

Original Research Paper

Wordnet and Ontology Based Query Expansion for Semantic Information Retrieval in Sports Domain

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Article history

Received: 12-04-2014

Revised: 15-07-2014

Accepted: 09-09-2014

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Abstract: Semantic Search has been a major longing factor from the envisage state of Semantic Web. Information on the Web is growing at a very rapid pace and has become quite voluminous over the past few years. Semantics of the query is not considered in Traditional Search system since it is a mere Keyword based Search. To increase the number of relevant documents retrieved, queries need to be disambiguated by looking at their context. A query expansion algorithm for Semantic Information Retrieval in Sports Domain (SIRSD) is proposed to do Semantic Search to improve search over large document repositories. This algorithm reformulates user queries by using Word Net and domain ontology to improve the returned results. Our proposal is illustrated with sample experiments showing improvements with respect to Traditional Search and providing ground for further research and discussion. SIRSD reduces the issue of semantic interoperability during the user query search. It has been inferred that there will be a higher value of average precision and recall for the Semantic Search system when compared to Traditional Search. The results show its effectiveness in generating a suitable number of query search with an accuracy of 87.1% compared to other competitors of generic search engines.

Keywords: Semantic Web, Semantic Search, Information Retrieval, Ontology, Query Expansion, Protégé, RDF

Introduction

World wide Web Consortium (W3C) defines the Semantic Web which is the data on the world wide web with the Semantic content. Information Retrieval is the process of obtaining Information Resources relevant to the query given by the end user. It is difficult to specify information that is needed using exact query terms.

The most critical language issue for retrieval effectiveness is the term mismatch problem: The indexers and the users do not often use the same words. This is known as the vocabulary problem. Furnas *et al.* (1987), compounded by synonymy (same word with different meanings, such as “apple” (‘company’ vs. ‘fruit’)) and polysemy (different words with the same or similar meanings, such as “tv” and “television”). Synonymy may result in a failure to retrieve relevant documents, with a decrease in Recall value. Polysemy may cause retrieval of erroneous or irrelevant documents, thus implying a decrease in Precision of retrieval.

Several approaches such as interactive query refinement, relevance feedback, word sense disambiguation and Search results Clustering have been proposed to deal the Vocabulary problem. The natural and the most successful technique is to expand the original query with suitable words that best capture the actual user intent. The accuracy of the search can be improved by Semantic Search which will understand the intent of the searcher and the contextual meaning of terms to generate more relevant results. Patil and Jadhav (2012) designed semantic information retrieval using ontology and SPARQL and its application to cricket domain.

Why Query Expansion

The query expansion is needed to abtain the ambiguity of natural language. The query expansion adds the meaningful terms to the initial query. New terms can be added by either manual or automatic or user-assisted. User expertise is required to make decisions on which terms can be added in the new query

in case of manual query expansion but the terms with highest weighting are added to the initial query in automatic query expansion. The system generates possible query expansion terms during the user assisted query expansion and the user selects which query to be included. The method which provide more contextual information in the result retrieved should be chosen as the term selection method. Bhogal *et al.* (2006) acquired the contextual information from relevance feedback, term co-occurrence and also it has been derived from knowledge models such as ontologies. The user's original query is augmented by new features with a similar meaning in automatic query expansion.

Query Expansion Using Relevance Feedback

Salton and McGill (1983) defined relevance feedback as a technique for modification of the initial query using words from top-ranked or identified relevant documents. It is an easier way of improving the retrieved document as opposed to the user having to construct a new query. The relevance feedback loop requires the user to enter an initial query which results in a display of ranked documents (usually titles/abstracts). From this display, the user makes relevant judgments and selects the relevant documents. The relevant terms from these documents are added to the initial query. An alternative to this is pseudo relevance feedback where the top ranked 'n' documents are assumed to be relevant. Terms from these documents are selected and used for expanding the query. Whether pseudo-relevance feedback or traditional relevance feedback is used, the term selection method is a key factor in the performance of expanded queries.

Query Expansion Using Ontology

Ontologies are applicable to domain independent retrieval such as web information retrieval and also more useful in specialized information retrieval tasks. They have also been used in query expansion. Ontologies are effectively formal and explicit specifications in the form of concepts and relations of shared conceptualizations. Wilson *et al.* (2012) learnt ontology from text. Ontologies range from general to domain-specific. Word Net and Euro Word Net are examples of a general ontology. Many domain-specific ontologies exist for example in the medical and legal domains. Word Net has been a popular general ontology used in the area of query expansion.

