Pseudo Relevance Feedback Technique and Semantic Similarity for Corpus-based Expansion

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Abstract: The adaptation of a Query Expansion (QE) approach for Arabic documents may produce the worst rankings or irrelevant results. Therefore, we have introduced a technique, which is to utilise the Arabic WordNet in the corpus and query expansion level. A Point-wise Mutual Information (PMI) corpus-based measure is used to semantically select synonyms from the WordNet. In addition, Automatic Query Expansion (AQE) and Pseudo Relevance Feedback (PRF) methods were also explored to improve the performance of the Arabic information retrieval (AIR) system. The experimental results of our proposed techniques for AIR shows that the use of Arabic WordNet in the corpus and query level together with AQE, and the adaptation of PMI in the expansion process have successfully reduced the level of ambiguity as these techniques select the most appropriate synonym. It enhanced knowledge discovery by taking care of the relevancy aspect. The techniques also demonstrated an improvement in Mean Average Precision by 49%, with an increase of 7.3% in recall in comparison to the baseline.

1 INTRODUCTION

Ambiguity is the foremost challenge in Query Expansion (QE) for the Arabic language. It is not uncommon for different possible translations of a word to have very different meanings, and because of its rich and complex morphology, Arabic is notorious for its morphological ambiguity (Attia, 2007). In light of the foregoing, it is clear that the developers of Arabic information retrieval systems need to consider the issues associated with QE. One of the main problems faced by web search engines in the Middle East is that there are very few Arabic Web pages with valuable information; this can clearly be seen in the case of the free encyclopedia (Wikipedia), which enables Internet users to create and edit different articles, where the Arab contribution does not even exceed 1%, at best (Al-Kabi et al., 2012). Furthermore, from their evaluation of Google queries based on language preferences (Al-Eroud et al., 2011), it can be concluded that, if an Arabic query is submitted in Arabic, and if there are many relevant popular pages in English, it is not justifiable for Google to retrieve such popular pages, even if they are in English and the query is in Arabic. Furthermore, Arabic users prefer to use English terms instead of Arabic ones in their queries. Although there is no QE technique that fully meets human requirements in finding a quick and efficient query, use of the expansion technique has become an urgent necessity due to the differences between the languages spoken in the world’s communities and the vast developments that have occurred in the World Wide Web since each technique demonstrates its own advantages and disadvantages.

In this work, a technique for query result expansion is introduced in which the query results, rather than the query, are expanded. In our proposed method, a query is submitted and the system retrieves a set of expanded related documents. For the queries, we used the TREC-2001 Arabic dataset corpus and its 25 queries. We expanded the relevant documents using an Arabic WordNet (AWN) (Al Ameed et al., 2006). We also used a Pointwise Mutual Information (PMI) method, using data collected via Information Retrieval (PMI-IR) as a corpus-based measure of the semantic similarity of words (Turney, 2011). Therefore, the contributions of this work are:

i. The proposal of an automatic corpus-based expansion technique. This technique uses a PRF expansion technique and the Arabic WordNet based on available synsets of terms
as a semantic resource to select expansion terms.

ii. The development of an effective approach that applies a PMI-IR semantic similarity measure with an automatic corpus-based expansion technique to select the most appropriate expansion terms in order to disambiguate word senses. The proposed Corpus-based Expansion using the Pseudo Relevance Feedback and Semantic Similarity (CEPRFSS) approach has improved the performance of the AIR framework.

The rest of the paper is organised as follows: Section 2 highlights the related work; Section 3 explains the methodology; Section 4 discusses the result; and finally, Section 5 concludes this study.

2 RELATED WORK

Previously mentioned methods in literature did not fully consider the semantic relations between terms and expanded terms in query, which has led to the failure of improving retrieval effectiveness (Hoseini, 2011; Jarrar, 2011) However methods involving automatic relevance feedback query expansion successfully improved retrieval effectiveness (Liu et al., 2004). Thus, we have proposed an approach to build and find the best co-occurrence relationship between terms.

In light of the foregoing, it is clear that QE can be and in fact has been addressed using a variety of different approaches. Here, the focus has been on the application of these approaches in AIR problems, taking special note of how these approaches might be suitable in dealing with specific features of the Arabic language. At the moment, the TREC 2001 corpus is the most popular in terms of Arabic text IR and is adequate enough to be considered for system evaluation. However, researchers face problems with stop-word lists in the pre-processing process since there is no standard one. So, researchers have resorted to creating their own lists. Text normalisation is another problem that affects AIR because there are no standard steps for researchers to follow; therefore, researchers create and follow their own normalisation steps.

