

# AN AUTOMATIC TEXT SUMMARIZATION USING LEXICAL COHESION AND CORRELATION OF SENTENCES

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## Abstract

Due to substantial increase in the amount of information on the Internet, it has become extremely difficult to search for relevant documents needed by the users. To solve this problem, Text summarization is used which produces the summary of documents such that the summary contains important content of the document. This paper proposes a better approach for text summarization using lexical chaining and correlation of sentences. Lexical chains are created using Wordnet. The score of each Lexical chain is calculated based on keyword strength, Tf-idf & other features. The concept of using lexical chains helps to analyze the document semantically and the concept of correlation of sentences helps to consider the relation of sentence with preceding or succeeding sentence. This improves the quality of summary generated.

In this paper we discuss a summarization method, which combines lexical chaining with correlation of sentences in which relation of a sentence with the preceding sentence is considered. Our experiments show that the inclusion of both these features improves the quality of summary generated.

**Keywords**— Text summarization, Wordnet, Correlation of sentences, Lexical chains

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## 1. INTRODUCTION

### 1.1 Motivation

These days, the number of Web pages on the Internet almost doubles every year as the information is now available from a variety of sources. It takes considerable amount of time to find the relevant information. Automatic Text Summarization will help the users to find the relevant information rapidly. It generates the summary of the document and one can read the summary and decide the relevance of the document to the information needed by the user.

### 1.2 Background Research:

Text summarization is the process of producing a condensed version of original document. This condensed version should have important content of the original document. Research is being done since many years to generate coherent and indicative summaries using different techniques. According to (Jones, 1993) the text summarization is described as two step process

- i) Building a source representation from the original document.
- ii) Generating summary from the source representation

Text summarization can be broadly classified into two types: Single document summarization and multi-document summarization. This paper focuses on single document summarization that generates summary of single document. The text summarization can be categorized into extractive and abstractive based on the nature of text representation in the summary.

Many methods have been proposed till now on generating a coherent summary. The earlier methods used only statistical methods that focused on term frequency [1] for choosing important sentences. These methods were not found to be efficient as it did not consider all the contexts of the word or identify semantically related terms known as cohesion.

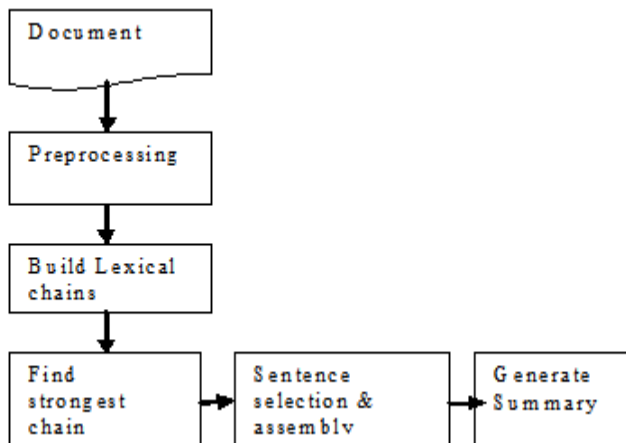
Then came methods which used semantic representation of the original document supported by a domain-specific knowledge base. Now a days text summarization is considered as a natural language processing task. Lexical chains a simplest form of lexical cohesion was introduced by Morns & Hirst[2]. But it was found that all possible senses of the word were not taken into account.

Berzilay & Elhadad [2] presented a better algorithm that constructs all possible interpretations of the source text using lexical chains. It is an efficient method for text summarization as lexical chains identify and capture important concepts of the document without going into deep semantic analyses. Lexical chains are constructed using some knowledge base that contains nouns and its various associations.

Our Algorithm is based on the method used above. We have used Wordnet to generate domain-specific extractive summary using Lexical chains for the nouns in the document. The algorithm segments the given content into sentences & then into tokens. These tokens are tagged using POS tagger. The Nouns are selected & for each noun in the segment, we consider its sense using Wordnet. Then we attempt to merge these senses into all of the existing chains in all possible ways, hence building every possible

interpretation of the segment. Next merge chains between segments that contain a word in the same sense in common. The algorithm then calculates score of lexical chains, determines the strongest chain and uses this to generate a summary. We have also used the concept of correlation of sentences to generate a good quality summary.. The terms that occur in the strongest lexical chains are considered as key terms and the score of sentences is calculated based on the presence of key terms in it. All the sentences are ranked based on their score and top n sentences are selected for inclusion in the summary. Then the correlation of sentences is checked and if any sentence has correlation with the previous sentence, then the previous sentence should also be included in the summary based on condition as shown in the algorithm below

## 2. ARCHITECTURE OF TEXT SUMMARIZATION



Preprocessing includes

- Segmentation
- Tokenization
- POS(part of speech tagging) at lexical level.
- Stemming.

## 3. LEXICAL CHAIN COMPUTING ALGORITHM

1. Input Original document for generating summary (.txt file).
2. Divide the document into sentences using segmentation.
3. Each sentence is divided into tokens using tokenizer.
4. These tokens are tagged using POS Tagger.
5. For each noun build the synsets.
6. For each sentence generate a map using 4 relations: Synonym, Hypernym, Hyponym, Merynym.
7. Calculate distance of each word from other related words.
8. Build Lexical chains using generated map.
9. Calculate each chain weight using values of distances of each word

10. Select longest chain i.e. best chain having highest chain weight
11. From the original document select sentences that have words in the best chain retaining their order of occurrence in the original document.
12. Pick top n sentences as summary based on the percentage of original document to be used for generating summary.
13. If the selected sentence starts with words : although, however, moreover ,also, this, those and that ,then they are related with the preceding sentence.
14. If the rank of the preceding sentence is equal to or greater than 70% of the rank of the selected sentence, then it is included in the summary. In this way correlation between sentences is maintained.

