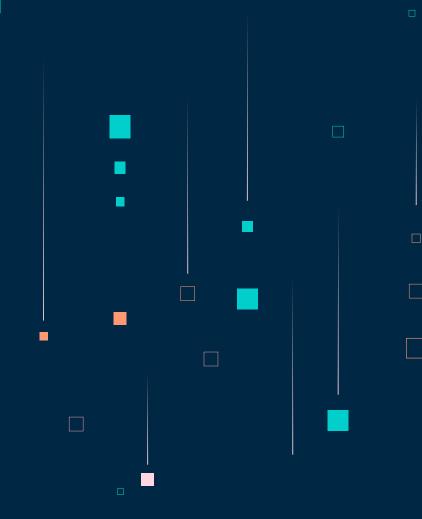
# INFORMATION RETRIEVAL PROJECT

Denis FESTA Salima MAMMA



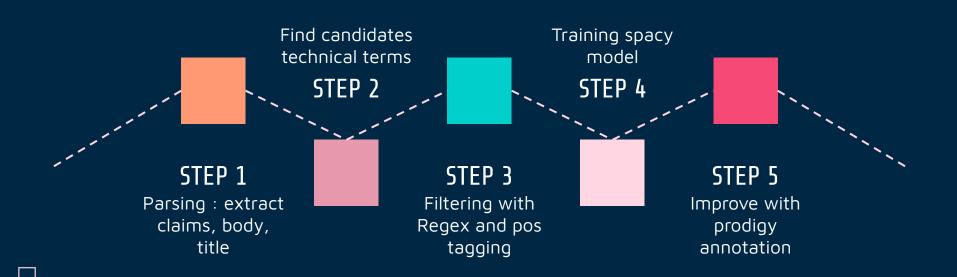
# **CONTENT OF THE PRESENTATION:**

- Our Technical terms extraction approach
- Our relation extractions approach
- Researches and further improvements

#### DOMAIN DESCRIPTION:

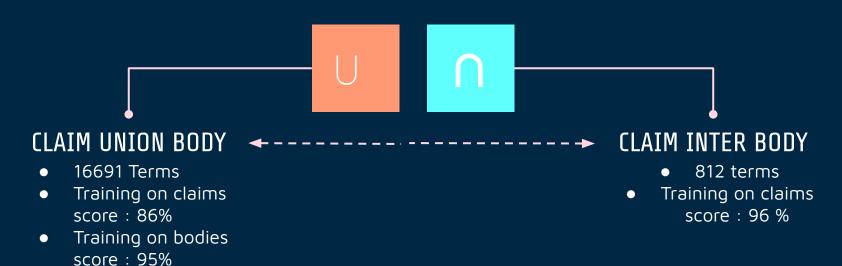
- Domain: H04N Pictorial communication
- Important topics covered : Electronic acquisition, scanning, reproduction of images, conversion of optical information into electrical signals, transmission, handling pictorial communication signals .. ect

# TECHNICAL TERM EXTRACTION PROCESS:



## TECHNICAL TERM EXTRACTION VOCABULARY:

#### DIFFERENT VOCABULARIES



#### PRODIGY ANNOTATIONS:

- Training: 26 sentences
- Gold Patent (Test): 107
- Performances on the gold patent:

#### before Prodigy:

- 'p': 0.6682832201745877
- 'r': 1.0
- 'f': 0.8011627906976745

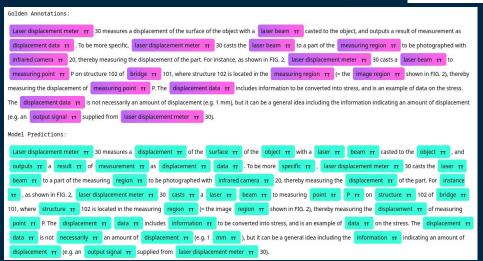
#### with Prodigy:

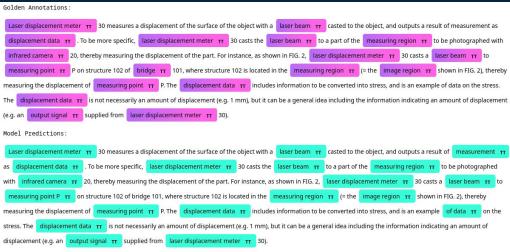
- -'p': 0.879948914431673
- -'r': 1.0
- -'f': 0.936141304347826