The stemming process is very important for highly morphological languages such as the Arabic language because it impacts AIR in terms of removing inappropriate affixes (Attar & Fraenkel, 1977). Here too, there is no standardisation as researchers use different set of affixes with different lengths. As we have seen in this study, different approaches and different challenges have led researchers to seek to achieve different sets of objectives, making it difficult to perform comparisons in many cases. However, difficulties in comparisons are also real in many of the studies. Most researchers’ evaluations focus on the expansion of their own collected texts. There is very little work on the expansion of standard corpora, particularly those that describe much of the information found on the Internet, where expansion is in demand (Menai & Alsaeedan, 2012; Otair et al., 2013). Usually, researchers test their expansion techniques using recall, precision, average recall, average precision, mean average precision, and the f-measure, which are evaluation measures commonly used in the research and development of IR technology. Query expansion techniques were tested as well (Mitra et al., 1998). All these processes were combined and incorporated into the different phases of the standard IR framework and the additional phases to support Arabic text retrieval were also taken into account in the enhanced framework. We used TREC 2001 corpus and queries based on previous researches (Larkey et al., 2002; Taghya et al., 2005).

3 METHODOLOGY

We used one Arabic test corpus, the Arabic Newswire from the Linguistic Data Consortium (LDC), which has also been used in recent TREC experiments. This corpus is composed of articles from the Agence France Presse (AFP) Arabic Newswire service. The source material was tagged using a TIPSTER style SGML and was transcoded to Unicode (UTF-8). The corpus includes articles from 13 May 1994 to 20 December 2000. The data is stored in 2337 compressed Arabic text data files. There are 209 Mbytes of compressed data (869 Mbytes uncompressed) consisting of 383,872 documents containing 76 million tokens of over approximately 666,094 unique words. The query set associated with the LDC corpus was created for TREC 2001. There are 25 topics with relevance judgements [14], available in Arabic, French and English, with Title, Description, and Narrative fields. We used the Arabic titles and descriptions as queries in our monolingual experiment. Queries, which are also called “topics” in the TREC 2001, have special SGML mark-up tags. A total of 25 queries were defined as part of the TREC 2001.

We applied a corpus-based expansion method using the PRF technique and semantic similarity using WordNet. The enhancement of the initial retrieved result is an important step in retrieving more
of the documents that did not appear in the initial result. The query expansion technique is an important approach that is used to enhance the performance of any retrieval system. An automatic query expansion (i.e., without user interaction) was also used. The expansion technique consists of two stages.

A. First Stage: Corpus-based Expansion using PRF (CEPRF)

The first stage is based on the PRF technique and is applied to the initial retrieved document result from the initial query submission. The first stage is called the Corpus-based Expansion using Pseudo Relevance Feedback (CEPRF) technique. The general procedure of this step is:

• Step 1: The terms are selected from the top n-ranked document after the query ‘Q’ is submitted for the first time.
• Step 2: The synonym(s) for all the terms from the top n documents are selected from WordNet.
• Step 3: The new query of the selected synonyms from step 2, Q’, is constructed.
• Step 4: The difference between the retrieved documents of Q’ and the result of the original Q are added to the final result.

B. Second Stage: Corpus-based Expansion using PRF and Semantic Similarity (CEPRFSS)

The second stage is a modified version of stage one, which is based on the PRF technique, the Arabic WordNet, and the semantic similarity measure. It is called CEPRFSS. The general procedure is:

• Step 1: The terms are selected from the top n ranked document after the query ‘Q’ is submitted for the first time.
• Step 2: The synonym(s) for all terms from the top n documents is (are) selected from WordNet.
• Step 3: The PMI_IR scores between each term and its corresponding synonym(s) are calculated.
• Step 4: The synonym of the highest PMI-IR score i.e., the highest semantic relationship is selected.
• Step 5: The new query of selected synonyms from step 4, Q’, is constructed.
• Step 6: The difference between the retrieved documents of Q’ and the result of the original Q are added to the final result.

The goal of this approach is to retrieve more relevant documents for the user query. To achieve this, the system provides an approach to deal with expansion terms. After checking the semantic similarity between the documents’ terms and their corresponding synonyms, only the highest similarity is accepted.