Materials and Methods

Traditional Method

Searching is done in three stages namely document indexing, term weighting, similarity coefficients in

classical search engines. In the first stage, document indexing is done. Importance of the term used within the document is calculated with the help of term frequency in the second stage. In the last stage, documents and queries are represented by vectors of term weight. Finally the retrieval is done by cosine similarity. In this method documents are represented poorly, search keyword must be precise, document with similar context but different term won't be retrieved. Salton and McGill (1983) proved that this method gives very low precisions and recall values.

Semantic IR Method

Traditional methods are unable to understand the meaning of the content. In semantic web, knowledge representation is very much easier. Information related to the specific domain is stored in the ontology. Vallet *et al.* (2005) used the semantic knowledge base instead of keyword-based index for efficient searching. Vallet *et al.* (2005) were not able to achieve greater performance.

Semantic Indexing

One of the techniques to improve precision and recall values is Semantic indexing. Semantic index is created on ontological data and keyword search is performed. Semantic indexing based on the lucene indexing extends the traditional full text index with extracted data. The document containing ontological data get higher rates while Ranking. Performance of the system is improved when compared to Traditional Search. Kara *et al.* (2010) were not able to achieve the 100% precision and recall values.

Term Co-Occurrence

Term co-occurrence refers to two or more terms that are situated next or near to each other in the source document. Smeaton and Rijsbergen (1983) have done experiments with new terms which is generated from sources such as maximum spanning trees and found very little improvement. Peat and Willett (1991) have used query terms that have high collection frequencies for query expansion. Since high frequency terms do not discriminate between relevant and non-relevant documents, the addition of these terms for query expansion is ineffective.

Problem Statement

One of the important concern in Search is to capture the Semantic of the user query. The indexers and the users need to use the related and synonym words instead of using the same words. This can be handled by expanding the original query with other words that best capture the actual user intent. System is implemented for the Sports domain. Data is

extracted from the Sports domain and stored in the ontology. Ontology is stored in OWL format which is based on RDF and RDFS. The objective of the work is to retain the user friendly nature of the search query as well as to improve the retrieval performance.

System Implementation

The overall architecture of the system is shown in Fig. 1. The system is fragmented into number of segments for better understanding. The integral segments within the system consist the phases like parsing of input query, extracting synonym words, ontology construction, keyword extraction from ontology, Semantic Search with expanded query using Google API. The following subsections describe these phases in more detail.

Parsing of Input Query and Extracting Synonym Words

The system user enters a query to fetch data on sports domain in natural language. The input user query need to go through the parser to analyze the given query grammatically. The parsing is done to analyze the query syntactically to determine every word in the query. The parsed output is stored in the file in the format shown as parsed sentence. The parsed tree form of query “name of football clubs in EEFA “is depicted in Fig. 2. The parsed words are shown in Fig. 4.

Tool Used: Stanford Parser

Related synonym sets of various words in the query are also obtained using Word Net API. Similar words for the query is also obtained from Similarity measure using the Equation 1: We define the document vector of D_i as $d_i = (d_{i,1}, \dots, d_{i,M})$, where d_i , j is the weight of the annotation of document D_i with I_j , if such annotation exists and zero otherwise. We define the extended query vector as $q = (q_1, q_2, \dots, q_M)$. Now the similarity measure of D_i for the query Q is computed as Equation 1:

$$Sim(D, Q) = \frac{\sum_{i=1}^M d_i \cdot q_i}{\sum_{i=1}^M |d_i| \cdot |q_i|} \quad (1)$$

The Semantic Similarity between the two words P and Q is also modeled as a function $sim(P, Q)$ that returns a value in range $[0: 1]$. The popular Co-Occurrence measure, Web Jaccard is used to compute Semantic Similarity using page counts.

$WebJaccard(P, Q)$ is defined in Equation 2:

$$WebJaccard(P, Q) = \begin{cases} 0, & \text{if } H(P \cap Q) \leq c, \\ \frac{H(P \cap Q)}{H(P) + H(Q) - H(P \cap Q)}, & \text{otherwise.} \end{cases} \quad (2)$$

Ontology Construction and Keyword Extraction from Ontology

A formal representation of knowledge as a set of concepts within a specific domain is an ontology. Strong ontology is constructed based on handful of information that is gathered by referring through various Sports websites. The PROTÉGÉ Tool is used in building a perfect ontological hierarchy and can be used to represent exact interconnected relationships between the classes using properties. RDF is used to represent information and to exchange knowledge in the web. The Sample RDF for Sports class is shown in Fig. 3. The class Hierarchy for sports domain is depicted in Fig. 6. The set of more related key words to the given user query is extracted from the sports ontology. At the end, the collection of words which are Semantically related to the Sports Domain are obtained.

