

# Review Paper: Sarcasm Detection and Observing User Behavioral

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## ABSTRACT

Sarcasm is a sort of sentiment where public expresses their negative emotions using positive word within the text. It is very tough for humans to acknowledge. In this way we show the interest in sarcasm detection of social media text, particularly in tweets. In this paper we study new method pattern based approach for sarcasm detection, and also used behavioral modelling approach for effective sarcasm detection by analyzing the content of tweets however by conjoint exploiting the activity traits of users derived from their past activities. By using the various classifiers such as Random Forest, Support Vector Machine (SVM), k Nearest Neighbors (k-NN) and Maximum Entropy, we check the accuracy and performance.

## General Terms

Sarcasm Detection, Patter Based Approach, User Behavioral modeling.

## Keywords

Sarcasm, Sentiment, SVM, KNN

## 1. INTRODUCTION

Social net-working websites have become a popular platform for users to express their feelings and opinions on various topics, such as events, or products. Social media channels have become a popular platform to discuss ideas and to interact with people worldwide area. Twitter is also important social media network for people to express their feelings, opinions, and thoughts. Users post more than 340 million tweets and 1.6 billion search queries every day [1] [2].

Twitter is a social media platform where users post their views of everyday life. Many organizations and companies have been interested in these data for the purpose of studying the opinion of people regards the political events, popular products or Movies. When a particular product is launched, people start tweeting, writing reviews, posting comments, etc. on social media such as twitter. People turn to social media network to read the comments, and reviews from other users about a product before they decide whether to purchase or not. If the user review is good for the particular products then the users are buy the product otherwise not. Organizations are also depends on these sites to know the response of users for their products and use the user feedback to improve their products [3]. Sentiment analysis is the opinion of the user for the particular things. Sentiment analysis is the extraction of feeling from any communication (verbal/non verbal). Two ways to express sentiment analysis.

1) Explicit sentiments: Direct expression of the opinion about the subject shows the presence of explicit sentiment.

2) Implicit sentiments: Whenever any sentence implies an opinion then such sentence shows the Presence of implicit sentiment (Indirect expression).

Sentiment analysis and opinion mining depends on emotional words in a text to check its polarity (i.e., whether it deals positively or negatively with its theme) [4]. Sarcasm is a type of sentiment where people express their negative feelings using positive word in the text. The example of this is "I love the pain of breakup". The love is the positive words but it express the negative feeling, such as breakup in this example. It is usually used to transfer implicit information within the message a person transmits. It is hard even for humans to recognize. Used Pattern based approach for detecting sarcasm on twitter. The definition of sarcasm is the activity of saying or writing the opposite of what you mean, or of speaking in a way intended to make someone else feel stupid or show them that you are angry.

## 2. LITERATURE REVIEW

In [3], authors show the interest in sarcasm delectation in the tweeter. For capturing real time tweets they use the Hadoop base framework, and processes that tweets they used the different six algorithms such as parsing based lexicon generation algorithm (PBLGA), tweets contradicting with universal facts (TCUF), interjection word start (IWS), positive sentiment with antonym pair (PSWAP), Tweets contradicting with time-dependent facts (TCTDF), Likes dislikes contradiction (LDC), these algorithm are used identifies sarcastic sentiment effectively. This method is more suitable for real time streaming tweets.

In [4], authors use the computational system it is use for harnesses context incongruity as a basis for sarcasm detection. Sarcasm classifier uses four types of features: lexical, pragmatic, explicit incongruity, and implicit incongruity features. They evaluate system on two text forms: tweets and discussion forum posts. For improvement of performance of tweet uses the rule base algorithm, and to improve the performance for discussion forum posts, uses the novel approach to use elicitor posts for sarcasm detection. This system also introduces error analysis, the system future work (a) role of numbers for sarcasm, and (b) situations with subjective sentiment.

In [5], authors used the machine learning approach to sarcasm detection on Twitter in two languages English and Czech. First work is sarcasm detection on Czech language. They used the two classifier Maximum Entropy (MaxEnt) and Support Vector Machine (SVM) with different combinations of features on both the Czech and English datasets. Also use the different preprocessing technique such as Tokenizing, POS-tagging, No stemming and Removing stop words, its use for finding the issue of Czech language.

In [6], authors have investigated characteristics of sarcasm on Twitter. They are concerned not just with identifying whether tweets are sarcastic or not, but also consider the polarity of the tweets. They also have compiled a number of rules which improve the accuracy of sentiment analysis when sarcasm is known to be present. Resercher have developed a hash tag tokenizes for GATE method so that sentiment and sarcasm found within hash tag can be detected more easily. Hash tag tokenization method is very useful for detection of sarcasm and checks the polarity of the tweet i.e. positive or negative.

In [7], authors are used two methods such as lexical and pragmatic factors that are use for differentiate between sarcasm from positive and negative sentiments expressed in Twitter messages. They also created corpus of sarcastic Twitter messages in which determination of the sarcasm of each message has been made by its author. Corpus is used to compare sarcastic utterances in Twitter to utterances that show positive or negative attitudes without sarcasm.

In [8], authors have developed a sarcasm recognizer to determine sarcasm on Twitter consists of a positive sentiment contrasted with a negative situation of sarcasm in tweets. They use novel bootstrapping algorithm that automatically learns lists of positive sentiment phrases and negative situation phrases from sarcastic tweets. They show that determine contrasting contexts using the phrases learned through bootstrapping.

