

SENTIMENT ANALYSIS OF IMDB MOVIE REVIEW

TY B.Tech. Mini Project Report

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submitted in partial fulfilment of the requirements for the award of the degree

of

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in

Computer Engineering & Technology

BY

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CERTIFICATE

It is hereby certified that the work which is being presented in the TY B.Tech. Mini Project Report entitled "Sentiment Analysis Of IMDb Movie Review", in partial fulfillment of the requirements for the award of the Bachelor of Technology in Computer Engineering & Technology and submitted to the School of Computer Engineering & Technology, Alandi(D), Pune, Affiliated to Savitribai Phule Pune University (SPPU), Pune is an authentic record of work carried out during an Academic Year 2020-2021, under the supervision of Mrs.Neha Hajare, School of Computer Engineering & Technology.

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ABSTRACT

Sentimental Analysis is a new variant in the research area. It basically refers to opinions or views of the different data that is being collected using surveys, comments and reviews over the web. The data set we have used in our project is from imdb movie rating website. Through this we can classify the review as positive or negative, a large amount of data is being generated daily. This will determine the polarity of reviews and these reviews will be classified as three types positive, negative and neutral.

LIST OF FIGURES

Fig. No.		Fig. Name	Page No.
Fig 3.1	Block Diagram		12
Fig 3.2	Use case Diagram		14
Fig 3.3	Sequence Diagram		15
Fig 3.4	Activity Diagram		16
Fig 4.1	Result 1		18
Fig 4.2	Result 2		18
Fig 4.3	Result 3		19
Fig 4.4	Result 4		19

CONTENTS

Acknowle	cknowledgements	
Abstract	Abstract	
List of Figures		iii
1.	Introduction	8
1.1	Motivation for the project	9
1.2	Problem Statement	
1.3	Objectives and Scope	
2.	Literature Survey	10
3.	System Design	12
3.1	Block diagram/ Proposed System setup	12
3.2	Use case Diagram	14
3.3	Sequence Diagram	15
3.4	Activity Diagram	16
3.5	Hardware & Software requirement	17
4.	Implementation and Results	17
4.1	Algorithm	17
4.2	Results	18
5.	Conclusion and Future scope	20
References		21

1.INTRODUCTION

What is Sentimental Analysis?

Sentiment analysis refers to the use of natural language processing, text analysis and computational linguistics to extract and identify subjective information in the whole sentence. It aims to determine the attitude of the speaker or writer with respect to some topic, in our case, with respect to movies. The attitude can be his or her judgement or evaluation. Affective state (i.e. emotional effect of writer when writing). The intended emotional communication (i.e. emotional effect the writer wishes to have on the reader). In the last few years there has been a rise in usage of social media such as blogs and social media networks such as Instagram, twitter, etc. which has fueled the interest in sentiment analysis.

Online opinion has turned into a kind of virtual currency with rapid increase of reviews, ratings, recommendations and other forms of expressions, for any business that is looking to market their products, identify new opportunities and manage their reputation. In order to automate the process of filtering the noise, understanding the conversations, identifying customers or viewers needs, many are now looking to the field of sentiment analysis. But the main problem of most of the sentiment analysis algorithms is that they use simple terms to express sentiment about a product or service. However, cultural factors, sentence negation, sarcasm, language ambiguity, use of slangs and differing context make it very difficult to turn the string of text into simple pro or con sentiment.

A fundamental task in sentiment analysis is classifying the polarity of given text. It focuses on whether the expressed opinion is positive, negative or neutral. Sometimes it goes beyond polarity and looks at emotional states such as "angry", "sad" and "happy".

Existing approaches to sentiment analysis can be grouped into four main categories. They are keyword spotting, statistical methods, and concept-level techniques. Keyword spotting classifies text by affect categories based on the presence of unambiguous affect words such as happy, sad, afraid, and bored. Statistical methods influence elements from machine learning such as latent semantic analysis, support vector machines, bag of words and Semantic Orientation. Concept-level approaches through the analysis of concepts that do not explicitly convey relevant information, but which are implicitly linked to other concepts that do so.

