

TWITTER SENTIMENT ANALYSIS OF MOVIE REVIEWS

Shalaka Ghanekar, Aishwarya Surve, Prachi Abhyankar

Student, Student, Professor/Guide

Information Technology,

Finolex Academy of Management and Technology, Ratnagiri, India

Abstract : Now a days, online social networks usage are pervasive. Mining text present in online social networks will be useful for predictive analytic. Predicting information from unstructured data present in social networks is a challenging research problem. Extracting, identifying or characterizing the sentiment content of the text unit using statistics and machine learning methods are referred as sentiment analysis or text analysis. In this work sentiment analysis using algorithms such as Decision trees and Support vector machines, which are machine learning algorithms will be demonstrated using tools like Anaconda, WEKA etc. Sentiment analysis using Support Vector Machines (SVM) showed high accuracy when compared to Naïve Bayes.

IndexTerms - Twitter sentiment; ensemble learners; twitter sentiment analysis.

I. INTRODUCTION

Now a day's online reviews has become an increasingly trendy approach for public to share their sentiments and opinions towards the product bought and services received. Online reviews become one of the most important part of any business today. Online reviews presents wealth of information on products and services, if the reviews are properly utilized then it is valuable for vendors in network and social intelligence so that business will be improved. A recent study focused on cost-effective values of online reviews and provides deep understanding between product reviews and their sales performance. People tend to read online reviews understanding the opinions and sentiments and trust them as much as they are recommended by their friends or families. Twitter, a social networking service plays significant role in social networking research. Tweets give rich information about movie, product, or service. Sentiments perform very important role for predicting future sales performance, a mix of good and even bad reviews will create a positive effect on the sales performance and sales prediction. Generally, a binary opposition in opinions is assumed. For/against, like/dislike, good/bad, etc. [4][5][6] Some sentiment analysis jargon are Semantic orientation and Polarity.

Twitter sentiment analysis is commonly treated as a (supervised) classification problem, where tweets are commonly categorized into four classes--positive, negative, neutral and irrelevant—based on the opinions expressed in a tweet. In practice, the irrelevant class is often consolidated with neutral, thus converting this to a ternary-classification problem. To this end, a wide collection of classification algorithms such as Support Vector Machine (SVM) and Random Forests has been applied. Recently, one sees the emergence of constructing ensemble learners for Twitter sentiment classification. Coupled with feature engineering, ensemble learners have shown to outperform individual learners. [1]

II. CHARACTERISTICS

Characteristics of tweets:

- Length: The determined probable length of a Tweet is 140 letters. Our training data is having a mean length of 82 characters.
- Data availability: Using twitter API we can amass lakhs of tweets for analysis purpose.
- Linguistic model: Users of twitter media are allowed to post their tweets in their native languages. The frequency of misspells is higher in twitter when compared with other domains.
- Domain: Twitter users can post their tweets from several media including PCs, Cell phones etc. Tweets are generated about a variety of things (actually no constraint) which differs from specific domains like movie reviews. We perform sentiment analysis which is the process of finding out whether a piece of statement is neutral, positive or negative and is also referred to as opinion mining that finds out the attitude or opinion of a speaker. To construct a sentiment analyser, the first thing is to get equipped with the correct approach. One such approach is machine learning where we can develop several approaches to classify opinions. We use supervised machine learning algorithms which allows automated aggregated feedback without the need of manual intervention. We use Support Vector Machines (SVM) and Naive Bayes (NB) classifiers to achieve high accuracy. Both of these approaches require training data.

III. RELATED WORK

A comprehensive survey in note that Naïve Bayes and SVM have been the two most commonly used algorithms for twitter sentiment analysis. As for ensemble learners, authors in describe a collection of ensemble classifiers consisting of SVM, Random Forests (RF), Logistic Regression and Multinomial Naïve Bayes (MNB). One of the best ensemble learners includes SVM, RF and MNB using features such as words and other lexical features. In, ensemble methods built on top of Naïve Bayes, Maximum Entropy, Decision Tree, k-Nearest Neighbors and SVM are discussed. The authors however did not evaluate the ensemble learners on Twitter data, but on user reviews. Note that both studies involve intensive feature engineering towards constructing competitive ensemble learners. Comparing our work with the above ensemble methods, this works considers off-the-shelf implementations of four relatively different classifiers, namely, MaxEnt and Naïve Bayes in the Mallet machine learning toolkit, SentiStrength and Pattern. No additional feature engineering is involved. This ensures simplicity and reproducibility of our algorithms. This work also studies how well these classifiers and our ensemble learners adapt to different training datasets. This will help address the lack of labeled data when performing supervised Twitter sentiment analysis. [1].