Query Expansion and the Semantic Search

Various queries are formed based on permutations and combinations of Similar words and the words obtained from the ontology. The expanded query of the input query is given in Fig. 5. The queries formed will be more refined and are sent to Google Search API which fetches the web links related to the user query.

Example:

The set of expanded queries is an input for the web Search which provide the corresponding web links to the given queries. The retrieved web links for the semantic search are obtained by means of the extracted domain keywords from the ontology. The algorithm for query expansion is as follows.

SIRSD Algorithm:

Input: End User Query

Output: Expanded Query

Method: Given a query,

1. Ontology is constructed for the Sports Domain.
2. Break the query into different word. Parse the query using Stanford Parser
3. For each word find the synonyms words using Word Net
4. Related words for the query word is extracted from the Ontology
5. The words from the step 4 and step 3 are combined using Permutation
6. The original query plus new words acquired by step 5 form a new query

Return the new expanded query

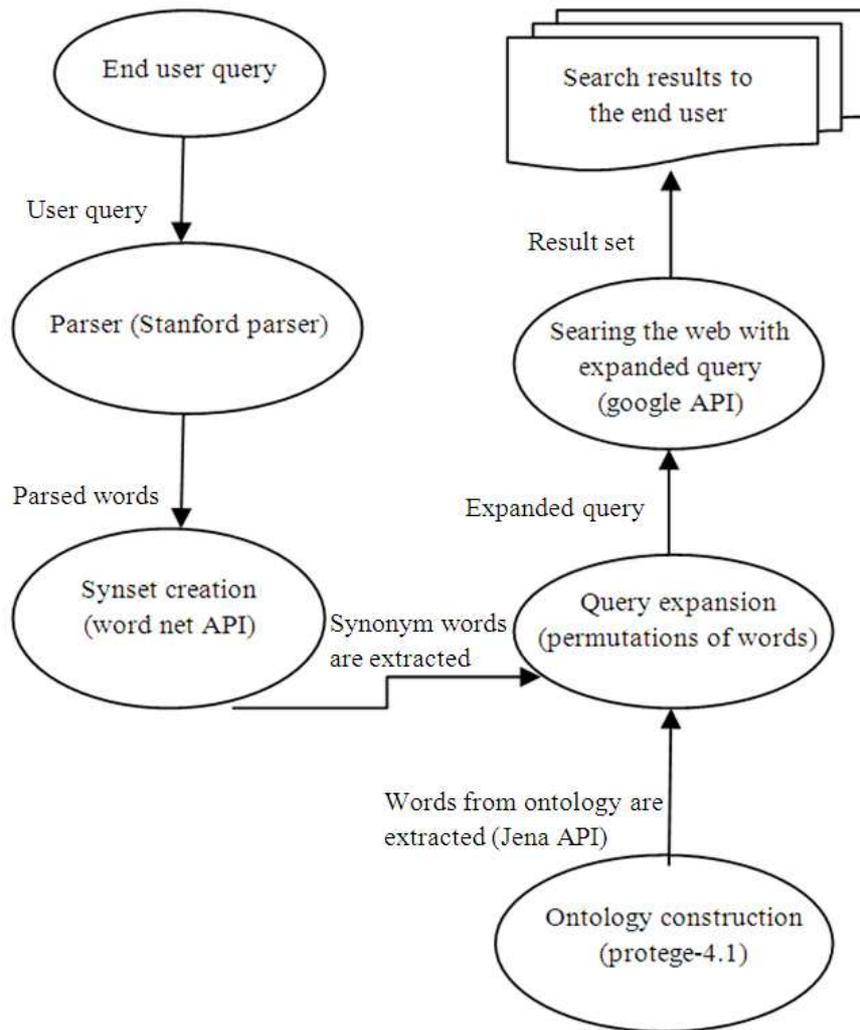


Fig. 1. System flow diagram

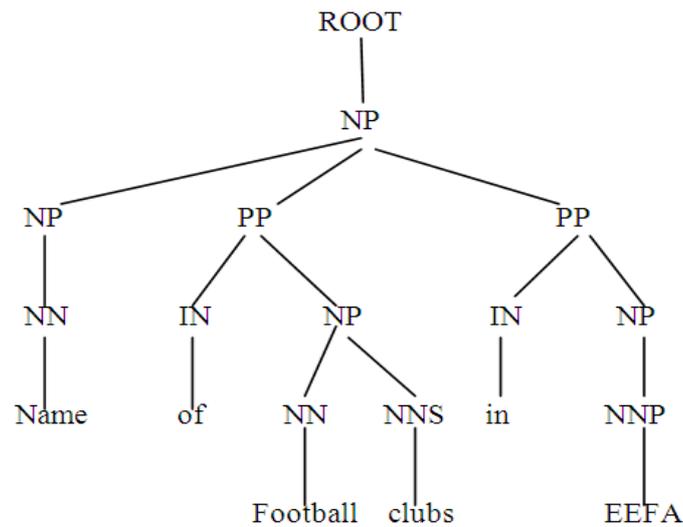


Fig. 2. Parsed Tree form of query

```
<!-- Ontology information -->
<owl: Ontology rdf: about = “ “>
< Rdfs: Label>sport vocabulary</ rdfs: Label>
<!-- Ontology information -->
<Owl: Ontology rdf: About = ““>
< Rdfs: Label>sport vocabulary</ rdfs: Label>
</Owl: Ontology>
<!--Class definitions-->
< Owl: Class rdf: ID = “team sports”>
< Rdfs: Sub class of rdf: Resource = “#sports”/>
< Rdfs: Label>team sports</rdfs: Label>
< Rdfs: Label xml: Lang = “pt”>esporte coletivo</rdfs:
Label>
< Owl: Disjoint with rdf: Resource = “#individual sports”/> <
/owl: Class>
```

Fig. 3. Sample RDF for sports class

User query: Name of football clubs in EEFA

Parsed words: (ROOT (NP (NP (NN name))
(PP (IN of) (NP (NN football))

Fig. 4. Parsed words of the input query

User query: Name of football clubs in EEFA

Expanded query: Name or list the football: Soccer clubs in
EEFA

Fig. 5. Expanded query of the input query

