

Text Information Retrieval Using Data Mining Clustering Technique

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Abstract

In the aspects of mining, it is used to extract the data's in the efficient manner to reduce the searching time of the user. In data mining cluster play an important role to group the information effectively. The current aspect is to clustering the sentence level text by using the Fuzzy Relational Clustering Algorithm .By means of that, it allows patterns to all clusters .we give a text in a sentence or a sentence that has to be relatively present in documents or a set of documents. An advantage of Fuzzy Relational Clustering Algorithm is that it assigns soft membership values, which can be interpreted as a measure of the degree to which an object belongs to each of the clusters. It is a generic fuzzy clustering algorithm that can be principle be applied to any relational clustering problem, and application to several non-sentence data sets has shown its performance to be comparable to Spectral Clustering. This algorithm that operates on relational input data. Thus the results of applying the algorithm to overlapping clusters of semantically related sentences.it operate the data in the form of similarities. Between the data and the data objects by using the novel fuzzy clustering pattern.

Key terms: Data mining, Clustering, Fuzzy Clustering, Spectral Clustering, Sentence level clustering.

Introduction

Sentence Clustering

Sentence clustering aims at grouping sentences with similar meanings into clusters; commonly, vector similarity measures, such as cosine, are used to define the level of similarity over bag-of-words encoding of the sentences. Then, standard clustering algorithms can be applied to group sentences into clusters. The main challenge for any sentence clustering approach is language variability, where the same meaning can also be listed in different technique.

Sentence level clustering is an application of text classification. The most common objectives in text classification are to classify texts into fairly objective categories such as topics; Clustering has become an increasingly important topic with the explosion of information available via the Internet. Clustering is important it used to group the information to reduce the searching time. Its ability to automatically group similar textual objects together enables one to discover similarity between the given inputs.

Methods used for text clustering include Market Analysis, Financial for costing, grouping data by using logics and simple rule-based systems among the text. In text clustering, it is important to gives the input text properly.

In this paper we focus the text clustering base on two features.

1. Generating relevant term clusters based on lexical semantic relatedness.
2. Projecting the sentence set over these term clusters.

Steps To Obtain Term Cluster:

In order to obtain term clusters, a term connectivity graph is constructed for the given sentence set and is clustered as follows:

1. Create initially an undirected graph with sentence-set terms as nodes and use lexical resources to extract semantically-related terms for each node.
2. Augment the graph nodes with the extracted terms and connect semantically-related nodes with edges. Then, partition the graph into term clusters through a graph clustering algorithm.

Projecting Sentence Ot Term Cluster

To obtain sentence clusters, the given sentence set has to be projected in some manner over the term clusters obtained in Step 1. Our projection procedure resembles unsupervised text categorization (Gliozzo et al., 2005); with categories represented by term clusters that are not predefined but rather emerge from the analyzed data:

1. Represent term clusters and sentences as vectors in term space and calculate the similarity of each sentence with each of the term clusters.
2. Assign each sentence to the best-scoring term cluster. (We focus on hard clustering, but the procedure can be adapted for soft clustering).

Literature Review

Pedrycz, W, Winnipeg, Man, Waletzky, J. et al.[1]proposed a “**Fuzzy clustering with partial supervision**” “Presented here is a problem of fuzzy clustering with partial supervision, i.e., unsupervised learning completed in the presence of some labeled patterns. The classification information is incorporated additively as a part of an objective function utilized in the standard FUZZY ISODATA.

Yuhua Li, David McLean, Zuhair A. Bandar, James D. O’Shea, and Keeley Crockett et al. [2]proposed a “**Sentence Similarity Based on Semantic Nets and Corpus Statistics**” This paper focuses directly on computing the similarity between very short texts of sentence length. It presents an algorithm that takes account of semantic information and word order information implied in the sentences. The

semantic similarity of two sentences is calculated using information from a structured lexical database and from corpus statistics. The use of a lexical database enables our method to model human common sense knowledge and the incorporation of corpus statistics allows our method to be adaptable to different domains. The proposed method can be used in a variety of applications that involve text knowledge representation and discovery.

