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Project Report
Generating Acrostics via Paraphrasing and Heuristic Search
DBPRO - Database Projects (WS 2014/2015)

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2.4.1 Word Insertion or Deletion

The idea around this operator is to insert words in the text or delete words from it, in order to insert new letters and accomplish the goal acrostic or to remove words and change the position of words inside the text.

To ilustrate the execution, consider the following text:

```
Ah ja, ich heisse Frederik Hoske und ich bin 13 Jahre. Ich kann nicht
vorstellen, weil ich kaum Deutsch sprechen kann. Trotzdem versuche ich
es. Ich habe zwei Geschwister Mein Bruder der 16 Jahre alt ist und
meine Schweseter ist elf.
```

After inserting the word *"mir"* in the sentence *"Ich kann nicht vorstellen"* in the first line and after breaking a line right before *"Trotzdem"* the algorithm can reach the acrostic *amt*. Note that the insertion of *"mir"* was crutial for the result, once that the letter *m* was not there.

```
Ah ja, ich heisse Frederik Hoske und ich bin 13 Jahre. Ich kann
mir nicht vorstellen, weil ich kaum Deutsch sprechen kann.
Trotzdem versuche ich es. Ich habe zwei Geschwister Mein Bruder der
16 Jahre alt ist und meine Schweseter ist elf.
```

The Word Insertion or Deletion operator takes as input a text. Then first it tries to insert a new word in each space and second tries to remove each word of the text. The condition to insert a new word w in the *i-th* space of the text is that w has to fit the context around the *i-th* space. It means that from the set of all possible words of the language, only a restricted subset can be inserted in this place. More specifically, the algorithm starts by taking for each space in the text n words around it as context − In our implementation in this context $n = 4$. This is a so called n-gram, an array of words. After this, the n-gram just taken is sent to the context database (which is in this implementation the NetSpeak API [2]), that returns the possible words that could be inserted in the required space. For each of these possible words a new version of the text is created with the word inside.

Analogously, for each word w in the text a n-gram including the words around it is created − In our implementation we take two words from each side, so here $n = 5$. w is then taken out of the n-gram, which is tested against the context database to check whether this n-gram is frequent enough in the language. If the answer is positive a new version of the text without w is created. Our implementation allows the adjustment of the minimum frequency cited above, but we set it to zero, so a broader set of deletions is executed.

The queries to the context database are made in form of HTTP requests to the netspeak web service using the NetSpeak API.

2.4.2 Synonyms

The synonym operator has the goal of changing words in the text for other words, which have similar meaning. In general the operator takes a text as input and generates a set of new texts, in which each text has a word replaced by a synonym that may eventually contain more than just one word.

In order to perform the replacements it is required a synonym dictionary, which is know as thesaurus. In our implementation we used Open Thesaurus [3], which is available for download for free. This data source is available as a plain text file, but as the dictionary is accessed many times during the execution of the algorithm, it easily becomes intractable to handle a text file as a database.

To solve this problem we decided to use a NoSQL database server [4], namely, Redis. Redis is an open source advanced key-value pair cache and store [5]. Into the database server we load once the data from the thesaurus in a structured way where, every word is added as a key that points to a set of synonyms. Thus can the application easily and efficiently find similar terms for a given word only by accessing this key.

Naturally it is then required that the Redis server is running and listening to requests when the application runs, and that it has been once loaded by our script with the data from the dictionary.

[EXAMPLE]

PROS

CONS

3 Evaluation of the Results

4 Summary of Findings

to better: size of ngram, other context databas, degenerating into breadth first, webservices slow, databases not good

References

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A Appendix