



Mining User Opinions on Different Product Aspects from Online Product Reviews

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Abstract: Product Manufacturer expects the customer to review their product once they have purchased. Due to the popularity gain of the e-commerce the reviews of the product is also increasing day by day in large number. There is no limitation in writing the reviews by the customer it all depends on the customer satisfaction and their likes. This makes it difficult for a product manufacturer to read the reviews, analyze the opinions and to keep track of product opinions. Thus, it aims to mine the product opinions with respect to its attribute and then rate the aspect. Most of the works before were using document level or sentence level of opinion mining. This work makes use of aspect based opinion mining (phase level or word level). It mainly focuses on mining user opinions on different product aspects from online product reviews and to identify whether the opinions are positive or negative. This methodology will enhance effectiveness of the time efficiency of aspect orientation.

Keywords: Aspect based opinion mining, Sentiments, Sentiment orientation.

I. INTRODUCTION

Data mining explores enormous volume of statement in order to resolve reality problems and it has strongly shaped the number of tools, methods and algorithm. The main objective is to manage a rich broadcast to utilize the patterns, rules and to get insight knowledge of the data mining practice. The users have the opportunity to phrase their opinions due to the disturbance created in e-commerce and broadcasting. Due to the opinion rich reviews that are available in social networking, online review sites, blogs and media helps in gaining the knowledge and use the technologies and can analyse the opinions of people [1]. The data mining is accessible for the social media which is used for different purposes and for the user given product reviews. Data mining can be referred as powerful techniques for analysing huge amount of accumulated data. The requirement is that small information and the repeatedly used data in data mining has the fancy to extract the information from the raw data and to figure it to be data.

A. Opinion Mining (Sentiment Analysis)

Opinion Mining makes use of web in order to extract the feedback of the users. It understands the reviews of the user with respect to attitude, emotions, opinions concerning the organization depending on attributes, entities, persons. Most of the services will be feed on blog posts,

home shopping sites or review sites where the user will reveal their opinion depending on the products. The consumers and also the producer can be able to recognize like what public visualize about the yield or services and also it should be useful for the both. The opinions will be extracted by making use of sentimental analysis and opinion mining from the user given product reviews. There are many different formats in which the reporting of the opinions of users are made using e-commerce sites. The monitoring, associating the opinions on the top of each other which is related in e-commerce sites then analyse and extract the opinions and sentiments is actual new challenge [2]. In order to manage the business a clear understanding of product opinion, brand new and high standing management is done by the opinion mining tools. These tools have a great scope to assist the user so that can expect the opinions or sentiments and make use of information technologies. The supervised learning algorithm is used to effectively build sentiment classifiers that demand a labelled announcement for a given domain. Thus the sentiments are expressed differently and it is valuable to translate it to new domain.

B. Three Levels of Opinion Mining

Opinion mining which is also called as sentimental analysis is an approach where the people opinions, attitude, sentiments and their emotions of the product are expressed. It mainly comes under the natural language processing. It is



virtually distinct item of tracking behaviour and gangs up to closed once. The hall of knowledge will actually analyse which case has feedback expressing, what are reviews that are commented and particular one who commented the reviews. There are three categories of opinion mining when taken into account:

1. **Document-level:** Document-level is a detailed list problem to get the predefined one after passing the input copy. It mainly expresses the opinions such as web reviews or questions in surveys. This document sometimes does not consider single opinions. Most of the time the document consists of many sentiments and it prevent the customer in drilling the data in order to get useful information. It is subjective to classify the opinions based on positive or negative or neutral of sentiment orientation [3]. The prediction of an opinion will classify whether it's being prosperous or not helpful. Likewise, opinion spam is useful to classify whether it is a spam or not spam.
2. **Sentence-level:** It is fragile level when compared with that of document level [4]. It is placed one after the other in sentence-level. It will be ranked in some criteria as specified based on the opinions identified. It accurately summarizes the opinions that aim to pick sentences which summarize the answers more closely. Finally, the answers' mining identifies the comparative sentences and then extracts the reference.
3. **Phrase-level:** Phase-level opinion mining is known as orientation based opinion mining [5]. It is a finer grained hit or miss and rapidly looks at opinion. The function of this directly of hit or miss is to capture sentiments on aspect items. Aspects which are nouns or phrases are called as vigorous aspects. e.g.: 'battery'. "The battery of laptop is fine". In order to make proper decisions aspect based opinion mining is used. It concentrates on the explicit aspects. The document level and sentence level cannot fetch proper detailed results.

