Services for Text Simplification and Analysis

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Abstract

We present a language technology service for web editors' work on making texts easier to understand, including tools for text complexity analysis, text simplification and text summarization. We also present a text analysis service focusing on measures of text complexity.

1 Introduction

Our research on digital inclusion has resulted in a variety of tools for making texts easier to understand, including an automatic text summarizer (Smith and Jönsson, 2011a; Smith and Jönsson, 2011b), syntactic (Rennes and Jönsson, 2015) and lexical (Keskisärkkä and Jönsson, 2013; Johansson and Rennes, 2016) simplification, and a number of measures of text complexity (Falkenjack et al., 2013; Falkenjack and Jönsson, 2014). In the project TECST[1] (Text Complexity and Simplification Toolkit) we developed a web service that integrated all these tools and made them easily available for producers of easy-to-read texts. The service can also be used by end users, i.e. anyone wanting to make a text easier to comprehend. Similar systems exist for other languages, such as Spanish (Saggion et al., 2015), Brazilian Portuguese (Scarton et al., 2010) and English (Lee et al., 2016).

The set of text complexity measures provided by the TECST service is limited to a subset of features that is meant to be easily understandable by non-linguists. The complete set of features has instead been made available in the separate service SCREAM[2]. All services presented in this paper can also be accessed through the SAPIS REST API (Fahlborg and Rennes, 2016). In what follows we will first present the features included in SCREAM and continue with the tools included in TECST.

2 Text analysis

All tools in TECST use Stagger (Östling, 2013) for tagging and MaltParser (Nivre et al., 2007) for parsing. We have also implemented support for the OpenNLP part-of-speech tagger (Morton et al., 2005) and older versions of MaltParser.

3 SCREAM – Text complexity features

SCREAM currently comprises 117 features.

3.1 Shallow features

Shallow text features are features that can be extracted after tokenization by simply counting words and characters. They include:

MWLC Mean word length calculated as the average number of characters per word.

MWLS Mean word length calculated as the average number of syllables per word. The number of syllables is approximated by counting the number of vowels.

MSL Mean sentence length calculated as the average number of words per sentence.

3.2 Lexical features

Our lexical features are based on categorical word frequencies extracted after lemmatization and calculated using the basic Swedish vocabulary SweVoc (Heimann Mühlenbock, 2013). SweVoc is somewhat comparable to the list used in the classic Dale-Chall formula for English (Dale and Chall, 1949). Though developed for similar purposes, special sub-categories have been added (of which three are specifically considered). The following frequencies are calculated:

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[1] `http://www.ida.liu.se/projects/scream/webapp/`

[2] `http://www.ida.liu.se/projects/scream/webapp/analysis/index.html`

SweVocC SweVoc lemmas fundamental for communication (category C).

SweVocD SweVoc lemmas for everyday use (category D).

SweVocH SweVoc other highly frequent lemmas (category H).

SweVocT Unique, per lemma, SweVoc words (all categories, including some not mentioned above) per sentence.

3.3 Morpho-syntactic features

The morpho-syntactic features concern a morphology based analysis of text. The analysis relies on previously part-of-speech annotated text, which is investigated with regard to the following features:

UnigramPOS Unigram probabilities for 26 different parts-of-speech tags in the document, i.e. the ratio of each part-of-speech, on a per token basis, as individual attributes. Such a unigram language model based on part-of-speech, and similar metrics, has shown to be a relevant feature for readability assessment for English (Heilman et al., 2007; Petersen, 2007; Dell'Orletta et al., 2011). The tag set used is collected from the Stockholm-Umeå Corpus (Ejerhed et al., 2006).

RatioContent The ratio of content words (nouns, verbs, adjectives and adverbs), on a per token basis, in the text. Such a metric has been used by for instance Alusio et al. (2010).

3.4 Syntactic features

These features are estimable after syntactic parsing of the text. The dependency based features consist of:

ADDD The average dependency distance in the document on a per dependent basis. A longer average dependency distance could indicate a more complex text (Liu, 2008).

