

A Survey On Application Of Information Retrieval Models Using Nlp

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Abstract: *Text mining accustomed establish the hidden patterns and to get a vital data from the great deal of unstructured knowledge it's going to written material also. Different field techniques like machine learning, visualization, text analysis, database technology, statistics, knowledge management, natural language processing are incorporated in text mining. It's thought about to be mostly booming area next to big data and artificial intelligence. There exist varied techniques, tools and applications in text mining. Information retrieval thought-about being a eminent application of text mining. Information retrieval could be a key technology for knowledge management. Information retrieval is the method of getting and presenting additional connected data from the biggest assortment of knowledge resources in step with to the user's need. This paper is a survey of discussing concerning varied application of Information Retrieval using NLP and available methods of information retrieval. This survey discusses concerning assorted models used in Information retrieval. Here we tend to conjointly list the research contribution and limitations of Information retrieval models found in various articles.*

Keywords: *Text mining, Information retrieval, NLP, Artificial Intelligence*

1. INTRODUCTION

Information retrieval is mostly thought-about as a subfield of computer science that deals with the illustration, storage, and access of data [16],[17]. Information retrieval is worries with the organization and retrieval of information from massive database collections [21],[17]. It deals with the search for information and also the illustration, storage and organization of knowledge. Information retrieval is concerned with search processes in which a user needs to identify a subset of information which is relevant for his information need within a large amount of knowledge. The main goal of information retrieval system (IRS) is to "finding relevant information or a document that satisfies user information needs". [17] To attain this goal, IRSs typically follows the process like Indexing, filtering and Searching. The above three are the basic process in information retrieval. Documents are unit described summarized format in indexing. All stop and common words are removed in filtering steps. Searching is performed within the documents based on users need. The paper is organized in the following way section 2 discusses literature review. Section 3 discusses about various available models in information retrieval. Section 4 discusses about merits and demerits of information retrieval. Section 5 research gap and challenges in information retrieval and Section 6 gives conclusion.

LITERATURE REVIEW

ArpitDeo et al., [6] presented system; to extract the text from web documents, all markup language tags are removed. Then stop words and special characters are removed from extracted text for solely meaning contents. TF-IDF thought is employed for feature selection. Currently PSO optimization technique is employed for identifying and refining the features set, these selected features are stored in a database that is employed for information retrieval method.

AkramRoshdi et al., [7] represented completely different indexing methods for reducing search space associated different searching techniques for retrieving an information. This tend to are providing the summary of traditional IR models.

Xiaolu Luet al., [3] conferred QUEST, a technique which will answer complicated questions directly from textual sources on, by computing similarity joins over partial results from totally different documents. This technique is completely unsupervised, avoiding training-data bottlenecks and having the ability to cope with rapidly evolving ad hoc topics and formulation vogue in user questions.

Zongcheng Jiet al., [18] proposed formalizing short text conversation as a pursuit drawback at the primary step, and using progressive information retrieval (IR) techniques to hold out the task. It have tendency to investigate the importance more over because the limitation of the IR approach. The experiments demonstrate that the retrieval-based model will create the system behave rather "intelligently", once combined with an enormous repository of conversation data from social media.

AchilleSouilia et al., [15] has extracted information from patent documents. He followed four steps such as selected relevant text area. Segmented the paragraph based on matching tags, for parsing XML tags were used. Linguistic phenomena were represented with finite automata. Author considered graph generation as a step before the final step. Author proposed a automated method for extraction of IDM concepts from patent document. Regarding performance analysis author received good precision score, but recall score seems to be very low. This research made him to consider only the partial output. He extends his view that his work has not attained the maturity. The author expressed that he tried to provide assistance for performing mining task easier in patent documents, rather than replacing.

ParulKalra Bhatia et al., [20] planned a survey done on completely different algorithms that are being worked thus far on PIR systems. Completely different algorithms are being employed to retrieve data within the PIR systems.

Said A. Salloumet al., [9] author did survey in text mining in NLP with Facebook and Twitter, the two media which connects people, where people can communicate, share their views through chats. Most of the data found to be in unstructured format. According to author he decided to consider the result of this research as a base for his future work. Author says that text mining can be classified in to text clustering, classification, association rule mining and trend analysis. Author also says that in nearing future many researches is possible with text mining.

N I Widiastuti [4] in this research conducted in domains of text mining, NLP and deep learning. More than 50 articles used from a various portal of scientific articles. Author says that there exist much research gaps to develop deep learning specifically in text mining and NLP

Anu Bajaj et al., [2] mentioned the applications of information retrieval with deep learning e.g., web search by reducing the noise and aggregation precise results, trend detection in social media analytics, anomaly detection in music datasets and image retrieval etc.