## 4. EVALUATION

Evaluation is the most important part of any research work. It helps to compare various techniques based on evaluation metrics.

This paper uses precision & recall [4,5,6]technique for evaluation which is based on statistical measures. Precision evaluates the proportion of correctness for the sentences in the summary whereas recall is utilized to evaluate the proportion of relevant sentences included in the summary.

### 4.1 Precision

$$\text{Precision} = \frac{\{\text{Retrieved sentences}\} - \{\text{Relevant sentences}\}}{\{\text{Retrieved Sentences}\}}$$

The higher the precision value, the better is the efficiency of the system in reducing irrelevant Sentences

### 4.2 Recall

$$\text{Recall} = \frac{\{\text{Retrieved sentences}\} - \{\text{Relevant sentences}\}}{\{\text{relevant sentences}\}}$$

Higher the recall value, better the efficiency of the approach in selecting only relevant sentences.

### 4.3 F-Measure

The weighted harmonic mean of precision and recall is called as F-measure

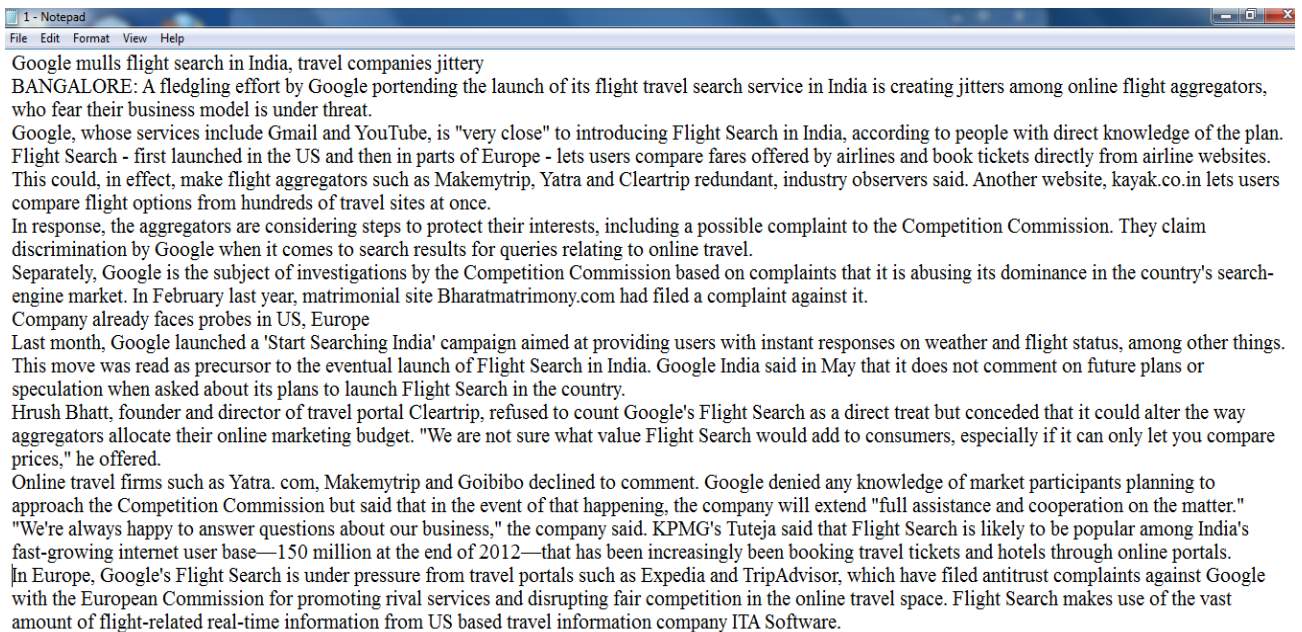
$$\text{F-measure} = \frac{2 \times \text{Precision} * \text{recall}}{\text{Precision} + \text{recall}}$$

## 5. EXPERIMENTAL RESULTS

Three documents are taken in news domain. The original document, manually generated summaries and summaries generated by the above approach are shown below. The

precision recall and F-measure are calculated for these three documents and they are compared with other two summarizers.

## Original Document 1



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Google mulls flight search in India, travel companies jittery  
BANGALORE: A fledgling effort by Google portending the launch of its flight travel search service in India is creating jitters among online flight aggregators, who fear their business model is under threat.

Google, whose services include Gmail and YouTube, is "very close" to introducing Flight Search in India, according to people with direct knowledge of the plan. Flight Search - first launched in the US and then in parts of Europe - lets users compare fares offered by airlines and book tickets directly from airline websites. This could, in effect, make flight aggregators such as Makemytrip, Yatra and Cleartrip redundant, industry observers said. Another website, kayak.co.in lets users compare flight options from hundreds of travel sites at once.

In response, the aggregators are considering steps to protect their interests, including a possible complaint to the Competition Commission. They claim discrimination by Google when it comes to search results for queries relating to online travel.