ADDS The average dependency distance in the document on a per sentence basis. A longer average total dependency distance per sentence could indicate a more complex text (Liu, 2008).

RD The ratio of right dependencies to total number of dependencies in the document. A high ratio of right dependencies could indicate a more complex text.

SD The average sentence depth. Sentences with deeper dependency trees could be indicative of a more complex text in the same way as phrase grammar trees have been shown to be (Petersen and Ostendorf, 2009).

Dependency type tag ratio Unigram probabilities for the dependency type tags resulting from the dependency parsing, on a per token basis, as individual parameters. This is viewed as a single feature but is represented by 63 parameters. These parameters make up a unigram language model and is comparable to the phrase type rate based on phrase grammar parsing used in earlier research (Nenkova et al., 2010).

VR The ratio of sentences with a verbal root, that is, the ratio of sentences where the root word is a verb to the total number of sentences (Dell'Orletta et al., 2011).

AVA The average arity of verbs in the document, calculated as the average number of dependants per verb (Dell'Orletta et al., 2011).

UVA The ratio of verbs with an arity of 0-7 as distinct features (Dell'Orletta et al., 2011). This is viewed as a single feature but is represented by 8 parameters.

TPC The average number of tokens per clause in the document. This is related to the shallow feature average number of tokens per sentence.

PreM The average number of nominal pre-modifiers per sentence.

PostM The average number of nominal post-modifiers per sentence.

PC The average number of prepositional complements per sentence in the document.

TokensPerClause The average number of tokens per clause in the document. This is related to the shallow feature average number of tokens per sentence.

PreModifiers The average number of nominal pre-modifiers per sentence.

PostModifiers The average number of nominal post-modifiers per sentence.

PrepComp The average number of prepositional complements per sentence in the document.

3.5 Text quality metrics

The three most used traditional text quality metrics used to measure readability for Swedish are:

LIX Läsbarhetsindex, readability index. Ratio of words longer than 6 characters coupled with average sentence length.

OVIX Ordvariationsindex, word variation index, which is essentially a reformulation of type-token ratio less sensitive to text length.

NR Nominal ratio, the ratio of nominal word, used to measure formality of text rather than readability, however, this is traditionally assumed to correlate to readability.

4 TECST

TECST consists of a subset of the features from SCREAM, the text simplifier STILLETT, and the text summarizer FRIENDLYREADER.

4.1 STILLETT

STILLETT is a rule-based tool for automatic text simplification in Swedish. StilLett was originally developed as an extension of CogFlux (Rybing et al., 2010) that included a set of text rewriting operations (Decker, 2003). The first version of STILLETT (Rennes and Jönsson, 2015) was extended to support additional rules; rewriting from passive to active tense, quotation inversion, rearrangement to straight word order, and sentence split. Due to the inefficiency of phrase based parsers, a new version of STILLETT was developed, now relying on dependencies, providing faster simplification. We are currently working on methods for the automatic extraction of simplification operations based on an aligned corpus of simplified and regular texts (Rennes and Jönsson, 2016; Albertsson et al., 2016). The automatically harvested rules will eventually be included in addition to the existing rule sets.

4.2 FRIENDLYREADER

FRIENDLYREADER (Smith and Jönsson, 2011a; Smith and Jönsson, 2011b) is the automatic text summarizer used in TECSTthat extracts the most important sentences in a text based on distributional semantics. It uses a word space model, in this case Random Indexing (RI) (Hassel, 2007;

Hassel, 2011) with pre-trained word vectors. Furthermore, to handle long sentences with many words, the mean document vector is subtracted from each of the sentence's word vectors before summarizing the vectors (Higgins and Burstein, 2007). FRIENDLYREADER does not directly use a vector distance metric to select sentences, instead it uses the Weighted PageRank algorithm to rank the sentences (Chatterjee and Mohan, 2007). In this case each vertex depicts a unit of text and the edges between the units represent a connection between the corresponding text units, c.f. TextRank (Mihalcea, 2004). Thus, the importance of a vertex within a graph considers global information from the entire graph, not only the local context of the vertices, as ranks are recursively computed so that the rank of a vertex depends on all the vertices' ranks.

