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Implementation on Sentiment Analysis of Bollywood Movies Reviews by Using ML-Classifier Algorithm

N. S. Magar¹, S. N. Deshmukh²

Department of CSE, Dr. Babasaheb Ambedkar Marathwada University, Aurangabad, (MH), India

ABSTRACT: Sentiment analysis is related to the analysis of human emotions and their opinions from the text. Analysis of the sentiment finds and justifies the sentiment of that person with respect to the content. Online sites and Social media contains the data in the form of Reviews, tweets, posts etc. The tweets in the twitter contents are maximum 280 character long. In this research it is tried to build sophisticated model for analyzing the review and tweets regarding Bollywood movies and Songs. As the classifier used are Naïve Bayes, Random Forest, and they are classified as hit, flop, and average by extracting the sentiment from each of the tweet.

KEYWORDS: Feature Vector, MLC, Naïve Bayes & Random Forest. Classifier, Movies Review etc

I. INTRODUCTION

As we know the social networking is online platform. Which connect People with each other using internet through the online media message, blog post, real time review, Google review, conversation forums, and many more. Various review as well as tweet are also available on this platform using which it is possible to predict the status of movie or the song before deciding to go for it. This all can be done with the help of sentiment analysis and opinion mining and techniques.

II. RELATED WORK

There are two techniques which are widely used to detect the sentiments from text are following.

- A. Symbolic techniques (ST)
- B. Machine Learning techniques (MLT).

A. Sentiment analysis using Symbolic Techniques

Suggested an approach for sentiment analysis called 'bag of words' [4]. In the mentioned approach, individual words are neglected and only collections of words are considered [6]. which determines an emotional matter in a sentence. Word Net is used for getting synonyms and distance metric to find the orientation of adjectives.

B. Sentiment analysis by using Machine Learning Techniques

In Sentiment analysis by using Machine Learning Techniques: machine learning algorithms allow computers to evolve behaviours based on empirical data from sensor or databases [16]. A major focus of machine learning research is to automatically learn to recognize complex patterns and make intelligent decisions based on data the difficulty lies in the fact that the set of all possible behaviours given all possible inputs is too large to be covered.

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III. PROPOSED WORK

Various techniques have been developed to do sentiment analysis of tweets and Reviews. In our Algorithms feature vectors is used. The following Functional Block Diagram Explain proposed system will work.it has two mode training domain & Its Testing domain

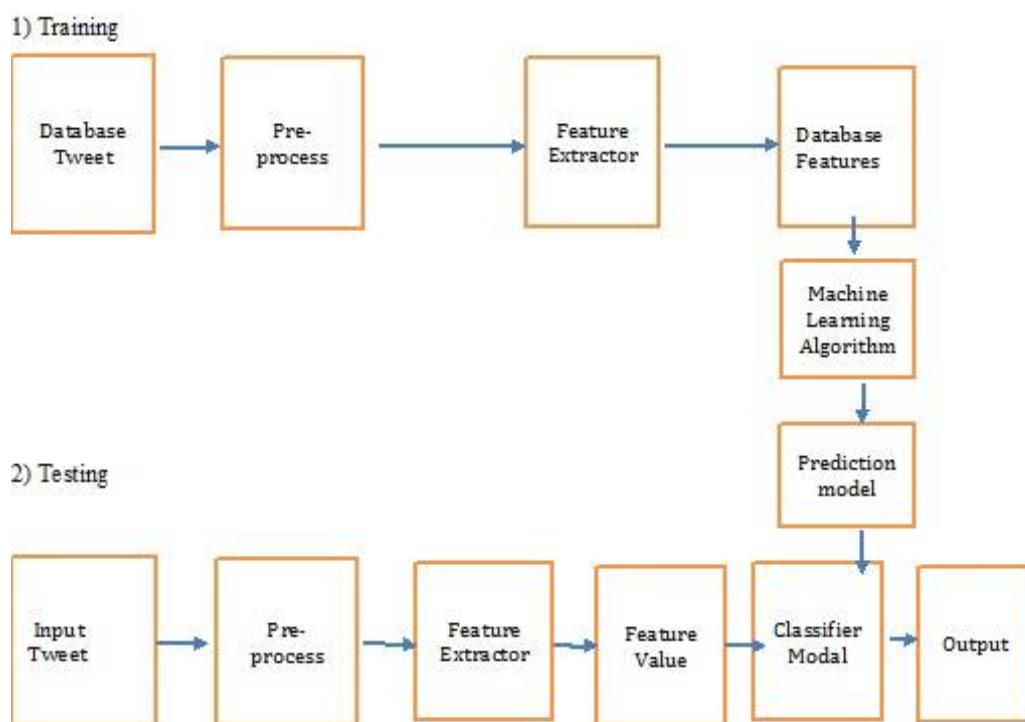


Fig1 : Block Dia gram of Propose System.

