Efficient Notes Generation through Information Extraction

C. Nalini, Shwetambari Kharabe, Sangeetha S

Abstract: For a specific topic we find several reference books and textbooks to refer and the content is much more than one needs, therefore at the time of revision or quick study we need the summary and short version of the whole content for revision, especially at the time of examinations.

There we expect to build up a semi-mechanized method to produce notes from English content records like Reference Books and Text books. The method talked about is viewed as a spearheading endeavor in the field of NLP (Natural Language Processing). This strategy has a wide scope in the instructive space. The procedure when executed as an application can be utilized by both employees and understudies.

Keywords: NLP, Segmentation, Parsing, Ontology

I. INTRODUCTION

Here we talk about condensing a content report into progressively reasonable archive with less substance which covers a large portion of the vital focuses and gives a reasonable thought regarding the entire record, similar to a synopsis[19],[21],[23].

For the most part, in the vast majority of the course readings, just 20-22% of the words contain the data you have to comprehend or require while overhauling the theme. They are known as watchwords. Furthermore, the staying 80% contains near no fundamental data in that capacity, which comprises of pronouns, connectives like "of", "has", "for", and so forth. The main motivation behind these words is to connect the catchphrases together to frame significant and reasonable sentences[1]. They are valuable for first time perusing, yet for amendments, simply the watchwords can take the necessary steps effectively, so overhauling only a little measure of content is much useful as perusing the entire content once more.

We achieve this by using various methodologies and tools like Parsing, ontology creation, segmentation etc which are discussed further[20],[22],[24]. For a given text document like a reference book we find too much information on a topic which is more than one needs or can handle, to tackle this problem we can make notes and study a topic, here we try to accomplish the generation of notes automatically with abstractive approach[2]. This application can be used to get a gist of a given document and can be used for quick study like revision and omit the information from a document which is not much useful.

To accomplish larger amount of precision in [3] archive order increasingly educational highlights of reports are considered. For this reason, for example, weight is doled out to HTML labels, which influences the proficiency of data recovery; these are characterized utilizing hereditary calculations. Archives arrangement happens at the degree of separated words, dislike old style works, the importance of each word here is characterized in connection to their useful highlights, which are the events of a word in the title, accentuation is given on words by methods for italic, intense text styles or its underlining and position of a word on the page. A DIG (Document Index Graph) calculation depends on diagram hypothesis and expressions and their loads are considered for making proposals in work [3].

In this methodology for content segmentation,[5] we utilize a productive direct message division calculation (called TSHAC). It thinks about both computational intricacy and division precision. The procedure of TSHAC has 4 stages. To begin with, Preprocessing of long message; tokenization, stopwords are evacuated, and stemming are led to build the jargon of the content. Content is then spoken to as vector after content preprocessing, every one of which speaks to a sentence inside the content. A piece of sentence similitudes are then processed to build the sentence-likeness lattice. At long last, a the ideal point limits are distinguished by the proposed algorithm.[5]

The strategy for extraction of significant sentences from the content to synopsis dependent on meaning of score of pertinence for every one of the sentences and called "sentence by sentence" is recommended in this segment[14],[16],[18].

The procedure to be utilized is space autonomous which makes it special from different strategies. From the assessment estimates applied to the method, we can say that the strategy helps in semi-computerized age of notes. The presentation thoroughly relies upon the cosmology gave as information. The more exact the philosophy creation, the more precise the yield will be[25],[27],[29].

Revised Manuscript Received on July 22, 2019.

C.Nalini, Department of Computer science and Engineering, Bharath Institute of Higher Education and Research, Chennai, Tamilnadu, India. Email: nalinicha2002@gmail.com

Shwetambari Kharabe, Department of Computer science and Engineering, Bharath Institute of Higher Education and Research, Chennai, Tamilnadu, India. Email: shweta.chiwhane@gmail.com

Sangeetha S, Department of Computer science and Engineering, Bharath Institute of Higher Education and Research, Chennai, Tamilnadu, India. Email: sangeethasathya@gmail.com
II. METHODOLOGY

- **File Input**
  - Input the file that is to be read and analyzed.
- **Pre-processing**
  
  In pre-processing module there are three main activities that take place namely (i) Word Tokenization (ii) Stopword Removal (iii) Stemming. In Word Tokenization we separate the input documents into separate unique individual words and sentences which are formerly known as tokens. For this we have used the NLTK tokenize library.
  
  In the Stopword Removal phase we segregate the stopwords from the tokens using the NLTK.corpus library. And in the stemming phase we bring down the words to their root phase. Eg: using will be stemmed as use[8],[10],[12].

A. Weight Calculation

\[ w_i = \frac{\text{termFreq}_i}{\sqrt{\text{tokenCount}}} \times (\log \frac{\text{docCount}}{\text{docFreq}_i} + 1) \]

Using the above we calculate the weights of each individual words and store the value in the database.

B. Sentence Similarity

In this module we compare every sentence with its consecutive three sentences. We calculate the similarity function of the sentences using the Euclidian distance formula. We have used the SK learn library for this.

C. Segmentation

In the segmentation module we club the similar sentences together within an array. This is done by using the value of the similarity function of the sentences.