- Better Recall by using class hierarchies and rules. For example, a query for water sports in Spain would return results in scuba diving, windsurf and other subclasses, in Cádiz, Málaga, Almería and other Spanish locations
- Better precision by reducing polysemic ambiguities using instance labels and classifications of concepts and documents

To evaluate the performance of the system we have sampled eight queries and have expanded them. Searching was done using the expanded query with the help of Google Search API. The performance measures of precision and recall have been depicted in the Equation 3 and 4.

Measures Used

Recall

$$RECALL = \frac{\text{Number of relevant links given by the system}}{\text{Total number of relevant links}} \quad (3)$$

The Recall calculated here is the relative recall in which performance is compared with relation to the Traditional Google Search Engine.

Precision

$$PRECISION = \frac{\text{Number of relevant links given by the system}}{\text{Total number of links retrieved}} \quad (4)$$

The measure precision calculated here is the number of correct information that the system returns (accuracy).

When this SIRSD system was tested, an improvement in performance was observed as a result of semantic analysis. But a small fraction of them had a similar performance with a Traditional Search Engine. The Table 1 gives the Sample Queries and Table 2. Shows the samples of the Performance levels of our system comparing it with Traditional Search systems.

After the detailed analysis it is found that this query expansion algorithm gives the better performance improvement. It is proved that the precision and recall vales are high for all the sample queries. It gives an average of 40% increase in Precision value and 20 to 30% increase in recall value.

The Fig. 7 depicts that the precision value of our system SIRSD is higher than the values obtained in Traditional Search Engines. The recall value of our system is also increased than the values obtained in Traditional Search Engines. It is depicted in the Fig. 8.

The recall and precision Value Vs all the eight queries for both Semantic Search and the Traditional Search was done and it is shown in Fig. 9.