Separately, Google is the subject of investigations by the Competition Commission based on complaints that it is abusing its dominance in the country's search-engine market. In February last year, matrimonial site Bharatmatrimony.com had filed a complaint against it.

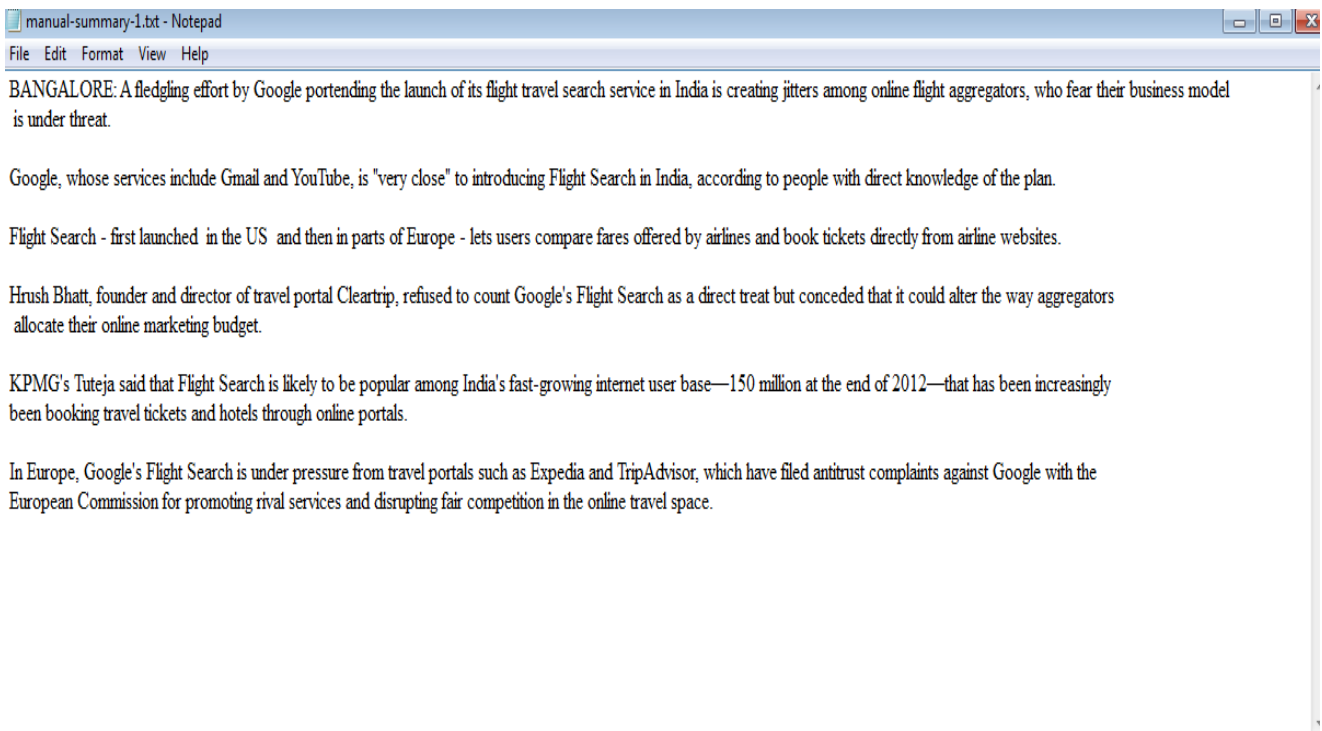
Company already faces probes in US, Europe  
Last month, Google launched a 'Start Searching India' campaign aimed at providing users with instant responses on weather and flight status, among other things. This move was read as precursor to the eventual launch of Flight Search in India. Google India said in May that it does not comment on future plans or speculation when asked about its plans to launch Flight Search in the country.

Hrush Bhatt, founder and director of travel portal Cleartrip, refused to count Google's Flight Search as a direct treat but conceded that it could alter the way aggregators allocate their online marketing budget. "We are not sure what value Flight Search would add to consumers, especially if it can only let you compare prices," he offered.

Online travel firms such as Yatra.com, Makemytrip and Goibibo declined to comment. Google denied any knowledge of market participants planning to approach the Competition Commission but said that in the event of that happening, the company will extend "full assistance and cooperation on the matter." "We're always happy to answer questions about our business," the company said. KPMG's Tuteja said that Flight Search is likely to be popular among India's fast-growing internet user base—150 million at the end of 2012—that has been increasingly been booking travel tickets and hotels through online portals.

In Europe, Google's Flight Search is under pressure from travel portals such as Expedia and TripAdvisor, which have filed antitrust complaints against Google with the European Commission for promoting rival services and disrupting fair competition in the online travel space. Flight Search makes use of the vast amount of flight-related real-time information from US based travel information company ITA Software.

## Ideal Summary of Document 1



manual-summary-1.txt - Notepad  
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BANGALORE: A fledgling effort by Google portending the launch of its flight travel search service in India is creating jitters among online flight aggregators, who fear their business model is under threat.

