



The nearness of the word doesn't before like delivers a negative score instead of a positive one, giving a general notion score of - 2.

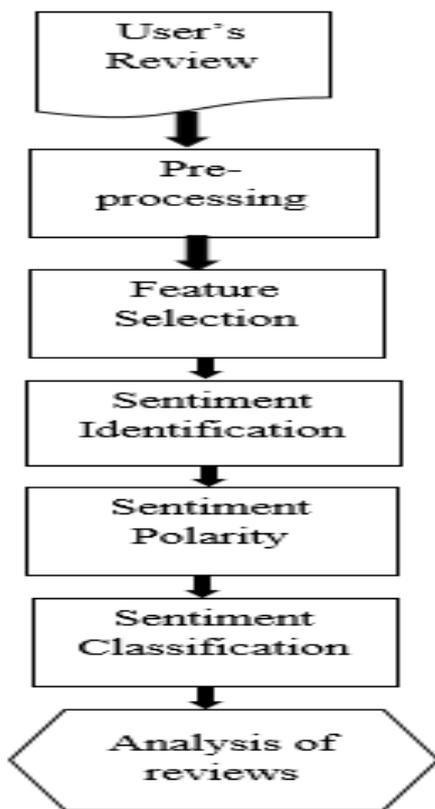


Figure 1: Sentiment Analysis process

## There are various levels of sentiment analysis

Has been discussed are as follows:

**A. Aspect level or Feature level:** Aspect level is the opinion mining and feature-based summary. The classification concerns the identification and extraction of product characteristics from source data. This type is used when we need sentiments about the desired appearance.

- Task1: identify and extract the features of objects that were commented on by a commentator (for example, a reviewer)
- Task2: determine if the views of functions are negative, positive or neutral.

**B. Document level:** Sentimental analysis at the document level classifies the entire opinion of the document based on different feelings about the product or service. This level classifies the opinion document as positive, negative or neutral.

- Task: Opinion Mining of whole document
- Classes: Positive, Negative and Neutral
- Assumptions: Each document emphasizes on a single entity and contains opinion from a single opinion holder.

**C. Sentence or phrase level:** The sentence level Opinion Mining identify whether each sentence having a positive, negative or neutral opinion for a brand or service. This type of sentence is used for reviews and comments that contain one sentence and written by the user. Various activities related to this:

- Task1: Identifying subjective/objective sentences  
Classes: Objective and Subjective
- Task2: Opinion Mining of Sentences.  
Classes: Positive or Negative

*Assumption:* A sentence holds only one opinion which may not always be true

**D. Word Level:** In the most recent works, the previous polarity of words and phrases was used to classify sentiment at the level of sentences and documents. Classification of words by sentiment mainly uses adjectives as characteristics, but adverbs. Two methods for automatically annotating sentiment at the word level:

- Dictionary-Based Approaches.
- Corpus-Based Approaches.

## II. RELATED WORK

Shidaganti, *et al.* [2] tweeter data was analyzed to collect the opinions of the users. Tweeter is mostly used for voicing their views bysmall note in reply to many types of merchandises, brands and celebrities along with political criticisms. Clustering and TF-IDF process was used.

Mumtaz, *et al.* [3] proposed machine learning based approach, which gave more accuracy than the existing lexical method.

Onam Bharti, *et al* [4] given a prologue to this interesting issue and to display a system which will perform slant examination on online cell phone audits by partner changed K implies calculation with Naïve Bayes characterization and KNN. We got a general characterization precision of 91% on the test set of 500 portable audits. The running time of the calculation was  $O(n + V \log V)$  for preparing and  $O(n)$  for testing, where  $n$  is the quantity of words in the records (direct) and  $V$  the size of the diminished vocabulary. It is a lot quicker than other AI calculations like Naïve Bayes characterization or Support Vector Machines which set aside a long effort to meet to the ideal arrangement of loads. The precision was similar to that of the present cutting edge calculations utilized for supposition arrangement on portable surveys.