Lu, Shin-Yee, Fu, King Sun, et al. [3] proposed an “**A Sentence-to-Sentence Clustering Procedure for Pattern Analysis**” Cluster analysis for patterns represented by sentences is investigated. The similarity between patterns is expressed in terms of the distance between their corresponding sentences. A weighted distance between two strings is defined and its probabilistic interpretation given. The class membership of an input pattern (sentence) is determined according to the nearest neighbor or k-nearest neighbor rule. A clustering procedure on a sentence-to-sentence basis is proposed

Xiaoyan Cai, Wenjie Li, You Ouyang, et al. [4] proposed a “**Simultaneous ranking and clustering of sentences: a reinforcement approach to multi-document summarization**” “Multi-document summarization aims to produce a concise summary that contains salient information from a set of source documents. In this field, sentence ranking has hitherto been the issue of most concern. Since documents often cover a number of topic themes with each theme represented by a cluster of highly related sentences, sentence clustering was recently explored in the literature in order to provide more informative summaries. Existing cluster-based ranking approaches applied clustering and ranking in isolation. As a result, the ranking performance will be inevitably influenced by the clustering result. In this paper, we propose a reinforcement approach that tightly integrates ranking and clustering by mutually and simultaneously updating each other so that the performance of both can be improved. Experimental results on the DUC datasets demonstrate its effectiveness and robustness.

Richard Khoury, et al. [5] proposed a “**Sentence Clustering Using Parts-of-Speech**” “Clustering algorithms are used in many Natural Language Processing (NLP) tasks. They have proven to be popular and effective tools to use to discover groups of similar linguistic items. In this exploratory paper, we propose a new clustering algorithm to automatically cluster together similar sentences based on the sentences’ part-of-speech syntax. The algorithm generates and merges together the clusters using a syntactic similarity metric based on a hierarchical organization of the parts-of-speech. They demonstrate the features of this algorithm by implementing it in a question type classification system, in order to determine the positive or negative impact of different changes to the algorithm

Extracting and Filtering Related Terms

In a number of lexical resources providing pairs of semantically-related terms; within the suggested scheme, any combination of resources may be utilized. Often resources contain terms, which are semantically-related only in certain contexts. E.g., the words *visa* and *passport* are semantically-related when talking about tourism, but cannot be

considered related in the banking domain, where visa usually occurs in its credit card sense.

In order to discard irrelevant terms, filtering procedures can be employed. E.g., a simple filtering applicable in most cases of sentence clustering in a specific domain would discard candidate related terms, which do not occur sufficiently frequently in a target-domain corpus. In the example above, this procedure would allow avoiding the insertion of passport as related to visa, when considering the banking domain.

Clustering the graph nodes:

Once the term graph is constructed, a graph clustering algorithm is applied resulting in a partition of the graph nodes (terms) into clusters. The choice of a particular algorithm is a parameter of the scheme. Many clustering algorithms consider the graph's edge weights.

To address this trait, different edge weights can be assigned, reflecting the level of confidence that the two terms are indeed validly related and the reliability of the resource, which suggested the corresponding edge (e.g. Word Net synonyms are commonly considered more reliable than statistical thesauri).

Fuzzy Clustering:

Data Clustering in the process of grouping the element in one class or one label. Fuzzy clustering is a class of algorithms for cluster analysis in which the allocation of data points to clusters is not "hard" (all-or-nothing) but "fuzzy" in the same sense as fuzzy logic. In fuzzy clustering (also referred to as soft clustering), data elements can belong to more than one cluster.

Problem Definition

General

The purpose of the system analysis phase of a project is to analysis the problem statement and finds the additional requirements in the project. In the system analysis phase, the project team defines the problem statement in the project.

Existing System

Existing system uses novel fuzzy clustering algorithm that operates on relational input data; the input data is form of data in the form of a square surrounding substance of data objects. The algorithm uses a graphical method of the input data, that perform in an Expectation-Maximization framework in which the graph centrality of an object in the diagram is interpreted as likelihood. Output of the algorithm applied to sentence clustering tasks demonstrate that the algorithm is capable of identifying overlapping clusters of semantically related sentences. The major disadvantage of this process is time and space complexity because of fuzzy clustering mechanism. The existing system uses page ranking mechanism to avoid time complexity but the result is not efficient.

Disadvantages of Existing System

The results often suffered from instability in the optimization algorithms that were used.

A limitation of existing approach is the high dimensionality introduced by representing objects in terms of their similarity with all other object.

Proposed System

The proposed system implements the fuzzy logic with the clustering algorithm like the existing algorithm but instead of using the page ranking mechanism the proposed system uses the N-GRAM preprocessing mechanism which includes stop word removal, stemming etc. to avoid the space and time complexity.

Advantage of Proposed System

- Relational clustering is achieved
- Accurate Search result
- Increase in search speed
- Identifying overlapping clusters.

Experimental Setup

Pre-Processing

This remove stop words such as prepositions, pronouns, articles, and irrelevant document metadata. Then, adopted a traditional statistical approach for text mining, in which documents are represented in a vector space model. A dimensionality reduction technique used known as Term Variance (TV) that can increase both the effectiveness and efficiency of clustering algorithms. TV selects a number of attributes (in our case 100 words) that have the greatest variances over the documents.

