Extraction of Drug-Drug Interactions from Biomedical Texts

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- Software
- 2 Learners
- 3 Experiments
 - Word Vectors
 - Learning curves
- 4 Results
- Conclusions



Software

- Base language: Python
- XML parser: ElementTree
- Pandas
- Numpy
- Gensim: Word2Vec
- NLTK
 - a. Word tokenizer
 - b. Part-of-Speech tagger
 - c. English stemmer
- Scikit Learn
 - a. Data preprocessing
 - b. Pipelines
 - c. SVM
- Keras: ANN



Method

- Task9.1 2 approaches
 - a. Embeddings: stemming, POS, puctuation removal, lowercase
 - b. Hand-crafted features:
 - Is the word capitalized?
 - Is the part of speech NN?
 - Previous word, following word word2vec
 - If there are some trigger words surrounding this word?
 - Does the word have more consonants than vocals?
- Task9.2
 - Hierarchical classification (binary + multiclass)

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Learners

- Training/Validation split: $(90\%, 10\%) \sim (133k, 14k)$ words
- Metrics:
- Feed-forward Neural Network
 - Architecture: [vector_size, 512, 256, 128, 3]
 - Activations: ReLU, Softmax
 - Dropout fraction: 0.5
 - Objective: Categorical cross-entropy
 - Solver: Adam
 - Number of parameters: 175363
- 2. Support Vector Machine
 - Kernel: radial basis function
 - C = 1
 - $\gamma = auto$
 - $\alpha = 0.001$



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Word Vectors

Vector size	Micro F1	Macro F1
10	0.9481	0.5041
20	0.9517	0.5435
50	0.9495	0.5215
100	0.9499	0.5232
200	0.9479	0.4855

- Vector size
- Vector type
- Preprocessing

Table: F1-score for different word-vector sizes.

Туре	Micro f1	Macro f1		
original	0.9499	0.5098		
stem	0.9599	0.6144		
original + PoS	0.9481	0.5105		
stem + PoS	0.9517	0.5435		

Table: F1-score for different word-vector types.

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Word Vectors

- Vector size
- Vector type
- Preprocessing

Preprocessing	Micro f1	Macro f1		
No	0.9599	0.6144		
Minmax(0,1)	0.9515	0.5207		
Standardize	0.9581	0.5897		

Table: F1-score for different preprocessing.

Loss/Accuracy

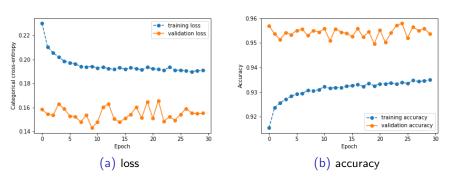


Figure: Learning curves

Micro F1/Precision/Recall score

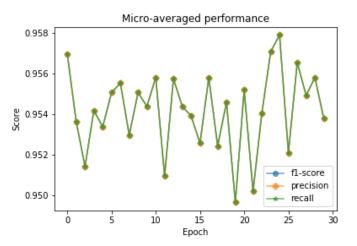


Figure: Micro scores.



Macro/Weighted F1/Precision/Recall score

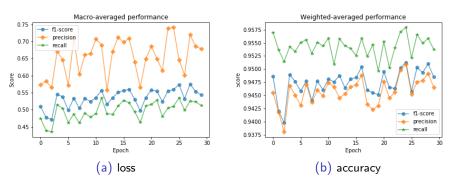


Figure: Macro and Weighted scores.



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Results

	Exact		Partial			
	Precision	Recall	F1	Precision	Recall	F1
DrugBank	0.61	0.43	0.5	0.61	0.5	0.55
MedLine	0.51	0.29	0.37	0.51	0.35	0.41
Both	0.56	0.35	0.43	0.56	0.41	0.48

Table: Results Task1 on gold test dataset.

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Conclusions

- Word vector size (too small, too big)
- stemming and PoS improved performance used individually, but not in conjuction
- Poor results (especially the recall)
 - Poor features?
 - Limited model capacity/expressivess?