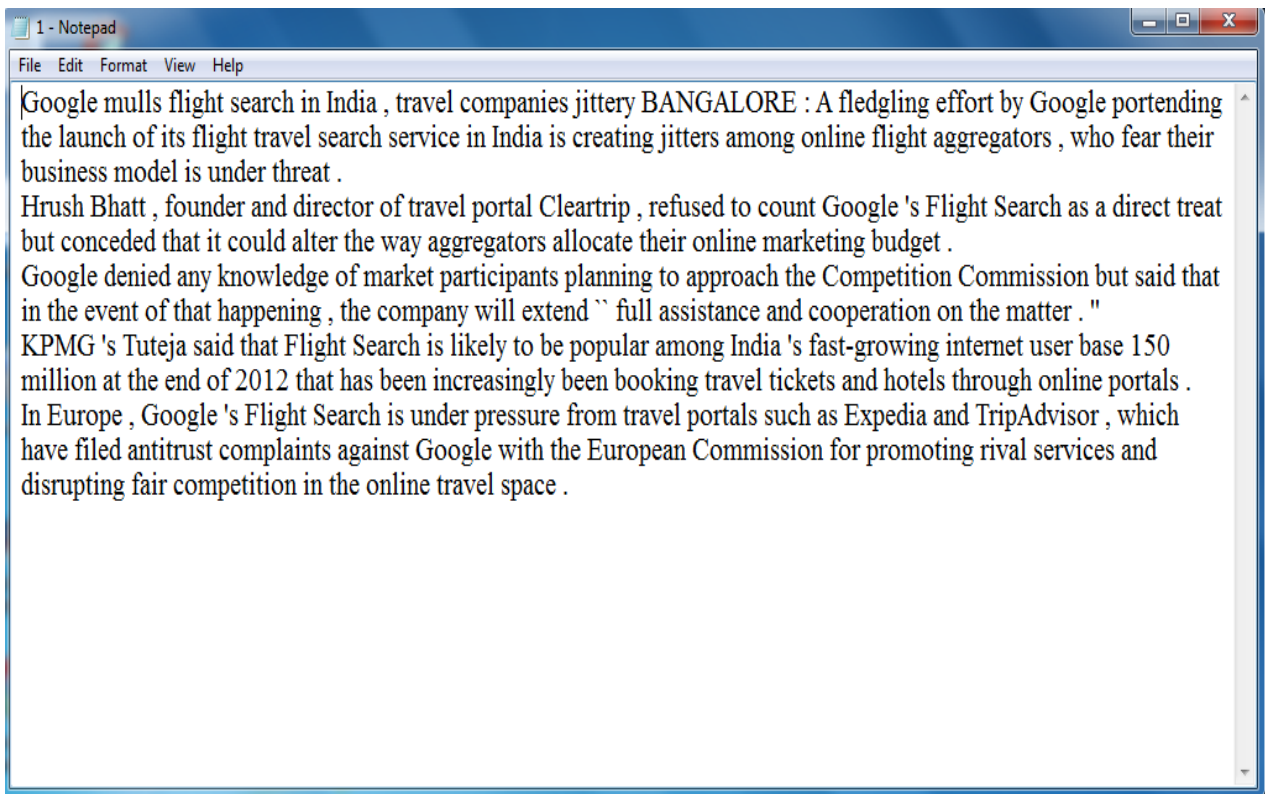
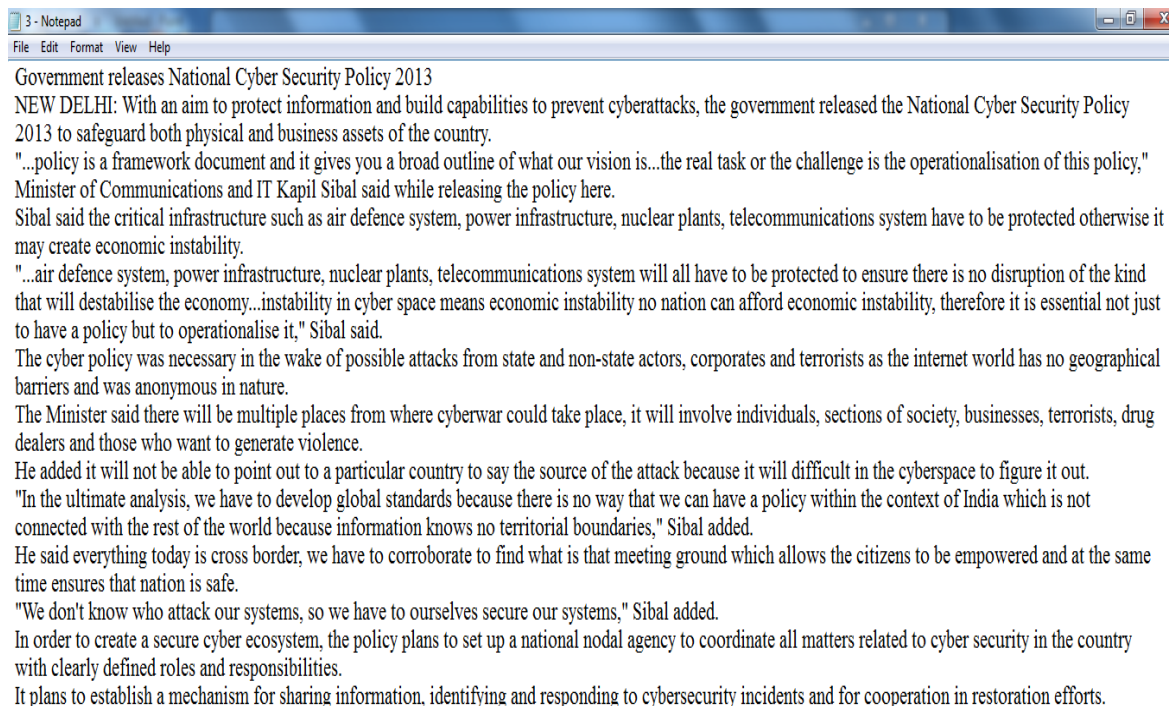
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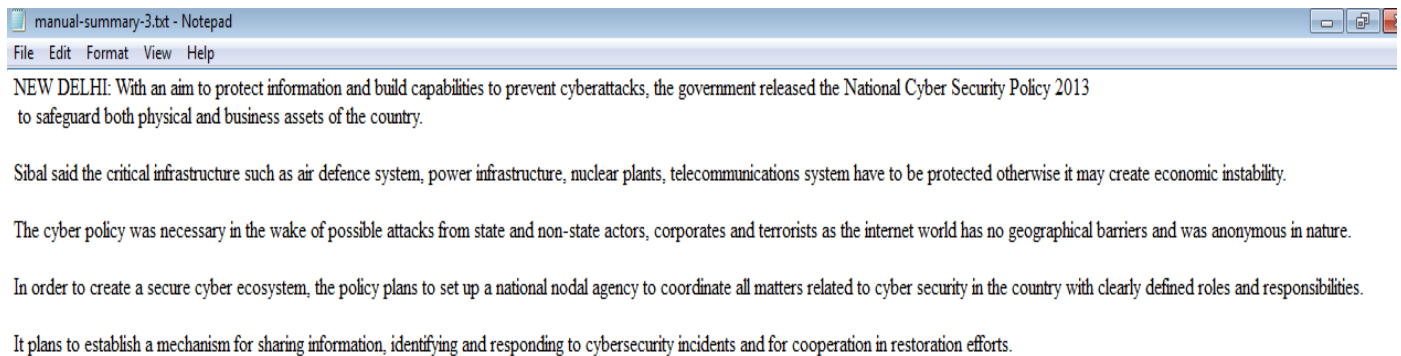