Twitter allows researchers to collect tweets by using a Twitter API. One must have twitter account to obtain twitter credentials like (i.e. APIkey, API secret, Access token and Access token secret) which can be obtained from twitter developer site. Then install a twitter library to connect to the Twitter API. Twitter has developed its own languageconventions. The following are examples of Twitterconventions. a) “RT” is an acronym for retweet, which indicates that the user is repeating or reposting.

- b) “#” stands for hashtag and is used to filter tweets according to topics or categories.
- c) “@user1” represents that a message is a reply to a user whose user name is “user1”.
- d) Emoticons and colloquial expressions or slang languages are frequently used in tweets
- e) External Web links are also frequently found in tweets to refer to some external sources.
- f) Length: Tweets are limited to 280 characters.

Creation of a Dataset

Since standard twitter dataset is not available for electronic products domain, we created a new dataset by collecting tweets real time. Tweets are collected automatically using Twitter API and they are manually annotated as positive or negative. A dataset is created by taking 15 tweets. These tweets are labelled as like 5 Hit, 5 Average and 5 Flops for getting dataset.



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Pre-process

The data pre-processing can often have a significant impact on the performance of a supervised ML algorithm. The steps that are carried out in pre-processing of data are as follows.

- a) Stop-words Removal: The commonly used words like a, an, the, has, have etc. which carry no meaning i.e. do not help in determining the sentiment of text while analysing should be removed from the input text.
- b) Punctuation Removal: Punctuation marks such as comma or colon often carry no meaning for the textual analysis hence they can be removed from input text. The mention of other accounts with the @ sign were also removed including any other symbol or special character such as &*""?!,: \$ % # () □ / +<> = [] n ^ _ { } | ~.

Feature Extraction

Scores are calculated based on sentiment dictionary which has positive, negative and neutral special words that define their respective sentiments.

Positive score:-Total positive score obtained by adding the positive scores of each positive adjective. Negative score:-Total negative score obtained by adding the negative scores of each negative adjective. Neutral score:-Total neutral score obtained by adding the neutral scores of each neutral adjective.

Classification

For classification there are two classifier Naïve Bayes and Random Forest. Output i.e. Bollywood movie performance is classified into 3 classes. Hit, Average and Flop Naïve Bayes
Random Forest

1) Naive Bayes Classifier

The main outcome by using Naive Bayes [11] classifier is that it has an ability to analyse each feature independently so which makes it very popular. The Probability function of Naïve Bayesian classifier is given as,

$$P\left(\frac{C}{x}\right) = \frac{P\left(\frac{x}{C}\right)P(C)}{P(x)} \dots\dots\dots(1)$$

- P (c|x) is the posterior probability of class (target) given predictor (attribute).
- P(c) is the prior probability of class.
- P (x|c) is the likelihood which is the probability of predictor given class.
- P(x) is the prior probability of predictor

2) Random Forest

It is a method for classification and regression of the other tasks, that operate by constructing a multitude of decision model system that is applicable for trees at training time and outputting at the class the probability function of random forest is given as .

$$\hat{f} = \frac{1}{B} \sum_{b=1}^B f_b(x') \dots\dots\dots(2)$$

- X' is a prediction for unseen sample.
- f_b is a regression tree.
- B is a training sample.

Learning Algorithm

Step 1: Create data files for the classifier.

(1.1) Create a file of tweets with their sentiment for each sentiment analyzer (test set).



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- (1.2) Apply the pre-processing operations
- (1.3) Create a dictionary of negative, positive and neutral labelled tweets for each sentiment analyser (training set).
- (1.4) Convert xlsx files i.e. excel sheet

Step 2: Build Naïve Bayes/Random Forest classifier model.

- (2.1) Create model of each analyser by providing the training set file.
- (2.2) Store the learning model

Testing Algorithm

Step 1: Load the test set file.

Step 2: Apply the pre-processing operation

Step 3: Find the sentiment score i.e. features

Step 4: Execute the model based classifier selection on the test set.