So our project, unlike others, aims to do sentiment analysis using lexical approach. In most of the string of text or sentences expressing the opinion about the movie, their thoughts which contain different writing habits, use of slang words, sarcasm, non-english words etc. so using lexical approach and using different libraries, we can easily counter those problems. Making use of tokenization our system would be easily identifying the main keywords in the sentences.

Then the sentence is examined for subjectivity, only the sentence with subjective expressions are retained and sentences which convey objective expression are discarded. In this manner the filtration and text preparation would do the rest work easy. Then using some common computation techniques sentiment detection would be done. Sentiment classification i.e. identification of the given sentiment is either positive or negative, is done

1.1 Motivations

Motivation for the project:- In the era of technology each and everything is going online. From as simple as booking bus or flight tickets to as complex as trading online. Every business is having their own websites on internet to provide different services such as, buy monthly grocery, make payments, book tickets, watch online movies etc. For these businesses, customers are god like. Customer's feedback / review/ opinions are very important to businesses, to analyse their growth, take closer look at their market, make changes accordingly and stand on the customer's demands.

Even industry like film industry, film makers are so much dependent on customers/consumers that they constantly need to keep a close eye on viewers likes/dislikes, opinions, feedbacks to make appropriate decisions while making films. This is where our project, Sentiment Analysis of Movie reviews kicks in.

1.21.2 Problem Statement

To perform Sentiment Analysis on IMdb reviews of Movies by lexicon approach and to identify whether review is positive, negative or neutral.

1.31.3 Objectives and Scope

To implement an algorithm for automatically classification of text into positive, negative and neutral.

To study and implement tokenization.

To study and implement lemmatization.

To calculate polarity value of review.

To calculate accuracy of the algorithm.

2. LITERATURE SURVEY

[1] Walaa Medhat of a School of Electronic Engineering, Canadian International College, Cairo Campus of CBU, Egypt has published (Ain Shams University) a paper based on Sentiment analysis algorithms and applications which can three main classification levels in SA: document-level, sentence-level, and aspect-level SA. They are presented an overview on the recent updates in SA algorithms and applications. s. After analyzing these articles, it is clear that the enhancements of SC and FS algorithms are still an open field for research.

The entity can represent individuals, events or topics. These topics are most likely to be covered by reviews. The two expressions SA or OM are interchangeable. These articles give contributions to many SA related fields that use SA techniques for various real-world applications. After analyzing these articles, it is clear that the enhancements of SC and FS algorithms are still an open field for research. They express a mutual meaning. They only represent the level on it so we are going for the next paper. Sentence-level SA aims to classify sentiment expressed in each sentence.

Sentiment Classification techniques can be roughly divided into machine learning approach, lexicon based approach and hybrid approach [69]. The Machine Learning Approach (ML) applies the famous ML algorithms and uses linguistic features. The Lexicon-based Approach relies on a sentiment lexicon, a collection of known and precompiled sentiment terms.

[2] Maite Taboada of a Simon Fraser University has published (Simon Fraser University) A paper based on Semantic orientation (SO) is a measure of subjectivity and opinion in text. It usually captures an evaluative factor (positive or negative). From these they are presented a word-based method for extracting sentiment from texts. This attribute to this criteria for selecting and ranking words, which include excluding ambiguous words and including fewer rather than more words. There exist two main approaches to the problem of extracting sentiment automatically. The lexicon-based approach involves calculating orientation for a document from the semantic orientation of words or phrases in the document. We have presented a word-based method for extracting sentiment from texts. Building on previous research that made use of adjectives, we extend the Semantic Orientation Calculator to other parts of speech.

We also introduce intensifiers, and refine our approach to negation. SO-CAL is applied to the polarity classification task, the process of assigning a positive or negative label to a text that captures the text's opinion towards its main subject matter. We show that SO-CAL's performance is consistent across domains and on completely unseen data.

