### ОБРОБЛЕННЯ СИГНАЛІВ І ЗОБРАЖЕНЬ ТА РОЗПІЗНАВАННЯ ОБРАЗІВ

Signal/Image Processing and Pattern Recognition

## ПРАЦІ \* PROCEEDINGS

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### Processing of Direct Speech in Belarusian Texts with NooJ

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#### Abstract

paper focuses on processing of direct speech in selarusian electronic texts for the purpose of audiobook cation. Usually, for creation of an audiobook, synthesis with one voice is used. It gives us perspective on the kelihood of making text-to-speech synthesis many-voiced, making audiobooks more approximate to the presentation of characters' unique speech features.

#### 1. Introduction

work on direct speech processing was started in 2013 by group of researchers from the United Institute of mormatics Problems of the National Academy of Sciences of selarus. The main goals of the study include the identification all cues in electronic texts and the identification and processing of cues with authors' text insertions for the aims of adiobook creation.

Actually, European scientists have already developed algorithms for character identification and automatic determination of the character's role with the help of NooJ matactic grammars [1]. As for the Slavic languages, the work Croatian scientists on the direct speech identification should be noted [2], though they do not consider the problem gender identification. Such programs for audiobook reating as MP3book2005 and AUDIOBOOK are also eveloped in this direction. They have special inbuilt units for beical analysis of dialogues, which can provide the marking of the characters' and author's words in a dialogical text. In AUDIOBOOK steps were taken to read dialogues in maracter, but the program does not cover all the cases. It mores the cue structures with more than one insertion of the mithor's words. In addition, it is not able to identify the meder of a character on such indicator as a "verb + masculine men" combination in the author's words:

-Трэба напісаць "яць", - адказвае вучань.

- We should write "яць", - the pupil (he) answers.)

Thus, the tasks confronting the authors include the agorithm development for direct speech processing to immalize as much syntactic structures of dialogical text as assible, and to identify automatically the gender of a immacter by the insertions of the author's words in the direct areacter. We also discuss the use of the developed algorithms in a TTS system.

# 2. The development of automated algorithm for direct speech and author's text identification

At the first stage we have selected texts in Belarusian and mentified all the paragraphs with direct speech. The found

paragraphs were separated according to the characters' gender and all the cues with author's text insertions were also marked. Then the cues were analyzed to define the syntactical direct speech structures and to detect gender indicators (such as past tense verbs and nouns with gender attributes) in author's text insertions.

The following syntactic structures were revealed in direct speech:

Direct speech apart from the author's text:

-C(!|!!|!!!|?|?!|...|.).

Direct speech followed by the author's text:

-C(,|!|!!|!!!|?|?!|...|.)-A(...|.)

Direct speech with one or more insertions of the author's text: -C(,|!|!!|!!!!??!!...|.) - A(,|...|.|:|.) - C(,|!|!!!!!!!??!!!...|.) - C(,|!|!!!!!!!??!!!...|.)).

The structures contain the following annotations: C – the words of a character (speaker), A - the author's text, brackets (,) – the beginning and the end of a choice set of punctuation marks, | - symbol or (separation of punctuation marks in a choice set).

On the basis of these findings the algorithm for direct speech identification was developed. The main idea is that only those paragraphs are taken into consideration that begin with a dash. After a dash being found, the following elements of the paragraph are alternatively defined as the character's words and the author's words. The algorithm's complexity consists in indicating of a set of characters that separate the character's part from the author's part.

Let us describe the developed algorithm:

1. Process the next paragraph TT of a text T. If TT = Ø, then go to Step 14, otherwise go to Step 2.

2. If TT begins with a dash, then go to Step 3,

otherwise - Step 1.

3. If a sequence of any number of SSw1 set's elements is found next, and at the end of which there is any element of SSp3 set, then go to Step 11, otherwise – Step 4.

4. If a sequence of any number of SSw1 set's elements with SSp2 set's elements placed between them (not several elements in succession) is found next, and at the end of which there is any element of SSp3 set, then go to Step 11, otherwise – Step 5.

