A Review on sentiment analysis techniques and their performance

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Abstract: In today's time various people give there reviews about their sentiments after shopping or even various social issues. Using large processing engines these reviews can be interpreted to generates the analysis for the positive and negative polarity. Various researchers has focused on different types of techniques and tools for the analysis purpose. These analysis will be useful for the purpose of decision making for different types of organizations. researches are based on studying the reviews on tourism, shopping, Govt. decisions etc.. With the help of lexical approach tokenize the reviews. Identify the positive negatives and neutral reviews by comparing them with established Ontology. This mined information will provide better view regarding the total reviews. Rather than studying all the individual reviews, it will be much better to have collective analysis by automated tool.

Key Terms: Ontology, Lexical, Reviews.

T. INTRODUCTION

Web produces enormous amount of data. This data grows every day. Large chunk belongs to people reviews for different products they had purchased previously through Ecommerce or even what they feels about the govt decisions. These reviews are regarding their experience of product purchases or opinion about the decision of any individual. These reviews sometimes can be false and influenced. Various people before purchase any thing on E-commerce wants to know the reviews of people. Now problem is it is very difficult to see this enormous data and understands the positivity and negativity of the different reviews. So There is hard requirement of certain automated tool, which can process this large amount of data and extract useful trend from all these reviews, which will be much easy to understand the percentage of positive and Negative reviews. In past decade or so E-Commerce industry on internet has

risen many folds. Online shopping has grown tremendously. Doing so will provides time efficient solutions. There is high chances that people get cost effective products, cost effective foods etc. Through online procedure shop. Later on they give reviews regarding the product and services they get once they experience in real. All the shopping websites also emphasis on people post experience. Because study shows that people purchase is influenced by these reviews. Once certain automated tool will be available will provides better way to search the relevant data and analysis with less time and cost. These automated tool will provides data graphically. A simple number on a rating system is not providing enough information, but neither a long review in which users express opinions about more than shopping features[1]. There are plenty of problems exists in such reviewing system like.

- 1. Long reviews sometimes become sense less.
- 2. Rating based reviews are less understandable.
- 3. Reviews some times are misleading.
- 4. Reviews some times are biased.
- 5. Reviews some times provides duality of views.
- 5. People some times has language problem in both reading and writing.

A system that will analysis the reviews and also extracting the useful analysis from the reviews data. This analysis will be useful for consumers who usually put the shopping based on people Reviews. But will be difficult to read each individual reviews. So a collective analysis will helps in understanding the things and put the general point of view.

I.1 Analysis Module

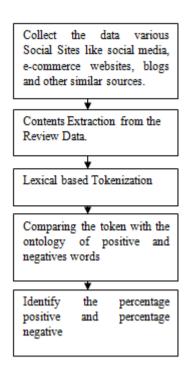


Figure 1. Review analysis Flow [1]

1.2 Methodology

The process of sentiment analysis includes various sequential steps. These steps are like

- data acquisition,
- b. data pre-processing and normalization
- feature extraction and representation, c.
- d. Labeling

1.3 Tools

Which having sentiment analysis various tools are used

- Hadoop(Hive).
- Matlab. 2.

II. **RELATED WORK**

SHAHID SHAYAA(2018) et. al: this paper focused on sentiment analysis for the big data. opinion mining and sentiment analysis for the Big data categories the people sentiment into different classes. Each class represents the people mood for the current aspect under consideration. This paper is focused on technical and non technical aspect of the system for the opinion mining.

Swati Redhu(2018) et. al: This paper focuses on the process of opinion mining and sentiment analysis for the Big data. The data collected is from different sources like social media, e-commerce, blogs, social media. This paper focuses on data acquisition, data pre-processing, Normalization and feature extraction and finally representation. This paper focuses on the different methods and techniques for the sentiment analysis and opinion mining.

DAVID ZIMBRA (2018) et al: This paper focuses on the process of classification for the sentiment analysis for different classes of the twitter data into different classes. The success rate of different classification is around 70%. 28 different commercial and academic system classifies different datasets into different classes. Paper focuses on the different trends and techniques for improving the results.

M.Trupthi(2017)et. al: This paper is focused on sentiment analysis of large number of reviews or tweet data on various social media sites. Large amount of data is being processed by the processing engine. This processing engine can be Hadoop etc. This can process the large amount of data to extract the useful patterns. Sentiment analysis in current paper is based on features.

Prerna Mishra(2016)et. al: This paper focus on sentiment analysis for the tweeter data. Paper has collected a data of the persons opinion on Modi ji's digital india campaign. The data collected is from various social media tools. First they have collected the sentiment of the persons. In second step classifying the data based on polarity of these sentences w.r.t. Positive, Negative, or neutral. From the analysis it is clear that the positive polarity is 50%, negative polarity as 20% and neutral polarity is 30%.