There are numerous research papers and studies that focus on sentiment classification for Twitter. These studies describe some interesting methodologies of detecting and identifying sentiment from twitter data. Supervised machine learning approach requires an annotated training dataset. The annotation was done based on the presence of either positive smiley emoticons such as ":)" or negative frowning emoticons such as ":(". The process of training the system on the negative and positive sentiment messages alone ignores the important effect of neutral sentiment tweets. This paper includes neutral and irrelevant tweets. Go's work strips emoticons from the training data, eliminating a rich source of sentiment information. This work includes emoticons for feature extraction. Go's work also considers the usefulness of different feature sets, including unigram, bigram, unigram + bigram, and parts of speech tags. The results of this study show that the unigram features work very well, with small improvement when bi-grams are also used. We also used unigram as a feature in this paper. Hu and Liu's system for customer review opinion mining uses a sentiment analysis component based on word lists generated for both positive and negative sentiment orientation. This paper also incorporated these sentiment word lists. According to Liu's work, word lists of sentiment orientation are useful but insufficient on their own to determine overall text sentiment. We are also not relying just on opinion lexicon and have used additional attributes as features for classification.[3]

Abbreviations and Acronyms

WEKA – Waikato Environment for Knowledge Analysis

SVM- Support Vector Machines

Equations [3]

Equation I - The Precision (P) is calculated as the ratio of instances that were predicted correctly with respect to the predicted size of the instances. It is calculated using the following equation:

$$\text{Precision} = \frac{TP}{TP + FP} \quad (3.0)$$

Equation II - The Recall (R), also known as the True Positive Rate (TPR) is the ratio of correctly predicted instances against the actual number of instances. It is represented using the following equation:

$$\text{Recall} = \frac{TP}{TP + FN} \quad (3.1)$$

Equation III - The accuracy (AC) is the calculated ratio of the total number of correct predictions against the total number of instances in the data set size. It is determined as follows:

$$\text{Accuracy} = \frac{TP + TN}{TP + TN + FP + FN} \quad (3.2)$$

Equation IV- F-score (F) is used to express the balance between the recall and precision of a classification run. The efficiency of the accuracy of a test is measured using its F-score. The F-score is calculated as the harmonic mean of recall and precision. It is expressed through following equation:

$$F1 - \text{Score} = \frac{\text{Precision} + \text{Recall}}{2} \quad (3.3)$$

IV. ALGORITHMS

SVM-

A. Pre-processing

The tweets are extracted for pre-processing. For Sentiment classification, numbers, stop words are not needed. They create noise in data. The pre-processing steps are removing numbers, punctuation, stop words and stemming. Stemming is the process of transforming the word into root word. [2][9][10]

B. POS Tagging

The tweets are POS tagged to extract the frequent adjectives, nouns and adverbs. These are the valuable adaptive sentiment words. The PMI-IR values are calculated for the sentiment words and are assigned to the tweets to get the feature vector. [2]

C. Learning with Unlabelled Data

The feature vectors with the labelled data are given to adaptive training algorithm. The classifier is trained on different topics. The unlabelled data U are chosen such that it has confidence threshold. These unlabelled data are predicted and augmented to L. The new features are updated to train the SVM for next iteration. [2]

D. Calculation of Evaluation of Metrics

The evaluation of the query results is calculated by using equation (I), (II), (III), (IV). The proposed model is evaluated by testing the model by providing the test data. The metrics are calculated for the model and results are found. [2]

E. Performance Analysis

Step length is the maximum number of most confident unlabelled data that can be chosen from the test dataset.. The graph shows that even the step length increases the performance of the algorithm better. As the training data size increases the new feature values are updated for accurate classification. [2][7][8]

There are four main advantages:

- It has a regularisation parameter, which makes the user think about avoiding over-fitting.
- It uses the kernel trick, so that you can build in expert knowledge about the problem via engineering the kernel.
- An SVM is defined by a convex optimisation problem for which there are efficient methods.
- It is an approximation to a bound on the test error rate, and there is a substantial body of theory behind it which suggests it should be a good idea.

The disadvantages are:

The theory only really covers the determination of parameters for a given value of the regularisation and kernel parameters and choice of kernel. In a way the SVM moves problem of over-fitting from optimising the parameters to model selection. [2]

