

# Sentiment Analysis of Consumer Reviews using Multi-Strategy Technique

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**Abstract** – Sentiment analysis is useful in industrial intelligence application atmosphere and recommender systems, as a results of it is a awfully convenient channel for the two ends of the availability to talk. at intervals the sentiment analysis, many ways in which and techniques were used, like machine learning, polarity lexicons, language method, and psychology scales, that verify different types of sentiment analysis, like assumptions created, methodology reveals, and validation dataset. Since internet has become an excellent provide of customer reviews, the realm of sentiment analysis (also noted as sentiment extraction, opinion mining, opinion extraction, Ad sentiment mining) has seen an outsized increase in academic interest over the previous couple of years. As known, the opinions of shoppers unit of measurement expressed in sentiment phrases. ancient machine learning techniques cannot represent the opinion of articles okay. This project proposes a multi-strategy sentiment analysis methodology with linguistics similarity to resolve the matter with partial phrase matching. Naïve scientist category identification is in addition applied to hunt out the chance of information distribution in varied class of information set. The project is meant pattern R Studio one.0. The committal to writing language used is R 3.4.4.

**Keywords**– partial phrase matching, Machine learning, opinion mining, opinion extraction

## I. INTRODUCTION

### 1.Sentiment Analysis

Sentiment analysis is discourse mining of text that identifies and extracts subjective data in supply material, and serving to a business to grasp the social sentiment of their complete, product or service whereas observation on-line conversations. However, analysis of social media streams is typically restricted to merely basic sentiment analysis and count based mostly metrics. this is often appreciate simply scratching the surface and missing out on those high price insights that square measure waiting to be discovered.

Thus what ought to a complete do to capture that low hanging fruit? With the recent advances in deep learning, the power of algorithms to analyse text has improved significantly. artistic use of advanced computer science techniques may be a good tool for doing in-depth analysis. we have a tendency to believe it's necessary to classify incoming client spoken language a couple of complete supported following lines:

- Key aspects of a brand's product and repair that customers care concerning.
- Users underlying intentions and reactions regarding those aspect.

These basic ideas once employed in combination become a really necessary tool for analyzing a lot of complete

conversations with human level accuracy. Sentiment Analysis is that the most typical text classification tool that associate and analyses an incoming message and tells whether or not the underlying sentiment is positive, negative our neutral. Intent associate analysis steps up the sport by analyzing the user's intention behind a message and characteristic whether or not it relates an opinion, news, marketing, complaint, suggestion, appreciation or question. Grammar Check Re-write Again Next.

### 1. About The Project

The Internet is presently not solely a crucial supply of knowledge, however conjointly a platform of expressing views and sharing experiences. During this network, we are able to simply collect reviews concerning merchandise or services. Sentiment analysis is beneficial in industrial intelligence application atmosphere and recommender systems as a result of it is a awfully convenient channel for the 2 ends of the provision to speak.

within the sentiment analysis, several ways and techniques were used, like machine learning polarity lexicons , tongue process, and psychology scales, that verify differing types of sentiment analysis, like assumptions created, technique reveals, and validation datasets. At present, sentiment analysis is created at 3 levels: word, sentence, and document, of that the sentence and therefore the document square measure typically employed in most current studies. The word

level, the basic, and consequently the a lot of important and more difficult level, however, is rarely studied. For Chinese as a language, really short sentiment phrases of 1 or 2 Chinese characters square measure most fuzzy in that means. ancient machine learning techniques can't represent this characteristic. therefore a replacement hybrid sentiment analysis is planned during this study, that comprehensively uses Zadeh's fuzzy pure mathematics, machine learning theory, and the technique supported polarity lexicons. It considers disjunctive conjunctions, like "(but)", "(while)", "(however)" etc.

Western students started the sentiment analysis earlier. They 1st determined the sentiment tendency of words or phrases and quantified them as a live of real values, which may be more accustomed verify the sentiment tendency of sentences and paragraphs. They analyzed the sentiment tendency. The 3 normal machine learning algorithms for sentiment analysis square measure NB (Naive Bayes), ME (MaxEnt, or most Entropy), and SVMs (Support Vector Machines). For simplicity of the experiment, we have a tendency to solely opt for NB and SVMs.

### III. SCOPE AND OBJECTIVES

1. To take exact phrase match for conditional probability. That is two phrases in two records need not match exactly during probability finding.
2. To check Skip grams (elimination of a word in phrase) during classification.
3. To check partial phrase during classification.
4. To mine opinion at word, sentence, and document levels, and gives sentiment polarities and strengths of articles.
5. To solve the problem with semantic fuzziness we initiate a multi-strategy sentiment analysis.

### IV. EXISTING SYSTEM

In existing system, data set is taken as records from Excel worksheet with category in second column. Preprocessing work such as stop word removal, stemming and Unicode removal are being done. Punctuation marks are removed. All characters are converted in lower case. Then two words, three words combinations are found out. If the count is above the given threshold among all the records, then the words are treated as valid phrases. These phrases conditional probability is found out among all categories which become Naïve Bayes Classification work. In addition, synonym words and phrases are kept in separate file. Before taken for probability finding, the phrases are replaced with synonym words so that two different phrases become same in semantic similarity/fuzziness concept.