5 SAPIS

SAPIS[3] (StilLett SCREAM API Service) is a back-end solution providing the calculation of text complexity features (SCREAM) and the application of simplification operations (STILLETT) on a remote server. SAPIS is able to present simplification feedback on a sentence level by identifying sentences where any of the rules in STILLETTis applicable. A textual feedback is returned for each sentence that matches any of the patterns given in the simplification rule sets.

SAPIS also provides simple access to part-of-speech tagging and dependency parsing.

6 Conclusions

We have presented a service that integrates a variety of tools aiming to make texts easier to understand. Current work focuses on corpus collection for STILLETT, and interaction design to improve usability and make the measures of text complexity easier to interpret.

Acknowledgments

References

Sarah Albertsson, Evelina Rennes, and Arne Jönsson. 2016. Similarity-based alignment of monolingual

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[3]http://www.ida.liu.se/projects/stillett/Publications/SAPIS_User_Manual.pdf

corpora for text simplification. In *Coling 2016 Workshop on Computational Linguistics for Linguistic Complexity (CL4LC), Osaka, Japan*.

Sandra Alusio, Lucia Specia, Caroline Gasperin, and Carolina Scarton. 2010. Readability assessment for text simplification. In *Proceedings of the NAACL HLT 2010 Fifth Workshop on Innovative Use of NLP for Building Educational Applications*, pages 1–9.

Nilhadri Chatterjee and Shiwali Mohan. 2007. Extraction-Based Single-Document Summarization Using Random Indexing. In *Proceedings of the 19th IEEE international Conference on Tools with Artificial intelligence – (ICTAI 2007)*, pages 448–455.

Edgar Dale and Jeanne S. Chall. 1949. The concept of readability. *Elementary English*, 26(23).

Anna Decker. 2003. Towards automatic grammatical simplification of swedish text. Master's thesis, Stockholm University.

Felice Dell'Orletta, Simonetta Montemagni, and Giulia Venturi. 2011. READ-IT: Assessing Readability of Italian Texts with a View to Text Simplification. In *Proceedings of the 2nd Workshop on Speech and Language Processing for Assistive Technologies*, pages 73–83, July.

Eva Ejerhed, Gunnel Källgren, and Benny Brodda. 2006. Stockholm Umeå Corpus version 2.0.

Daniel Fahlborg and Evelina Rennes. 2016. Introducing SAPIS - an API service for text analysis and simplification. In *The second national Swe-Clarin workshop: Research collaborations for the digital age, Umeå, Sweden*.

Johan Falkenjack and Arne Jönsson. 2014. Classifying easy-to-read texts without parsing. In *The 3rd Workshop on Predicting and Improving Text Readability for Target Reader Populations (PITR 2014), Göteborg, Sweden*.

Johan Falkenjack, Katarina Heimann Mühlenbock, and Arne Jönsson. 2013. Features indicating readability in Swedish text. In *Proceedings of the 19th Nordic Conference of Computational Linguistics (NoDaLiDa-2013), Oslo, Norway*, NEALT Proceedings Series 16.

Martin Hassel. 2007. *Resource Lean and Portable Automatic Text Summarization*. Ph.D. thesis, ISRN-KTH/CSC/A–07/09-SE, KTH, Sweden.

Martin Hassel. 2011. Java Random Indexing toolkit, January 2011. `http://www.csc.kth.se/~xmartin/java/`.

Michael J. Heilman, Kevyn Collins-Thompson, Jamie Callan, and Maxine Eskenazi. 2007. Combining Lexical and Grammatical Features to Improve Readability Measures for First and Second Language Texts. In *Proceedings of NAACL HLT 2007*, pages 460–467.

Katarina Heimann Mühlenbock. 2013. *I see what you mean. Assessing readability for specific target groups*. Dissertation, Språkbanken, Dept of Swedish, University of Gothenburg.

Derrick Higgins and Jill Burstein. 2007. Sentence similarity measures for essay coherence. In *Proceedings of the 7th International Workshop on Computational Semantics (IWCS), Tilburg, The Netherlands*.