The majority of the statistical text classification research builds Support Vector Machine classifiers, trained on a particular data set using features such as unigrams or bigrams, and with

or without part-of-speech labels, although the most successful features seem to be basic unigrams

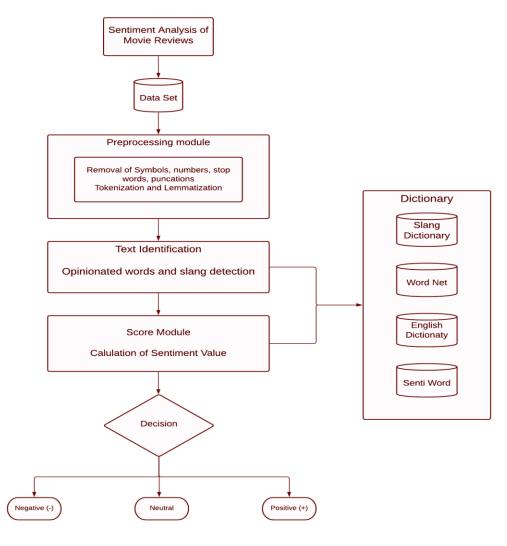
[3] Karthik Konar 1MCA Student, Dept. of Computer Engineering, NMIMS Mukesh Patel School of Technology Management & Engineering, Vile Parle(West) Mumbai has published (NMIMS Mukesh Patel School of Technology Management & Engineering, Vile Parle(West) Mumbai) in their paper which can be based on Sentimental analysis is that particular domain, where you try to understand human emotions with the help of a software. Human emotions are in written form, and we classify those sentiments as positive, negative, neutral. This paper provides a detailed comparison of various applications of sentiment analysis which was implemented using different approaches such as lexicon, machine learning, and VADER sentiment analysis.

There are various applications of sentimental analysis such as Review classification, Product review mining, etc. In review classification, we classify the user reviews into 3 categories i.e. positive, negative, and neutral. For e.g.: There are a lot of customers, who post a lot of reviews, how do we know the sentiment of each customer? This is where we can use sentimental analysis and classify the reviews to be positive, negative, neutral. Based on the experiment performed we found that VADER sentiment analysis provides the highest accuracy when compared with the lexicon and machine learning approach.

VADER Sentiment analysis approach to classify the accuracy of positive and negative reviews and the time taken for execution is very less when compared with the lexicon and machine learning approach.

3. SYSTEM DESIGN

3.1 Block Diagram



(Figure 3.1 Block Diagram)

For our project we have proposed the block diagram that explains how the data undergoes the execution in the whole program along with detailed explanation .

Dataset

It contains a record of reviews of different movies. It which contains over 5000 reviews with their respective sentiment for future analysis

Preprocessing Module

In this section preprocessing steps are carried out on a dataset. It includes data cleaning, tokenization, lemmatization which is explained later on in detail.

Text Identification

This module performs tasks where words are identified and classified with the help of dictionaries.

Dictionary

It contains sentiment resources which helps for Text Identification. These sentiment resources(dictionaries) contain positive and negative sentiment words. Some misspelled words are also included in the lexicon as they appear frequently in the social media text.

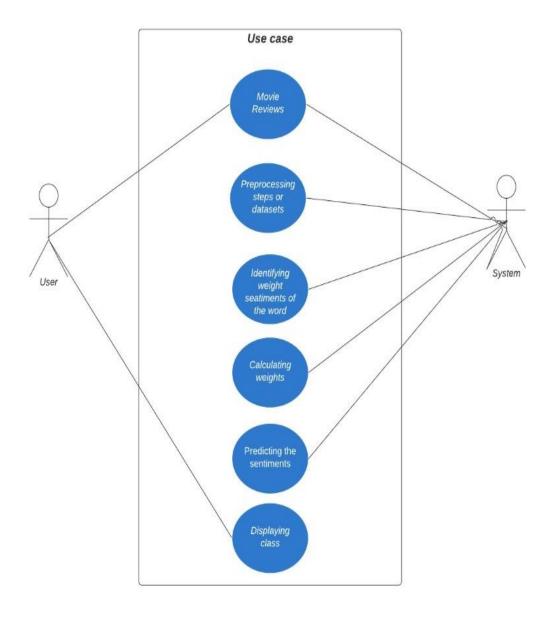
Score Module

It will help us to calculate the sentiment value of the review with the help of a dictionary which has sentiment values for words. It calculates positive, negative and neutral sentiment value a review separately and then it would calculate the final sentiment score

