Part-Of-Speech Taggers Features In French Learner Texts

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Abstract—This paper presents investigation result of the comparative analysis of TreeTagger, CoreNLP and SpaCy as part-of-speech taggers for processing texts in French. Acting in the framework of a Learner Texts Corpus Creation Project a group of researchers from the Institute of Foreign Languages together with the Institute of Mathematics and Information Technologies of Petrozavodsk State University analyzed the above mentioned tagging tools mainly being focused on defining the most effective one.

I. INTRODUCTION

Corpus technologies are actively developing and find their application in various fields, from linguistic researches to their use in the foreign languages teaching techniques. [1], [2]

The most important component of the corpus in any language is a tagset, which allows the morphological analyzer (morphologizer) automatically process linguistic objects in a formalized form [3].

The main objective of our research is to select the most suitable tagger for processing a text in French. To tackle this problem we:

- analyzed 30 learner texts in French using the most famous and affordable taggers TreeTagger, CoreNLP and SpaCy.

- highlighted the main errors (separation of words, highlighting parts of speech, etc.) made by each of the specified taggers.

-compared the tagging results and highlight the tagger with the least number of errors

- selected the most appropriate tagger to use in work with the corpus of learner texts in French using theanalysis results

II. MAIN PART

During the research work we analyzed 30 learner texts of different types such as essays, motivation letters, topics, article evaluation using TreeTagger, CoreNLP and SpaCy taggers. All the texts under analysis were written by the students of the Institute of Foreign Languages of Petrozavodsk State University. The analysis of the texts made it possible to obtain each tagger tools comprehensive evaluation. Anna Zhestkova Petrozavodsk State University Petrozavodsk, Russia zhest1806anna@gmail.com

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TreeTagger [4] is the only tool with tags in French among the three part-of-speech taggers under consideration which proves to be its main advantage comparing to other tools. Considering this part-of-speech tagger as a possible version of a tool for working with texts, it should be taken into account that it is not very efficient from the point of view of visual perception.

The reasons for that can be mentioned as the follows: first, the tagger font color is black, second, the tags are located directly above the words making the text less readable so you have to read the whole text while paying a special attention to the superscripts, and finally, the body of the text is not split into separate sentences, which also complicates the process of checking the tagger.

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Fig.1. Screenshot of the tagged text

CoreNLP - one of the main advantages of this tagger is its visual presentation. So due to this peculiarity all the sentences in the text beginning on a new line are also numbered. This

COMPARISON OF PART-OF-SPEECH TAGGERS

feature simplifies the word search. Moreover, this immediately allows the reader to spot/identify/ the sentence types presented in the text. At that tags for tokens are highlighted in colour, which makes it more convenient to define parts of speech.

Nevertheless it should be noted that the texts tagging is done in English making the analysis on the one hand more simple, but imprecise at the same time.

1	82° BETINGAN BEGNI 9555 2017 629 BOOM EXECT ACP 1000M 830 BOO EXECT BEDINGON EXECUTION BANGT De nos jours II existe beaucoup de livres , de genres littéraires différents , d'auteurs ,
2	전전적 16788 207 8008 8/30 207 203 207 1058 201 8008 201 8008 201 8008 201 800 828 2020 C'est pourquoi c'est assez facile de choisir quelque livre intéressant pour le line .
10	영양에 사진은 실입 EDANCTESSM ESEM ESEM ESEM EXEM NO. AND NO.
4	ACC RATE REAL TELL MACE RAVE REAL REAL REAL REAL REAL REAL REAL REA
100	DED 100,000 BOOMD FUNCTI DEVID 19500 19500 19500 19500 19500 19500 19500 19500 19500 19500 19500 19500 19500 1 Tout d'abord , je me détends quand je m'immerge dans le monde de littérature .
6	HENN PROM VERSE ALCOLOR AND
7	ARE ADV EXANCIDED 1800 MI ADV MARSH ARE MARSH BARDI 1800 MI COOMU DED 1800 MI EAACID ARE MARCH ARE MARCH AND COOMU ARE EXANCID De plus , la lecture m'aide à élargir mes hortsons et mon vocabulaire , à parier plus logique et mieux .
8	2016 ENACE ACC MORE DEL MOZA ACC EST MOZA 202 MOZA ENACE 202 ENACE 202 MOZA ENACE 202 MOZA ENACE 202 ENACE 202 ENACE 202 MOZA ENAC
9	A28 MOMBAR A28 DEL MOMBA HA28 DEL MOMBA PARCI 25500 HA36 MEMBA SOCIAL DEL MOMBA HA36 DEL MOMBA HA36 DEL MOMBA HA En dépt de ces arguments pour la lecture , certains sont convaincus que la littérature est le passe-temps inutile .
10	REGIO INERSI SCICILU RECIN INFORM INFORM INFO INFORMATION INFORMATIONI INFORMATICI INFORMATICI INFORMATICI INFORMATICI INFORMATICI I INFORMATICA INFORMATICA INFORMATII INFORMATICA INFORMATICA INFORMATI
11	EED NOOM MEEN SCONDEED NOOM VEEN ACCIED NOOM ARE DED NOOM PROVINSEN AND PEON VEEN AND ECONUS AND COOLU VEEN BED NOOM PAYED Ces gens estiment que la lecture fait de le mai à les yeux c'est pourquoi cela peut être dangereux et endommage notre santé .
12	REF BEED MOUN RAP DED MOUN FANCT FROM HAN WERB SCHUFFON VERB RAV REF MOUN SCHUF AUCH MOUN Malgré ces visions contre la lecture , je suis persuadée qu'il fait plus de biens que mai