The current release of AWN contains 11 270 Arabic synsets (vs. 115 000 synsets for English WordNet) and 23 496 Arabic words (vs. 200 000 words for English WN). It also contains entries that are named entities (1142 synsets and 1648 words). The AWN ontology contains different relations between its items such as hyperonymy/hyponymy (supertypes/subtypes relations), synonymy, meronymy/holonymy part/whole relations), etc. Our semantic expansion approach uses synonyms. The idea of this research is to apply our semantic expansion process to acquire new documents related to the top-ranked documents. Then, we add these documents to the documents’ retrieved result to store data relevant to the user query. In this study, we built a stem-semantic relationship via the following steps:

Step 1: After normalisation and tokenisation, the tokenised, non-stop-word words from each document are sent to WordNet so the corresponding synonym(s) can be searched. Simultaneously, the same word is sent to the stemmer for stemming. If any synonym(s) is available for this word in WordNet, we save the synonym(s) in a database, as shown in Table 1.

Table 1: Sample of document terms’ stem with the available synonym(s) from WordNet.

<table>
<thead>
<tr>
<th>Document Name</th>
<th>Term</th>
<th>Synonym</th>
</tr>
</thead>
<tbody>
<tr>
<td>19980923_AFP_ARB_0118.txt</td>
<td>غريب (strange)</td>
<td>شاذ (odd/typical)</td>
</tr>
<tr>
<td>19980923_AFP_ARB_0118.txt</td>
<td>عجيب (odd/typical)</td>
<td>أجنبي (foreign)</td>
</tr>
<tr>
<td>19980923_AFP_ARB_0118.txt</td>
<td>عبد (slave)</td>
<td>عجيب (wondrous)</td>
</tr>
<tr>
<td>19980923_AFP_ARB_0118.txt</td>
<td>عبد (slave)</td>
<td>مستخدم (servant)</td>
</tr>
</tbody>
</table>

Step 2: To select the most appropriate semantically-related synonym for the document term, we used a semantic similarity measure to find the value of similarity between each term, followed by a corpus-based semantic similarity measure to find the corresponding synonym. Turney (2011) suggested using the Pointwise Mutual Information with data collected via Information Retrieval (PMI-IR), as an unsupervised measure for the evaluation of the semantic similarity of words (i.e., the semantic similarity between the term and the corresponding synonym). These data are based on word co-occurrence using the counts collected over a large
The semantic similarity measure between the term and the corresponding synonym scores are computed as follows:

- Given two words \( w_1 \) and \( w_2 \), their PMI-IR is measured as:
  \[
  \text{PMI-IR}(w_1, w_2) = \log_2 \frac{p(w_1 \& w_2)}{p(w_1)p(w_2)}
  \]  
  (1)

where \( w_1 \) is the term, \( w_2 \) is the corresponding synonym, and \( p(w_1) \) is the number of times the term \( (w_1) \) appears in the corpus. \( p(w_2) \) is the number of times the synonym \( (w_2) \) appears in the TREC. \( p(w_1 \& w_2) \) is the number of times both (term, synonym) appear together in the TREC-2001.

- After the PMI-IR is calculated for each term and its corresponding synonym(s), the synonym of the highest PMI-IR score is selected. Table 2 shows the synonyms of the terms that have the highest PMI-IR score, previously presented in Table 1.

<table>
<thead>
<tr>
<th>Term (( w_1 ))</th>
<th>Synonym (( w_2 ))</th>
<th>PMI-IR</th>
</tr>
</thead>
<tbody>
<tr>
<td>غريب</td>
<td>عجيب</td>
<td>4.969</td>
</tr>
<tr>
<td>عبد</td>
<td>جارية</td>
<td>0.512</td>
</tr>
</tbody>
</table>

The Arabic WordNet was used to provide the semantic relationship between the terms in the expansion process.

### 4 RESULTS

In the experiment, the retrieval result of a typical baseline IR system (Atwan et al., 2014) is compared against the retrieval achieved by stemming, and after expansion. Four experiments were executed. Based on previous work (Atwan et al., 2014), two runs were carried out: Baseline without stemming (Baseline); and Baseline with a Light10 stemmer with a combined stop-words list (LCS). The other two runs that were executed were the Corpus-based Expansion using Pseudo Relevance Feedback (CEPRF); and the Corpus-based Expansion using Pseudo Relevance Feedback and Semantic Similarity (CEPRFSS) runs. The CEPRFSS used PMI-IR as the semantic similarity measure whereas the CEPRF did not. Recall and precision measures were used to evaluate all four runs. The retrieval results were analysed by calculating the differences between the different runs using the Mean Average Precision (MAP) and F-measure techniques.

