

TERMS INTERRELATIONSHIP QUERY EXPANSION TO IMPROVE  
ACCURACY OF QURAN SEARCH

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For my beloved mother and father



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## ABSTRACT

Quran retrieval system is becoming an instrument for users to search for needed information. The search engine is one of the most popular search engines that successfully implemented for searching relevant verses queries. However, a major challenge to the Quran search engine is word ambiguities, specifically lexical ambiguities. With the advent of query expansion techniques for Quran retrieval systems, the performance of the Quran retrieval system has problem and issue in terms of retrieving users needed information. The results of the current semantic techniques still lack precision values without considering several semantic dictionaries. Therefore, this study proposes a stemmed terms interrelationship query expansion approach to improve Quran search results. More specifically, related terms were collected from different semantic dictionaries and then utilize to get roots of words using a stemming algorithm. To assess the performance of the stemmed terms interrelationship query expansion, experiments were conducted using eight Quran datasets from the Tanzil website. Overall, the results indicate that the stemmed terms interrelationship query expansion is superior to unstemmed terms interrelationship query expansion in Mean Average Precision with Yusuf Ali 68%, Sarawar 67%, Arberry 72%, Malay 65%, Hausa 62%, Urdu 62%, Modern Arabic 60% and Classical Arabic 59%.

## ABSTRAK

Sistem capaian Al-Quran menjadi alat untuk pengguna mencari maklumat yang diperlukan. Mesin pencari adalah salah satu enjin carian yang paling popular yang berjaya dilaksanakan untuk mencari pertanyaan ayat yang relevan. Namun, cabaran utama mesin pencari Al-Quran adalah kesamaran kata, khususnya kesamaran leksikal. Dengan munculnya teknik pengembangan pertanyaan untuk sistem capaian Al-Quran, prestasi sistem pengambilan Al-Quran mempunyai masalah dan masalah dari segi mendapatkan maklumat yang diperlukan pengguna. Hasil teknik semantik semasa masih kekurangan nilai ketepatan tanpa mempertimbangkan beberapa kamus semantik. Oleh itu, kajian ini mencadangkan pendekatan pengembangan pertanyaan melalui hubungan antara istilah yang berdasar untuk meningkatkan hasil carian Al-Quran. Lebih khusus lagi, hubungan istilah dikumpulkan dari kamus semantik yang berlainan dan kemudian menggunakannya untuk mendapatkan akar kata tanpa imbuhan. Untuk menilai prestasi cadangan permintaan hubungan antara istilah yang dicetuskan, percubaan dilakukan menggunakan laman set al-Quran dari laman web Tanzil. Secara keseluruhan, hasilnya menunjukkan bahawa pengembangan pertanyaan hubungan antara istilah tanpa imbuhan adalah lebih unggul daripada pengembangan pertanyaan istilah antara hubungan yang berimbuhan dalam Purata bagi Ketepatan Purata dengan Yusuf Ali 68%, Sarawar 67%, Arberry 72%, Bahasa Melayu 65%, Hausa 62%, Urdu 62% , Bahasa Arab Moden 60% dan Bahasa Arab Klasik 59%.

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## LIST OF SYMBOLS AND ABBREVIATIONS

AP	- Average Precision
AR	- Average Recall
TS	- Term Selection
WASS	- Word Ambiguities Semantic Search
MAP	- Mean Average Precision
BEP	- Break Even Point
MRR	- Mean Reciprocal Rank
BM25	- Best Matching
TFID	- Term Frequency Inverse Document Frequency
VSM	- Vector Space Model
LSI	- Latent Semantic Indexing
PLSI	- Probabilistic Latent Semantic Indexing
Word2Vec	- Word to Vector
ALQS	- Apache Lucene Quran Search
P@K	- Precision at K
F-Measure	- F1 Measure
NDCG	- Normalized Discounted Cumulative Gain
CBOW	- Continuous Bag of Word
FAQ	- Frequency Ask Question
SAS	- Structure Activity Relationship
PC	- Pearson Correlation
RMSE	- Root Mean Squared Error
R@K	- Recall at Rank K
DCG	- Discounted Cumulative Gain
RDF	- Resource Description Framework

