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Frequently Asked Questions Retrieval for Croatian Based on Semantic Textual Similarity

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- Frequently Asked Questions (FAQ) databases are a popular way of getting domain-specific expert answers to user queries.
- A FAQ database consists of many question answer pairs (FAQ pairs). In larger databases it can be difficult to manually find a relevant FAQ pair.
- Automated retrieval is challenging
 - Short texts cause keyword matching to perform poorly
- The goal of this work is to build a FAQ retrieval system for Croatian

Outline



- Data set
- Retrieval model
- Features
- Results
- Conclusion

- From the web we obtained the FAQ of Vip a Croatian mobile phone operator (1222 unique FAQ pairs)
- Ten annotators were asked to create 12 queries each
- The annotators were then asked to paraphrase the queries
 - Turn into a multi sentence query
 - Change the syntax
 - Substitute some words with synonyms
 - Turn into declarative sentence
 - Combination of the above

- For each set of paraphrased queries we retrieve potentially relevant documents using a pooling method (including keyword search, phrase search, tf-idf and language modeling)
- The annotators were asked to review the retrieved set, assigning a binary relevance score to each retrieved FAQ pair. To reduce bias the pairs are presented in random order
- FAQ pairs not retrieved by the pooling method were assumed to be irrelevant

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The annotated data set includes:

- A list of queries
- A list of relevant FAQ pairs for each query
- Additional metadata (i.e. categories of FAQ questions and information about annotators)
- The data set is freely available for research purposes (takelab.fer.hr/data/faqir)
- We focus only on queries which have at least one answer (327 of them)

- We frame the FAQ retrieval task as a supervised machine learning problem.
- A classifier(SVM) is trained on annotated data:
 - Input a query and a FAQ pair
 - Output a binary relevance decision and a confidence score
- The classifier decision itself is not directly used, rather, the results are ordered by classifier confidence
- A variety of semantic similarity metrics is are used as features

• The coverage of text T1 with words from T2:

$$no(T1, T2) = \frac{|T_1 \cap T_2|}{|T_1|}$$

- The ngram overlap feature is the harmonic mean of no(T1,T2) and no(T2,T1)
- It is calculated on unigrams and bigrams between the user query and both the FAQ question and FAQ answer

- To account for varying importance of words they can be weighted using information content (ic)
- The weighted coverage of text T1 with words from T2:

$$wno(T_1, T_2) = \frac{\sum_{w \in T_1 \cap T_2} ic(w)}{\sum_{w' \in T_1} ic(w')}$$

The weighted ngram overlap feature is the harmonic mean of wno(T1, T2) and wno(T2, T1)

- Cosine similarity between query and FAQ pair bag-of-words vectors
- The elements of the vectors are weighted using tf-idf
- The FAQ pair is considered a single document (no distinction between the question and answer parts)

Features – LSA

- LSA derived word vectors ([Karan et al., 2012]) from the HrWaC corpus ([Ljubešić & Erjavec, 2011])
- The vector of a text T is derived compositionally ([Mitchell & Lapata, 2008]):

$$v(T) = \sum_{w \in T} v(w)$$

- The similarity of texts is given by the cosine of their vectors
- Computed between the user query and both the FAQ question and FAQ answer
- Weighted variant:

$$v(T) = \sum_{w_i \in T} ic(w_i)v(w_i)$$

Features – ALO

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- Aligned lemma overlap ([Šarić et al., 2012])
- Given texts T1 and T2 greedily align words:
 - Find the most similar (LSA similarity) pair of words and remove them from futher consideration
 - Repeat until all there are no more words to pair up

■ Calculate similarity for each pair (*ssim* = LSA similarity)

$$sim(w_1, w_2) = \max(ic(w_1), ic(w_2)) \times ssim(w_1, w_2)$$

Calculate the overall similarity

$$alo(T_1, T_2) = \frac{\sum_{(w_1, w_2) \in P} sim(w_1, w_2)}{\max(length(T_1), length(T_2))}$$

- Question classification data set containing 1300 questions ([Lombarović et al., 2011])
- Question classes: numeric, entity, human, description, location, abbreviation
- Using document frequency the most frequent 300 words and 600 bigrams are selected as features
- SVM 80% accuracy
- The classifier outputs for the user query and FAQ question are included as features

- Query expansion dictionary
- Motivated by brief analysis of system errors. Aims to:
 - Mitigate minor spelling variances
 - Make similarity of cross-POS or domain specific words explicit
 - Introduce rudimentary world knowledge useful for the domain
- The dictionary includes a list of rules in the form word - expansionword1, expansionword2, ...
- In total there are 53 entries in the dictionary

Query expansion examples

Query word	Expansion words
face ograničiti (<i>to limit</i>) cijena (<i>price</i>) inozemstvo (<i>abroad</i>) ADSL	facebook ograničenje (<i>limit</i>) trošak (<i>cost</i>), koštati (<i>to cost</i>) roaming (<i>roaming</i>) internet

- Classifier performance is evaluated using the F1 measure
- FAQ retrieval system performance is evaluated using standard IR metrics:
 - Mean Reciprocal Rank (MRR)
 - Mean Average Precision (MAP)
 - R Precision (RP)
- All metrics are calculated using a 5 fold cross validation over the 327 available user queries.
- A baseline FAQ retrieval system is based on tf-idf

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Features used in the models

Feature	RM1	RM2	RM3	RM4	RM5
NGO	+	+	+	+	+
ICNGO	+	+	+	+	+
TFIDF	_	+	+	+	+
LSA	_	_	+	+	+
ICLSA	_	_	+	+	+
ALO	_	_	+	+	+
QED	_	_	—	+	+
QC	_	_	-	_	+

Results

Classification results

Model	Ρ	R	F1
RM1	14.1	68.5	23.1
RM2	25.8	75.1	37.8
RM3	24.4	75.4	36.3
RM4	25.7	77.7	38.2
RM5	25.3	76.8	37.2

FAQ retrieval results

Model	MRR	MAP	RP
Baseline	0.341	21.77	15.28
RM1	0.326	20.21	17.6
RM2	0.423	28.78	24.37
RM3	0.432	29.09	24.90
RM4	0.479	33.42	28.74
RM5	0.475	32.37	27.30

Most frequent causes of error:

- Lexical interference a non relevant FAQ pair can still have high lexical overlap
- Lexical gap lack of lexical overlap
- Semantic gap reasoning and/or world knowledge are required
- Word matching errors informal spelling variations

- Presenting the entire ordered list puts an unnecessary burden on the user
- The list can be shortened using different cutoff criterions:
 - FN first N
 - MTC measure criterion
 - CTC cumulative measure criterion
 - RTC relative measure criterion
- A better criterion will yield higher recall with less retrieved documents

Results





Conclusion and future work take[lab];

- A FAQ retrieval engine was built based on supervised machine learning using semantic similarity features
- Deceivingly high or low word overlap remains a problem, a possible solution is to use syntactic information
- The query expansion dictionary proved quite beneficial. The generation of expansion rules could be automated by analysing query logs collected over a longer time span ([Cui et al., 2002], [Kim & Seo, 2006])
- from a practical perspective, work on scaling up the system to large FAQ databases is required

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Thanks for your attention! take[lab];

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