#### 1. Drawbacks Of Existing System

Exact phrase match is taken for conditional probability. That is two phrases in two records must match exactly during probability finding.

Skip grams (elimination of a word in phrase) is not checked during classification. Partial phrase is not checked during classification.

### V. PROPOSED SYSTEM

In proposed system, like existing system, data set is taken as records from Excel worksheet with category in second column. Preprocessing work is carried out. Then words combinations are found out and valid phrases are gathered. These phrases conditional probability is found out among all categories which become Naïve Bayes Classification work. In addition, synonym words replacement is also made. Moreover, partial phrases like two words in one sentence and three words in other sentence are also treated as same phrases during naïve bayes classification.

#### 1. Advantages Of Proposed System

- Exact phrase match is not taken for conditional probability. That is two phrases in two records need not match exactly during probability finding.
- Skip grams (elimination of a word in phrase) is checked during classification.
- Partial phrase is checked during classification.

### VI. MODULE DESCRIPTION

#### 1. Download Twitter Data

in this module, twitter data is downloaded using 'twitter' package in which two or more search words are given such as laptop, tablet and mobile. all the three contents are saved in 'laptoptwitter.csv', 'tablettwitter.csv' and 'mobiletwitter.csv' file. the first column contains twitter tweet posts and second column with laptop, tablet or mobile.

#### 2. Preprocess Twitter Data

in this module, twitter data is preprocessed using 'tm' package in which stemming, stopword removal and url link removal is carried out.

#### 3. Sentiment Words File Creation

in this module, a csv file created in which sentiment phrase, category and sentiment value is being added as records. the category is one of laptop, tablet and mobile. the sentiment value is from -5 to +5 based on importance.

#### 4. Two Adjacent Words Phrase Combination

In this module, Twitter Data is converted into two words phrases such as first word and second word as one phrase, second word and third word as next phrase and so on for all tweets. These phrases are checked with sentiment value records taken from 'sentimentvalues.csv' created in previous module. If the phrase is matched with sentiment phrase then sentiment value of the corresponding category is taken and added. For all tweets, mobile category's positive and negative score is found out and displayed. Likewise tablet and mobile

categories are also prepared. Then conditional probability of these phrases in all the three categories are found out and displayed.

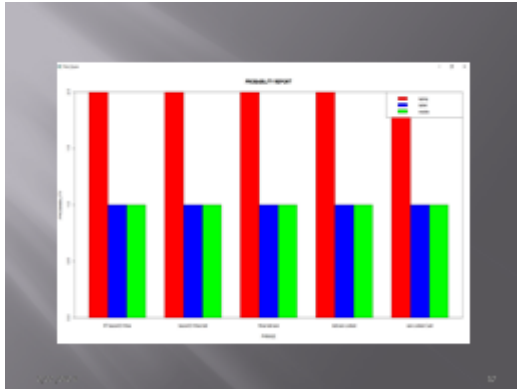


Fig.1 polarity values of top five words

### 5. Three Adjacent Words Phrase Combination

In this module, Twitter Data is converted into three words phrases such as first word, second word and third word as one phrase, second word, third word and fourth word as next phrase and so on for all tweets. These phrases are checked with sentiment value records taken from 'sentimentvalues.csv' created in previous module. If the phrase is matched with sentiment phrase then sentiment value of the corresponding category is taken and added. For all tweets, mobile category's positive and negative score is found out and displayed. Likewise tablet and mobile categories are also prepared. Then conditional probability of these phrases in all the three categories are found out and displayed.

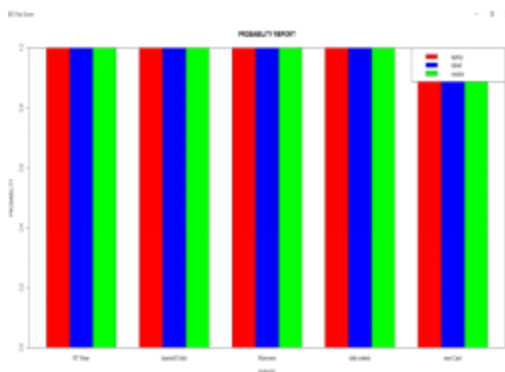


Fig .2 positive and negative values of the data

### 6. Missed Word Phrase Combination

In this module, phrases are formed with middle word deletion from the previous module phrases. These phrases are checked with sentiment value records taken from 'sentimentvalues.csv' created in previous module. If the phrase is matched with sentiment phrase then sentiment value of the corresponding category is taken and added. For all tweets, mobile category's positive and negative score is found out and displayed. Likewise tablet and mobile categories are also prepared. Then

conditional probability of these phrases in all the three categories are found out and displayed.

## VII. CONCLUSION AND ENHANCEMENT

A new technique for the calculation of polarities and strengths of sentiment phrases is planned during this project, that may be accustomed analyze linguistics similarity of sentences even with partial phrase matching. It uses a likelihood worth, instead of a customary worth for the polarity strengths of sentiment phrases, compared with the standard strategies.

According to the polarities and strengths of these phrases, it proposes multi-strategy sentiment analysis technique severally supported NB. notably, within the technique supported NB, it considers oppositive conjunctions. the tactic may be used for the sentiment analysis of documents. The practicability and effectiveness of the tactic is evidenced. In future, this project could concentrate on however the similarity may be realized victimization emoticons and Unicode character representations.

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