

# A Survey on Sentiment Analysis and Topic Modeling

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#### ABSTRACT

	Sentiment Reason Mining is an emerging research area in this era of social	
Article Info	media. Sentiment Reason Mining aims to resolve two problems: first is finding	
Volume 9, Issue 2	the reason of a sentiment, and second is interpreting sentiment variations. Time	
Page Number : 149-154	and Event where sentiment is being changed is also an important factor.	
	Aspect-Based methods, Supervised Learning, Topic Modeling, and Data	
Publication Issue :	Visualization etc. can be used for finding the reason of a sentiment. VADER	
March-April-2022	Sentiment Classifier can be used for sentiment of tweets. LDA is topic Modeling	
	algorithm. In this research paper we have reviewed some the research work	
Article History	performed for this purpose. We have reviewed various research work which	
Accepted : 20 March 2022	have used social media content as dataset. TF/IDF feature extraction is used in	
Published: 30 March 2022	most of the work. Sentiment Detection tools VADER and Text Blob are also	
	discussed in our work.	
	Keywords : Topic Detection, interpreting sentiment variations, opinion reason	

## mining, Sentiment Analysis, Sentiment Reasoning, Sentiment Spikes, Topic Model, Artificial Intelligence, Machine Learning, LDA, VADER

#### I. INTRODUCTION

Opinions related to products, services or any topic on social media and other platforms are very useful tools for stack holders of the businesses as well as other entities. A huge quantity of reviews and format of reviews are main challenge for analysis from these reviews. Topic modeling from the text is also an important research subdomain for this purpose. To finalize the topic from big amount of short text is a challenging task. Other challenge in this research is unstructured data and multiple languages. Algorithms like LDA[1] can be used for topic modeling in this case. Many researchers have opted for various versions of LDA.

#### II. SENTIMENT ANALYSIS

Sentiment analysis can be defined as a process that automates mining of attitudes, opinions, views and emotions from text, speech, tweets and database sources through Natural Language Processing (NLP)" [1]. Ther are various methodology to measure sentiment analysis; among them most of work using TF/IDF. Mostly it works using classification techniques. It classify the text into multiple class like 'Positive', 'Negative' and 'Neutral'. Following Figure shows the example of Sentiment Analysis. Mostly sentiment

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analysis extracts text from it source eg. Social Media, preprocess like stop word removal etc. tokenize the remaining text and then apply algorithm for classification.

## LDA

The Linear Discriminant Analysis or LDA is a margins, column width, column spacing and other features.



## Figure 1 Example of Sentiment Analysis

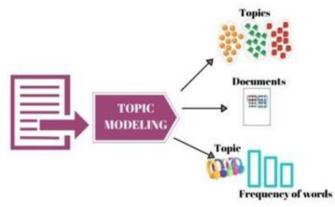


Figure 3. Topic Modeling

Various techniques are used for sentimental analysis. As per research work[1] sentiment analysis can be performed using any of method (supervised, unsupervised or semi supervised) in machine learning

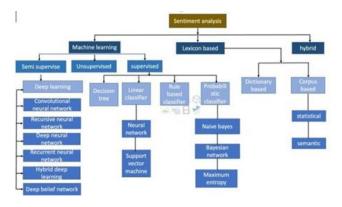
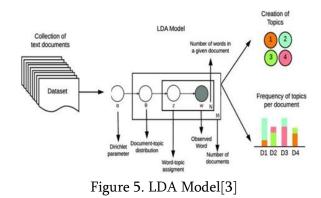


Figure 4. Main Sentimental Analysis Techniques [1]

## TOPIC MODELING

Topic Modeling is a technique in NLP that is used for extraction of topic from various text or documents. In other word we can also say that it is clustering of text with similar content in particular subject. In topic modeling mostly tokenized words are applied. Various preprocessing techniques are also applied like Stop Word removal, non- asci characters are removed etc. your final paper, check that the format conforms to this template. Specifically, check the appearance of the title and author block, the appearance of section headings, document topic modeling algorithm It is dimensionality reduction technique. This method is used as a pre-processing step in Machine Learning and applications of pattern and topic classification. The main goal of LDA is to extract the features in higher dimensional space onto a lower- dimensional space in order resources and dimensional costs. Following example shows relation how topic modeling is done using LDA.





#### VADER

VADER (Valence Aware Dictionary and sEntiment Reasoner) is a lexicon and rule-based sentiment analysis tool. This tool is specifically attuned to sentiments expressed in social media. VADER is used for sentiment analysis of text which has both the polarities i.e. positive/negative. It does not require any training data. VADER can be used in multi domain also

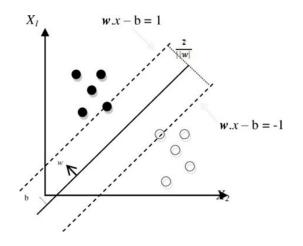
#### TextBLOB

TextBlob is a library for processing text content for sentiment extraction. This method provides a simple API for diving into common natural language processing (NLP) tasks. The NLP process contains partof-speech tagging, noun phrase extraction, sentiment analysis, classification, translation, and more. Polarity is calculated in TextBlob. For sentiment polarity score it uses Tokenization.

#### SVM

SVM is a supervised method for machine learning algorithm. Use of SVM algorithm are in both classification or regression challenges. As we know classification of documents is predicting a label/group/topic and Regression is predicting a continuous value. SVM performs classification by finding the hyper-plane that differentiates the classes.

