

Dual-FOFE-net Neural Models for Entity Linking with PageRank



Feng Wei, Uyen Trang Nguyen, Hui Jiang Dept. of EECS, York University, Canada {fwei,utn,hj}@cse.yorku.ca

Simple and Efficient Neural Linking Nets



This paper presents a simple and computationally efficient approach for entity linking, which achieves state-of-the-art performance on both TAC2016 and TAC2017 task datasets.

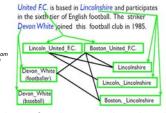
PageRank based Distillation



PageRank based distillation not only enhances the candidate coverage, but also speeds up the whole model.

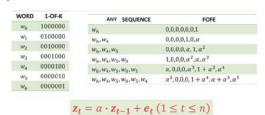
Limitations of conventional candidate generation module:

- Too many candidates
- Too much noise



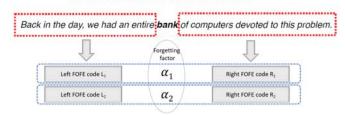
We rank the candidates of each mention based on their numbers of outbound links to all the recognized mentions in the same document, and keep the top τ candidates for each mention.

Simple and Efficient Dual-FOFE-net

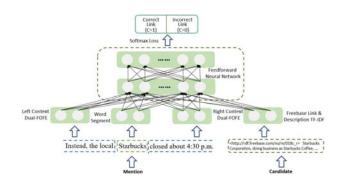


- Fixed-size ordinally-forgetting encoding (FOFE) codes, as an alternative sequence embedding representations, can almost uniquely encode any variable-length sequence of words into a fixed-size representation without losing any information.
- A smaller forgetting factor is helpful for representing the positional information of all words in the sequence; a larger forgetting factor is useful for maintaining the long-term dependency of past context.

Encode the context with dual-FOFE:



- Given a mention, its left and right contexts are encoded by dual-FOFE, and projected into a dense vector.
- The principle idea of dual-FOFE is to generate augmented FOFE encoding codes by concatenating two FOFE codes using two different forgetting factors.



We propose a simple and computationally efficient feed-forward neural ranking model, compared with recurrent neural networks or convolutional neural networks, by using dual-FOFE codes as input.

Results on TAC-KBP Task Datasets

		[17]		Our proposed	
		TAC Rank 1		models	
		NERLC	CEAFmC	NERLC	CEAFmC
Triling	gual	64.7	66.0	65.9	67.1
Engli	$_{ m sh}$	66.6	67.6	67.7	69.0
Chine	ese	65.0	70.2	66.4	70.7
Span	ish	61.6	63.5	62.5	64.4

On the TAC2016 dataset, our model outperforms the best system by 1.2% (NERLC F_1) and 1.1% (CEAFmC F_1), respectively, in terms of the overall trilingual entity linking performance.

	[33]		Our proposed	
	TAC Rank 1		models	
	NERLC	CEAFmC	NERLC	CEAFmC
Trilingual	67.8	70.5	68.0	70.9
English	66.8	68.8	67.2	69.3
Chinese	71.0	73.2	71.6	72.4
Spanish	65.0	68.9	64.8	69.3

On the TAC2017 dataset, our model achieves higher F_1 scores, by 0.2% (NERLC F_1) and 0.4% (CEAFmC F_1), respectively, compared to the best system in terms of the overall trilingual entity linking performance.