OWL	- Ontology Web Language
LSI-VSM	- Latent Semantic Indexing-Vector Space Model
PLSI-VSM	- Probabilistic Latent Semantic Indexing-Vector Space Model
Word2Vec	- Word to Vector
SVD	- Singular Value Decomposition
LDA-VSM	- Latent Dirichlet Allocation
RI-VSM	- Random Indexing-Vector Space Model
EM	- Expectation-maximization Algorithm
WSD	- Word Sense Disambiguation
JDK	- Java Development Kit
IDE	- Integrated Development Environment
TSV	- Term Selection Value
TI-TSV	Terms Interrelationship - Term Selection Value
sITI-TSV	- Statistical Linguistic Terms Interrelationship Term Selection Value
WslTI-TSV	- Weighting Statistical Linguistic Terms Interrelationship Term Selection Value
TIQE	- Terms Interrelationship Query Expansion





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## **CHAPTER 1**

### **INTRODUCTION**

#### **1.1 Background Study**

The increase in demand for search engine has become a worldwide phenomenon. Today, many countries of the world are witnessing the growth of search engine users. These users use the search engines not only for locating web pages but also for research purposes such as finding research articles and papers. Along with search engine users, there is also a drastic growth in the number of search engines [1]. This led to improvements of such search engines to produce relevant search results to users and minimize irrelevant ones. The effect of this is that users' information needs largely depend on the search engines precision. The search engine is an information retrieval application area that plays significant roles in ensuring relevant information is obtained from available collections. In particular, the query is used to search available or needed information within a document. Many search engines are now available ranging from simplest to more complex ones. For instance, desktop search [2], personal search [3], emails search [4], enterprise search [5] and mobile search [6] [7] are an important search engine that is currently used by many users (personal, organizations and government). The users use these search engines to search for relevant information. Web Search has been the most successful search engine that has been used in different applications to search for information. Such information can come from different sources, especially web pages. This is however not satisfactory because of the inability of a search engine to understand the query and bring out needed information [1]. For a web search engine to display the relevant documents based on users need, additional terms must be given to a query to

properly understand the meaning behind that query. For instance, documents that match a query with additional terms can be more likely to be relevant to the user.

Query expansion is one of the techniques used to improve search engine performance by adding additional information to a user query. Recently, query expansions have attracted the attention of many researchers and practitioners. This led to an increasing number of query expansion techniques and algorithms. Much of these query expansion techniques have been examined in [8] and [9] to add additional terms to the query. For example, the study [8] compared and contrast different query expansion techniques in terms of their datasets and query processing algorithms. It presents a survey of about 57 years' periods. In addition to that, the study [9] provides useful information on how to query expansion techniques based on word sense disambiguation, fuzzy rules and term weighting. Zhou et al [10] concentrated on how additional terms can be added to users profiles for better query expansion. The query and its weight can be used to measure the relevance of a document. A similar paper [11] takes account of user profiles to get relevant information. According to Liu et al [12], word embedding can be used to generate a list of words so that word similarities can be chosen. Other studies considered either Wikipedia [13] or query log [14] to improve search effectiveness. However, the query expansion still suffered a vocabulary mismatch [15], specifically in terms of word ambiguities of term selection [16] and term weighting [17] which lead to a lack of better search quality results. Consequently, linguistics and ontology methods have emerged among others as approaches of query expansion that concerned to improve performance.

The holy Quran is the holy book of guidance where laws, commandments, codes for their social and moral behaviors are laid down to Muslims. It contains a comprehensive religious philosophy. The language of the Holy Quran is Arabic but was translated into various languages all over the world. The holy Quran is divided into 114 surahs or chapters and each chapter consists of individual ayaat or verses. There are in total 6326 verses in the Holy Quran. The text of the Holy Quran has remained unchanged over the past 1440 years. The emerging Quran search on a Quran search system can be used to enhance a variety of task, such as searching for a specific verse. It can also enable the discovery of related verses of different keywords by allowing expanding query with different terms. A query can be expanded with synonyms of terms based on terms relationship. Afzal and Muktar

[18] suggested a Quran English WordNet solution as a linguistic approach to short queries, especially when linking to semantic similarities. Their work achieved significant improvement. Although semantic similarity proves effective, Moawad, Alromima and Elgohary [19] stated that absences of semantic resources were identified in many languages and as such alternative semantic approaches need to be developed. Bentrchia, Zidat and Marir [20] examined the possibility of using semantic relatedness instead of only semantic similarities. This paper used only semantic relatedness to obtain a relationship between two concepts using “AND” conjunctive. However, Lashkari, Bagheri and Ghorbani [21] identified based on the literature that depending only on semantic resource would not improve better search performance. Integrations of other methods with semantic were suggested to yield better results. This thesis utilizes ontology method-based query expansion to expand the query. An ontology method-based query expansion is the main tool for selecting query expansion term from knowledgebase based on their relatedness to the query [22].