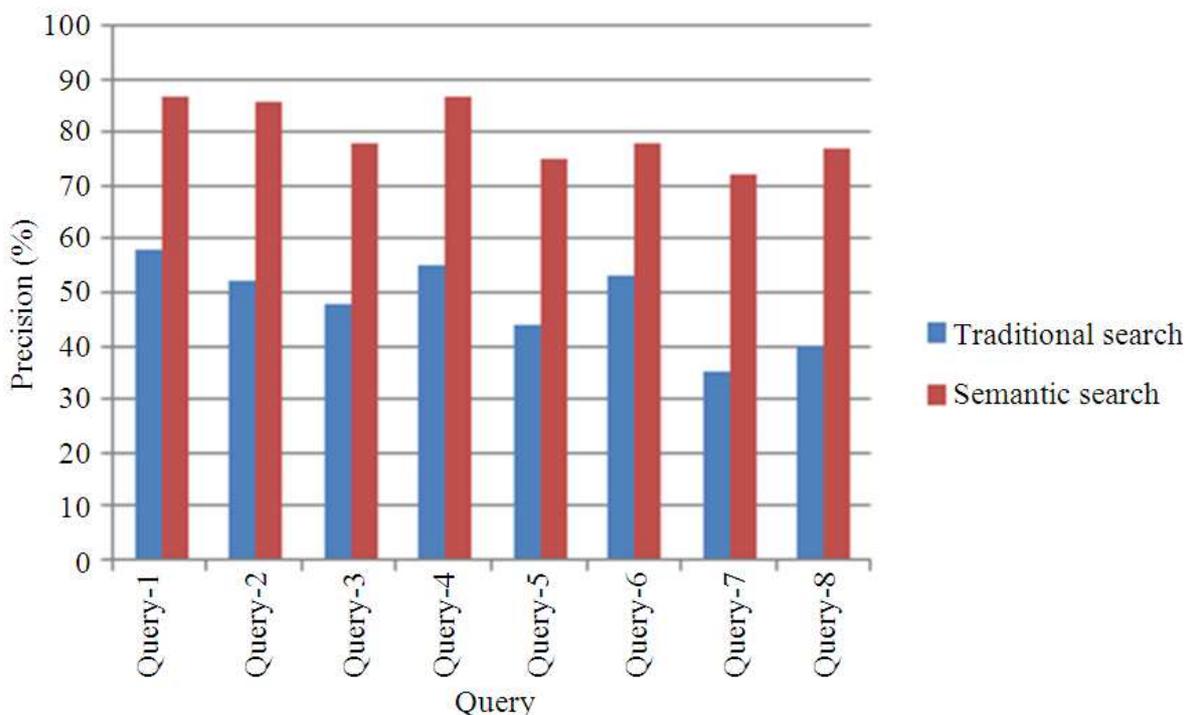


Fig. 7. Precision value of SIRSD and Traditional Search

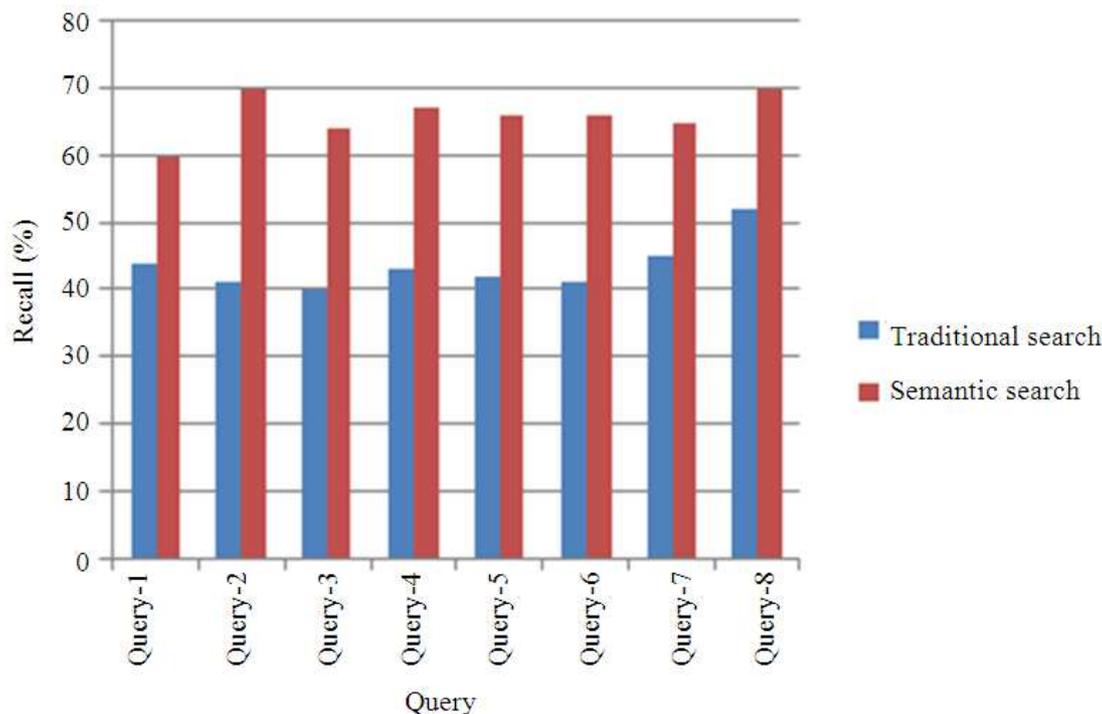


Fig. 8. Recall value of SIRSD and Traditional Search

Table 1. Evaluation queries

Query No.	User query
Query-1	List the team name in EEFA
Query-2	Events of olympics
Query-2	Persons who won medal in olympics
Query-4	Persons who won medal in chess
Query-5	What are the upcoming sports events in Europe?
Query-6	What is sports concepts?
Query-7	200 meter players
Query-8	Martial arts sports

Table 2. Precision and recall values of SIRSD and Traditional Search

Query no	Traditional Search		Semantic search by SIRSD	
	Precision (%)	Recall (%)	Precision (%)	Recall (%)
Query-1	58	44	87	60
Query-2	52	41	86	70
Query-3	48	40	78	64
Query-4	55	43	87	67
Query-5	44	42	75	66
Query-6	53	41	78	66
Query-7	35	45	72	65
Query-8	40	52	77	70

Discussion

In this study, it is found that the precision and recall value of our system is higher than the Traditional Search system. Ra *et al.* (2013)

constructed a military ontology for semantic data processing in the Army Tactical Command Information System (ATCIS). It can be used for constructing intelligent military information systems. Li *et al.* (2012) implemented a query expansion algorithm to improve the performance of image retrieval system by expanding the original query based on the information from top ranked images.

Chen *et al.* (2012) expanded the query using relevance feedback fusion to enhance IR effectiveness. Linguistic and structural information of Feedback documents have not been exploited to improve the performance. Faiz *et al.* (2012) proposed a learning information Retrieval system based on a semantic annotation process with contextual exploration. To improve the results, a machine learning technique is applied to sort the results according to their similarity with the request term. Kumar *et al.* (2010), developed an ontology for education domain using Protégé and retrieve University information.

