

# Performance Comparison of Feature Selection Algorithm in Opinion Data Mining

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Abstract - To determine the opinion of any person experiencing any services or buying any product the usage of Sentiment Analysis, is a common practice in the field of text mining. It is a process of using computation to identify and categorize opinions expressed in a piece of text. Individuals post their opinion via reviews, tweets, comments or discussions which is our unstructured information. Sentiment analysis gives a general conclusion of audits which benefit clients, individuals or organizations for decision making. The primary point of this paper is to compare different feature selection algorithm to identify the maximum accuracy in different type of dataset. It is an approach to determine two or more methodologies collation. The fundamental commitment of this paper is to experiment whether combined use of feature Selection and existing classification methodologies can yield better accuracy.

Keywords - opinion mining, Feature Selection Algorithm sentiment analysis, opinion extraction

## I. INTRODUCTION

There has been a ton of research work that is being done on supposition mining and notion examination. The main inquiry that emerges as a primary concern is that what is supposition mining and opinion investigation? The second thing that one may think would be what is their centrality? Conclusion is a man's point of view around an issue or a protest.. In this way sentiment mining is the method used to separate clever data in view of a man's conclusion from crude information accessible on web. While conclusion investigation break down the extremity of opinion(positive or negative).

One may think the need of supposition mining and assumption examination as for what reason would one have to think about somebody's sentiment. The appropriate response exists in the expanding utilization of web by individuals for looking about different items, news, and most recent data and so on. Today individuals are likewise setting their remarks and conclusions via web-based networking media with the goal that they can be seen by other individuals as well. Study has demonstrated that such conclusion likewise influence the general population perusing those suppositions [6,7].

So the comments identified with an item or issues are to be broke down by the related association so they can enhance in view of the comments of individuals. Assessment mining and estimation examination are utilized to concentrate such comments and dissect them based on its extremity individually. In this paper we endeavor to get teamed up assessment with the assistance of opinion examination. The proposed idea is clarified

with the assistance of a contextual investigation. The contextual investigation depends on the sentiment about understudies separated from the comments given by the instructors in light of their execution. The comments are handled with nostalgic investigation and sentiments (positive or negative) are uncovered.

# 1. Advantages

Opinion mining and estimation investigation not just finds an application in online comment locales however they likewise can be utilized as sub-segment innovation in suggestion frameworks [1]. Opinion mining clubbed with slant examination can be utilized to deliver less of negative criticism and a greater amount of positive feedback[1,9]. They can likewise be utilized as a part of recognizing unfriendly remarks or web journals over the web by the administration.

Opinion mining and opinion examination additionally assist business expert with analyzing their item comments and after that take viable measures for positive comments. Another utilization of conclusion mining is in legislative issues. It can be utilized for distinguishing individuals' comment about government hopefuls remaining for races. It can likewise be utilized for investigating individuals' comment about an administration policy.

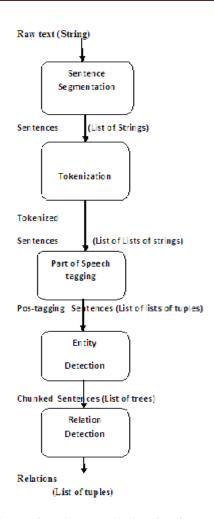


Fig.1 Steps of sentiment analysis using feature selection algorithm.

## II. LITERATURE SURVEY

Many calculations have been proposed keeping in mind the end goal to comprehend and execute opinion mining and assessment investigation. Scientists have created models for distinguishing the extremity of words, sentences and entire record [2]. Various instruments are likewise accessible now for opinion extraction, assessment examination and sentiment rundown. There have been inquires about with respect to advancement for better calculations for such devices.

Pang et al. [1] played out a broad review of in excess of three hundred thesis by covering applications, regular difficulties for feeling examination, significant errands of assessment mining viz., conclusion extraction, assumption arrangement, extremity assurance, and rundown.It analyzed movie reviews by using different machine-learning techniques to determine if it will be as powerful as various other classification techniques to classify

sentiments. By varying input features, the best classification accuracy scores around 77%.

**KuiLiang and Chen [2]** proposed calculation for opinion extraction, supposition rundown and following the assessment which might be utilized for different dialects. The conclusion extraction calculation takes estimation of sentiment holder into thought while in this paper the estimation of supposition holder is taken to be one.

**O'Leary et al.[13]** displayed an overview on blog mining, which incorporates presentation on blog inquiry and mining, sort of sites to be broke down, unit and kind of sentiments to be removed from online journals, and their applications.

Montoya et al.[8] recorded some open issues alongside accomplishments got up to this point in the territory of subjectivity examination and estimation investigation. Tang et al. [14] examined four issues identified with conclusion mining, i.e., subjectivity characterization, word notion grouping, archive assumption order and supposition extraction. For subjectivity arrangement, they featured some methodologies like similitude subordinate, NB classifier, Numerous NB classifier, and cutbased classifier

**Feldman** [15] concentrated on five particular issues in the field of SA: Report level SA, sentence-level SA, viewpoint based SA, relative SA and, feeling dictionary securing. They likewise recorded some open issues like SA of synthesis explanation, programmed element acknowledgment, dialog on multi-substance in same survey, mockery recognition and subjectivity arrangement at better level.

**L. Qiu, H. Rui** [10] proposed report level investigation which has advantage of better level of characterization. He utilized a basic model with various parameters for report level examination. He utilized Viterbi's calculation for understanding the models with various parameters or contentions.

Wilson's [12] approach was to recognize relevant extremity for a huge subset of slant articulations. This approach was a stage level supposition examination. Hu's and Dave's exploration centered around separating assessment from comments. Hu's [8] examine was an item highlight based research. It went for removing item includes and gave item based synopsis.