EXPLANATION:

- 1. Here the dataset is first imported from the database
- 2. Preprocessing steps are carried out on a dataset. It includes data cleaning, tokenization, lemmatization
- 3. It identifies the words based on adjectives and classified with the help of dictionaries.
- 4. Then with the help of score module, calculate sentiment value
- 5. The review will be classified into positive, negative, neutral based on sentiment value

3.2 Use case diagram



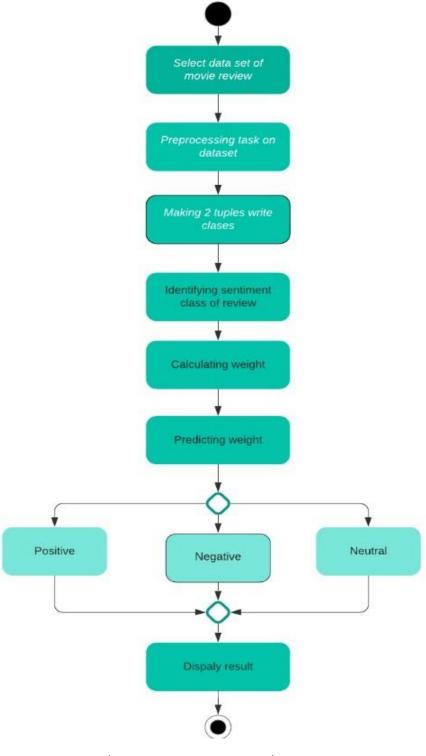
(Figure 3.2 Use case Diagram)

3.3 Sequence diagram



(Figure 3.3 Sequence Diagram)

3.4 Activity diagram



(Figure 3.4 Activity Diagram)

3.5 Hardware & Software requirement:

- Windows 10 Machine
- (8 GB RAM, 256 SSD)
- Anaconda (Virtual Environment)
- Jupyter Notebook

4. IMPLEMENTATION DETAILS

4.1 Algorithm analysis:

- Step 1: Importing the data set
- Step 2: Removing single letters and numeric from the reviews
- Step 3: Snippet to tag the parts of speech for each review
- Step 4: Narrow down each review with only adjectives and verb forms
- Step 5: Making a list of tuples which has review and its respective class
- Step 6: Separating the data set by each class i.e. "Positive", "Negative", "Neutral"
- Step 7: Merge the reviews in each class in to three single lists
- Step 8: Splitting the words in each list to get the word frequencies
- Step 9: Finding the frequencies of the unique words in each class
- Step 10: Snippet to set the dictionary size
- Step 11: Identifying the similar words for each class in each review
- Step 12: Calculating the score of positive, negative and neutral words in each review
- Step 13: Predicting the class based on the score

4.2 Result

Creating a data frame with respective weigths In [39]: UL = pd.DataFrame() In [40]: UL['POS']= PW_1 In [41]: UL['NEG']= NGW_1 In [42]: UL['NEU']= NUW_1 In [43]: UL['truth']=A['sentiment'] In [44]: UL['PRE'] = A['review'] In [45]: UL.head(5) Out[45]: POS NEG NEU 0 1.00 0.92 0.0 positive One of the other reviewers has mentioned that . 1 1.00 0.81 0.0 positive A wonderful little production.

The.. 2 1.00 0.86 0.0 positive I thought this was a wonderful way to spend ti... 3 0.72 1.00 0.0 negative Basically there's a family where a little boy ...

(Figure 4.1 Result 1)

Predicting the class based the weights.