Currently, the term selection assumptions are mainly concentrated on the probability of word occurrences. For instance, Jagendra et al [16] approach mainly assumed the chance of word occurrence. The term selection depends closely on how words relate to each other. Zhou et al [10] proposed a personalized query expansion. However, query expansion approaches are significantly improving the search engine performance in terms of relevance feedback, synonyms, ontology or related technique like semantic mapping. The combination of expert judgment, word synonyms and stemming require researchers’ attention. Although some researchers such as Wasim et al. [23] try to address word ambiguities in term selection challenges by utilizing relevant assessment. Still, the basis of the work combines relevant feedback and word synonyms on biomedical document retrieval. Singh and Sharan [24] presented different term selection methods using relevant feedback. They combined the relevance feedback with a rank combination to improve search performance. This thesis is concerned with the Terms Interrelationship (TI) based on semantic search. Terms Interrelationship can effectively express different term relationships and improvement of query expansion.

## 1.2 Research Motivation

Terms interrelationship (TI) in semantic search can be used in query expansion through adding word meanings. Using TI semantic search in query expansion has been a topic of research from both researchers and practitioners. However, most of the studies such as Singh and Sharan [24] are focused on the pseudo-relevance feedback technique, which demands selecting terms for expansion from top retrieved documents. In the absence of such term selection methods, TI semantic search is used with synonyms [25]. Holy Quran is the holy book revealed by Allah to serve as guidance to Muslims. In the Quran search system; there was a need for a query expansion for expanding Quran query terms, which will ensure retrieving relevant verses (ayats) and chapters (surahs) of the holy Quran. The goal is to expand relevant query term with the similar term meaning to retrieve relevant information need. It does not necessarily consider the query and document terms to be the same because terms may be different but contained the same meanings. In Quran search query expansion, adding the semantic meaning of a term is interesting because Quran search always involves verse meaning to have term selection. For instance, Quran 7:31 said, “O children of Adam, take your adornment at every Masjid, and eat and drink, but be not excessive. Indeed, He likes not those who commit excess”. The word “Masjid” is ambiguous because it may be a time or place of prayers (Yusuf Ali), place of worship (Pickthal) or Kabah at the time of Tawaf. Thus, adding the meaning of a term in query expansion might improve the query expansion results. Therefore, word ambiguities need to be addressed to properly understanding this query. Because of word ambiguities present in a query, commonly used technique, such as pseudo-relevance feedbacks are not appropriate for the Quran search system. Therefore, the first contribution of this thesis deals with the problem of providing term selection for query expansion with word ambiguities. The thesis proposed TI in semantic search solution for this problem.

Besides the presence of word ambiguities, a challenging aspect of term selection in the Quran search system is that the system-retrieved results may contain many irrelevant documents. In other words, the system retrieved results not only about term selection but also relevant to the query. For instance, a system may contain different term selection methods in query expansion but still contain many irrelevant results. Such situations can arise not only in the Quran search system but



also in also other systems. For a similar domain, Moaward et al [19] has proposed a bi-gram method of query expansion, where query terms are expanded with n-gram for effective results. They take into account improving relevant search results but not semantic search. Moreover, such improvement requires a better term selection method for word ambiguities in query expansion. This thesis identifies improving query expansion for TI in semantic search.

### 1.3 Problem Statement

Term selection is a challenge to query expansion, which refers to selecting terms that are similar to the query terms. The idea is that more documents that are relevant can be retrieved through query expansion if proper terms are selected similar to the query terms. Addressing the term selection required addressing word ambiguities [4]. Lexical ambiguity is more commonly addressed by researchers [5] [4] that concern on word ambiguities in a query which still led to many irrelevant documents retrieved. In addition, there are many different word ambiguities in a query. Hence, there is a need to address lexical ambiguities that are commonly found in Quran search to minimize irrelevant documents retrieval. Terms relationship in natural language understanding is addressing word ambiguities for semantic search that employs one semantic dictionary or WordNet. However, fewer concentrations concern about the terms relationship from more than one WordNet. Having stated the challenges of term selection in terms of lexical ambiguities, the researchers are now able to formulate the main problem in query expansion as:

***How can the lexical ambiguities of term selection are exploited to produce the best possible query expansion for Quran system***

As a semantic search method, term relationship was used to capture different Quran terms of lexical ambiguities that could guide term selection in query expansion. The thesis presents meaningful definitions of relationships that apply to lexical ambiguities that may arise when selecting a term for query expansion. Terms interrelationship algorithms of lexical ambiguity will be designed to provide accurate query expansion and evaluated in Mean Average Precision (MAP), Mean Reciprocal



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