Encoder-Attention based Automatic Term Recognition

Sampritha H. Manjunath, John P. McCrae

Insight SFI Research Centre for Data Analytics Data Science Institute, National University of Ireland Galway

sampritha.manjunath@insight-centre.org

Introduction

- Creation of a binary classifier using deep learning approach to identify terms and non-terms.
- Wikipedia titles are used as positive examples and weak n-grams (of length 1-4) are used as negative examples for • Automatic term recognition
- Transformer's BERT is used for word embedding.
- Encode, Embed, Attend and Predict (EEAP) . approach is used to design the neural network for classification

Methodology

- The project can mainly be divided into four major tasks: data pre-processing, candidate term extraction, training data preparation, and modeling.
- Extract n-grams (of length 1-4) and filter them based • on scores obtained by TF-IDF. This gives us the candidate terms.
- Train the EEAP model along with BERT embeddings lacksquareto predict whether the given n-gram is a term or non-term













label

True



Confusion matrix for ATR classification

EA-ATR model is evaluated against the ATR4S[1] model.

Comparison - EA-ATR(A) vs ATR4S(B)					EA-ATR model	
Dataset	A precision	B precision	A accuracy	B accuracy	F1-score	Recall
GENIA	0.8045	0.7760	60%	24%	0.7460	0.6955
Krapivin	0.6345	0.4279	62%	42%	0.7612	0.9511

Conclusion

- The model performs 28% better than the ATR4S [1] base model.
- This method has the potential to be used as a multilingual model as it does not require any annotations.
- The experiments are a clear example of a deep learning model being applied to NLP tasks by reducing the repetitive computational requirement for each dataset and to extract automatic terms more precisely.

[1] Nikita Astrakhantsev. ATR4S: toolkit with state-of-the-art automatic terms recognition methods in Scala. Language Resources and Evaluation, 52(3):853–872, 2018.