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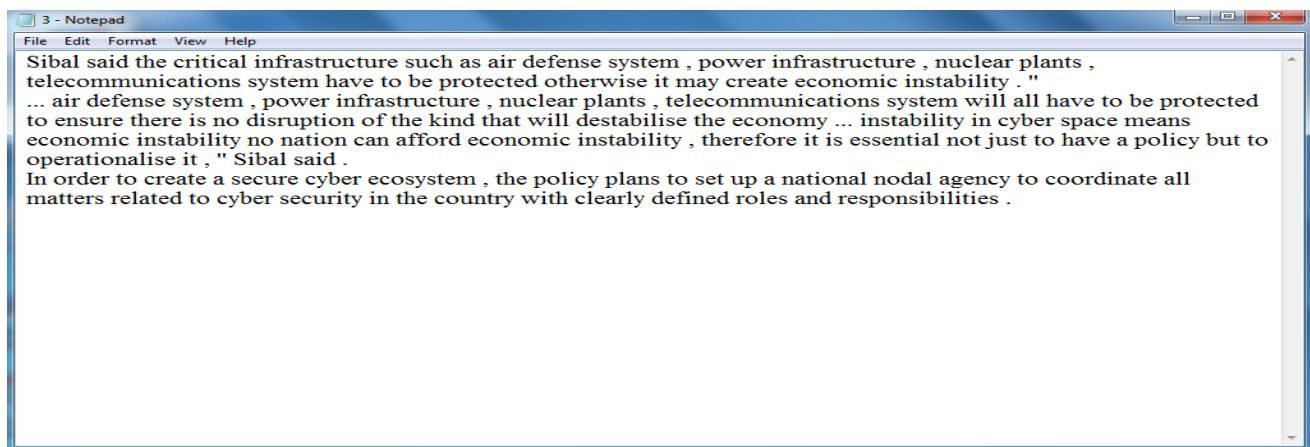
**Summary of Document 1 generated by our Summarizer****Original Document 2**