Vida Johansson and Evelina Rennes. 2016. Automatic extraction of synonyms from an easy-to-read corpus. In *Proceedings of the Sixth Swedish Language Technology Conference (SLTC-16), Umeå, Sweden*.

Robin Keskisärkkä and Arne Jönsson. 2013. Investigations of Synonym Replacement for Swedish. *Northern European Journal of Language Technology*, 3(3):41–59.

John Lee, Wenlong Zhao, and Wenxiu Xie. 2016. A customizable editor for text simplification. In *Proceedings of COLING, Osaka, Japan*.

Haitao Liu. 2008. Dependency distance as a metric of language comprehension difficulty. *Journal of Cognitive Science*, 9(2):169–191.

Rada Mihalcea. 2004. Graph-based ranking algorithms for sentence extraction, applied to text summarization. In *Proceedings of the ACL 2004 on Interactive poster and demonstration sessions*, ACLdemo '04, Morristown, NJ, USA. Association for Computational Linguistics.

Thomas Morton, Joern Kottmann, Jason Baldridge, and Gann Bierner. 2005. Opennlp: A java-based nlp toolkit.

Ani Nenkova, Jieun Chae, Annie Louis, and Emily Pitler. 2010. Structural Features for Predicting the Linguistic Quality of Text Applications to Machine Translation, Automatic Summarization and Human–Authored Text. In E. Krahmer and M. Theune, editors, *Empirical Methods in NLG*, pages 222–241. Springer-Verlag.

Joakim Nivre, Johan Hall, Jens Nilsson, Atanas Chanev, Gülşen Eryigit, Sandra Kübler, Svetoslav Marinov, and Erwin Marsi. 2007. MaltParser: A language-independent system for data-driven dependency parsing. *Natural Language Engineering*, 13(2):95–135.

Robert Östling. 2013. Stagger: an open-source part of speech tagger for swedish. *Northen European Journal of Language Technology*, 3.

Sarah Petersen and Mari Ostendorf. 2009. A machine learning approach toreading level assessment. *Computer Speech and Language*, 23:89–106.

Sarah Petersen. 2007. *Natural language processing tools for reading level assessment and text simplification for bilingual education*. Ph.D. thesis, University of Washington, Seattle, WA.

Evelina Rennes and Arne Jönsson. 2015. A tool for automatic simplification of swedish texts,. In *Proceedings of the 20th Nordic Conference of Computational Linguistics (NoDaLiDa-2015), Vilnius, Lithuania,*.

Evelina Rennes and Arne Jönsson. 2016. Towards a corpus of easy to read authority web texts. In *Proceedings of the Sixth Swedish Language Technology Conference (SLTC-16), Umeå, Sweden*.

Jonas Rybing, Christian Smith, and Annika Silvervarg. 2010. Towards a Rule Based System for Automatic Simplification of Texts. In *Swedish Language Technology Conference, SLTC, Linköping, Sweden*.

Horacio Saggion, Sanja Stajner, Stefan Bott, Simon Mille, Luz Rello, and Biljana Drndarevic. 2015. Making it simplext: Implementation and evaluation of a text simplification system for spanish. *ACM Transactions on Accessible Computing*, 6(4).

Carolina Scarton, Matheus de Oliveira, Arnaldo Candido, Jr., Caroline Gasperin, and Sandra Maria Aluísio. 2010. Simplifica: A tool for authoring simplified texts in brazilian portuguese guided by readability assessments. In *Proceedings of the NAACL HLT 2010 Demonstration Session*, HLT-DEMO '10, pages 41–44, Stroudsburg, PA, USA. Association for Computational Linguistics.

Christian Smith and Arne Jönsson. 2011a. Automatic Summarization As Means Of Simplifying Texts, An Evaluation For Swedish. In *Proceedings of the 18th Nordic Conference of Computational Linguistics (NoDaLiDa-2010), Riga, Latvia*.

Christian Smith and Arne Jönsson. 2011b. Enhancing extraction based summarization with outside word space. In *Proceedings of the 5th International Joint Conference on Natural Language Processing, Chiang Mai, Thailand*.