Step 5: Save results in the output file

IV. RESULTS AND ANALYSIS

The proposed system was developed on MATLAB R2013 a using GUI. By using the twitter account it created the application for our project and the valid credentials of twitter account it is known as Json file and the input database that get from Twitter account. And will obtain the result which will be shown in TABLE 1 & Bar Graph 1

TABLE 1. Comparative Analysis for ML-Classifer

| <i>Movie Name</i> | <i>No. of Tweets</i> | <i>Naïve Bayes</i> | | | <i>Random Forest</i> | | |
|--------------------|----------------------|----------------------------|---------------------|-----------------------|---------------------------|---------------------|-----------------------|
| | | <i>Evaluation Time (S)</i> | <i>Accuracy (%)</i> | <i>Error Rate (%)</i> | <i>Evaluation Time(S)</i> | <i>Accuracy (%)</i> | <i>Error Rate (%)</i> |
| <i>Sanju</i> | <i>30</i> | <i>143.29</i> | <i>86.20</i> | <i>13.79</i> | <i>118.81</i> | <i>89.65</i> | <i>10.34</i> |
| <i>Race3</i> | <i>35</i> | <i>88.77</i> | <i>88.57</i> | <i>11.42</i> | <i>91.23</i> | <i>91.42</i> | <i>8.57</i> |
| <i>Raazi</i> | <i>40</i> | <i>140.96</i> | <i>90.00</i> | <i>10.00</i> | <i>133.43</i> | <i>95.00</i> | <i>5.00</i> |
| <i>Soorma</i> | <i>25</i> | <i>68.94</i> | <i>88.00</i> | <i>12.00</i> | <i>71.70</i> | <i>92.00</i> | <i>8.00</i> |
| <i>Blackmail</i> | <i>30</i> | <i>91.08</i> | <i>86.66</i> | <i>13.33</i> | <i>93.32</i> | <i>93.33</i> | <i>6.66</i> |
| <i>Missing</i> | <i>25</i> | <i>71.61</i> | <i>84.00</i> | <i>16.00</i> | <i>73.20</i> | <i>92.00</i> | <i>8.00</i> |
| <i>Octomber</i> | <i>16</i> | <i>47.06</i> | <i>81.25</i> | <i>18.75</i> | <i>48.96</i> | <i>93.75</i> | <i>6.25</i> |
| <i>102 not out</i> | <i>32</i> | <i>103.49</i> | <i>87.51</i> | <i>12.50</i> | <i>106.16</i> | <i>93.75</i> | <i>6.25</i> |
| <i>Average</i> | <i>29.10</i> | <i>85.45</i> | <i>86.52</i> | <i>13.47</i> | <i>92.10</i> | <i>92.61</i> | <i>7.38</i> |

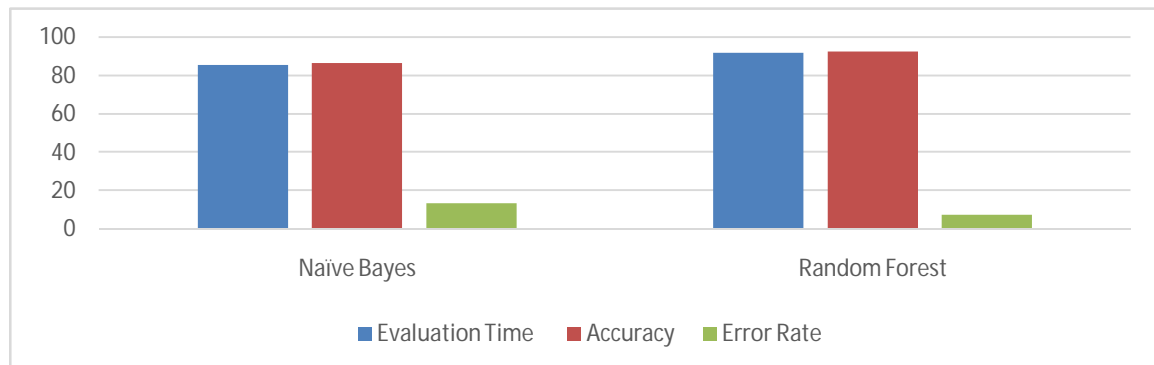
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BarGraph 1. Performance of classifiers



IV. CONCLUSION

We can observe that the MLT is very easier and efficient than other techniques. So MLT are easily applied to any online social sites for analysis of sentiments. Considered the above data, it can be concluded that works better the Naive Bayes or Random Forest. However, final conclusion on above fact is that the prediction methods used can be suggested for getting the prediction as it is giving accurate data. The Random Forest Result (classifier Accuracy, Evaluation Time and Error Rate) is better than the Naive Bayes classifier. So we can increase the accuracy of classification as we increase the training data.

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