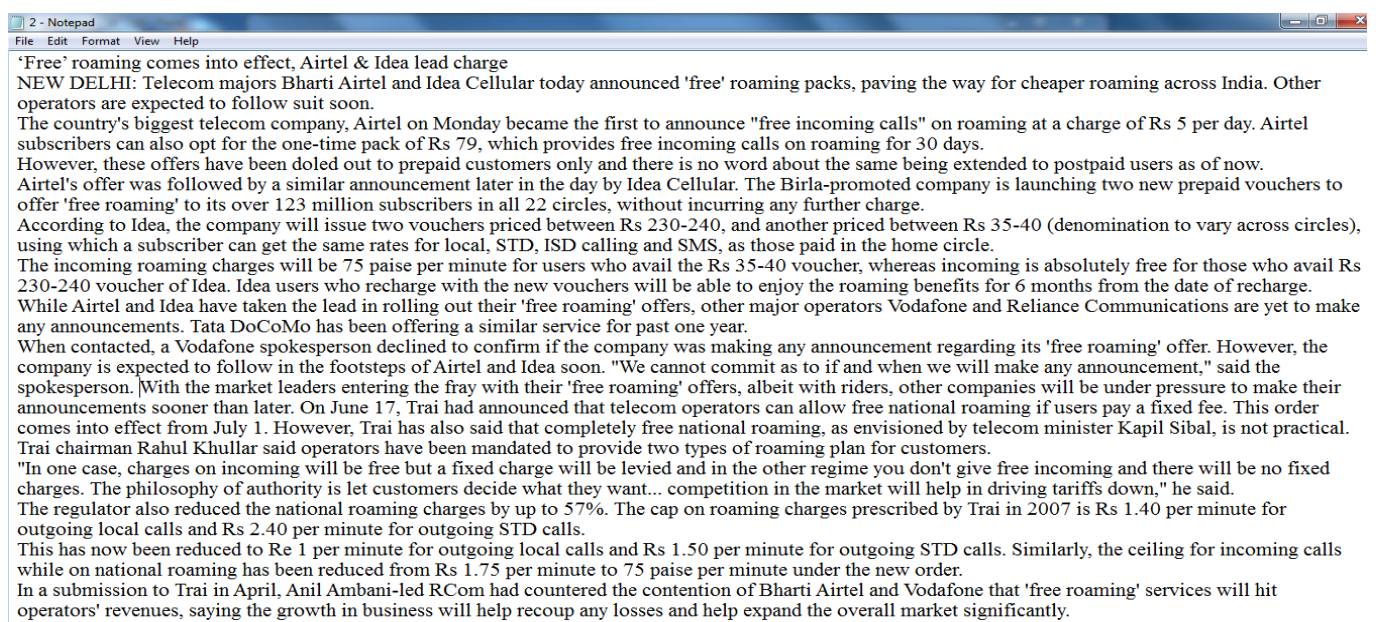
## Ideal Summary of Document 2



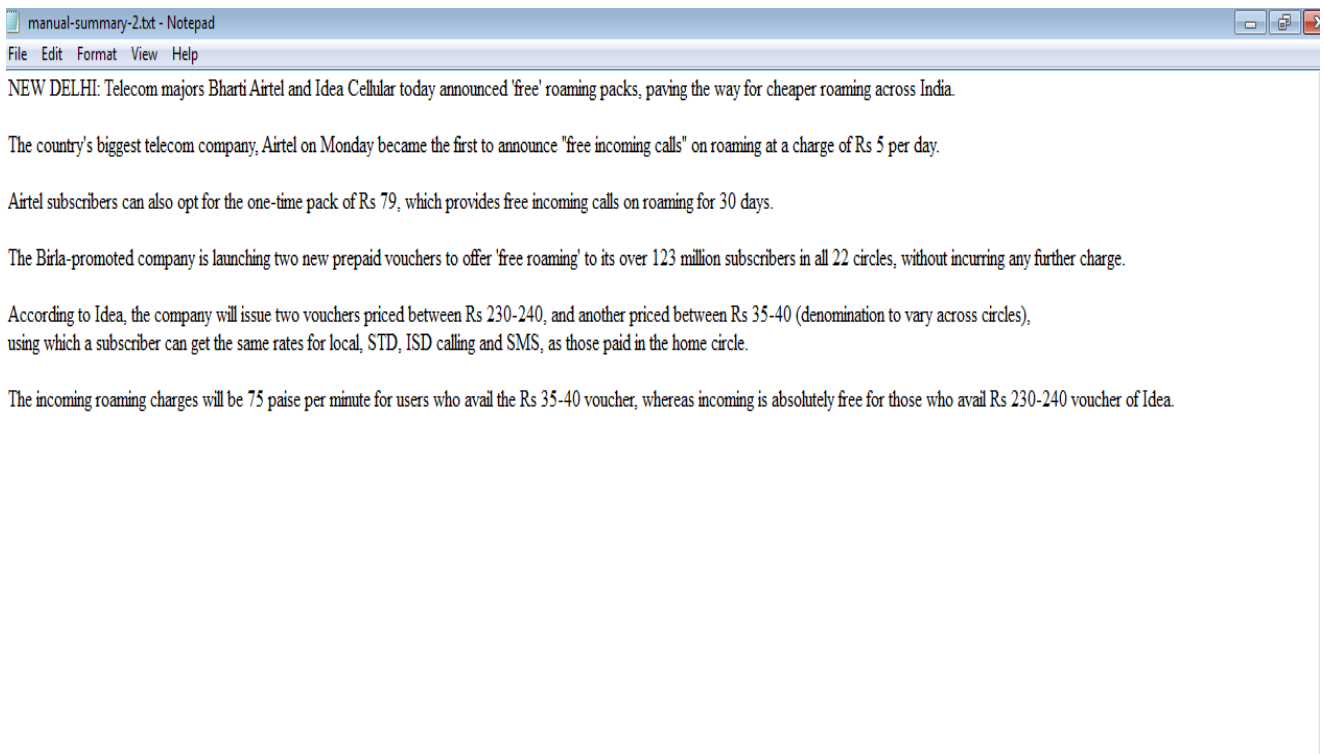
## Summary of Document 2 generated by our summarizer



