

## TEXT CLASSIFICATION ON TWITTER DATA

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**Abstract:** Sentiment analysis is a classification problem where the main focus is to predict the polarity of words and then classify them into positive or negative sentiment. Two types of Classifiers are mainly used, namely lexicon-based and machine learning based. The first one include SentiWordNet and Word Sense Disambiguation while the second one include Multinomial Naive Bayes (MNB), Logistic Regression(LR), Support Vector Machine(SVM) and RNN Classifier. In this paper, existing datasets have been used, the first one from “Sentiment140” from Stanford University, consisting of 1.6 million tweets and the other one originally came from “Crowdfower’s Data for Everyone library”, consisting of 13870 entries, and both datasets are already categorized as per the sentiments expressed in them. Textblob, Sentiwordnet, MNB, LR, SVM and RNN Classifier are applied on the above dataset and a comparison is drawn between the results obtained from above mentioned sentiment classifiers, classifying tweets according to the sentiment expressed in them, i.e. positive or negative. Also, along with the machine learning approaches, an ensemble form of MNB, LR and SVM has been performed on the datasets and compared with the above results. Further the above trained models can be used for sentiment prediction of a new data.

**Index Terms – Sentiment Analysis, Classifiers, lexiconbased, machine learning, Multinomial Naive Bayes, Logistic Regression, Textblob, SentiWordNet.**

### I. INTRODUCTION

Sentiment Analysis which means to analyze the underlying emotions of a given text using Natural Language Processing (NLP) and other techniques to extract a significant pattern of information and features from a given large corpus of text. It analyses the sentiment and attitude of the author towards the topic of the subject

mentioned in the text. This text can be a part of any document, post on social media or from any database source. Sentiments can be classified as objective or subjective, positive or negative or neutral. This classification can be either lexicon-based or machine learning based. Lexicon based classification makes use of already existing dictionary which has pre-

assigned scores to each word and those scores are used to calculate the overall sentiment expressed in the sentence whereas in machine learning based classification, a model is trained using some ML algorithm using some labeled data and then use that model to predict a class for a new text. Twitter is nowadays easily one of the most popular micro blogging platforms and millions of users express their views publicly on Twitter making it a rich source of information on public opinions and thus, helpful in sentiment analysis on any topic. Here, lexicon and machine learning based approaches have been incorporated to reveal the prevailing sentiments of tweets. Textblob, SentiWordNet and Word Sense Disambiguation are giving the correct sense of a word in a given context for Lexicon-based Sentiment analysis while among the machine learning based algorithms MNB, LR, SVM and RNN Classifier have been used. In this paper, a comparison has been presented in terms of accuracy in predicting the sentiment of a given tweet. An ensemble approach has also been implemented on the datasets which involve majority voting of MNB, LR and SVM and the results are being

compared with the rest of the approaches.

## II. EXISTING SYSTEM

Sentiment analysis is a classification problem where the main focus is to predict the polarity of words and then classify them into positive or negative sentiment. Classifiers used are of mainly two types, namely lexicon-based and machine learning based. The former include SentiWordNet and Word Sense Disambiguation while the latter include Multinomial Naive Bayes(MNB), Logistic Regression(LR), Support Vector Machine(SVM) and RNN Classifier.

## III. PROPOSED SYSTEM

Lexicon and machine learning based approaches have been employed to reveal the prevailing sentiments of tweets. Textblob, SentiWordNet and Word Sense Disambiguation are giving the correct sense of a word in a given context for Lexicon-based Sentiment analysis while among the machine learning based algorithms MNB, LR, SVM and RNN Classifier have been used. The project involved analyzing the design of few applications so as to make the application more users friendly. To do so, it was really important to keep the navigations from one screen to the other well-ordered and

at the same time reducing the amount of typing the user needs to do. In order to make the application more accessible, the browser version had to be chosen so that it is compatible with most of the Browsers.

### **RELATED STUDY**

Sentiment analysis is an application of natural language processing. It is also known as emotion extraction or opinion mining. This is a very popular field of research in text mining. The basic idea is to find the polarity of the text and classify it into positive, negative or neutral. It helps in human decision making. To perform sentiment analysis, one has to perform various tasks like subjectivity detection, sentiment classification, aspect term extraction, feature extraction etc. This paper presents the survey of main approaches used for sentiment classification. The wide spread of World Wide Web has brought a new way of expressing the sentiments of individuals. It is also a medium with a huge amount of information where users can view the opinion of other users that are classified into different sentiment classes and are increasingly growing as a key factor in decision making. This paper contributes to the sentiment analysis for customers' review classification which is helpful to