5. If a sequence, starting with any element of SSp1 followed by any number of SSw1 set's elements, is found next, and at the end of which there is any element of SSp3 set,

then go to Step 11, otherwise - Step 6.

6. If a sequence, starting with any element of SSp1 followed by any number of SSw1 set's elements with SSp2 set's elements placed between them (not several elements in succession), is found next, and at the end of which there is any element of SSp3 set, then go to Step 11, otherwise – Step 7.

7. If a sequence of any number of SSw1 set's elements with SSp1 set's elements placed between or after SSw1 set's elements (not several elements in succession) is found next, and at the end of which there is any element of SSp3 set, then go to Step 11, otherwise - Step 8.

8. If a sequence, starting with any element of SSp1 followed by any number of SSw1 set's elements with SSp1 set's elements placed between or after SSw1 set's elements (not several elements in succession), is found next, and at the end of which there is any element of SSp3 set, then go to Step 11, otherwise - Step 9.

9. If a sequence of any number of SSw1 set's elements with SSp1 set's elements placed between or after SSw1 set's elements (not several elements in succession) and SSp2 set's elements placed between SSw1 set's elements (not several elements in succession) is found next, and at the end of which there is any element of SSp3 set, then go to Step 11, otherwise - Step 10.

10. If a sequence, starting with any element of SSp1 followed by any number of SSw1 set's elements with SSp1 set's elements placed between or after SSw1 set's elements (not several elements in succession) and SSp2 set's elements placed between SSw1 set's elements (not several elements in succession), is found next, and at the end of which there is any element of SSp3 set, then go to Step 11, otherwise - Step 1.

11. If a dash is found next, then go to Step 12, otherwise conclude that TT belongs to direct speech and go to Step 1.

12. If a sequence of any number of SAw1 set's elements is found next, and at the end of which there is any element of SAp2 set, then go to Step 14, otherwise - Step 13.

13. If a sequence of any number of SAw1 set's elements with SAp1 set's elements placed before, between or/and after them is found next, and at the end of which there is any element of SAp2 set, then go to Step 14, otherwise -Step 1.

14. If a dash is found next, then go to Step 4, otherwise conclude that TT belongs to direct speech and go to

15. The end of the algorithm.

Within the given algorithm, Steps 3-10 correspond to the speaker's words identification and Steps 12-13 correspond to the author's words identification. For the description of direct speech processing, the following sets of elements were used: SSw1, SSp1, SSp2, SSp3. Respectively, to describe the author's word's identification the SAw1, SAp1 i SAp2 sets were used.

Let us describe the elements of the sets being used. SSw1 and SAw1 sets include the rules of possible speaker's and author's words formation and have the following filling:  $SSw1 = SAw1 = {\langle WF \rangle, \langle NB \rangle, \langle WF \rangle \#-\#\langle WF \rangle}, where$ WF is any word form, and NB is any number. SSp1, SSp2, SSp3, SAp1 and SAp2 sets represent a range of different punctuation marks and are filled as follows:

 $SSp1 = \{ \wedge'' /, /, \backslash'' /, / ... /'' /, \wedge'' /, / ... \backslash'' /, /... /'' /, \wedge'' ... /, \wedge ... /'' /, ... /'' /, \text{Normalization} \}$ ^" -/, ^"?/, /?!\"/, /!\", \"/, /!\" \"/, ^")/, ^" (/, /.../, /.../, /"/);; /!!!/, /)./, /),/, \tau:/, /(/, /)/, /) -/, /) -/};

SSp3 = {1.1, 1,1, 1!1, 1?1, 1...1, 1?!1, 1...1, 1!..1, 1?..1, 1).1, \(\Lambda\)"?1, /...\"/, /...\"/, \\"./, \\"./};

 $SAp1 = \{/./, /,/, /.../, \land:/, /(/, /)/\};$ 

 $SAp2 = \{/./, /./, /.../, /)./, \wedge:/\}.$ 

The developed algorithm was implemented with the help of a computational linguistic tool NooJ, notably with its Belarusian module. NooJ allows to develop syntactical and morphological grammars and to test them on a large number of texts. Thus, the authors have developed a NooJ syntactic grammar DS All for automated identification of all paragraphs containing direct speech (Fig. 1). Its main parts graphs Speaker (Figure 2) and Author (Figure 3) - serve to identify respectively the character's words and the author's words.

