NEWS ARTICLE SUMMARIZATION AND CLASSIFICATION

PROJECT REPORT

Submitted for CAL in B.Tech –Natural Language Processing(CSE4022)

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ABSTRACT

Every day, people rely on a wide variety of sources to stay informed -- from news stories to social media posts to search results. Being able to develop Machine Learning models that can automatically deliver accurate summaries of longer text can be useful for digesting such large amounts of information in a compressed form. Automatic systems based on extractive summarization techniques select the most significant sentences of one or more texts to generate a summary. The amount of information available electronically is growing. So there is increasing demand for automatic method for text summarization. Text summarization is process of reducing the size of text document while preserving its information content. Providing concise article to the new generation where people don't have the time to read about news article these summarized content can provide information as well as knowledge and keep the younger generation aware.

1. Introduction:

1.1 **Objective and goal of the project**

In today's world Big data is widely used for publishing news on the internet or website. To thoroughly understand an event, we have to read massive reports and keep clues in mind, which is very difficult and usually results is a one sided interpretation. To see an occasion, individuals need to seek different reports, coax out pieces of information, along these lines to make a honest judgment dependent on their learning and feelings. While, as an expansive number of news reports keeps on being opened on the Web, particularly for mainstream occasions, the assignment of seeing all parts of the occasion turns into a troublesome. In this way, we fall back on automatic summarization system, which can gather gigantic reports appropriated in the entire Web, and give a brief outline of what occurred. Keeping people aware of all the news all around the world and mostly the younger generation who lead a digital life is the main objective and the goal of the project. Creating awareness in today's younger generation is very important because they are going to be the future for the nation.

2. Methodology: Experimental/Simulation

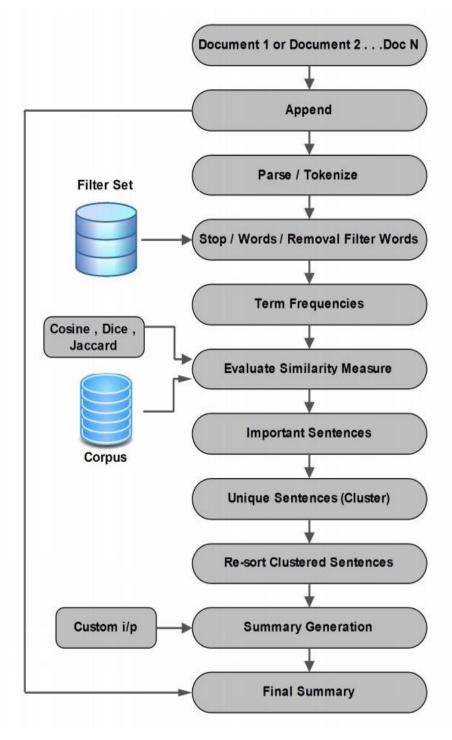


Fig. Proposed methodology for news article summarization.

The process of building the project is as follows –

Data Cleaning:

- Selection of news article from the internet. In the first step news articles which are required to be summarized are given by the user. We have taken the dataset and cleaned the dataset by removing the contractions words (like I'm, we'll), apostrophe, exclamation other symbols, numbers, question marks etc.
- **Append and Tokenization** Given a character sequence and a defined document unit, tokenization is the task of chopping it up into pieces, called tokens, perhaps at the same time throwing away certain characters, such as punctuation. News articles are appended and then the file content is tokenized into individual word.
- Lemmatization and Removal of Stop Words –After word is divided into smaller tokens we find out the root of a word. Lemmatization helps us to achieve the root forms (sometimes called synonyms in search context) of inflected (derived) words. A lemmatizer, a tool from Natural Language Processing which does full morphological analysis to accurately identify the lemma for each word. Doing full morphological analysis produces at most very modest benefits for retrieval. There are words which give us little or no context related information and such words are also removed from the list

Classification:

For classification we used random forest algorithm. It is a supervised classification algorithm. As the name suggest, this algorithm creates the forest with a number of trees. It is ensemble algorithm. *Ensembled algorithms* are those which combines more than one algorithms of same or different kind for classifying objects. Random forest is a meta estimator that fits a number of decision tree classifiers on various sub-samples of the dataset and uses averaging to improve the predictive accuracy and control over-fitting. The sub-sample size is always the same as the original input sample size.

Given the training dataset with targets and features, the decision tree algorithm will come up with some set of rules. The same set rules can be used to perform the prediction on the test dataset.

It works in four steps:

- 1. Select random samples from a given dataset.
- **2.** Construct a decision tree for each sample and get a prediction result from each decision tree.
- **3.** Perform a vote for each predicted result.
- **4.** Select the prediction result with the most votes as the final prediction.

We have taken test size to be 20% and test size to be 80% with the number of estimators=300 and depth of tree to be 150 for the better classification of dataset. Once the model is trained we are storing the model in the pickle file for using the trained model on the other dataset with help of library function job.lib.dump(joblib.dump(value, filename, compress=0, protocol=None, cache_size=None)) which returns The list of file names in which the data is stored. If compress is false, each array is stored in a different file.