Rule-based approaches attempt to identify sarcasm through specific evidences. These evidences are captured in terms of rules that rely on indicators of sarcasm. Focus on identifying whether a given simile (of the form ‘\* as a \*’) is intended to be sarcastic. They use Google search in order to determine how likely a simile is. They present a 9-step approach where at each step rule; a simile is validated using the number of search results. Strength of this approach is that they present an error analysis corresponding to multiple rules [9]. The hash tag sentiment is a key indicator of sarcasm. Hash tags are often used by tweet authors to highlight sarcasm, and hence, if the sentiment expressed by a hash tag does not agree with rest of the tweet, the tweet is predicted as sarcastic. They use a hash tag tokenizer to split hashtags made of concatenated words [6].

In [11], we have studied the different methods.

Following are the method for sarcasm detection on twitter.

### 1) Feature extraction

This method are used for annotating the data, it contain three categories.

- Sarcasm as wit: when used as a wit, sarcasm is used with the purpose of being funny.
- Sarcasm as whimper: when used as whimper, sarcasm is employed to show how annoyed or angry the person is.
- Sarcasm as evasion: it refers to the situation when the person wants to avoid giving a clear answer, thus, makes use of sarcasm.

### 2) Sentiment-related Features

It extracts sentimental components of the tweet and counts them. Positive emotional content (e.g. love, happy, etc.) and negative emotional content (e.g. hate, sad, etc.). Calculate the ratio of emotional words.

$$p(t) = \frac{(\& \cdot PW + pw) - (\& \cdot NW + nw)}{(\& \cdot PW + pw) + (\& \cdot NW + nw)} \quad 1$$

t=tweet, pw=positive words, nw =negative words, PW=highly emotional positive words, NW= highly emotional negative words, & =weight bigger than 1.

### 3) Punctuation-Related Features

It displays behavioral aspects such as low tones, Facial gestures or exaggeration. These aspects are translated into a certain use of punctuation or repetition of vowels when the message is written.

- Number of exclamation marks
- Number of question marks
- Number of dots
- Number of all-capital words
- Number of quotes

### 4) Syntactic and Semantic Features

It refers to the situation when the person wants to avoid giving a clear answer, thus, makes use of sarcasm.

- Use of uncommon words
- Number of uncommon words
- Existence of common sarcastic expressions
- Number of interjections
- Number of laughing expressions

### 5) Pattern-Related Features

Pattern is defined as an order sequence of words. Divide words into two classes: a first one referred to as CI containing words of which the content is important and a second one referred to as GFI containing the words of which the grammatical function is more important.

### 6) A behavioural modelling approach

In this method content to study how to develop a systematic approach for effective sarcasm detection by not only analyzing the content of the tweets but by also exploiting the behavioral traits of users derived from their past activities [10].

## 3. PATTERN BASED APPROACH

Sarcasm is a type of sentiment where public expresses their negative emotions using positive word in the text [4]. It is usually used to convey implicit information within the message a person transmits. Sarcasm may be used for different purposes such as criticism. It is very difficult for humans to understand. Recognizing sarcastic statements can be very useful to improve automatic sentiment analysis of data collected from different websites or social networks. Sarcasm is when a person says something different from what he means. Pattern based approach is used for detecting sarcasm on twitter.

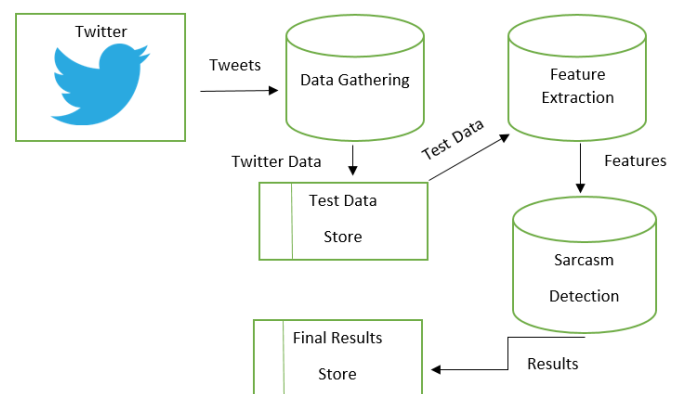


Fig 1: block diagram of sarcasm detection on twitter

#### **4. CONCLUSION**

Sarcasm detection research has grown significantly in the past few years, necessitating a look-back at the overall picture that these individual works have led to. This paper surveys approaches for automatic sarcasm detection. We have studied the different method for sarcasm detection; we also studied the pattern based approach for sarcasm detection. In this paper, the methods are used to detect sarcasm or as well as check the behavioral approach of the user, the method make used different component of the tweet, and also by using of Part-of-Speech tags to extract patterns characterizing the level of sarcasm of tweets. By using #sarcasm collect all the sarcastic tweets. In this way we discuss the different method such as Feature extraction, Sentiment-related Features, Punctuation-Related Features, Syntactic and Semantic Features, Pattern-Related Features, behavioural modelling approach for detection of sarcasm in the tweet. By using different algorithm or classifier such as Random Forest, Support Vector Machine (SVM), k Nearest Neighbors (k-NN) and Maximum Entropy, check the accuracy and performance. In future scope these approaches will show good results.

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