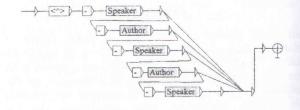


Figure 1: A general view of the DS All grammar

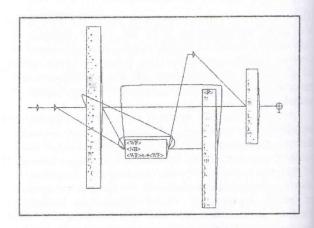


Figure 2: Syntactic subgraph Speaker of the DS\_All grammar

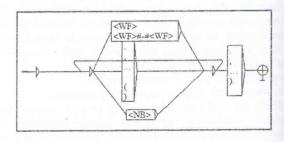


Figure 3: Syntactic subgraph Author of the DS\_All grammar

The resulting grammar can be applied sequentially to any Belarusian electronic text through NooJ, and the program function Locate Pattern allows to view the results of the grammar applied in the form of concordance (Fig. 4).

Before	Seq.	After
гўчына.	- Вось бачыце, шкада толькі, што вы ад нас далёка, а то б	- A xle
ськаю?	<ul> <li>А хіба тут няма каму гэтай справай заняцца? Вось мой к</li> </ul>	Сахан
ияліся.	- Не, я ўжо зусім страціў там ласку, дзякаваць Богу.	Айцец
заў ён.	- Апрача таго, я чуў, што ў яв жаніх ёсць ужо.	- LJI 141
ць ужо.	- Ці мала на свеце дурняў, - зноў дадаў а. Кірыл.	Матуи
a csaši	<ul> <li>Ну, то што? Хіба жаніхам свіней не падкладаюць?</li> </ul>	- Fara
цаюць?	- Гэта было б не па-хрысціянску.	- Зато

Figure 4: The results of applying the DS\_All grammar to a Belarusian text

## 3. The development of automated algorithms for character's gender identification from the author's text

In the Belarusian language singular past tense verbs may have gender attributes, for example, nanpaeiÿ 'he corrected' and exasana 'she said'. As such verbs may often occur in author's commentaries to direct speech, as well as some nouns having gender attributes, they may serve as gender indicators and be considered suitable for character's gender identification.

For the purpose of gender identification we have modified the algorithm mentioned above, namely in Steps 12-13 we used one more set with gender indicators. Then on the basis of the grammar DS\_All two separate grammars were developed – one for masculine gender identification (DS\_M), and one for feminine gender identification (DS\_F) (see [3] for more details). For this purpose, we worked further the graph Author and added resources for gender identification (Figure 5). In the Figure 5, one can see the subgraph VERBSmasculine. It includes the list of masculine verbs, which were selected at the stage of manual marking of texts in the Belarusian and Russian languages. The similar list of merbs was created within the subgraph VERBSfemenine for feminine gender identification.

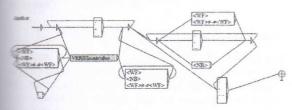


Figure 5: The subgraph Author, DS\_M

In order to use the outputs of the grammars in TTS system SAPI 5.1, it is necessary to adapt a text to a SAPI TTS XML Therefore, to select an appropriate speech synthesizer, syntactic grammar should provide annotations of the blowing kind:

OICE Required="name=[a synthesizer's name in TTS

A text for synthesis...

«VOICE».

Thus, in the Figure 6 one can see, that the speech special special special and AlesiaBel will be respectively special to the character's words (Speaker) and to the author's special (Author).

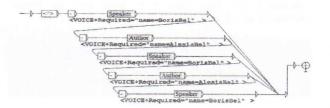


Figure 6: The DS\_All grammar after being adapted for SAPI 5.1

For example, after being processed by the DS\_M and DS\_F grammars, a dialogical text will be annotated as in the Figure 7. A female voice AlesiaBel is applied to the author's words, and voices ElenaBel and BorisBel are used for the female and male characters' words. Such annotation allows to input texts into the TTS system SAPI 5.1, where the indicated voices switch over are automatically (Figure 8).