Kotov and Zhai (2012) presented a study of the methods leveraging the concept net knowledge base for query expansion to improve the search results for poorly performing (or difficult) queries. When the initial search results are of very bad quality and other techniques such as relevance feedback and pseudo relevance feedback become ineffective. Luo *et al.* (2012) have done a Explicit Semantic Analysis (ESA) for query expansion.

Tanapaisankit and Song (2012) developed a personalized search system Query In Context (QIC), that was trying to enhance the individual search by incorporating user preferences in query expansion, capturing meanings embedded in documents. They were trying to explore other techniques to incorporate the semantic properties of terms in user queries.

Anh *et al.* (2012) proposed an approach of word sense disambiguation by creating a new model of semantic Network. They have summarized how to add semantics into a query in Vietnamese language based on Ontology of Object Member Property (OOMP) by improved Similar Noun Phrase Expansion (iSNPE) algorithm.

Gandhi and Srivatsa (2006) constructed an algorithm to improve the accuracy of Sessionizers in web usage analysis. Kanimozhi and Christy (2013) presented a system by incorporating ontology and SPARQL for semantic image annotation. Gandhi and Srivatsa (2007), done a Study report to resist web users against malicious activities using web logs.

Magesh and Thangaraj (2013), developed methods to retrieve the images from ontology and compare the image retrieval performance by using SPARQL query language, decision tree algorithm and Lire which is an open source image search engine. The SPARQL query language is used to retrieving the semantically relevant images using keyword based annotation and the decision tree algorithms are used in retrieving the relevant images using visual features of an image.

Sathianesan and Sankaranarayanan (2013) proposed blog summarizer to check the similarity of each sentence in a blog, sort the order of sentence based on the similarity of the text with query word

and reduces the number of sentences. This produces better relevant and summarized number of blogs compared to various search engines. SBMF confirms the hundred percentage relevancy compared to other blog search engines.

