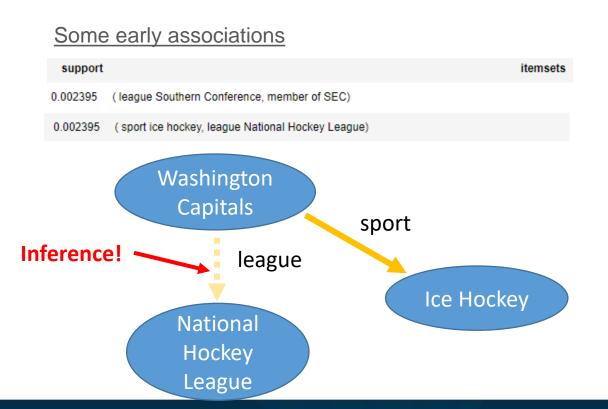
Date: October 2022 ▼



Current Progress

Estimated Total

\$13.50



Name 	Type	▽ Created		▽	Mes	sages available	∇
Articles.fifo	FIFO	10/20/2022, 17:04:48 EDT 100158			158		
Name	▼	Instance ID		Instance state	9 ▼	Instance type	▽
Article Queue Worker	i	i-091ebaa6da955fdaa		○ Terminate	d @ Q	t2.large	
Relation Extraction Worker	i	i-0ab1850889ef24422		⊘ Running	@ Q	c5.xlarge	
Relation Extraction Worker	i	i-0af5a562cb22ab67e		⊘ Running	@ Q	c5.xlarge	
Relation Extraction Worker	i	i-0d9e48489f40dcf72		⊘ Running	@ Q	c5.xlarge	
Relation Extraction Worker	i	i-0bf6d65705a462b48		⊘ Running	@ Q	c5.xlarge	
Relation Extraction Worker	i	i-0915356135ede0d01		⊘ Running	@ Q	c5.xlarge	
Relation Extraction Worker	i	i-0ca912cea7d0f9f5b		⊘ Running	@ Q	c5.xlarge	
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Relation Extraction Worker	i	i-06771b02375ff6f29		⊘ Running	@ Q	c5.xlarge	
Relation Extraction Worker	i	i-0ca674c2d2d2d77f6		⊘ Running	@ Q	c5.xlarge	
Relation Extraction Worker	i	i-09daf15fce53d4b36		⊘ Running	⊕Q	c5.xlarge	



a. (John_E_Blaha birthDate 1942_08_26) (John_E_Blaha birthPlace San_Antonio) (John_E_Blaha occupation Fighter_pilot)

b. John E Blaha, born in San Antonio on 1942-08-26, worked as a fighter pilot

Future Work

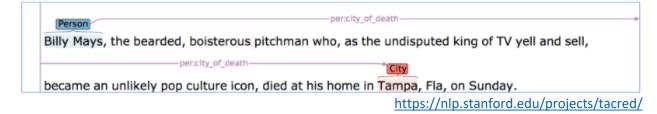
https://webnlg-challenge.loria.fr/

- Generate a new set of training data
- Start with generated triples by determining how often entities and verbs/predicates appear in the same sentence
- Use transformer to generate sentences from generated triple
- This would give the algorithm a more direct triple to sentence training

dbp:yearpro • 1996 (xsd:integer)

https://dbpedia.org/page/Tiger_Woods

Woods turned professional in 1996 https://en.wikipedia.org/wiki/Tiger_Woods





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