# EXAMPLE OF ANNOTATED TERMS IN TEXT:

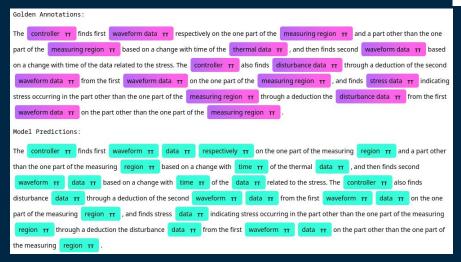
#### Before Prodigy

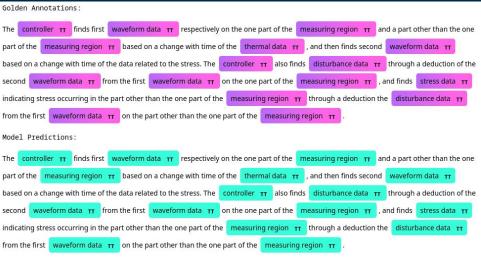




#### With Prodigy

# EXAMPLE OF ANNOTATED TERMS IN TEXT:





#### SOME IMPROVEMENTS TO DO:

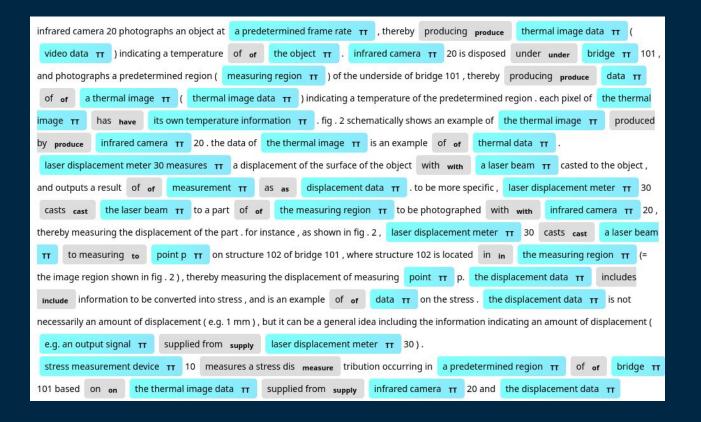
- Training on more patents with prodigy
- Handling special cases like hyphens
- Improve detection of compound terms ( 3 4 ngrams )



## RELATION EXTRACTION PIPELINE:

Get noun chunks overlapping with detected TT Extract predicates (verbs and prepositions only ) Filter relations (similarity score, co-occurence) Classification of relations into categories

#### RELATIONS EXTRACTED BEFORE FILTERING:



#### RELATIONS FILTERING

#### **EMBEDDING**

Apply Bert Embedding for "source" and "target" (all-mnpet-base-v2)

Compute cosine similarity

Filter relations using a threshold

#### CO-OCCURRENCE

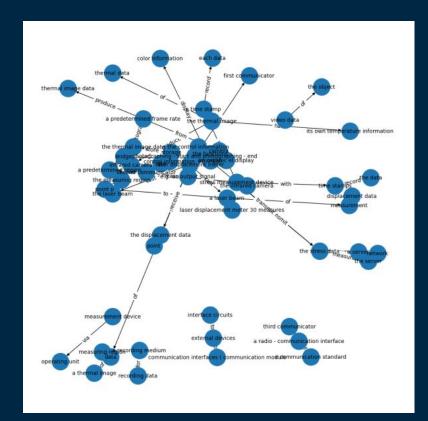
Compute the co-occurrence matrix of the technical terms

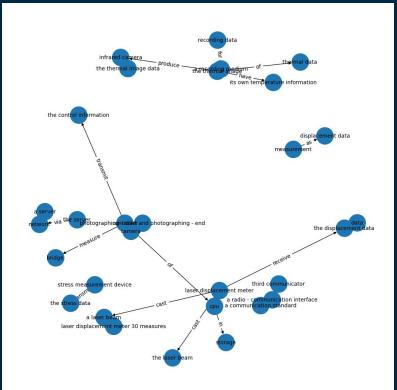
Filter the relations using a threshold

Combine it with the embedding



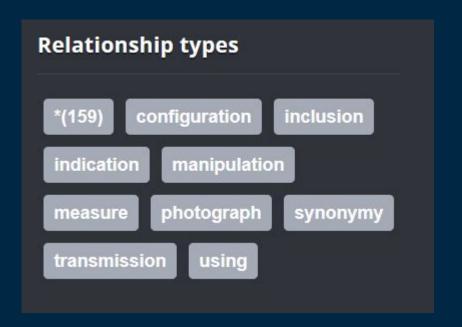
# RELATIONS BEFORE AND AFTER FILTERING:



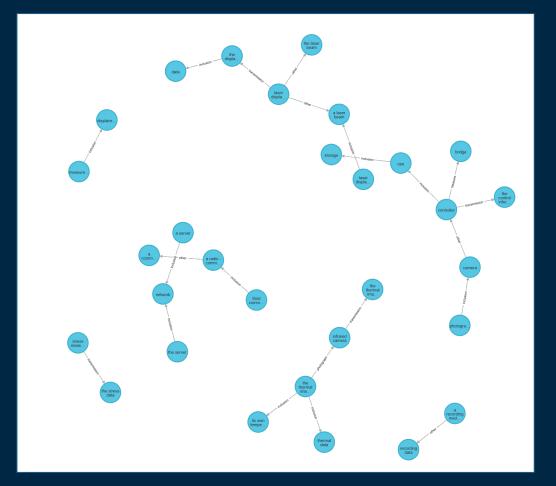


#### RELATIONS CLASSIFICATIONS:

- Done manually
- Used NEO4J to visualize and query using cypher in order to get more understanding on the semantic of relationships
- Remove irrelevant relations ( "other" category)



# NEO4J GRAPH:





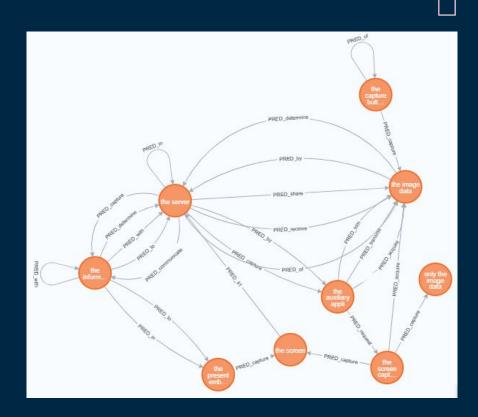


#### **SOME ISSUES**

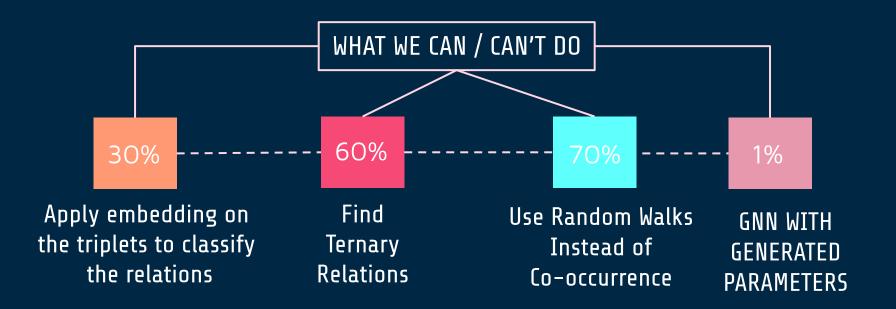
- One predicate can have different meanings and describe different relationships.
- Lemmatization leads to the loss of direction in relations
- We may lose some important relations with the co-occurrence threshold.

#### We tried and didn't work:

 Coreference resolution; due to the style of writing in patents



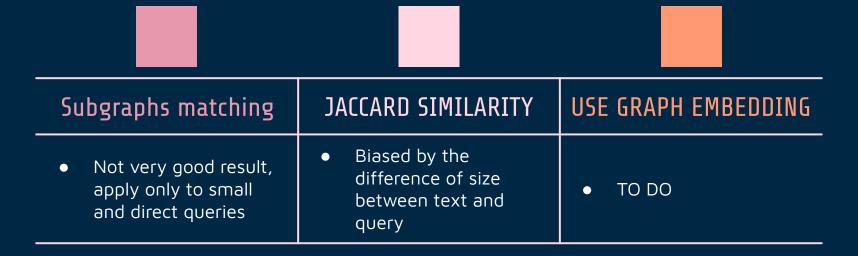
### **SOME IMPROVEMENTS:**



#### GP - GNN:

- Article Name : Graph Neural Networks with Generated Parameters for Relation Extraction
- How it works :
- 1- Encoding Module:
- -> Combine positional embeddings with word embeddings (Bi-LSTM)
- -> Extract representation of entity pairs (LSTM, GRU)
- -> Generate edge parameters using MLP to produce a transition matrix
- 2- Propagation Module: Learn Node representations layer by layer by taking into account the transition matrix
- 3- Classification Module : Make predictions about the relationship between entities

# SOME TRIES ON THE QUERYING SYSTEM





# CONCLUSION





#### References

- Graph Neural Networks with Generated Parameters for Relation Extraction <a href="https://paperswithcode.com/paper/graph-neural-networks-with-generated">https://paperswithcode.com/paper/graph-neural-networks-with-generated</a>
- An Analysis of Relation Extraction within Sentences from Wet Lab Protocols

  <a href="https://cse.cs.ucsb.edu/sites/default/files/publications/an\_analysis\_of\_relation\_ext\_raction\_within\_sentences\_from\_wet\_lab\_protocols.pdf">https://cse.cs.ucsb.edu/sites/default/files/publications/an\_analysis\_of\_relation\_ext\_raction\_within\_sentences\_from\_wet\_lab\_protocols.pdf</a>
- Word2Vec and Semantic Similarity using spacy | NLP spacy Series | Part 7 –
   Data Science Duniya (ashutoshtripathi.com)