**Kim and Hovy** [3] in their first model chose a point and broke down slant of comments utilizing word assessment classifier with word net. The second model utilized likelihood of assessment words.



# III. PROPOSED ALGORITHM

This paper proposes an algorithm based on the Feature Selection Algorithm. In the existing work polarity of remarks word by word in a sentence was not considered. The proposed work has been explained with help of a Feature Selection Algorithm. Many case has been considered wherein a set of data taken. The algorithm generates a numeric value for the opinion. If the opinion value are high the opinion are considered positive. Lower opinion value represents negative remarks.

The algorithm analyses the remarks word by word [2]. Sentiment words are identified and a combined value is given to each sentence. A database is maintained to identify the sentiment words. The database along with the sentiment word saves an associated value for the opinion word. The value assigned to each sentiment word is based on how much strong, or weak sentiment is being used. The value ranges from zero to ten. If a sentiment word emotes strongly positive opinion higher is its value in the database. A sentiment word that represents strong negative opinion lower is its value in the database.

When a sentence is analyzed, for each sentiment word found in the sentence, its opinion value is fetched from the database. Then the collaborated opinion value of that sentence is estimated. If there is negation in a sentence the value of opinion score is decreased/increased by a certain amount [2]. These are some different feature selection method with different dataset. The following is the algorithm that is being used. Select a feature selection method from the list:

- ILFS(Indoor Location Finding System)
- InfFS(Infinite feature Selection)
- ECFS(Eigenvector Centrality for Feature Selection)
- Mrmr(Minimum Redundancy Maximum Relevance)
- FSV(Feature Selection ConcaVe)
- MCFS(Multi Cluster Feature Selection)
- RFE (Recursive Feature Elimination)
- Fisher score Feature Selection.

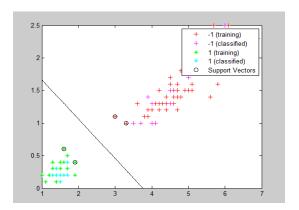
### 1. Algorithm-

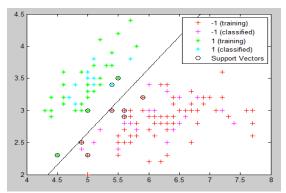
- Step-1 Feature Selection Methods in Opinion Data Mining
- Step-2 Select a feature selection method from the list
- Step-3 Load the data and select features for classification
- Step-4 Extract the sots a class
- Step-5 Randomly partitions observations into a training set and a test and set using stratified holdout
- Step-6 Number of features.
- Step-7 Infinite Latent Feature Selection
- Step-8 MCFS: Unsupervised Feature Selection for Multi-Cluster Data
- Step-9 For unsupervised feature selection, you should tune this parameter k, the default k is 5.
- Step10 Features Selection via Eigen vector Centrality

- Step-11 Regularized Discriminative Feature Selection for Unsupervised Learning
- Step-12 BASELINE Sort features according to pair wise correlations
- Step-13 Feature Selection and Kernel Learning Based Clustering
- Step-14 Use a linear support vector machine classifier
- Step-15 Selection method.

## IV. RESULT AND SIMULATION

The following are some of the cases considered for calculating and representing opinion value by Feature Selection Algorithm.





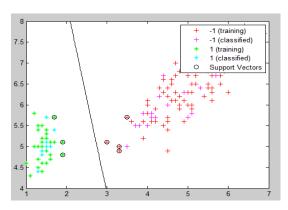


Fig.2 (a) Based on [2],(b) Based on [10], (c) Based on [14]

Table1 Result comparison by using different feature selections algorithm.

Method	X-Train	Y-Train	Accuracy	Error-
	Size	size	,	Rate
ILFS	120 4	120 1	99.28%	0.72
	30 4	30 1		
Inf-FS	120 4	120 1	98.46%	1.54
	30 4	30 1		
ECFS	120 4	120 1	98.26%	1.74
	30 4	30 1		
Mrmr	120 4	120 1	99.05%	0.95
	30 4	30 1		
FSV	120 4	120 1	99.2%	0.8
	30 4	30 1		
Mcfs	120 4	120 1	98.3%	1.7
	30 4	30 1		
RFE	120 4	120 1	98.98%	1.02
	30 4	30 1		
Fisher	120 4	120 1	98.59%	1.41
	30 4	30 1		

Distinctive relevant investigations are considered wherein give remark around a data and an ordinary feeling regard is figured. The count differentiates each word and feeling and nullification in the database. The count is realized in view of score designated to each thought word in the database. They cooperated supposition is evaluated by calming remarks word by word and after that executing the computation proposed. The surveyed supposition regard for any data can be utilized while giving assurance. Proposal may be given to a data according to the collaborated estimation regard. Among the current element determination calculations.

A few calculations includes just in the choice of applicable highlights without thinking about repetition. Dimensionality increments pointlessly on account of excess highlights and it likewise influences the learning execution. What's more, a few calculations select important highlights without thinking about the nearness of loud data. Nearness of uproarious data prompts poor learning execution and builds the computational time. Our examination reasons that there is requirements for a compelling bound together system for determination which ought to include in the choice of best element subset with no excess and uproarious data. It ought to be connected for a wide range of data and it ought to likewise ready to scale up with expanding dimensionality

#### V. CONCLUSION

The algorithm for calculating collaborated opinion value based on Feature Selection Algorithm. Different contextual analyses are considered wherein give comment about an information and a normal opinion esteem is figured. The calculation contrasts each word and opinion and refutation in the database. The calculation is actualized based on score appointed to every notion word in the database. They worked together opinion is assessed by relieving comments word by word and after that executing the calculation proposed. The assessed supposition esteem for any information can be used while giving determination. Suggestion might be given to an information as per the teamed up sentiment esteem.

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