II. RELATED WORK

In recent years Opinion analysis is become more common as many researches are doing work on this topic. Due to the analyses of customer reviews taken from the sites such as flip kart, Amazon, eBay, big bazaar etc it can be stated as beneficial analysis. Previous work was mainly on

the document level and sentence level of opinion mining which was actually misleading the product manufacturer. Our work is different as we are interested in opinions expressed on each product aspect rather than the whole review.

Patra B.G; Mukherjee N; Das A; Mandal S; Das D; Bandyopadhyay S [6] It is a novel approach having the dataset of restaurant and laptops and collects the proper features and also its opinions. This method is based on the help of different features and Conditional Random Field using machine learning algorithm. It follows Supervised Classification approach.

The methodology of this work starts with the POS tagging using Stanford Core NLP tool to parse each of the sentences. Then to identify the aspect the 'be' verb rule is used followed by inanimate words, dependency noun phrases, Post determinant word, dependency based sentiment used. The drawback of this work is that identification of aspect terms is failed in case of unstructured sentences or long incorrect sentences. The Stand ford Parser does not work well so it identifies incorrect boundaries. It also fails in identifying the polarity as many aspect terms are present.

Yan Li; Hui Wang; Zhen Qin; Weiran Xu and Jun Guo, [7] this method is Two -step estimation method. The 2 goals are to extract the features, cluster them as aspects then analyse its reputations. It mainly follows 2 steps:

1. Confidence Value will be challenged based on the features and clusters.
2. Challenge to conquer the Global reputation of the review for each aspect.

This method first calculates the prior Sentiment scores and then extracts the candidates features based on parser Stanford core NLP. Using pattern bootstrapping the collocation of opinions and product aspects can be extracted. So the two measures to fetch the features and patterns they are consider as Popularity and Reliability. Later the features clustered into aspects using the Word Net to group into similar category. The compactness identifies the similarities of cluster items and then texture as a quality then the lighter ones are removed out. Finally the methodology considers the dynamic sentiment ambiguous adjectives (DSAA) for opinion words and then captured global aspect reputations. The drawback of this work is it does not link accurately right opinions of aspects when a single review text may have



positive opinion on an aspect and negative opinion on another aspect of the product. Thus results show the incorrect linking of aspects.

Jintao Du; Wen Chan; Xiangdong Zhou, [8] this method is aspect identification method using the translation based technique. It mainly captures the relations that are modified then by using translation model the identification of aspect is done. Then the methodology follows with predicting each review with the helpfulness so the performance will be improved. The monolingual alignment model will be improved when the mining of aspect and opinion words are fetched rapidly. Then estimation of aspect candidate confidence with respect to bipartite graph improves the performance. The results of the experiments will states that the approach is more effective when compared with some baselines and the methods of state-of-the-art. The drawback of this work has less than perfect nature of corpus so the translation quality will be affected even the performance will become low. It also fails on incorrect linking of opinions.

Marrese-Taylor, Edison, Juan D. Velasquez and Felipe Bravo-Marquez, [9] this method is Aspect based opinion mining method. The dataset used here is tourism product reviews based on its opinions identified. The methodology follows with aspect identification techniques based on NLP, POS tagging and noun extraction. Then the sentiment prediction is done by using the rule based approach and the orientation is done using sentiment word dictionary. The aspect orientation rules are followed to get the results. The dataset taken here is Hotel and Restaurants corpus in order to mine the aspects. The drawback of this work is that the opinion word extraction is time consuming in large text as the opinion is searched word by word. Poor performance is resulted in getting the frequent nouns and noun phrases.

Pang, Bo and Lillian Lee, [10] this method mainly follows Opinion orientation technique. The focus is mainly on the new techniques of the sentiment application which gives a good rise when compared to that of traditional analyses. This work is mainly on the issues such as privacy, impacts on economic and manipulation using the opinion oriented access a proper rise can be obtained by collecting the text. This method also provides a benchmark datasets if any further analysis have to be made. The drawback of this work is that it needs to improve time efficiency of opinion orientation.

V. S. Jagtap and Karishma Pawar [11] this method is sentiment classification method. This method will understand the information in review text and then focused to mine the opinions of that text. Sentiment analysis is process where the people's emotions, attitudes are extracted as opinions with respect to organization or the attribute. The impact of decision making has made the addiction for the customers in making the choice for online shopping or any events. This method mainly works with sentence, phrase and also with document level of sentimental analysis. The goal is to analyse the polarity of sentences in 3 categories such as positive, negative and neutral in sentiment classification. The drawback of this work is it lack in identification of features indicating whether sentences are on-topic (which is a kind of co-reference problem). There is need to improve the time efficiency.

