AUTOMATIC TEXT SUMMARIZATION BASED ON PRAGMATIC ANALYSIS

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Abstract- The rapid growth of online information has encumbered the user with colossal amount of information. It is difficult to access large amount of data. This problem has increased the research in the field of automatic text summarization. Automatic text summarization is a technique where the text is input to the computer and it returns the clipped and concise extract of the original text and also sustains the overall meaning and main information content. In this paper, text summarization technique is designed for the documents having the fixed format. The proposed system generates the summary of the fixed format documents by analyzing all the different parts of the documents. The system consists of five stages. In first stage each sentence is partitioned into the list of tokens and stop words are removed. In second stage, frequency usage is counted for each word. In third stage, assign POS tag for each weighted term and Word sense disambiguation is done. In the fourth stage, pragmatic analysis is performed. After Pragmatic Analysis, summarized sentences will be store in a database.

Index Terms- POS tagging, Pragmatic Analysis Text Summarization, Wordnet

I. INTRODUCTION

Internet has developed the lives of many enthusiastic discoverers and researchers. Now-a-days, with the increasing demand of the internet huge volumes of information are arising continuously [1].Users finds it thorny to find the anticipated information swiftly and precisely [10]. It is useful for the users to extract the main content from the original document instead of the original document. Automatic Text Summarization automatically extracts the main content from the original document correctly and comprehensively. The language of the summary extracted should be consistent and efficient. Automatic Text summarization can be categorized into two- extraction and abstraction [11].Extraction means to select the phrases or sentences having the highest score from the original text and combined to obtain the new shorter text without changing the source text. Abstraction means to probe and interpret the text by using linguistic methods. Mostly, extraction method is used to produce the summary in automated text summarization system. During the last twenty years, several researchers addressed that the automated part of speech tagging is a well known problem. Part of speech tagging is a technique for elucidation of the lexical categories. In POS tagging, a felicitous tag is assign to each word

of the sentence.POS tagging is broadly used for lexical text analysis.POS tagging is a important task for the activities in natural language processing [7]. In POS tagging, it takes a sentence as input and assigns a unique tag to each word in the sentence. It is a firm conjecture that when it comes to index term extraction, the nouns carry most of the sentence meaning. Index term is a term that catches the essence of the sentence. In an ideal case, index terms should give excellent semantic representation. A subtask of POS Tagging is noun extraction in which every noun either proper or common is identifying in a document. Nouns are used as a most important feature to express the meaning of the text in natural language processing applications like text summarization, information retrieval, information extraction etc.

POS Tagging has a variety of techniques .It has two approaches-Supervised POS tagging and Unsupervised POS Tagging [4]. A pre tagged corpora is needed for Supervised Tagging Technique while it is not required in Unsupervised Tagging Technique.

Both techniques can be of two types: - Stochastic and Rule Based. Rule Based Technique requires a context rule for POS Tagging [5]. In rule based approaches, tags are assigned to the ambiguous and unknown words by using the contextual information. Stochastic Technique uses a Hidden Markov Model[9].The states mostly symbolize the POS Tags. The contingencies are calculated from the tagged corpus and the untagged corpus in order to calculate the most likely POS tags for the word of the sentence. Stochastic Training techniques can be categorized into Supervised and Unsupervised Stochastic Technique. Supervised stochastic technique requires only the pre tagged data in a huge volume to achieve high level of accuracy. On the other hand, Unsupervised stochastic technique does not need pre tagged data and it uses the computational methods to do the automatic word groupings or make tag sets and based on these groupings, estimate the probabilistic values required by stochastic taggers.

II. PROPOSED METHODOLOGY

The following figure represents the diagram of the propose system. Proposed model has the following stages:-

A. Database:

Firstly make a repository that will act as a source of the system. This repository contains the files on which text summarization

International Journal of Scientific and Research Publications, Volume 2, Issue 5, May 2012 ISSN 2250-3153

can be done. From this repository, one paper at a time is taken on which apply our first step i.e. tokenization.

B. Tokenization:

Each sentence is partitioned into the list of tokens. It is a part of the lexical analysis.

C. Stop Word Removal:

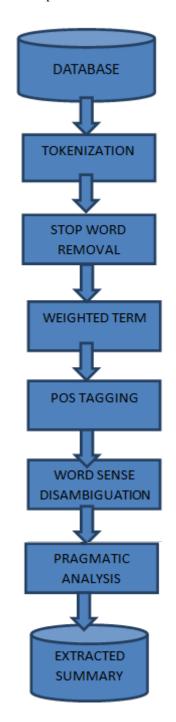
Some words are insignificant that are highly used in the English sentences. They exist in superiority in the documents. Therefore, they do not provide the actual idea about the theme of the text. While scoring the sentences, these words can be discarded. For example- articles like 'a', 'an', 'the'; 'by' come into sight mostly in all the text but does not include much semantic information. As we already generated the list of tokens, now remove the stop words from that list and store these words into a separate file.

List of some stop words considered are:-

a	ourselves
about	out
above	over
after	own
again	same
against	shan't
all	she
am	she'd
an	she'll
and	she's
any	should
are	shouldn't
aren't	SO
as	some
at	such
be	than
because	that
been	that's
before	the
being	their
below	theirs
between	them
both	themselves
but	then
by	there
can't	there's
cannot	these
could	they
couldn't	they'd
did	they'll
didn't	they're
do	they've
does	this
doesn't	those
doing	through
don't	to
down	too
during	under
each	until
	-

few	up
for	very
from	was
further	wasn't
had	we
hadn't	we'd
has	we'll
hasn't	we're
have	we've
haven't	were

List 1: Stop Words



D. Weighted Term:

Frequency usage of each word is calculated after removal of stop words. The words having the higher frequency are the weighted terms. Frequency can be defined as the number of times term is repeated in the document. For example- a term occurring 10 times in a document is much more relevant than the term occurring 2 times.