Method based on automatic query expansion uses Pseudo-Relevance Feedback technique and word sense disambiguation using WordNet was applied. In this case, we did not expand the original query itself; we expanded the retrieved documents instead. The concept behind this step is to automatically find more relevant data for the user query without user intervention.

The results demonstrate the usefulness of the integration of the expansion technique with the AIR. The output results obtained from the experiments consisted of recall and precision measurements. Figure 1 shows the 11-point precision graph for Baseline, LCS, CEPRF, and CEPRFSS.

<table>
<thead>
<tr>
<th>Runs</th>
<th>Metrics</th>
<th>Value</th>
<th>F-Measure</th>
</tr>
</thead>
<tbody>
<tr>
<td>Baseline</td>
<td>MAP</td>
<td>0.162</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Recall</td>
<td>0.468</td>
<td>0.241</td>
</tr>
<tr>
<td>LCS</td>
<td>MAP</td>
<td>0.255</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Recall</td>
<td>0.488</td>
<td>0.335</td>
</tr>
<tr>
<td>CEPRF</td>
<td>MAP</td>
<td>0.249</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Recall</td>
<td>0.498</td>
<td>0.332</td>
</tr>
<tr>
<td>CEPRFSS</td>
<td>MAP</td>
<td>0.293</td>
<td></td>
</tr>
<tr>
<td></td>
<td>Recall</td>
<td>0.502</td>
<td>0.370</td>
</tr>
</tbody>
</table>

In the evaluation phase, four runs were tested. Based on previous work, a total of four runs were performed, namely, (1) baseline without stemming; (2) LCS; (3) CEPRF; and (4) CEPRFSS. Our expansion technique was used with a PMI-IR measure for synonym selection. The results obtained from the system are then used to achieve recall and precision. The retrieval results were analysed using MAP, which calculates the differences between the four runs as shown in Table 3. The best results were obtained when the semantic similarity measure was used to select the best synonym.
The performances of the F1 measure for LCS and CEPRF were quite similar. The reasons for these results are because: (1) the expansion is recall-oriented and the CEPRF directly uses AWN synonyms in the expansion process, which may bring about uncertainty. For example, the term ‘زنبّط،’ which means ‘sort’ has many synonyms such as ‘كون،’ which means ‘universe’ and ‘format’. Because of this, the expansion terms will affect the number of retrieved documents (it will be increased) but the similarity to the queries will be low. So the new related documents to the queries will have low similarity; (2) the elimination of stop-words improves the precision, which reduces the appearance of the most frequent words and the ones that do not have meaning.

The best results were obtained when the semantic similarity measure was used to select the best synonym. The results show that our proposed approach is effective in expanding the results and disambiguating word senses. This automatic expansion technique (CEPRFSS) based on PRF and semantic similarity measure using AWN achieved the best result in comparison to the other two systems i.e., LCS and CEPRF.

Building the Stem-Semantic relationship process based on the terms in the TREC-2001 and the synonyms in the Arabic WordNet with a semantic similarity measure has improved AIR performance. The use of the Arabic WordNet and the semantic expansion is in line with the work of other researchers. However, this improvement in expansion is limited to the available synonyms of the terms in the TREC-2001 and are further restricted to the small size of the Arabic WordNet relations. Experiments that have used the expansion technique showed that the overall results exhibited improvement in retrieval effectiveness in terms of MAP by 49% and without degradation in recall compared to the baseline. In addition, the result in terms of recall was also improved by 7.3%.

5 CONCLUSIONS

This work presents two proposed approaches as automatic expansion techniques. The first approach is based on corpus and the PRF technique using the Arabic WordNet to select the expansion terms, in which the relationship among the candidate expansion terms and the corresponding synonyms are identified using corpus-based semantic similarity measurements that are based on their co-occurrence distributions. The second approach is that of automatic query expansion, for which we jointly use the Arabic WordNet and the PRF technique with semantic similarity measurement to confirm the newly expanded query terms. To overcome the limitation of the semantic synonym selection from WordNet, a corpus-based semantic similarity measurement is also used. The expansion approach has an important function in this system; it is the heart of this retrieval system, utilising knowledge regarding the synonyms from the Arabic WordNet based on available synsets of terms as a semantic resource to select expansion terms and expanded terms so as to add suitable and relevant documents for the user query. The CEPRFSS is an effective approach that applies the PMI-IR semantic similarity measure with the automatic corpus-based expansion technique to select the most appropriate expansion terms to disambiguate word senses. Overall, this approach has improved the AIR performance.

REFERENCES