Fig.2. Screenshot of the tagged text

SpaCy

When working with this tagger, you need to take into account the fact that its list of tags in English is the same as that one of the CoreNLP tagger on the one hand, at the same time the total error amount is bigger than the number in the above described part- of-speech taggers.

ADP DET NOUN PRON VERB ADV ADP NOUN ADP ADJ NOUN De nos jours il existe beaucoup de livres PUNCT de genres littéraires ADJ NOUN NOUN ADJ ADJ ADY NOUN AUX ADV ADJ différents PUNCT d' auteurs PUNCT C' est pourquoi c' est assez facile ADP NOUN ADV NOUN ADJ ADP DET VERB UNCT ADV ADP ADV de choisi quelque livre intéressant pour le lire PUNCT Tout de même PUNCT jil y a un peu de gens qui sont intéressés à la lecture ADV ADP PRON AUX DET ADV ADP NOUN PRON AUX VERB ADP DET ADJ il y a un peu de gens qui sont intéressés à la lecture ADV ADP PRON DET ADJ AUX DET ADV DP DET NOUN verb ADV VERB ADV VERB ADV PUNCT PRON NOUN SCONJ PRON préférées PUNCT Tout d' abord PUNCT je me détends quand je NOUN NOUN ADD DET ADJ ADP ADJ DUNCT CAL se semble DET NOUN ADJ EDT ADJ ADP ADJ DUNCT CLA se semble DET NOUN ADJ EDT ADJ ADP ADJ DUNCT CLA se semble DET NOUN ADJ CONJ ADD ADJ DUNCT CLA se semble QUENCT La lecture PUNCT De plus PUNCT CLA se semble OCCONJ ADV CCONJ DET ADJ ADP VERB NUM ADJ ADP au thérapie psychologique PUNCT DE plus PUNCT C A parler plus logique CCONJ ADV ADV ADV ADP ADJ ADP VERB ADV ADJ élargir mes horisons et mon vocabulaire PUNCT AD PARD NOUN ADP DET ADJ ADJ DET NOUN ADD DET ADJ ADD PUNCT EN dépit de ces NOUN ADP DET NOUN PUNCT Enfin PUNCT en lisant j' apprends des faits de ADJ PUNCT LENGT ADJ ADP VERB NOUN ADP DET ADJ arguments pour la lecture PUNCT in dépit de ces NOUN ADP DET NOUN PUNCT En dépit de ces NOUN ADP DET NOUN NOUN DET NOUN SCONJ PEN NOUN ADP DET d' activités plus plaisants PUNCT Intéressants et éducatifs PUNCT Ces NOUN ADV NOUN NOUN DET NOUN NOUN DET NOUN NOUN AUX ADV gens estiment que la lecture fait du mal aux yeux c' est pourquoi PRON VERB AUX ADJ CONJ VERB DET NOUN NOUN AUX ADV gens estiment que la lecture FUNCT Intéressants et éducatifs PUNCT Ces NOUN ADP DET NOUN NOUN DET NOUN NOUN AUX ADV gens estiment que la lecture fait du mal aux yeux c' est pourquoi PRON VERB AUX ADJ CONJ VERB DET NOUN NOUN AUX ADV gens estiment que la lecture FUNCT MES SCONJ PRON NOUN AUX ADV visions contre la lecture PUNCT per

Fig.3. Screenshot of the tagged text

It should be also noted that the number of tokens in the same text varied depending on the used part-of-speech tagger. The data is presented in diagram 1.

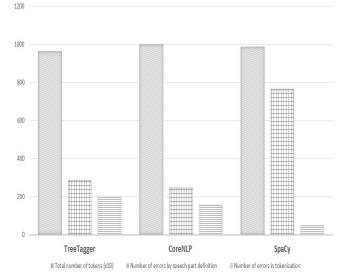


Fig.3. Diagram I

So we found out the indicator of the total number of tokens in the TreeTagger was the lowest for the errors in tokenization are often encountered due to abbreviated articles and pronouns which are written with an apostrophe followed by a word ending in a vowel and which are not separated into certain tokens. As you can see in Table II, TreeTagger, compared to CoreNLP and SpaCy, makes the most of word separating errors (2.1% versus 1.6% and 0.5%).