E. POS Tagging:

It is a process of assigning a felicitous tag to each word of the sentence[8]. A large lexical database for English language named "Wordnet" can be used for POS tagging[6]. Wordnet consists of most English nouns ,verbs, adverbs and adjectives which are grouped into sets of cognitivesynsets, each showing a distinct concept. Wordnet is systematized by meaning that means the words are semantic similar which are in close propinquity[2].

F. Word Sense Disambiguation:

It is a process of choosing pertinent senses of the word in a given context[3]. It removes the ambiguity of the word.

It is important for NLP applications like information retrieval, machine translation, part of speech tagging and Text processing.

Word sense disambiguation consists of 2 steps:-

- 1. Identify all the different senses for each word congruent to the text.
- 2. It involves assigning relevant sense for each word in context.

Word Sense Disambiguation can be reached by 2 approaches-Shallow approaches and Deep approaches. Access to the comprehensive body of world knowledge is assumed in Deep Approach. But these approaches are not favorable in practice because access to body of knowledge is possible in very limited domains. If knowledge exists, then this approach is better than the shallow approaches.

Shallow approaches consider the surrounding words and not even try to understand the text. These are not as stronger as Deep approaches but gives better results in practice because of limited knowledge domain[12].

G. Pragmatic Analysis:

Pragmatics can be defined as the study of the meaning in context. It is an effort to get the intended meaning of the text[13].Pragmatic Analysis is only performed on scientific or fixed format articles[14].After POS tagging, pragmatic analysis is performed which analyzes the weighted terms.

This analysis gives a fixed format to the files and the sentences containing the weighted terms will be stored in a database. These are the summarized sentences. Any format file is converted into a fixed format by pragmatic analysis. Thus, the main content of the document can be identified to compose the summary by analyzing the pragmatic function.

H. Extracted Summary

After pragmatic analysis, the summarized sentences get stored in a database. This summary is the result of text summarization. Hence,the file taken at the first stage is summarized into meaningful text through this process.

III. RESULTS

The system break up the given text into tokens as shown in table I,then stop words are removed from the list of tokens,shown in tableII.Word Frequency of each word is calculated after stop word removal,shown in fig 3.Then,POS tag is assigned to the weighted terms by using Wordnet as shown in fig 4. Word Sense Disambiguation is performed while doing the Pragmatic Analysis. During Pragmatic Analysis,maincontent of the document is identified to compose the summary.

Table I: Tokenization

I. hi	
II. How	1
III. Are	
IV. You	
V. Dear	-
VI. You	
VII.	Are
VII. VIII.	looking
IX. good	1
X. you	
XI. are	
XII.	comfortable
XIII.	this
XIV.	
XV.	of
XVI.	environment
XVII.	are
XVIII.	you
XIX.	going
XX.	somewhere
XXI.	what
XXII.	you
XXIII.	want
XXIV.	dear
XXV.	The
XXVI.	person
XXVII.	you
XXVIII	
XXIX.	looking

Table II: After Stop word Removal

XXX.	Dear
XXXI.	Good
XXXII.	Comfortable
XXXIII.	Kind

International Journal of Scientific and Research Publications, Volume 2, Issue 5, May 2012 ISSN 2250-3153

XXXIV	Environment		
XXXV.	XXXV. Somewhere		
	XXXVI. Dear		
XXXVI	XXXVII. person		
XXXVI	II. Don't		
XXXIX	Here		
XL.	Want		
XLI.	Sir		
XLII.	Sun		
XLIII.	Rises		
XLIV.	east.		
XLV.	Kind		
XLVI.	attitude		
XLVII.	Fine		
XLVIII.	Dear		

C:\Program Files\J Please enter a tex File has 18 words.	t file name: Lice	bin≻java GeneratedTokens ense.txt
WORDS	FREQUENCY	
attitude comfortable dear dont east environment fine good here kind person rises sirthe somewhere sun	1 3 1 1 1 1 1 2 1 1 1 1 1	

Figure 3: Showing words with their frequency.



Figure 4: Part of speech tagging

Here are all the possible senses of 'customer' with pos: 'n'
Sense: 99984659 Description: someone who pays for goods or services Synset: customer client
Here are all the possible senses of 'information' with pos: 'n'
Sense: 96634376 Description: a message received and understood Synset: information info
Sense: 95816287 Description: knowledge acquired through study or experience or instruction Synset: information
Sense: 97237561 Description: formal accusation of a crime Synset: information

ī.

Figure 5: Word Sense Disambiguation

Pragmatic Analysis is performed to extract the summary in a fixed format. When we perform the word sense disambiguation, all the synsets are generated. The sentences containing the synonyms of the words are removed .i.e repetitive sentences are removed and the main content is extracted and get stored in a database.

IV. CONCLUSION

In this work, we proposed an automatic text summarization approach by pragmatic analysis. We used supervised POS tagging approach and implemented it in Java. By pragmatic analysis, the intended meaning of the text is obtained and the text summary is generated and stored in a database. The systems produced the high quality compressed summary and provide better results than the manual summarization.

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