Calculating the number of clusters:

This project implements the approach consists of getting a set of data partitions with different numbers of clusters. A set of partitions result directly from multiple runs of a partitioned algorithm (e.g., K-means) starting from different numbers and initial positions of the cluster prototypes.

Clustering techniques

In this work partitioned K-means and K-medoids, the hierarchical Single/Complete/Average Link, and the cluster ensemble based algorithm known as CSPA. For instance, K-medoids is similar to K-means. However, instead of computing centroids, it uses medoids, which are the representative objects of the clusters.

Removing Outliers

Removing Outliers makes recursive use of the silhouette. Fundamentally, if the best partition chosen by the silhouette has singletons (i.e., clusters formed by a single object only), these are removed. Then, the clustering process is repeated over and over

again—until a partition without singletons is found. At the end of the process, all singletons are incorporated into the resulting data partition (for evaluation purposes) as single clusters.

Experimental Results

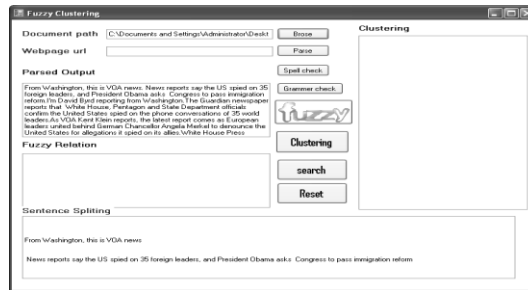


Figure 2: Pre-Processing Step

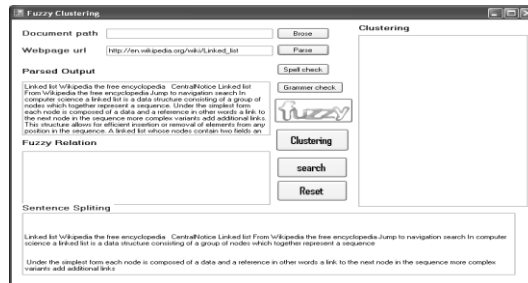


Figure 3: Calculating the number of clusters

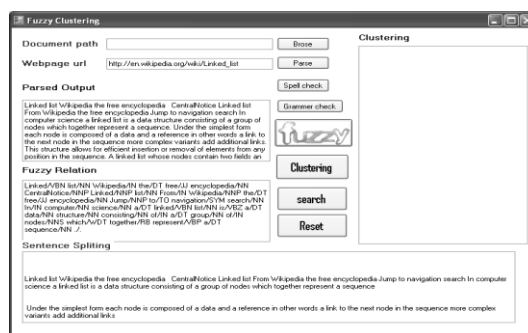


Figure 4: Clustering techniques

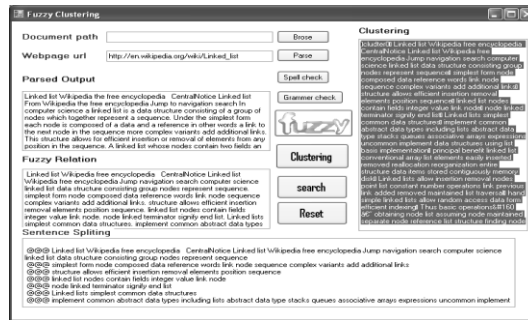


Figure 5: Cluster Formation



Figure 5: Removing Outliers

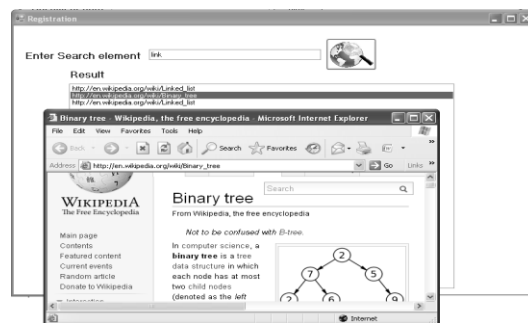


Figure 6: Clustering Search

Conclusion

The results we have presented show that the algorithm is able to achieve superior performance to benchmark Spectral Clustering and k-Medoids algorithms when externally evaluated in hard clustering mode on a challenging data set , and applying the algorithm to a recent news article has demonstrated that the algorithm is capable of identifying overlapping clusters of semantically related sentences. Has shown its performance to be comparable to Spectral Clustering and k-Medoid benchmarks. The algorithm can also be used within more general text mining settings such as user

query intended for text mining. Compare with any existing clustering algorithm, performance of the proposed system will ultimately depend on the input given by the user.

Future Enhancement: Any such improvements are orthogonal to the clustering model, and can be easily integrated into it.

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