<VOICE Required="name=ElenaBel">— Бацька вады, </VOICE> <VOICE Required="name=AlexiaBel">
- жолеми сказала Жайла.</VOICE>
<VOICE Required="name=AlexiaBel">
- Бацька вод, </VOICE> <VOICE Required="name=AlexiaBel">
- паправід Алесь.</VOICE>
<VOICE Required="name=BlenaBel">>— Вось так і Днапро пачанаецца кодзе.</VOICE>
<VOICE Required="name=ElenaBel">>— Жоная вада, </VOICE> <VOICE Required="name=AlexiaBel">
- сказала Лия.</VOICE>
- пава апусцілься на колені і зламала пальчыкный крыштальную паворхию.

Figure 7: The sentences from the Table 1 after being annotated with VoiceXML tags

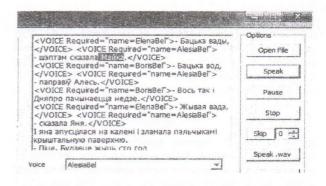


Figure 8: The speech synthesis of the annotated sentences

#### 4. Multi-coloured marking of a text

One more application for the described annotation is the multi-coloured marking of a text for visual presentation of the author's words and of the female and male characters' words. Such marking may be used by an editor to quickly analyze direct speech in a text and to select an optimal number of speech synthesizers or speakers.

To provide the multi-coloured marking, the authors have developed the VoiceXmlToColorReplacer software. The program process VoiceXML-files and allows converting VoiceXML-tags of speech synthesizers into HTML-tags with different styles of direct speech visual presentation.

After a text passes through the VoiceXmlToColorReplacer software, the character's cues and the author's insertions are marked with different colours: namely, author's words (AliesiaBel) are in black, the male

character's cues (BorisBel) - in blue, and the female character's cues (ElenaBel) - in red (Figure 9).

ралях, виды, колупан сказала Майха к балыка вод, паправіў Алесь - Вось тяк і Дняпро пачынавцца недзе. «Жышя ядла» сказаля Якл. 1 яна апусціпася на калені і аламала пальчыкамі крыштальную паверхню. - Піце Будзеца жыць сто год...

Figure 9: A fragment of a text with multi-coloured marking of direct speech

#### 5. Evaluation

Initially, a training text corpus with 106 000 word usages was used in developing the grammar. Then, in the process of testing, the experts have collected a test text corpus with 23 867 word usages. According to the performance evaluations, the total number of cues in the corpus was equal to 481 (N=481). Among them 233 cues include the author's text insertions, where 165 cues belong to male characters, 68 cues belong to female characters.

The quantity of all cues found by the algorithm DS\_All (be) is L=462; the number of those which have been correctly processed is M=461. The calculations have showed the following results for DS\_All (be): precision  $\approx$  99,5 %, recall  $\approx$  95,8 %, and F-score  $\approx$  97,6 %.

The quantity of all cues found by the algorithm DS\_M (be) is L=145; the number of those which have been correctly processed is M=143. The calculations have showed the following results for DS\_M (be): precision  $\approx$  98,6 %, recall  $\approx$  86,6 %, and F-score  $\approx$  92,2 %.

The quantity of all cues found by the algorithm DS\_F (be) is L=58; the number of those which have been correctly processed is M=67. The calculations have showed the following results for DS\_F (be): precision  $\approx$  98,2 %, recall  $\approx$  83,8 %, and F-score  $\approx$  90,4 %.

#### 6. Conclusion

In the process of character gender identification on the author's text insertions, rather good operating results were obtained. Moreover, the developed algorithm showed itself suitable for the use in combination with a TTS system and later may be applied in audiobook creation with reflecting the unique speech characteristics of characters.

However, text processing at the paragraph level is not sufficient for character gender identification in all cues. There are a lot of cues without author's text insertions, that is why now we face the task of gender identification directly from the character's words, and the most significant challenge is to provide text-level gender identification through the analysis of the text going before and after the cues. Moreover, further work needs to be done to create dictionary resources with verbs-indicators identifying, to expand the punctuation base (dash and quotation types, etc) and the test corporas.

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