Challenges to find in online product reviews

- The opinion word extraction is time consuming in large text as the opinion is searched word by word.
- In single review text by user, may have positive opinion on an aspect and negative opinion on another aspect of the product. Thus it fails in linking right opinions to aspects as given by the user.

III. METHODOLOGY

The solution to detect the mining opinions on different product aspects contains mainly 4 modules:

1. Automatic sentence boundary detection and extraction.
2. Sentence tokenization and POS Tagging.
3. Aspect and Opinion extraction.
4. Opinion Orientation.

A. System Architecture Description

The proposed system uses customer reviews to extract aspect and mine whether given is positive or negative opinion. A user review collected is a review of a product or service made by a customer who has purchased the product. The architectural overview for working model of the proposed system is shown in figure 1. Each review is split into individual sentences. A review sentence is given as input to data pre-processing. Next, it extracts aspect in each review sentence. Sentence boundary, tokenization and pos tagging are data pre-processing. The Opinion Extraction is done by extracting the opinion words from the POS tagged sentences which are forms of Adjectives. These opinion words are checked against the word net whether the opinions are positive or negative and with adjectives even adverbs will be checked. Then the Opinion Orientation for each



aspect extracted, will compute the number of positive and negative words occurred.

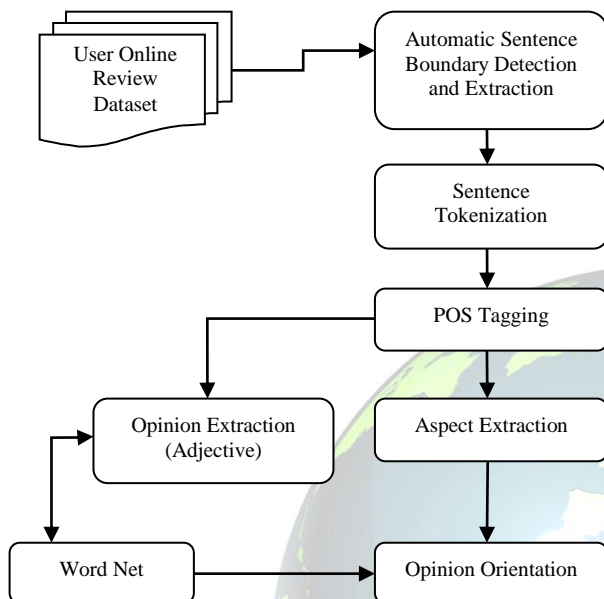


Fig 1: Proposed System Architecture

1. Sentence Extraction

The algorithm accepts the input of product user review from the dataset file. The dataset selected might be from Tacobel or the McDonalds restaurant. The reviews of the restaurant are taken from particular once which is in document level format. The document level will be difficult when compared to sentence level. So then it loads boundary detector to detect the sentence from the paragraph review. In order to do so it uses Apache Open NLP Model for detecting sentences for all possible boundaries. Using this tool which is inbuilt, the sentence extraction can be easily fetched. Thus, it finally splits the sentence reviews using proper boundary marks.

Input: Raw Review Text
Output: Sentences from the review text
// Read reviews from the dataset file ReviewText = File.Read("path of the file in the local system")
// Load the trained English sentence boundary detector model Model= ApacheOpenNLP.LoadSentenceBoundaryModel

("path of the English model file")

```

// Detect sentence boundaries
Boundaries[]=
Model.MarkSentenceBoundaries(ReviewText)

// Split review text into sentences using the boundary marks
Sentences[ ] = ReviewText.Split (Boundaries[ ])
  
```

2. Sentence Tokenization and POS Tagging

The algorithm accepts the input from sentence list. The sentence from the user review will be POS tagged with size of number of sentence extracted. It uses Apache Open NLP POS tag Model for initialize the tag. Then for each sentence list depending POS will be tagged. The sentence is having single space as delimiter to separate word. It combines tokens and POS tags in order to make sentence. Thus, it finally adds the POS tag sentence to the list.

