RoST-DL: Romanian Short Text Classification

Integrating Deep Learning for NLP in Romanian Psychology

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Introduction

We propose a Deep Learning based system for classifying Romanian short sentences in the context of a psychotherapy of anxiety and depression study (**PsiTAD***).

Three datasets of answers tested:

- emotions
- thoughts
- behaviors

Goals and Methods Used

Goals

- 1. Initial analysis of the **fastText** embeddings on the Romanian datasets.
- 2. Implementation and comparison of 3 model architectures
- 3. Test the capabilities of transfer learning for this problem: embeddings trained on large corpus: small datasets(~500 instances each)

Methods Used

- pre-trained Romanian embeddings using fastText
- PCA and averaging of all the sentences / class - inital exploration and data analysis
- Pandas, Keras, Tensorflow

Models:

- CNNs for sentence classification [2]
- GRUs
- BiLSTMs

Heavy usage of *dropout* is employed to avoid overfitting on the small datasets.

Data Analysis

Reasoning: good word embeddings should show some polarisation between the classes of the datasets.

Due to the n-grams for the word embeddings, even **OOV** words can be used with confidence (e.g. *disconfort*, which should be *discomfort*).

1. Top 5 most similar words for the target word (emotions)

Rank	Input word					
	depresie	disconfort	tristete			
1	depresia	discomfort	tristetea			
2	deprimare	inconfort	neliniste			
3	anxietate	stres	tristetii			
4	depresive	anxietate	dezamagir e			
5	neajutorare	iritare	deznadejde			
5 neajutorare iritare deznadejde Averaging all the sentences						

2. Closest words to the averages of all sentences (emotions)

Rank	Label				
	1	0			
1	nesiguranta	cred			
2	neliniste	dar			
3	incordare	simt			
4	teama	spun			
5	stresata	chiar			

Ideal: show closeness to some domain-specific words in the positive examples, when computing the avg. sentence.

Implementation & Results

- CNN
 - o varying filter sizes, 1D conv layers
- dropout of 60%
- o ADAM optimizer. Params: 0.6, 0.99.
- GRU
 - two layers, the first creating a more abstract representation
 - 50% dropout
 - o ADAM optimizer. Params: 0.9, 0.99.
- BiLSTM
 - Bidirectonal initial layer, 2nd layer uses the features produced by the 1st
 - 50% dropout and ADAM as above.

[Table 3] presents the classifier performance on all 3 datasets. Both LSTMs and GRUs have similar results, with LSTMs slightly better in all cases except "thoughts".

CNNs provide more consistent results across runs, when compared to recurrent models.

Conclusions

- Some insights into the quality of pretrained word embeddings for the Romanian language
- Thorough comparison between 3 real datasets from the Faculty of Psychology, using 3 different architectures
- Future roll-out as an automatic labelling system, easing the job of psychologists.

s citipioyed .		3. Classifier performances (30 runs each)					
agitatie	Dataset	Metric	CNN	GRU	BiLSTM		
tristete oboseala, agitație tristete vinovatie	Emotions	Std. Dev. Acc	1.97	3.31	3.90		
plans tristete putina teama, nesiguranta		Max. Acc.	88.09	92.85	95.23		
vinovatie, tristete antiqipata agitata, anxioasa dezamagire, suparare,		Mean. Acc.	83.9	86.66	89.88		
anexios , frica ,tahicardıe ,ameteli ,vedere ir தசிந் <mark>துகள்ளின் இதிருந்துக்கு இது ம</mark> ுறு அது இது இது இது இது இது இது இது இது இது இ	iac, tra	Std. Dev. Acc	3.88	5.24	6.42		
stress, anxietate, nervozitate, o anumita nervozit anxioasa tensionata, tematoare	re, ametita Behavors	Max. Acc.	90.47	88.09	92.85		
presjune pe creier - ir spatele fruntii nod in stomac; (este o senzatie de neliniste ce ra familie cu sotia dezamagire p m-am simtit fara pofta de nimirazាទាំខ្សាំងទូរ៉ុងទៀត់ង្កាំ	e	Mean. Acc.	86.19	82.26	81.71		
mi-am rezervat ziua de duminica pentru a t prea mare presiune si stres pe capul meu (la	recupera	Std. Dev. Acc	1.30	2.05	2.52		
am o stare confuza, parca sunt ametit simt de multe ori ca ma integpa in zona inimii	Thoughts	Max. Acc.	83.33	83.33	83.33		
simt de multe ori ca ma inteapa in zona inimii devin stresat și ma mananca pielea de pe frunte. cu familia la o piscina acoperita la care eram num raceste tot corpul. simt o teama co		Mean. Acc.	80.95	79.66	79.64		

Visualizing

PCA is performed on the sentences (obtained from averaging the embeddings of each one) [Figure A]

References

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