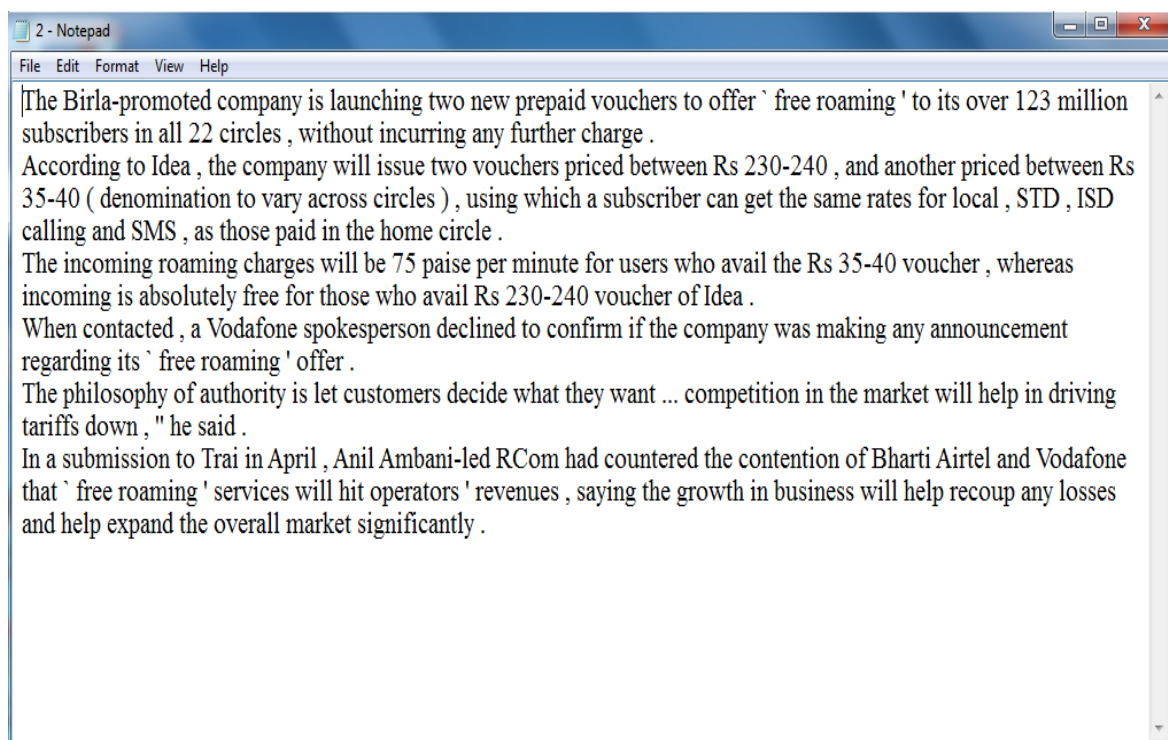
## Original document 3



### Ideal summary of document 3

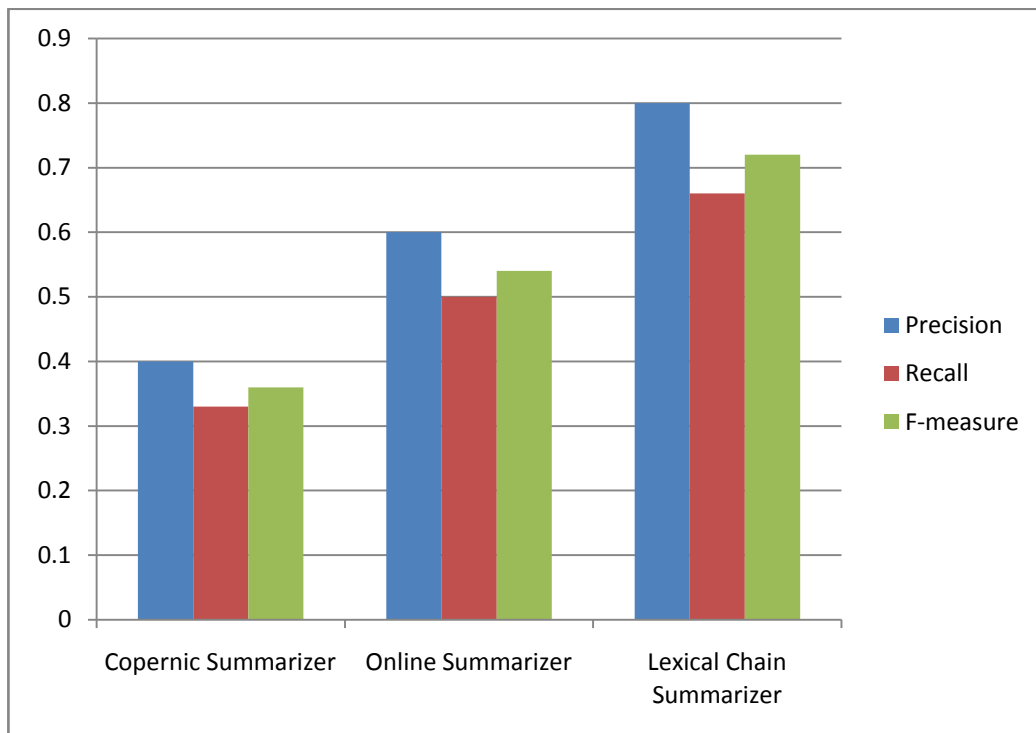
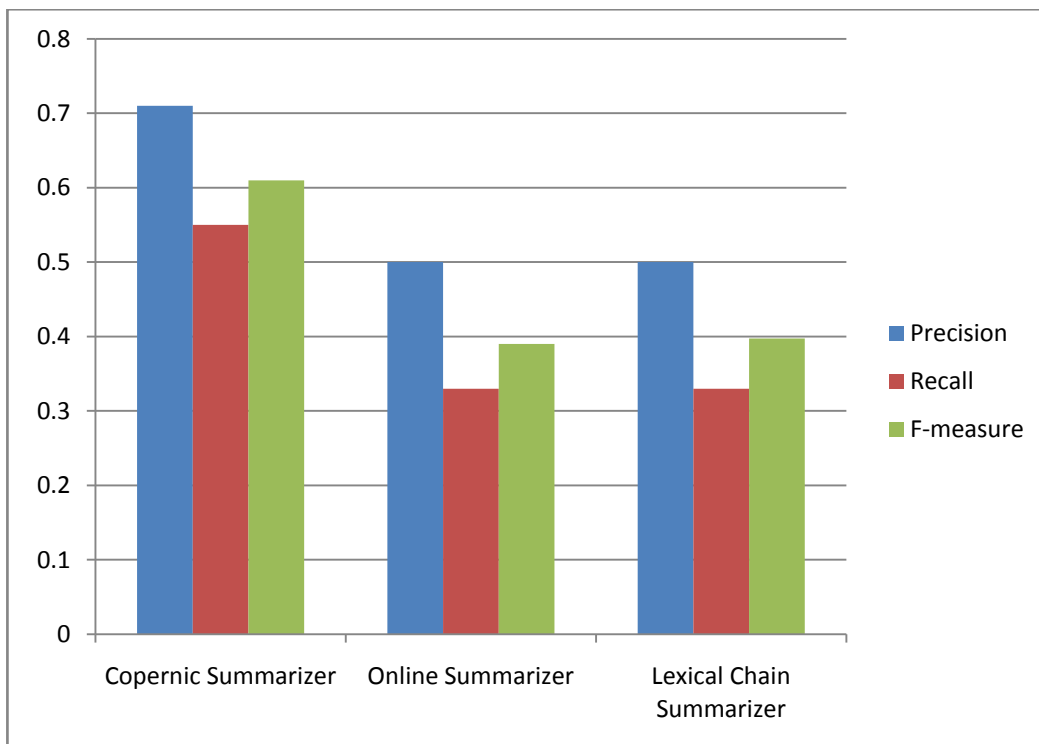