V. ARCHITECTURE DIAGRAM

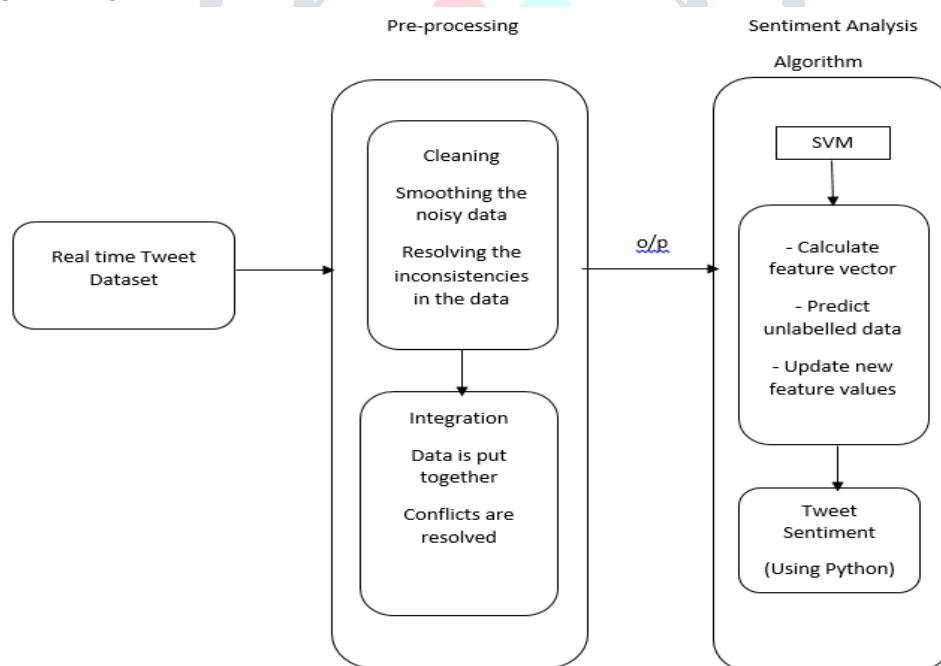


Fig 5.1. Architecture diagram

VI. IMPLEMENTATION

Extraction:

Twitter Archiver lets you easily save tweets for any search keyword or hashtag in a Google Spreadsheet. Enter a search query, or hashtag, and all matching tweets are automatically saved in the Google Sheet. Use boolean search or include advanced Twitter search operators to create more complex queries.

Twitter Archiver can be used for saving tweets around trending hashtags, conference tweets, your brand mentions, geo-tagged tweets, and more. It polls twitter every hour and pulls all the matching tweets into the Google Spreadsheet.

Pre-processing:

1. Load the raw text:
The text is small and will load quickly and easily fit into memory. This will not always be the case and you may need to write code to memory map the file. Tools like NLTK will make working with large files much easier.
2. Split into tokens:
Split the document by white space, including “ ”, new lines, tabs and more. We can do this in Python with the split () function on the loaded string.
3. Convert to lowercase:
Converting tweets to lowercase will ease the work of pre-processing.
4. Remove punctuation from each token:
Python provides a constant called string.punctuation that provides a great list of punctuation characters.
5. Filter out remaining tokens that are not alphabetic
6. Filter out tokens that are stop words:

Stop words do not contribute to the deeper meaning of the phrase. They are the most common words such as: “the”, “a”, and “is”. After applying SVM algorithm we will get the final output i.e. Classification of tweets.

SVM:

1. Install scikit_learn, nltk, scipy, numpy.
2. Import sys, time, re, nltk, joblib.
3. Text feature set X and feature values;
4. Non-text feature set X1 and feature values;
5. L: labeled tweets containing K sentiment classes on mixing topics;
6. U: unlabeled tweets on topic
7. t: classification confidence threshold;
8. l: the maximum number of unlabeled tweets selected in one iteration;
9. M: the maximum number of co-training iterations
10. Loop {
 - a. Repeat
 - b. [Adapting to unlabeled data]
 - c. Calculate confidence scores
 - d. Select the l most confident unlabeled tweets
 - e. Move them with predicted class labels from U to L
 - f. [Adapting to features]
 - g. Calculate the current feature vector X
 - h. Select the c most significant and topic adaptive sentiment words for each class
 - i. Update the feature values }
11. End loop
12. Train the multiclass SVM C on the features consists of X and X1 using augmented labeled data L.
13. Return X X1 C [3]
14. After execution, the positive tweets are labelled 1 and negative tweets are labelled 0.

Abbreviations and Acronyms

SVM- Support Vector Machine

IV. RESULTS AND DISCUSSION

4.1 After Extraction

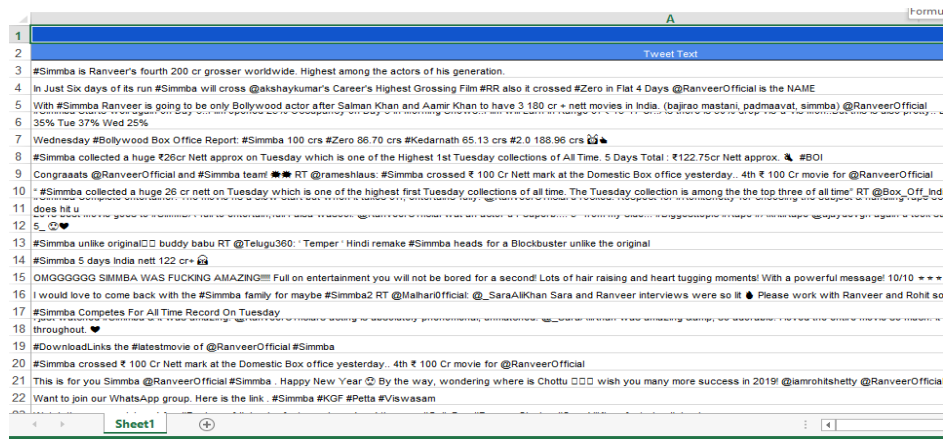


Fig. 4.1: Extracted tweets

4.1 After Pre-processing

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4.2 After sentiment analysis

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Fig. 4.2 Analysed tweets

VII. ACKNOWLEDGMENT

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