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### Study on Various Sentiment Analysis Techniques

Kartik Chamoli<sup>1</sup>, Nishant Vashist<sup>2</sup>, Harsh Sharma<sup>3</sup>, Dr. Manjot Kaur Bhatia<sup>4</sup>

1, 2, 3, <sup>4</sup> Jagan Institute of Management Studies, Rohini Sector-5

### I. INTRODUCTION

Today the scale with which social media produces data is huge. The huge volume of data generated by single social media user is massive. People nowadays share their ideas, learn from each other and share their opinions online. People use websites like Facebook, twitter, google+ etc to share their opinions with like minded people and to form a community. The data generated can used by data scientists and machine learning engineers to analyse the opinions of the people whether they are positive, negative or neutral. All of the likes, comments, popular tags on various social media websites can be used to know the general sentiment of a public towards a particular topic to gain important insights about human behaviour and thinking.

Knowing the people sentiments is know one of the most important factors for every organization to target them better and to know how they can influence the user's decision. Sentiment of a user basically consists of opinions, sentiments, appraisals, attitudes, and emotions. Analysing all these sentiments is a challenging task in itself. The major part of sentiment analysis is the retrieval of sentiments from the textual information by processing, searching and analysing the vast amount of data present. The social media algorithm works on sentiment analysis 24\*7 to only display those things to the user which will keep them longer on their website.

We will focus majorly on opinion mining techniques using machine learning and lexicon-based approaches along with the evaluation metrics. We will get an overview on the already available techniques like Naïve Bayes, Max entropy and support vector machine and analyse the challenges with them on huge twitter dataset.

### II. LITERATURE STUDY

Paper Title	Problem discussed/Technique	Framework setup	Result	Drawback
Vishal A. Kharde,	Author research shows that	Sentiment analysis		The research lacks
S.S. Sonawane	how different sentiment	methods like SVM,	A set of synonyms	visual
Sentiment Analysis	analysis techniques give	EWGA,	are offered after	representation of the
of Twitter Data: A	different accuracy and	CLMM, active	semantic analysis	different datasets
Survey of	precision on huge twitter	learning and SFA are	that have similarity	used for sentiment
Techniques	dataset.	applied to various	and show the	analysis.
		twitter datasets.	polarization of the	
			social media	
			content.	
Abdullah Alsaeedi ,	Discussion and review of	Ensemble approaches	Biasness of tweets	The research needs
Mohammad Zubair	various sentiment analysis	and supervised machine	when analysed	a high level of
Khan	along with the mathematical	learning	using supervised	mathematical
A Study on	techniques used to represent		and unsupervised	background and
Sentiment Analysis	them.		machine learning.	could not be
Techniques of				understood by a
Twitter Data				layman.
	The paper focused majorly on	Unigram, Bigram, N-	Built a model for	the study needto
Faizan,	the 5 stages of sentiment	gram, POS tagging,	the analysis of	improve on these
Twitter Sentiment	analysis.	Subjective, objective	feeling using KNN	limitations and
Analysis		features	algorithm with	applying the scope
			unigram, bigram	method and should
			and ngram features.	give more
				weightage on
				mental and social
				health problems.



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$Omar\ Y.\ Adwan(oxtimes),$	The paper gave an overview	Sentiment Features	The overview of	The paper doesn't
Marwan Al-Tawil,	on the sentiment classification	(SENF), Syntax Based	twitter sentiment	give us a deep
Ammar M. Huneiti,	phases and their	Features(SYNF),	analysis was given	insight on the
Razan H. Al-Dibsi,	subcategories. The phases are	Semantic Features	by analysing over	different approaches
Rawan A. Shahin,	pre-processing, feature-	(SEMF), Unigram	40 articles and	and how they can be
Abeer A. Abu Zayed	extraction, feature	Based	mentioning the	applied in the real
Twitter Sentiment	selection(filtering),	Features(UGF),N-gram	different	world on practical
Analysis	classification.	Features (NGF), Top	approaches used in	huge datasets.
Approaches: A		words features (TWF).	each one of them.	
Survey				

### III. PROCESSING THE DATASET

A tweet posted by a user contains a variety of opinions about users views expressed in different ways. The dataset used for sentiment analysis is usually labelled into two classes i.e. positive and negative polarity to represent the two extreme of the human emotions. The positive and negative polarity makes it easy to perform sentiment analysis. The raw data although contain a lot of redundancy and unwanted data that can cause problems to our algorithm while calculating the sentiment. We need to do some sort of pre-processing to make out data finer before passing it to an algorithm for processing.