Ahsan Mahmood et al., [7] has implemented Named Entity Recognition (NER) to identify peoples, locations and other entities considered to be one of the fundamental tasks of NLP. Author proposed a knowledge extraction framework to extract named entities from an Urdu book named Hadith. User can extract information through queries from this framework. Author intended to propose a framework and to apply vector space model in his future work.

Charu Virmaniet al., [8] has discussed various NLP approaches necessary for monitoring social network. Author had discussed various open source NLP libraries in this work. Most importantly author revealed many challenges in NLP like parts-of-speech, short context, noisy contents, extracting information about entities. Author had made his contribution by building smart system for analyzing social network information.

Sagar Gharge et al., [10] had contributed his research work in detecting malicious tweet using NLP. This system was developed using “Weka” tool and uses SVM machine learning algorithm in classifying the tweet as spam or not spam. In order to evaluate the accuracy of the system, author had chosen 1000 randomly tweet samples, out of which almost 60% were identified to be legitimate tweet and rest were found to be malicious tweet.

Ben He et al., [23] proposed Okapi BM25, logical model, language model, latent semantic indexing are used for document ranking in IR systems. Term proximity information is found to enhance performance of retrieval function. It's used for XML document retrieval that has a simulating internal structure within the sort of tree. To model this kind of information retrieval system, term proximity is calculated during a distributed environment with totally different resources.

Mourad Sarrouiet al., [11] presented in natural language processing based mostly retrieval system the query within the commencement is fed in to question processing stage. It extracts entity information from the query terms and alternative semantic data. Then the question is classified consistent with the domain, topic and also the entity of focus.

Florian et al., [22] the researchers presented a visual classifier training for text document retrieval with accurate filters. The active learning model reduces the effort of labeling but the support of multi-labeling is not focused.

Shaidah Jusoh [5] had made survey in NLP to identify application, techniques and challenging issues in NLP applications. Author used various papers for his survey from various databases such as Scopus, IEEE explorer and from Google. Where he found information extraction is the one and only most prominent NLP applications. He reviewed question answering system and automated text summarization in his research under NLP applications. He extends his views that ambiguity problem that exists in many languages found to be NLP challenge. Author listed information retrieval, query processing, advanced web search some NLP applications in his research contribution. The future technologies of NLP Chatbot, smart search, text mining, images, audios and videos, visualization of information from text documents like emails, Sms, and reviews and so on were highlighted in this paper.

Mahmoud Othman et al., [13] proposed his research work in opinion mining using NLP approach. Author proposed a system which classifies opinion in a document or set of documents. Different types of opinion were identified by the system includes categories such as opinionated, comparative, superlative and non-opinionated. Author tested his system with nearly 4000 sentences and evaluated his system with standard metrics such as precision, recall and F-score. Author had decided to focus in multi-language opinion mining in his future research work.

Sweta P. Lende et al., [14] developed question answering system which extracts information automatically for the asked questions. Author includes documents related to

education using NLP techniques. Author concludes that question answering information retrieval seems to be more complex when compared to other information retrieval system.

INFORMATION RETRIEVAL MODELS

An Information Retrieval (IR) model specifies the details of the document representation, the query representation and the retrieval functionality. The fundamental IR models can be classified into Boolean, vector, probabilistic and inference network model [24].

Boolean Model

In this model, the query is depicted by Boolean expression of terms and therefore the terms are unit connected with Boolean operators. The model may be explained by thinking of a query term as an unambiguous definition of a set of documents. As an example, the query term C merely the set of all documents that's indexed with the term C. The Boolean model permits for the use of operators of Boolean algebra, AND, OR and NOT, for query formulation, however has one major disadvantage: a Boolean system is notable to rank the came back list of documents [26]. Within the Boolean model, a document is related to a set of keywords. Queries are also expressions of keywords separated by AND, OR, or NOT. The retrieval perform during this model treats a document as either relevant or irrelevant [24]. In Figure 1, the retrieved sets are visualised by the shaded areas.

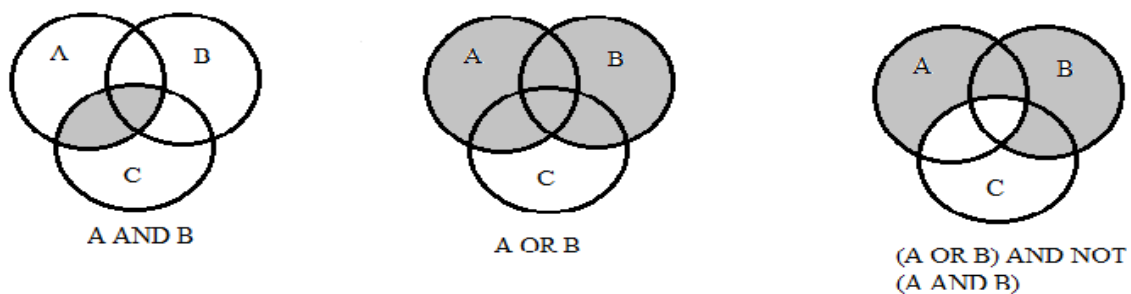


Fig 1. Boolean Combinations of sets visualized as Venn diagrams

Vector Space Model

In this model, the word and phrases are known as terms and these terms are represented in form of vectors. The vector space model can best be characterized by its attempt to rank documents by the similarity between the query and each document [27]. In the Vector Space Model, documents and query are represented as a Vector and the angle between the two vectors are computed using the similarity cosine function. Similarity Cosine function can be defined as:

Where,

$$sim(d_j, q) = \frac{d_j \cdot q}{\|d_j\| \cdot \|q\|} = \frac{\sum_{i=1}^N w_{i,j} w_{i,q}}{\sqrt{\sum_{i=1}^N w_{i,j}^2} \sqrt{\sum_{i=1}^N w_{i,q}^2}} \dots\dots\dots(1)$$

Documents and queries are represented as vectors

$$d_j = (w_{1,j}, w_{2,j}, \dots, w_{t,j})$$

$$q = (w_{1,q}, w_{2,q}, \dots, w_{t,q}) \dots\dots\dots (2)$$

Vector Space Model have weight scheme referred to as tf-idf weighting. These weights have a term frequency (tf) factor measuring the frequency of incidence of the terms within the document or query texts associated degree an inverse document frequency (idf) factor measuring the inverse of amount of documents that contain a query or document term [26].

Probabilistic Model

It assesses the chance of significance to the query. The documents are unit ordered by decreasing probability of their significance known as probability ranking principle. The foremost vital characteristic of the probabilistic model is its decision to rank documents by their probability of connexion in a given query [28]. Documents and queries are depicted by binary vectors $\sim d$ and $\sim q$, every vector part indicating whether or not a document attribute or term happens within the document or query, or not. Rather than probabilities, the probabilistic model uses odds $O(R)$, where R suggest the "document is relevant" and \bar{R} means "document is not relevant" $O(R) = P(R)/1 - P(R)$ [26].

Inference Network Model

The documents are modelled using the inference process in the inference networks. The documents are graded in line with the term strength. In this model, document retrieval is modeled as an inference process in an inference network. [29] Most techniques employed by IR systems will be enforced underneath this model. Within the simplest implementation of this model, a document instantiates a term with sure strength, and also the credit from multiple terms is accumulated given a query to figure the equivalent of a numeric score for the document. From an associate degree operational perspective, the strength of internal representation of a term for a document will be thought of as the weight of the term within the document, and document ranking within the simplest sort of this model becomes almost like to ranking within the vector space model and also the probabilistic models represented on top of. The strength of internal representation of a term for a document isn't outlined by the model, and any formulation will be used.

Query based model

Fig. 2 shows Information Retrieval system architecture [25], [19]. In this figure, the user who needs Information issues a query (**userquery**) to the retrieval system through the query operations module. The retrieval module uses the document index to retrieve those documents that contain some query terms. Compute relevance scores for them, then rank the retrieved documents according to the scores. The ranked documents are then presented to the user. The document collection is also called the text database, which is indexed by the indexer for efficient retrieval [25],[19]. The objective of any IR system is to produce a list of relevant documents to the user information need or query provided by the user. The user usually needs relevant documents even if the exact terms he/she used in the provided query were not present in these documents.