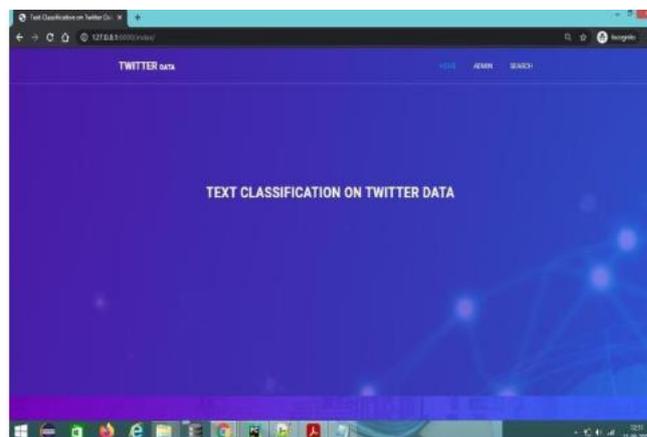
analyze the information in the form of the number of tweets where opinions are highly unstructured and are either positive or negative, or somewhere in between of these two. For this we first preprocessed the dataset, after that extracted the adjective from the dataset that have some meaning which is called feature vector, then selected the feature vector list and thereafter applied machine learning based classification algorithms namely: Naive Bayes, Maximum entropy and SVM along with the Semantic Orientation based WordNet which extracts synonyms and similarity for the content feature. Finally we measured the performance of classifier in terms of recall, precision and accuracy. One of the most important parts of running business successfully is analyzing customer's opinion and sentiments. In this paper, the paragraph of sentences given by the customer is accepted and after extracting each and every word, they are checked with the stored (database has been maintained here) parts of speech, articles and negative words. After checking against the database, CFG is used to validate proper formation of the sentences. Each sentences are delimited by `.' or `?' or `!'. Emotions are detected as - positive, negative or neutral sentence. There are 3

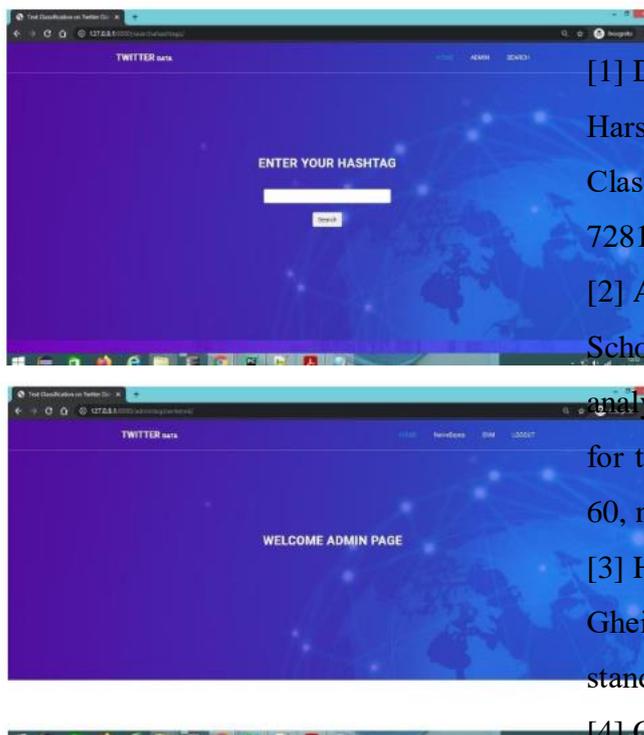
types of cases: 1. If the paragraph contains more positive sentences than negative, then overall result will be positive. 2. If the number of negative sentence is greater than positive sentence, then the overall result is negative. 3. If there are same numbers of positive and negative sentences in the input paragraph, then the result is neutral and if a sentence has been entered that is a normal statement neither positive nor negative, that will be also considered as neutral.

### WORKING METHODOLOGY

Decision Tree algorithm has become one of the most used machine learning algorithm both in competitions like Kaggle as well as in business environment. Decision Tree can be used both in classification and regression problem. A decision tree is arriving at an estimate by asking a series of questions to the data, each question narrowing our possible values until the model get confident enough to make a single prediction. The order of the question as well as their content is being determined by the model. In addition, the questions asked are all in a True/False form. This is a little tough to grasp because it is not how humans naturally think, and perhaps the best way to show this difference is to create a real decision tree

from. In the above problem  $x_1$ ,  $x_2$  are two features which allow us to make predictions for the target variable  $y$  by asking True/False questions. For each True and False answer there are separate branches. No matter the answers to the questions, we eventually reach a prediction (leaf node). Start at the root node at the top and progress through the tree answering the questions along the way. So given any pair of  $X_1$ ,  $X_2$ . One aspect of the decision tree I should mention is how it actually learns (how the 'questions' are formed and how the thresholds are set). As a supervised machine learning model, a decision tree learns to map data to outputs in what is called the training phase of model building.





### Conclusion

Various techniques for both lexicon-based and machine learning based, have been applied in this project and the results are compared. It has been observed that for a totally new data/text machine learning based models trained over a related data are much more accurate than the classification based on standard dictionaries. This is because of the fact that the text that's being observed i.e., the tweets are highly informal and do not use the standard grammar rules or the spelling and thus the data here is highly unstructured. The comparison results can be clearly observed among different machine learning algorithms also. As of now, among the algorithms used, RNN is observed to have the highest accuracy.

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