Vadivu and Hopper (2012), developed a Knowledge framework for the Indian Medicinal Plants (KIMP) that includes the ontology creation, user interface for querying the system. Jena is used to build semantic web applications and SPARQL Protocol and RDF Query Language (SPARQL) is used to retrieve various query patterns. This system achieved the automated mapping by considering lexical and edge based relatedness.

After the detailed analysis of the Fig. 7 to 9 we can infer that there is higher value of average precision and recall for the Semantic Search system when compared to Traditional Search. It depicts that our system has a better performance and accuracy in retrieving the results than the generic Search Engines. The average relative recall value of SIRSD system is also higher compared to that of the Traditional Search system. The higher value denotes the best coverage of our system compared to the generic search engines. Table 3 shows the average precision and recall values of both SIRSD and Traditional Search for the sample queries taken.

Table 3. Average precision and recall of SIRSD and Traditional Search

Average	SIRSD (%)	Traditional search (%)
Precision	80	48
Recall	66	44

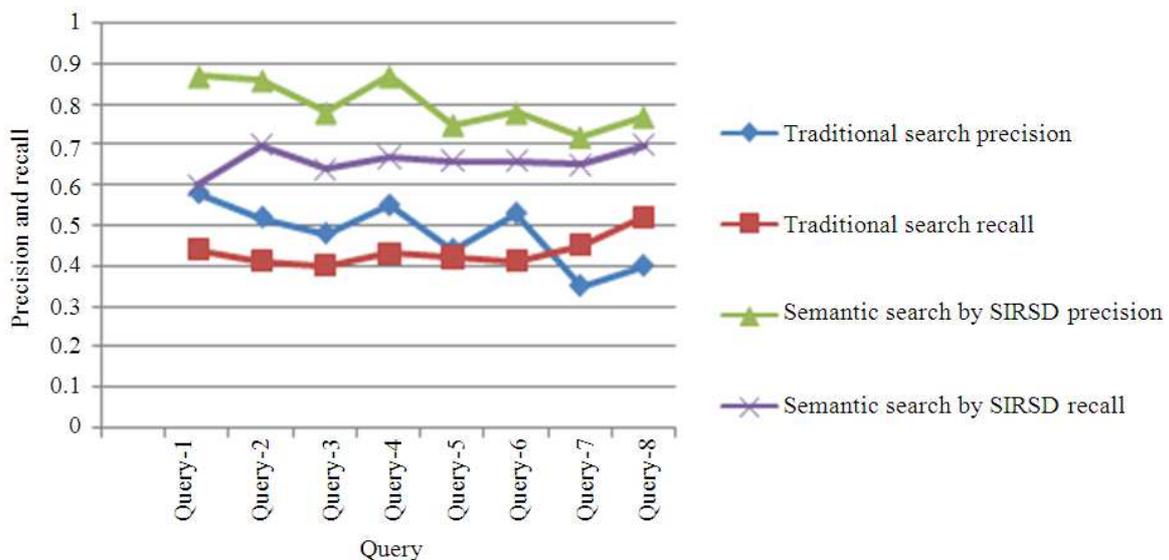


Fig. 9. Comparison of semantic search with Traditional Search (Precision and recall Vs queries)

Conclusion

Semantic relevant Information has been retrieved as a result of our SIRSD system. The system is successfully implemented and using the technology like Ontology, OWL, information extraction using Jena API. Query Expansion was introduced and Semantic Search was done which overcomes the drawback of Traditional Search. Considerable improvement in the performance of the system using domain specific information extraction is observed. As shown in the result section, it can achieve better Recall and Precision values for various domain and for multilingual databases. Other domain can be successfully implemented by having changes in the domain ontology and information extraction in line with sports domain. Other modules can be easily used in the new implementation without any changes. SIRSD reduces the issue of interoperability during the search of user need. Our ongoing research work is to use Semantic Similarity measures in query expansion, where a user query is modified using Semantically Similar words to improve the relevancy of the search. Multiliguality can be achieved by storing the Semantic Information from multiple languages.

Acknowledgement

We thank our management for timely support and help for the research. The authors would like to thank Mr C. Babu Rajan who helped in correcting the grammatical errors. We are also would like to thank the reviewers for their detailed, accurate and helpful comments.

Author's Contributions

All authors equally contributed in this work.

Ethics

This article is original and contains unpublished material. The corresponding author confirms that all of the other authors have read and approved the manuscript and no ethical issues involved.

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