SVM can work in linear and non-linear challenges. Here x is a feature vector and w.x-b=1 is hyper plane. Flow for SVM algorithm for Sentiment Analysis.



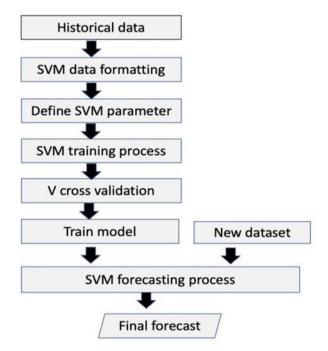


Figure 6. SVM Forecast for Classification Flow

#### **III. LITERATURE STUDY**

In this research paper we have reviewed work related to sentimental analysis and topic modeling. Various researchers have used different techniques and approaches. Each technique has its own pro and cons. We have also discussed strong features and limitations of the same work.

In research work[1] Fuad Alattar and Khalid Shallan have presented a Filtered-LDA (FLDA) method of interpreting sentiment variations on Twitter. Their approach utilizes cascaded LDA Models[1] with multiple tuning of hyper parameters to get candidate reasons those are cause of sentiment changes. After that they applied a filter to remove content that is about old topics. At last they extract Emerging Topics that are interpretable by a human. Finally, a novel Twitter's sentiment reasoning dashboard is introduced to display the most representative tweet for each candidate reason. In their work they have applied VADER[16] for sentiment analysis. For training purpose they have applied Airlines Dataset from Twitter.



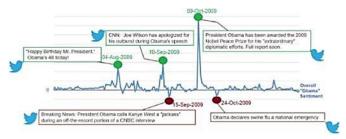


Figure 7. Sentiment Reasons for Obama [1]

As an output they have applied some analysis in the change of sentiments.



Figure 8. Sentiment Change by Dates [1]

In another research work[2] Prasoon Gupta et. al, have applied sentiment analysis for the tweets. The have used tweets by Indian citizens and applied NLP and machine learning classifiers. They have collected tweets from April 5, 2020 to April 17, 2020, a total of 12 741 tweets having the keywords "India lockdown". For sentiment analysis they have applied TextBlob and VADER lexicons, and preprocessed using the natural language tool kit provided by the Python.

In research work[3] by Christopher Whitfield et. al, proposed a custom Named-entity Recognition (NER) system and a Latent Dirichlet Allocation (LDA) method for topic modeling on a Reddit corpus. NER is a sub- task of Information Extraction (IE) which aims to classify certain named-entities found within an unstructured body of a text corpus. Researchers in work[4] also applied LDA for sentiment analysis topic modeling. They have applied the system over twitter dataset. They have extracted topics like Anger, Fear, Joy etc.

In research work[5], they conducted analysis over twitter dataset. They also applied LDA for topic modeling. Researchers proposed an in-depth analysis of the general debate from the Italian Twitter community. The time period selected as the lockdown period.

In research work[6] we have reviewed a work related to LSA (Latent Sementic Analysis). In the work they have applied LSA,PLSA and LDA. They have generated Non- negative Matrix factorization. They have applied TF/IDF feature vectors. In their work only bi-gram vector tested for urdu tweets.

In another work[7], they have analyzed sentiments based on volume of the text, location, frequent words and peoples emotion. They have also applied LDA for topic modeling. In their work they have not tried to get emerging topic.

In research work[8] different approach with Hybrid inverse document frequency with Fuzzy K- Means clustering algorithm is applied.

In research work[9] LDA with SNA (Social Network Analysis) is applied and achieved a good result. Researchers in [10,11,12] have applied LDA for topic modeling. In research work[13], they incorporate a background model that captures spatio-temporal fluctuations of non-event tweets. In research work[14] content based method for personalized tweet and followed recommendation is applied and Knowledge graph model is applied. They have not extracted latent features. In research work[15], LDA was applied for topic modeling to infer the different topics of discussion, and Valence Aware Dictionary and sentiment Reasoned was applied for sentiment analysis.



### IV. COMPARATIVE STUDY

This section will go through the most commonly used methods and algorithms in sentiment analysis and topic modeling.

#### Table 2 Methods Overview

Method/Tools	Pros	Cons
VADER[1,3,1 5]	Better Option for Small Text like Social	Not accurate for complex data
	Media. Easy to Train	Does not recognize
	Less Complexity	context.
TF/IDF[1,2,4]	Similarity calculation is simple.	Position of word is not considered.
	-	No sentiment is measured.
LDA[1,3,4,5,6, 15]	The LDA is a generative model, but in text mining, it introduces a way to attach topical content to text documents. Each document is viewed as a mix of multiple distinct topics.	Latent Dirichlet allocation when run on different datasets, LDA suffers from "order effects" i.e. different topics are generated if the order of training data is shuffled.
K-Means Clustering[8]	K-means is hard clusteringmodel It can be used with large dataset.	Can not provide better result when single context contain multiple topics.

## V. CONCLUSION

We have reviewed some of the research work related to topic modeling and sentiment analysis and found some basic issues related to optimized topic modeling After reviewing the research work we can summarize that LDA with VADER can be a better option particularly when Social Media text is being used for sentiment analysis.

As a part of future work Hybrid models of Sentiment analysis and topic model and transfer learning can be studied and applied.

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