



Query Based Text Summarization

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Abstract: Now a day's, The advancement of technology implies the large amount of online textual Information, this leads to a need for Text Summarizers which can provide important textual data from large data source like www, into a well-structured document formats. It becomes an alive analysis activity to specify the related data that is retrieved from huge set of documents which are established from various sectors like social media. Text summary is the action of precise the documents by protecting the key concepts of the text. This process can be accomplished via extractive summary and abstractive summary. In this paper, we suggested summarization technique that is to be applied on online information depending upon the User required Document.

Index Terms: Abstractive Summary, Extractive Summary, Information Retrieval, Text Summarizer.

I. INTRODUCTION

An innovation in Natural Language Processing with machine learning algorithms for language processing was initiated in the late 1980's. This process was implemented due to discouragement of set of corpus syntax which regulates the machine learning methods to process the language. Some of the primitive usage of machine learning algorithms such as decision trees, hard if-then rules are similar to existing hand-written rules. Increasing analysis has focused on statistical methods, which generate easy, probability decisions based on real-valued weights to the features based on input data. Recently, analysis has progressively focused on unsupervised and semi-supervised learning algorithms which are able to learn from data that has a combination of annotated and non-annotated data. In General, this function is more difficult than supervised learning, and it typically produces less accurate results for given input data. Increasing the availability of online data has demand in the area of automated text summarization within the NLP group [1]. The difficulty has been labelled from different outlook, in various domains and models. Text summary is the process of converting a shorter form of the given larger textual information which contains only vital information and thus, it helps the user to understand the text in a little time period. The main advantage is it reduces user's time in searching the important details in the document [2]. There are mainly two ways to summarize textual documents.

1. Extractive Way: It involves selecting phrases and sentences from the original text and including it in the final summary.
2. Abstractive Way: The Abstractive method involves achieving exclusively new phrases and sentences to grab the meaning of source document.

II. METHODOLOGY

In this section we describe the each and every step that we perform while we achieve our task to accomplish our objectives. The scenarios are as follows

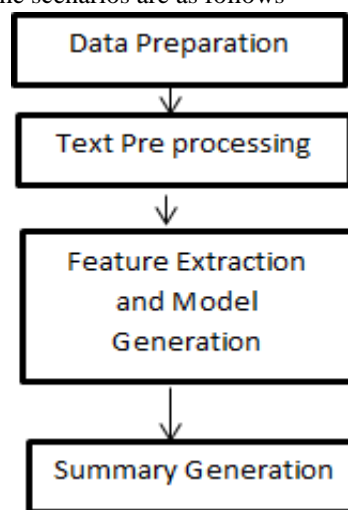


Fig I. Flow chart

1. Data Preparation

The prior step of our project is data preparation. Here data is collection from large amount among multiple sources through a query from online resources [3].

2. Text Pre-processing

The further step after data collection is text pre-processing [4]. As we collect data in the form of text we have to pre-process it. Text pre-processing is a vital part in all text processing functions. It contains some vital parts such as tokenization, Removal of stop words, stemming.

A. Tokenization

It basically refers to splitting a larger body of text into smaller lines, words or even creating words of non-English language. There are two types of tokenization [5]:
Word Tokenization: splitting of given information into meaningful words.

Sentence Tokenization: splitting of information into meaningful sentences [6]. Now, we are using sentence and word tokenizers for feature extraction demonstration.

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B. Removal of Stop Words

The words that are appeared repeatedly and regularly which doesn't make any sense [7] like "a", "an", "the" and so on are referred as stop words. So, in this step we can remove those words from the text.

C. Stemming

Stemming is a process that identifies root words, which diminishes the words into their roots [8]. For example, the words "is", "am" and "are" are converted to their root, that is, "be".

3. Feature extraction and Model Generation

It is the process of identifying key features from huge amount of dataset. In this paper, the most explanatory sentences are analyzed and extracted from the given text. The main goal is to choose which sentences are necessary [9][10]. In order to attain, some key features should be extracted from each sentence and there weights are calculated depends on its feature values. Then the important sentences are selected to be present in the summary. Important sentences can be preferred by using word weights and sentence weights using vectors concept [11][12].

A. Word Vector

It is numbered vectors that describe the word meaning. The main advantage of characterizing words as vectors is that they add itself to mathematical operations. It helps to minimize the loss of its predictions. It can be describe words as complex, stable floating point numbers where semantically identical words are charted to proximate points in geometric space. (i.e, it is defined as a row of real valued numbers where each point identifies as a dimension of its word meaning and where semantically identical words have similar vectors is called Word Vector [12]).

B. Sentence Vector

It is defined as an assignment of each sentence to its index and it can be considered as index to any other word [2][3]. The following figure describes the process in detail.

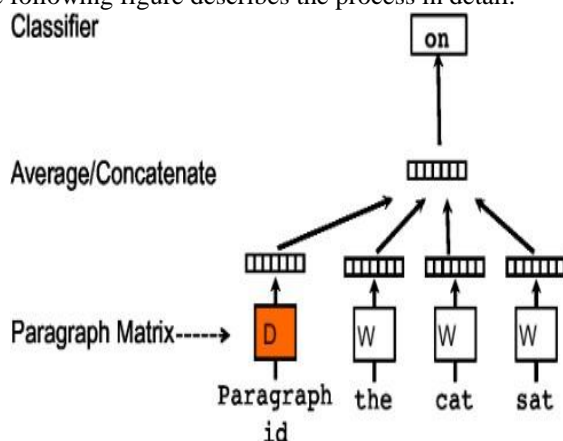


Fig II: Sentence vector

Substantially, every paragraph (or sentence) is graphed to a rare vector, and the combined paragraph and word vectors are helpful to predict the next word.

depend on their weights. Topmost sentences are preferred to produce the summary.

III. RESULTS

Let us consider an example of user required query is Text summarization techniques when we perform the same example on proposed model as follows:
 Enter a query sentence: Text summarization techniques
 Performing Google search for query: "Text summarization Techniques"
 Extracted text length from web for your given query: 1098
 Calculated length of summary: 274

IV. CONCLUSION

In this paper we study various models to perform summarization. The main protest of text summary is not only balancing query related data word weights and sentence weights can be calculated based on context. In order to improve the scope of generating the summary from set of multiple domains.

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4. Summary Generation

The further step after the completion of features extraction is summary generation; all over sentences are identified

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