Summary: Sentence Generation

- Generation of List of Frequent Words(BOW) After eliminating stop words the term-frequent data and inverse document frequency is calculated from text documents and frequent terms are selected which are used to generate text document summary for which we used the concept of Bag of Words. It refers to the representation of text which describes the presence of words within the text data. The intuition behind this is that two similar text fields will contain similar kind of words, and will therefore have a similar bag of words. Further, that from the text alone we can learn something about the meaning of the document.
- Cosine Similarity: Similarity measure is evaluated using cosine algorithm and important sentences are generated. Cosine similarity functions in way that it, generates metric that can be used to measure how similar the documents are irrespective of their size. Mathematically, it measures the cosine of the angle between two vectors projected in a multi-dimensional space. The cosine similarity is advantageous because even if the two similar documents are far apart by the

Euclidean distance (due to the size of the document), chances are they may still be oriented closer together. The smaller the angle, higher the cosine similarity. The cosine similarity helps overcome this fundamental flaw in the 'count-the-commonwords' or Euclidean distance approach. Unique sentences are clustered and re-sort and finally, the summary is generated.

- Word Embedding: The other approach we used is Word embedding's and the word mover distance. It is one of the most popular representation of document vocabulary. It is capable of capturing context of a word in a document, semantic and syntactic similarity, relation with other words, etc. Using the Glove library, we are taking vectors for a word for which there is list of vectors of length 100, consisting of words which are common in syntax and semantic
- Word Mover Distance: It uses word embedding to calculate the distance so that it can calculate even though there is no common word. The assumption is that similar words should have similar vectors. Retrieve vectors from any pre-trained word embedding models. It can be Glove, word2vec, fast text or custom vectors. After that it using normalized bag-of-words (nBOW) to represent the weight or importance. It assumes that higher frequency implies that it is more important.

The tokens are given to the word present in the sentence. These sentences are then ranked on the basis of the tokens given, after which the sentences are taken in account for the proper summary. We have also compared the Tf-Idf and Word Embedding's summarization on the basis of features.

Update Details in Database - When the summary is generated then its details is stored in the database and is available to the user for information analysis.

Setup Web Service - A web service to provide summary of given text documents, will be set up. The Web Service client will send request message consisting of document then the server sends the summary as the response message.

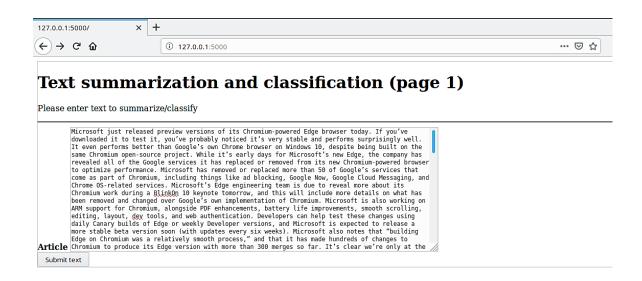
Finally, the summarized articles are deployed on a webpage where each individual user preferences will have noted over the time and a user will receive only those news article suggestions which he/she likes to read the most. This is an recommendation system also. This innovation in the project will keep the users interest in the news and current affairs which is what my ultimate aim is.

Execution steps:

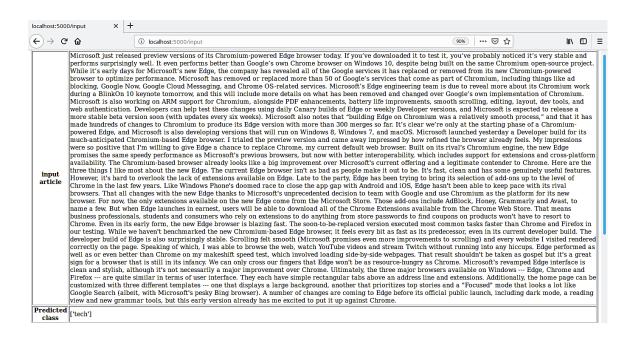
Step 1: Export the python file to FLASK APP and use flask run to execute.

```
dhruvsgarg@dhruvsgarg:~/NLPProject/BBC-Dataset-News/model$ export FLASK_APP=model2.py
dhruvsgarg@dhruvsgarg:~/NLPProject/BBC-Dataset-News/model$ flask run
 * Serving Flask app "model2.py"
 * Environment: production
    WARNING: Do not use the development server in a production environment.
    Use a production WSGI server instead.
 * Debug mode: off
```

Step 2: After the models are loaded, the Flask web app can be accessed from the browser. We input the text to summarize/classify.



Step 3: The HTML form when submitted invokes the Python script and the classification of the article is done. Here, the Random Forest Classifier model correctly classifies the News article as one of "Technology news"



Step 4: The summary of the input article is generated through 3 different methods as shown below.



Fig. The above images show the entire workflow from news article submission to predicted class and extractive summary generation.

3. Results and Discussion

An interactive Web App (Flask) is developed, the classification accuracy for various types of news articles using Random forest classifier was found to be 95 %.

This classification model generalized well and was able to correctly predict the category of all the news articles given to it. We used two approaches for feature extraction, Bag of word based TF-IDF vectorizer word embedding based Glove. Text summarization was successfully implemented using the two different features and comparison could be made. It was found that Word embedding feature produce a much better summary, when compared to bag of Word features. The Word embedding based summarization was able to use the context and semantic information to produce richer summarize text.

4. Conclusion

My innovation in the project to build was to provide targeted news article to individual user according to their reading habits and interests. This will always keep the user engaged to the app and would always want to come back and read more news articles that he/she likes according to their preferences. Various methods of extractive approach have emerged in the past. But it is hard to say how much greater interpretive sophistication, at sentence or text level contributes to performance. Without the use of Natural Language Processing, the generated summaries may not be much accurate in terms of semantics. If the input documents cover multiple topics, it becomes difficult to generate a balanced summary.

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