Cristian Bucur(2015) et al: This paper has proposed a robust system for analysis the review data taken from various reviewing sites. These reviews sites are specifically related to tourism. The researcher is of point of view that the analysis and representation in the advantage of both customer and service provider. Because customer can choose best at best price. Service provider can better there service. Because it is always learning to read direct reviews about your services. You can always fine tune the services if certain shortfall exists. But simultaneously he is of point of view that review should not be biased. Because it leads to wrong analysis.

Aliaksei Severyn(2015) et al: According to this researcher large amount of data is produced every day on the internet. This data is regarding reviews of the data. There requires robust model which can have transformation procedure for these reviews data. So that data can be analysis and be converted to well under stand able way. This is the data related to you-tube videos data. These reviews are the people personal views on the reviews

Kamal MAROUCHE(2015) et al: Today's environment is related to competition. Organization compete amongst themselves for attaching more and more customer. Because they have online stores. People do visit there stores repetitively. So companies wants to attach their customers to them by taking there online reviews. They carefully analysis the reviews data and put it in right perspective. Once the robust analysis is complete they put corrective actions to fine tune their services. So that they can attract new customer. The robust tool for analysis purpose will helps in having system more understandable and automatic

Montoyo, Martínez-Barco, & Balahur(2012). Majority of researchers are based on opinion mining for standardized text. This standardized data includes shopping reviews and news data Amazon data, and various type of other data. But the problem is the words and sentences used into these reviews are not standardized reviews. It will be very difficult to process this language containing informal words. But one thing in common is that all the reviews data is in English.

III Methodology

- data acquisition
- data pre-processing
- normalization
- feature extraction
- Representation,
- labeling

IV. **Comparison Table**

Author Name	Name of Paper	Technique	Constraints
Kamal AMAROUCHE et.	Product Opinion Mining	Classification based	The approach can be
al(2015):	for Competitive	technique. Classify on the	further enhanced to
	Intelligence	basis of Machine learning,	increase the percentage of
		Lexicon Approach.	true positive and false
	1341		negatives.
Cristian Bucur(2015)	Using Opinion Mining	It is based on classification	Further there requires to
	Techniques in Tourism	and aspect based approach	improve the performance
			by using ontology.
J. Islam(2016) et. al.	An Effective Approach of	This paper has applied the	This research can be
	Intrinsic and Extrinsic	approach of NLP that is	further enhanced to
	Domain Relevance	natural language	consider two thresholds
	Technique for Feature	processing. It also uses	for better analysis.
	Extraction	Intrinsic and Extrinsic	
	in Opinion Mining	Domain Relevance.	
S. S. Sohail(2015) et. al	A model for enhancement	Any user which provide	Till now large amount of
	in wearable technology	feedback will be useful	information domains are
	based on opinion mining	aspects for many	untouched. Which can be
	technique	processing and analysis	further enhanced and
		techniques.	processed in further
			researches.

S. Ahmed	A Novel Approach for	Used machine learning	Work can be further
	Sentimental Analysis and	algorithms like Naïve	enhanced by selecting the
	Opinion Mining based on	Bayes, SVM and	methods based on
	SentiWordNet using	Multilayer Perception	threshold like info gain
	Web Data	(MLP)	and chi square tests.
Dhanalakshmi V.(2016) et	Opinion mining from	comparative performance	The work can be further
al.	student feedback data	study of the algorithms	enhanced by considering
	using supervised learning	like SVM, Naïve Bayes, K	the more suitable
	algorithms	Nearest Neighbor and	techniques for
		Neural Network classifier.	categorization. These
			analysis tools will be
			based on SVM and Naïve
			Bayes

V. CONCLUSION

Studying different papers it is clear that reviews on different products by customer in different ways influence the purchase of the customers. Different ways of reviews can be star rating, Reviews in sentences, or even yes or no. Different researchers has studied these reviews in different ways. Also they have studied the effect of reviews on the future purchase of other customers. Various researchers has emphasis on processing the reviews data. So that any user rather than seeing the enormous data can see the processed data and make their mind more easily and effectively. Authors has emphasis the point that certain automated tool is required which can process this enormous data and extract the use full data. This useful data will be presented in useful and understandable way. Simultaneous to the positives there are various negatives. These negatives are due to biased reviews and wrongly presented reviews. So large amount of research further can be done, which can helps in removing these constraints and provides further reviews in useful way.

VI. FUTURE WORK

Study of various research techniques has given the procedure of sentiment analysis for given set of reviews given by the people on different subjects. These techniques has different success rate for the identification of true positive, false negative etc.. Further the technique can be enhanced for the sentiment analysis for the review data based on lexical analysis and generating the ontology of large number of the positive and negative sentences. This will enhanced the success rate of the generating the true positive and true negatives.

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