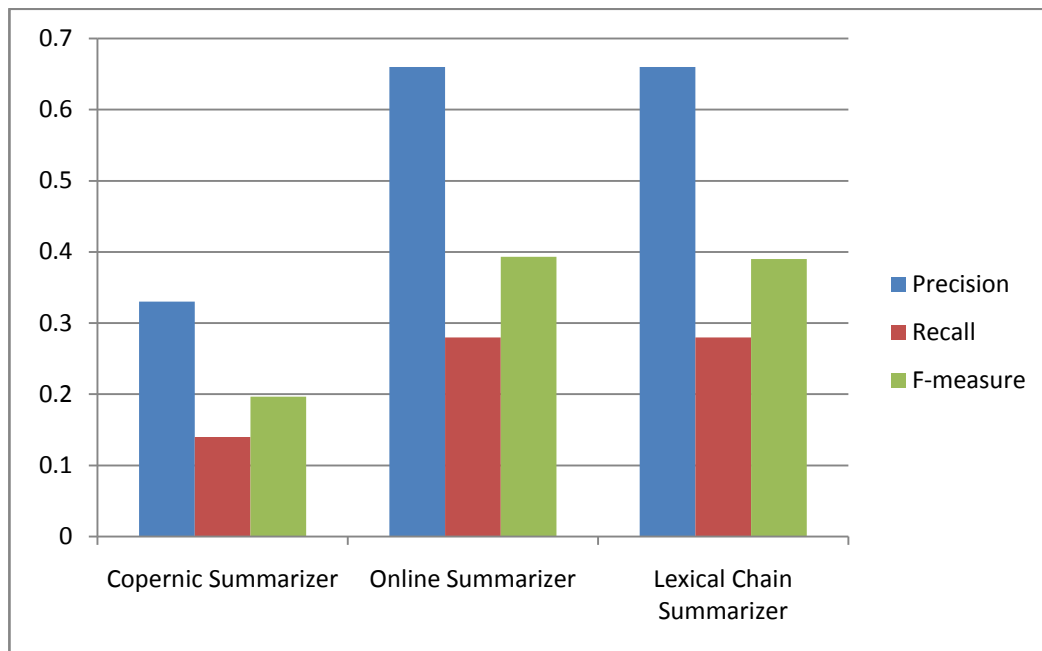
### Generated Summary of Document 3



## 6. COMPARISION

This paper considers online summarizer from freesummarizer.com[7], Copernicus summarizer and our summarizer using lexical chains of sentences for comparison. The above three documents are used as input to all the three summarizers. The precision, recall and F-measure are used as performance measures for summary generated.

**Document1****Document2:**

**Document3:****7. CONCLUSIONS**

[7] [www.freesummarizer.com/summarize/](http://www.freesummarizer.com/summarize/)

It is seen that for document 1 and document 3 our summarizer performs better than Copernicus summarizer. and online summarizer. For document 2, It performs equally as online summarizer but less efficient than Copernicus summarizer. Our summarizer is better as it also considers the semantic analysis of the document & correlation of sentences for generating the summary.

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