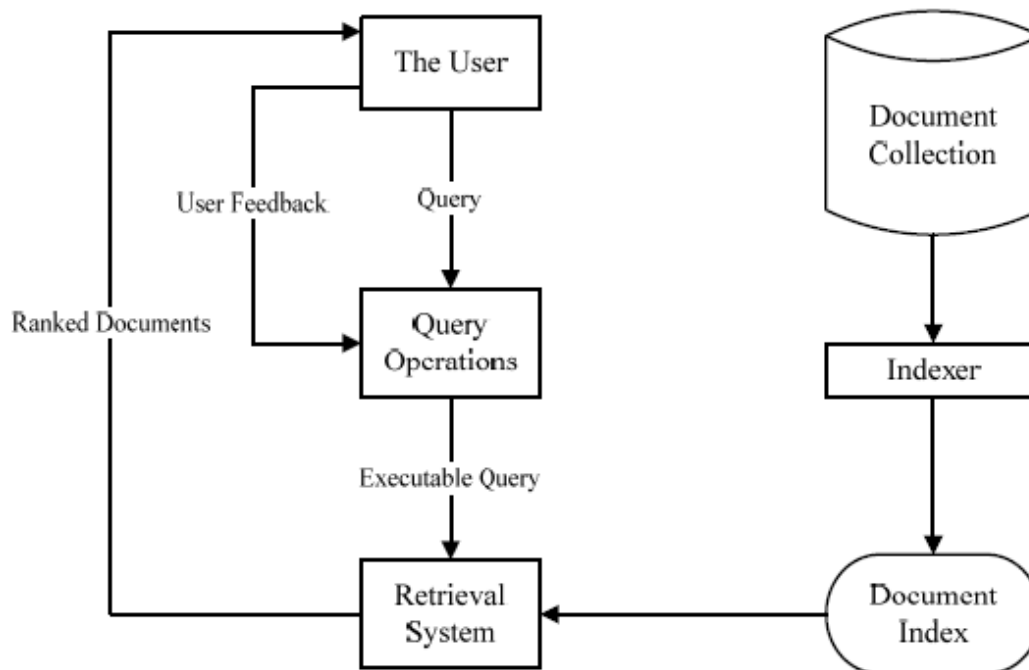


Fig 2. Information Retrieval system architecture

2. RESEARCH CONTRIBUTION AND LIMITATIONS OF INFORMATION RETRIEVAL

Authors	IR Models used	Purpose of Information Retrieval	Method summary and Research Contribution	Limitations
B. Piwowarski, 2016	Probabilistic model	To learn the term weight directly from the occurrences of terms in documents	This has the potential to capture patterns of occurrences that are linked with the importance of a term in a document and a collection.	By not computing term weights for frequent bi or tri-grams.
Dwaipayan Roy et al., 2016	Vector Space model	To develop a similarity metric that makes use of the similarities between the individual embedded word vectors in a	This approach to represent queries and documents as sets of embedded word vectors. A document is diagrammatical as a chance density perform of a	Proposed method gives MAP by up to 5:77%, in comparison to standard text-based language model similarity on

		document and a query.	combination of Gaussians.	test collection.
Chiranjeevi H S et al.,2020	TF-IDF (Term Frequency-Inverse Document Frequency) Model	A continual Convolution Neural network (RCNN), based mostly text data retrieval system which efficiency retrieves the text documents and data for the user query.	The proposed methodology is not limited to a domain and text document retrieval system. Performance results are extremely promising for organizations that want to extend the customer service effectively.	Nil
Anu Bajaj et al., 2020	Statistical model	The retrieved documents are ranked with the estimate of importance of document for a particular query	Deep learning automatically select raw, heterogenous, high dimensional data, without manual selection. The automatic word detection and topic allocation supported LSTM-RNN for document retrieval.	Nil
Anjali Goyal et al., 2017	SVM-TF-IDF	The information retrieval used for bug report assignment.	Results of the analysis showed a rise of up to 12.8% with in the efficiency for the top-5 list size in the information retrieval based technique	An information retrieval based technique which is not considers the time-based expertise computation
Kamal Sarkar et al., 2015	TF-IDF	This paper presents an evaluation and an analysis of some selected information retrieval models for Bengali	IR models are often enhanced with adding the power to acknowledge recognize synonyms and/or recognizing	Nil

		monolingual information retrieval task	similar phrases.	
ChahinezBenkoussas et al., 2015	probabilistic model	This work tackles the problem of book recommendation in the context of INEX (Initiative for the Evaluation of XML retrieval) Social Book Search track	A proposed a novel approach based on Directed Graph of Documents (DGD). It exploits social link structure to enrich the returned document list by traditional retrieval model	It is not explore citation links between scientific documents

RESEARCH GAP and CHALLENGES in INFORMATION RETRIEVAL

RESEARCH GAP in INFORMATION RETRIEVAL with NLP

- Some of the research gaps listed was extracted from the existing work
- To focus Computation in terms of weights for frequent bi or tri-grams.
- To increase MAP of Vector space model up to the MAP of standard text-based language model
- To consider time-based expertise in computation of information retrieval
- To include citation links while exploring scientific documents
- To focus in multi-language opinion mining in his future research work.
- To develop deep learning specifically in text mining and NLP

CHALLENGES in INFORMATION RETRIEVAL with NLP

- Some of the challenges in NLP are
- Finding Parts- of-speech
- Working with Short context
- Information retrieval from Noisy contents
- Extracting information about entities.
- Ambiguity problem that exists in many languages found to be popular NLP challenge
- Question and answering (query based) information retrieval seems to be more complex and challenging when compared to other information retrieval system.
- Most of the social network data are in unstructured format. Handling un-structured data for information retrieval in NLP found to be more complex and Challenging.

3. CONCLUSION

Information retrieval in NLP found to be one of the booming research areas. Information retrieval model helps to extract necessary information. This model can be used for different applications. Each model has its own merits and demerits. It provides many new research ideas for researchers. Future research focus of information retrieval can be in using Deep learning specifically in text mining and NLP. Research in text mining can be classified in to text clustering, classification, association rule mining and trend analysis in near future. In future

many researches are possible with text mining.

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