CoreNLP, on the contrary, splits parts of merged articles into different tokens, therefore, the total number of tokens is much higher than in other taggers. In SpaCy, the percentage of errors in the division of words with the total number of tokens is the smallest, based on it, we can conclude that this indicator is more accurate in this tagger. Thus, if tokenization is the prevailing goal of the text analysis, then SpaCy can be effective in the process of choosing a part-of-speech tagger.

TABLE II .POS TAGGERS COMPARISON

Percent /indicators (of the total number of tokens)	TreeTagger	CoreNLP	SpaCy
Percent of errors by speech part definition	3%	2,5%	7,8%
Percent of errors in tokenization	2,1%	1,6%	0,5%
Total percent of errors	5,1%	4,1%	8,3%

Therefore, with this error percent criterion (determining parts of speech in 3 taggers) TreeTagger and CoreNLP have

similar indicators: 3% and 2.5%. Whereas in SpaCy error percent is 7.8%.

The lowest percent of the total number of errors and the number of tokens, 4.1% is characteristic for CoreNLP tagger, as for the highest one it is for SpaCy namely 8,3%; and as to TreeTagger this percent is 5.1%. Consequently, the TreeTagger and CoreNLP part-of-speech taggers are more accurate in defining part of speech and allocating tokens.

We have compiled the following diagram III and TABLE III with marked erorrs using one of the texts (essays) analyzed through all three taggers as an example.

Wor d	Errors by speech part definition	Correction	Total number	Errors in tokeniz ation	Total number				
Treetagger:									
d'auteurs	adj	prp+nom	12	d'auteurs	8				
c'est	nam	pro:per+ver:pres		c'est(x2)	_				
c'est	ver:subi	pro:per+ver:pres		d'abord					
interresant	ver:ppre	adj		m'immerge					
tire	nom	ver:infini		m'aide					
m'immerge	nom	pro:per+ver:pres		j'apprends					
mlaide	nom	pro:per+ver:pres		qu'il					
(en) lisant	adj	ver:ppre		d'activités					
j'apprends	nom	pro:per+ver:pres							
qu'il	nom	kon+pro:per							
d'activités	ver:pper	prp+nom							
c'est	nom	pro:per+ver:pres							
CoreNLP:									
108	det	pron	13	au therapie	4				
ď	det	adp		des faits					
C	propn	pron		du mai					
est X4	aux	verb		aux yeux					
c'	noun	pron							
quelque	det	adj							
intéressé	verb	adj							
Ĵ	det	pron							
apprends	noun	verb							
convaincus	verb	adj							
sont X2	aux	verb			_				
notre	det	pron							
persuadé	verb	adj							
SpaCy									
genres	adj	noun	32	-	0				
littéraires	noun	adj			_				
d'	num	adp							
C'	adj	pron							
C'(x2)	noun	pron							
choisir	noun	verb							
quelque	adv	pron							
(pour) le	det								
(lire)		pron							
Tout	adv	pron							
le cture	adj	noun							
Quant	adv	adp							
littérature(x3)	adj	noun							
ď	verb	adp							
abord	aðv	noun							
détends	noun	verb							
m'(x2)	noun	pron							
immerge	noun	verb							
monde	adj	noun							
semble	adj	verb			_				

TABLE III

III. CONCLUSION

Due to our research analysis of TreeTagger, CoreNLP and SpaCy as part-of-speech taggers for processing texts in French we chose the CoreNLP tagger as the most effective tool in use for our project. CoreNLP proved to have almost the same number of errors in determining parts of speech, moreover, it made fewer errors in separating words and it turned out to be more illustrative at work tan other tools.

Finally, it can be summarized as follows: using TreeTagger, CoreNLP and SpaCy universal taggers it is necessary to take into account that every one of them has its merits and shortcomings in terms of various kinds of errors. So when choosing a tagger the latter must always be carefully considered.

In 2020 the Department of German and French Languages of the Institute of Foreign Languages and the Center of Artificial Intelligence of Petrozavodsk State University initiative was the starting point for learner texts corpus creation project in German and French languages with the subsequent development of various virtual applications based on it. This project on the one hand, provided teachers of a foreign language with efficient tools of checking a large number of written works, on the other hand, it gave the students the opportunity to improve their writing skills while completing student assignments [5].

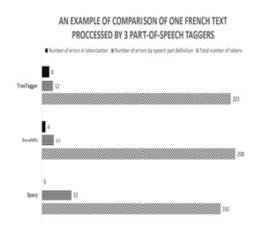


Fig.4. Diagram III

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