4 1.00 0.76 0.0 positive Petter Mattei's "Love in the Time of Money" is...

```
In [46]: for i in range(500):
                                  if (UL.POS[i]-UL.NEG[i] > 0.03):
                                           UL.PRE[i] = 'positive'
                                  if (UL.POS[i]-UL.NEG[i] < -0.03):
                                            UL.PRE[i] = 'negative'
                                  elif (UL.POS[i]-UL.NEG[i] \leftarrow 0.03 and UL.POS[i]-UL.NEG[i] \rightarrow= -0.03):
                                            UL.PRE[i] = 'neutral
                        F:\anaconda\lib\site-packages\ipykernel_launcher.py:3: SettingWithCopyWarning:
                       A value is trying to be set on a copy of a slice from a DataFrame
                        See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-ve
                            This is separate from the ipykernel package so we can avoid doing imports until
                       F:\anaconda\lib\site-packages\lip\kernel_launcher.py:5: Setting\lib\text{Setting\limbol}\text{Setting\limbol}\text{Warning:}

A value is trying to be set on a copy of a slice from a DataFrame
                        \textbf{See the caveats in the documentation:} \ \text{https://pandas.pydata.org/pandas-docs/stable/user\_guide/indexing.html \# returning-a-view-vertical part of the documentation of the
                        rsus-a-copy
                        A value is trying to be set on a copy of a slice from a DataFrame
                        See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-view-ve
                            import sys
In [47]: UL.head(5)
Out[47]:
                                POS NEG NEU
                                                                                                PRE
                         0 1.00 0.92 0.0 positive positive
                          1 1.00 0.81 0.0 positive positive
                         2 1.00 0.86 0.0 positive positive
                         3 0.72 1.00 0.0 negative negative
                         4 1.00 0.76 0.0 positive positive
```

(Figure 4.2 Result 2)

```
In [48]: pd.crosstab(UL.truth,UL.PRE)
Out[48]:
             PRE negative neutral positive
             truth
          negative
                      257
          positive
                              14
                                    223
In [49]: print (accuracy_score (UL.truth,UL.PRE))
         0.96
In [50]: print (recall_score(UL.truth,UL.PRE, pos_label= 'negative', average='micro'))
         0.96
         F:\anaconda\lib\site-packages\sklearn\metrics\_classification.py:1321: UserWarning: Note that pos_label (set to 'negative') is
         ignored when average != 'binary' (got 'micro'). You may use labels=[pos_label] to specify a single positive class.
           % (pos_label, average), UserWarning)
In [51]: UL.to_csv("ultimate.csv", sep=',', encoding='utf-8')
```

(Figure 4.3 Result 3)

```
In [1]: import pandas as pd
In [10]: import numpy as np import pandas as pd
         import matplotlib.pyplot as plt
 In [4]: A = pd.read_csv('C:/Users/Mangesh Shingare/ultimate.csv')
 In [5]: A.head()
 Out[5]:
             Unnamed: 0 POS NEG NEU
                      0 1.00 0.92
                                    0.0 positive
                      1 1.00 0.81
                                    0.0 positive
                     2 1.00 0.86
                                    0.0 positive
                      3 0.72 1.00
                                    0.0 negative negative
                     4 1.00 0.76 0.0 positive positive
In [22]: A.head().plot.bar(y=['POS','NEG'], x='Unnamed: 0')
Out[22]: <matplotlib.axes._subplots.AxesSubplot at 0x190b5198a88>
           1.0
           0.8
           0.4
           0.2
```

(Figure 4.4 Result 4)

5. CONCLUSION & FUTURE SCOPE

This paper about Sentiment Analysis of IMDb Movie Reviews using Lexicon approach. We have presented a Dictionary-based method for extracting sentiment from texts The algorithm provides a clear idea about the importance and working of the Sentiment predictions. It provides accuracy up to 96%

Sentiment analysis is a uniquely powerful tool for businesses that are looking to measure attitudes, feelings and emotions regarding their brand. To date, the majority of sentiment analysis projects have been conducted almost exclusively by companies and brands through the use of social media data, survey responses and other hubs of user-generated content.

The future of sentiment analysis is going to continue to dig deeper, far past the surface of the number of likes, comments and shares, and aim to reach, and truly understand, the significance of social media interactions and what they tell us about the consumers behind the screens.

Our project tries to identify the sentiments value of the review classifying it into positive or negative review. Later on we can create GUI interface which make ease the task of classifying the review.

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