D. Sentence Formation

In sentence formation module we score the sentences based on the value of similarity function and the weights. The sentences above the threshold value are then selected. And then using ontology we make the sentence meaningful by connecting the sentences. This whole process makes the sentences extractive[13],[15],[17].

E. Notes

This is the final output of the system. it is the summary of the input document.

III. LITERATURE REVIEW

Here we discussed the literature review of existing techniques:

- **Okul Prasad, Harish Mathivanan, Madan, Jayaprakasam, T.V.Geetha** [1] they propose a semi automated technique that generates slide presentation from a text document which is given as an input. A Model is proposed which contains 13 modules after implementing which we get a slide presentation using the bullet points extracted from the input document.
- **Alguliev, R.M, Aliguliyev, R.M** [2] propose a text summarization method that creates text summary by assigning relevant score to each sentence and extracting sentences from the original documents. While summarization this method considers weight of each sentence in the document. The relevance score of a sentence is calculated through its comparison with all the other sentences and with the document title by cosine measure.
- **Ji-Wei Wu, Judy C.R. Tseng, Wen-Nung Tsai** [3] propose an efficient linear text segmentation algorithm based on hierarchical agglomerative clustering. The proposed linear text segmentation algorithm is implemented without any auxiliary knowledge base, parameter setting, and user involvement. This technique provides comparable segmentation accuracy with several well know linear text segmentation algorithms.

IV. COMPARITIVE ANALYSIS

<table>
<thead>
<tr>
<th>No</th>
<th>Paper Name</th>
<th>Author</th>
<th>Method Proposed</th>
<th>Semantics</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>Efficient Text Segmentation using Linear Text Segmentation Algorithm</td>
<td>Ji-Wei Wu, Judy C.R. Tseng, Wen-Nung Tsai</td>
<td>A proposed linear text segmentation algorithm without any auxiliary knowledge base, parameter setting, and user involvement. This technique provides comparable segmentation accuracy with several well known linear text segmentation algorithms.</td>
<td>The algorithm can be applied in a wide range of applications, including text categorization, document summarization, and information retrieval.</td>
</tr>
<tr>
<td>2</td>
<td>Efficient Text Segmentation using Linear Text Segmentation Algorithm</td>
<td>Ji-Wei Wu, Judy C.R. Tseng, Wen-Nung Tsai</td>
<td>A proposed linear text segmentation algorithm without any auxiliary knowledge base, parameter setting, and user involvement. This technique provides comparable segmentation accuracy with several well known linear text segmentation algorithms.</td>
<td>The algorithm can be applied in a wide range of applications, including text categorization, document summarization, and information retrieval.</td>
</tr>
<tr>
<td>3</td>
<td>Efficient Text Segmentation using Linear Text Segmentation Algorithm</td>
<td>Ji-Wei Wu, Judy C.R. Tseng, Wen-Nung Tsai</td>
<td>A proposed linear text segmentation algorithm without any auxiliary knowledge base, parameter setting, and user involvement. This technique provides comparable segmentation accuracy with several well known linear text segmentation algorithms.</td>
<td>The algorithm can be applied in a wide range of applications, including text categorization, document summarization, and information retrieval.</td>
</tr>
<tr>
<td>4</td>
<td>Efficient Text Segmentation using Linear Text Segmentation Algorithm</td>
<td>Ji-Wei Wu, Judy C.R. Tseng, Wen-Nung Tsai</td>
<td>A proposed linear text segmentation algorithm without any auxiliary knowledge base, parameter setting, and user involvement. This technique provides comparable segmentation accuracy with several well known linear text segmentation algorithms.</td>
<td>The algorithm can be applied in a wide range of applications, including text categorization, document summarization, and information retrieval.</td>
</tr>
<tr>
<td>5</td>
<td>Efficient Text Segmentation using Linear Text Segmentation Algorithm</td>
<td>Ji-Wei Wu, Judy C.R. Tseng, Wen-Nung Tsai</td>
<td>A proposed linear text segmentation algorithm without any auxiliary knowledge base, parameter setting, and user involvement. This technique provides comparable segmentation accuracy with several well known linear text segmentation algorithms.</td>
<td>The algorithm can be applied in a wide range of applications, including text categorization, document summarization, and information retrieval.</td>
</tr>
</tbody>
</table>

V. CONCLUSION

The proposed procedure is area autonomous, so we can concentrate notes from any report with no assistant information base which makes it remarkable from different strategies[7],[9],[11]. From all the assessment estimates applied to the method, we
presume that the procedure produces notes utilizing a semi-robotized notes age framework. The presentation absolutely subject to the cosmology gave as info. The more precise the metaphysics, the more exact the yield or the produced notes will be.

REFERENCES


AUTHORS PROFILE

C.Nalini Professor, Department of Computer science and Engineering, Bharath Institute of Higher Education and Research, Chennai, India

Shwetambari Kharabe Research Scholar, Department of Computer science and Engineering, Bharath Institute of Higher Education and Research, Chennai, India

Sangeetha S Assistant Professor, Department of Computer science and Engineering, Bharath Institute of Higher Education and Research, Chennai, India