Pre-processing of tweet include following points.

- 1) Remove all URLs (e.g. www.yzy.com), hash tags (e.g. #topic), targets (@username)
- 2) Try to correct the spellings in the tweet especially if there are any repeated characters.
- 3) Try to replace all the emojis and emoticons with their sentiment.
- 4) Remove all the punctuations, symbols and numbers from the tweet as they have no significance
- 5) Remove Stop Words
- 6) Try expanding the acronyms used in the tweets by using an acronym dictionary as the machine learning algorithm may not understand acronyms.
- 7) Remove non English tweets as they can change the correctness and preciseness of the algorithms and can act as bad input.

### IV. FEATURE EXTRACTION

The pre-processed dataset has numerous distinctive properties. In the point birth system, we prize the aspects from the reused dataset. These aspects are used to decipher the positive and negative opposition in a judgment which is useful for determining the opinion of the individualities using models like unigram, bigram (18).

Machine learning is used for representing the major features of texts and documents. These are known as feature vectors and these features .Some exemplifications features that have been reported in literature are

### A. Words And Their frequentness

Unigrams, bigrams and n- gram models with their frequency counts are considered as features. There has been further exploration on using word presence rather than frequentness to better describe this point. Pangetal showed better results by using presence rather of frequentness.

### B. Speech Tags

Speech like adjectives, adverbs and some groups of verbs and nouns are good pointers of subjectivity and sentiment. We can induce syntactic reliance patterns by parsing or reliance trees.

### C. Opinion Words And Expressions

Piecemeal from specific words, some expressions and expressions which convey sentiments can be used as features. Eg: cost someone an arm and leg.

### D. Position Of Terms

The position of a term with in a textbook can affect on how important the term makes difference in overall sentiment of the textbook.

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### E. Negation

Negation is an important but delicate point to interpret. The presence of a negation generally changes the opposition of the opinion.

### V. MODEL EVALUATION

One of the most common and appropriate technique used for evaluation of a classifier is through confusion matrix.

The confusion matrix is given in a generalized form below.

	Predicted class1	Predicted class 2
Actual class 1	True positive(tp)	False
		negative(fn)
Actual class 2	False positive(fp)	True
		negative(tn)

By applying this technique we can derive the generalized evaluation parameters. These parameters include:

1) Accuracy: The classifier accuracy indicates how accurately the classifier can predict the result. It can be calculated using the below formula accurately the classifier has predicted the result. It can be calculated using the formula: accur(a) = tp + tn/tp + tn + fp + fn

2) *Precision:* It represents how often the prediction by the classifier when it indicated true is correct. The formula for precision is represented below.

precision(p) = tp/tp + fp

3) Recall: Recall represent the true positive rate of the classifier.

Formula for recall is:

recall(r) = tp/tp + fp

4) F1 Score: It indicates the weighted average of recall and precision. Formula for recall is:

F1 score =2p. r/p-r

### VI. CONCLUSION

In this paper, we provide a survey and comparative study of existing techniques for opinion mining including machine learning and lexicon-based approaches, together with cross domain and cross-lingual methods and some evaluation metrics. An attempted was made to compare the different techniques and outcomes of algorithms performance. Research results show that machine learning methods, such as SVM and naive Bayes have the highest accuracy and can be regarded as the baseline learning methods, while lexicon-based methods are very effective in some cases, which require few effort in human-labelled document. We also studied the effects of various features on classifier. We can conclude that more the cleaner data, more accurate results can be obtained. We can focus on the study of combining machine learning method into opinion lexicon method in order to improve the accuracy of sentiment classification and adaptive capacity to variety of domains and different languages.

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