Input: Sentence List
Output: POS Tagged Sentences
// Initialize the POS Tag Sentence Array with the size of no.of sentences extracted POSTaggedSentences[]= StringArray[Sentences[].size]
// Initialize the POS Tagger Model POSTaggerModel= ApacheOpenNLP.LoadPOSTaggerModel("path of the English POS tagger model")
// For each sentence in the Sentence List find the POS Tag FOR each sentence in Sentences[]
START
sentenceTokens[] = Sentence[i].Split(" ") // Uses single space as delimiter to separate words
posTags[]= POSTaggerModel.Tag(sentenceTokens[])
// Combine tokens and posTags for all tokens in order to make POS Tagged Sentences
String posTagSentence
FOR each tokens in sentenceTokens[]
START
posTagSentence.add(sentenceTokens[k]+" "+posTags[k])



```
END
// Add the POS Tagged Sentence to the list
posTaggedSentences[i] = posTagSentence
END
```

3. Aspect and Opinion Extraction

The algorithm accepts the input from POS tagged sentences. The POS sentence extracts only nouns and adjectives for the respective aspects and opinions. It will check the POS tagged words in the sentences then for each tokens, it will extract if its nouns or adjectives. If aspects are nouns then NN tags are used to represents Nouns. If opinions are adjectives then JJ tag are used to represents adjectives. Then, finally it links the aspects with the opinion list extracted in a sentence.

```
Input: POS Tagged Sentences
Output: Aspects and their Opinion Words List

// parse through each POS Tagged Sentences and extract
Nouns for Aspects and
//Adjectives for Opinion Words
// Initialize Map Data Structure to Store Aspects and its
Opinions
Map<String, <List<String> > AspectsAndOpinions

// Extract Aspects and Opinions
FOR each POSTaggedSentences[ ] // Loop #1
START
Sentence = POSTaggedSentences[i]
Tokens[ ] = sentence.split(" ") // Extract POSTagged
Words in the sentences

FOR each token in Tokens[ ] // Loop #2
START
IF(Tokens[i] contains "NN") // Aspects are Nouns.
The NN tag represents Nouns
THEN
Aspect = Tokens[i])
END IF

IF(Tokens[i] contains "JJ") // Opinions are
Adjectives. The JJ tag represents Adjectives
THEN
OpinionList.add (Tokens[i])
END IF
// Link Aspect with the opinion list extracted in a
sentence
AspectsAddOpinions.put(Aspect, OpinionList)
```

```
END // Loop #2
END // Loop #1
```

4. Opinion Orientation

The algorithm accepts the input from Aspect opinion words. The aspect opinion contains nouns and adjective needs to be properly oriented with polarity. It will extract the opinions and it will be checked against the word net. The Word Net actually contains list of negative and positive words of natural language processing. Depending on result of each aspect opinions in module 3, further process of opinion orientation whether it is positive or negative is obtained. This process will be orientated for each opinion with respective aspect. Once the orientation is completed then the total score of sentiments will be calculated and separate overall view of positive rating and negative rating will be obtained. Then, finally it orientation of overall product will be analysed.

```
Input: AspectList and OpinionList
Output: Opinion Orientation and Summary

// Initialize Opinion Counts
OverallPositive = 0
OverallNegative = 0

FOR each Aspect in Aspect List
FOR each OpinionWord in OpinionList
OpinionOrientation=
SentimentAnalyser(OpinionWord)
IF OpinionOrientation = POSITIVE THEN
Aspect.Positive = Aspect.Positive + 1
OverallPositive = OverallPositive + 1
ELSE
Aspect.Negative = Aspect.Negative + 1
OverallNegative = OverallNegative + 1
END IF
END FOR
END FOR

// Print the Aspect Level Opinion
FOR each Aspect in Aspect List
Print Aspect, Aspect.Positive, Aspect.Negative
END FOR
// Print Overall Product Opinion
Print OverallPositive, OverallNegative
```



V. RESULTS

The main output of the proposed system is to show the mining opinions on the online product reviews. The comparison results from the survey done are having some disadvantages so this work makes sure to keep count of all the opinions when compared to previous work and to improve its time efficiency. This section accepts Tacobel restaurant dataset as input and proceeds with further module functionality on this review. It shows the comparison of old work and new present work.

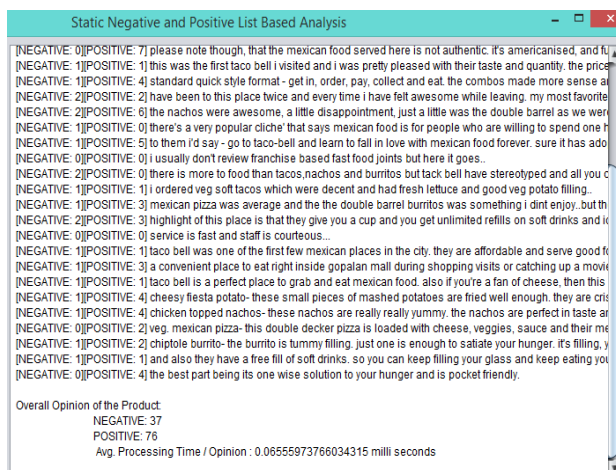


Fig 2: Opinion Orientation in Static based Analysis

The figure 2 explains the analysis using the Static Opinion database which is before work. The process is dealing with static database where both the positive and negative words are collected.

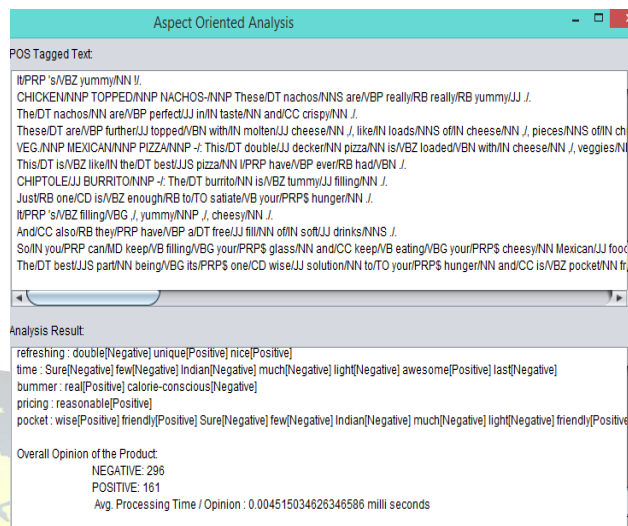


Fig 3: Opinion Orientation in Aspect Oriented Analysis

The figure 3 explains the present existing analysis of the Aspect Oriented Analysis. This process is dealing with data pre-processing steps and then Word Net which is actually used to check the adjectives against word net. Using this we will get a clear view whether the opinions fetched is positive or negative.

Table I

Comparison of Positive, Negative opinions, time efficiency gained with Static Analysis and Aspect Oriented Analysis using Tacobel Dataset

Sl. no	Analysis	Positive Opinion	Negative Opinion	Time
1.	Static based Analysis	76	37	0.06555ms
2.	Aspect Oriented Analysis	161	296	0.00451ms

The table I numerical values are fetched from the values taken from the figure 2 and figure 3. The result executed there is from the dataset review Tacobel. Obtaining the values from these figures this table is obtained.

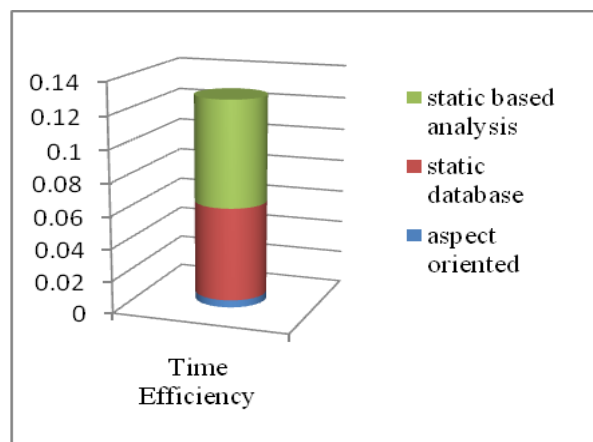


Fig 4: Analyses of time efficiency with given sample ratings in table 1

The figure 4 it shows the variation of ratings with respect to time efficiency taken from table 1. The proposed work has less amount of process time when compared to other two. As seen in above figure first one represents the Aspect oriented analysis and other two are using the static based database. Thus, a clear declaration is obtained as seen in figure that proposed work aspect orientation obtains very less time and has good time efficiency when compared with others.

VI. CONCLUSION

In this technique, we determine the mining opinions on different product aspects from online reviews. The extracted opinions are checked with respect to its polarity and then determine the rating of positive and negative of product based on the weight of opinions on different aspects. The time efficiency of aspect orientation in this work is improved with respect to online product reviews. It shows efficient results of time efficiency when compared to other works of aspect mining on online product reviews. It also shows high searching rate with respect to positive and negative words.

The future work in mining opinions can be achieved by improving the problems such as misleading of opinions and pronoun resolution. Along with other data and details of product in this work, the algorithm can be improved for further more efficient collection of positive and negative polarity. The proposed work can be used for real time if it is automatically connected to electronic commerce with help of application platform interface. Thus, monitoring of any new comments of product can be detected.

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