**Bo, Pang, et.al** [5] considered the issue of collection reports not by topic, anyway by as a rule estimation, e.g., choosing if a review is sure or negative. Using film studies as data, we find that standard AI procedures convincingly beat human-conveyed baselines. Regardless, the three AI procedures we used Naive Bayes, most outrageous entropy gathering, and support vector machines) don't execute too on assessment portrayal as on standard subject based characterization. The experts shut by taking a assumption that parts that make the presumption gathering issue all the all the more testing.

**Ellen Spertus, et.al** [6] portrayed a few ways to deal with fire acknowledgment, including a model framework, Smokey. Smokey assembles a 47-elementfeature vector dependent on the linguistic structure and semantics of each sentence, joining the vectors for the sentences inside each message. A preparation set of 720 messages was utilized by Quinlan's C4.5 choice tree generator to decide highlight based guidelines that had the option to accurately order 64% of the blazes and 98% of the non-flares in a different test set of 460 messages. Extra strategies for more noteworthy exactness and client customization are likewise examined.

Agarwal, *et al.*[7] did work on semi-structured, structured and unstructured data. Domain-based clustering using cosine and Jaccard similarity index was used. Cosine index gave fast results due to steady form in comparison. Jaccard index was used in composite form of scheme to find the similarity.

Zahrotun, *et al.*[8] examined several clustering techniques. Clustering used for classification of data which usually fits to same class. Cosine, Jaccard similarity and their combination was used for value-based similarity.

Basari, *et al.* [9] discussed the concept of opinions-based mining which referred to the application of computational linguistics, natural language processing (NLP) and mining of the text in order to classify the good or bad movie on the basis of message opinion. SVM mainly presents a supervised method of learning which helps in analyzing the data and recognizes the patterns used for the purpose of classification. Abdul-Mageed, *et al.*[10] presented work on standardized version of Arabic data for the case of sentiment analysis. Here, a set of data was collected and further it performs an automatic classification of step where the process of tokenization was performed over the data. The process of two-stage classification was performed. The results obtained shows that the used approach works in an efficient and effective manner. A. M. Popescu, *et al.*[11] presented a method of unsupervised extraction of information that was used in extracting the review-based opinions. This kind of work was done using the following steps. In the first case, the product-based features were identified and secondly, the opinions associated with the product were identified and in the third case, the opinion's polarity was identified. Finally, the method proposed was rank and the opinions were based on analysis of strength. The semantically built orientation was obtained using an approach named relaxation-labelling approach. The results based on recall and precision method of the proposed approach effectively shows the effectiveness in identification of sentiments of an individual. X. Song, *et al.*[12] analysed the social-media based usage of the platform through the process of micro blogging and it extracts the informational data from them. The experts have analysed that the users of the system mainly use the platform of social media for updating their routine-based tasks and to know what is used by them on daily basis. Such type of analysis was based on distinct type of micro blogs platforms in distinct kind of geographical area based on time.

[15] proposed a deep learning based approach for Medical Diagnosis. In this paper Function Approximation (FA) is modified by Convolutional Neural Network (CNN). [16] proposed A Novel Weighted Class Based Clustering for Medical Diagnostic Interface. [17] proposed Gaussian Kernel Approximation for Medical Diagnostic Interface.

### III. MACHINE LEARNING TECHNIQUES FOR TWITTER SENTIMENT ANALYSIS

#### A. K-Nearest Neighbors (KNN)[18]

K-Nearest Neighbor Technique (KNN) is used for classification. This classification of data is performed on basis of the majority voting of the Neighbors [18].

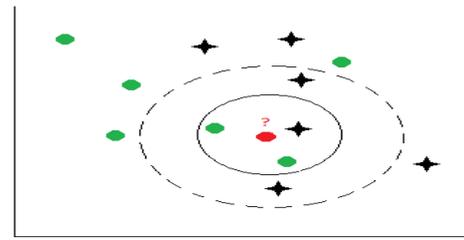


Fig. 2: KNN Used for Classification

Process of KNN explained in Fig2 where a red input point wants to be classified. Then according to KNN, if value of K is 1 then this point will be allocated to class of star which is closest one. If the value of k is 3 then the class of data point will be green circle.

#### B. Decision Tree

Decision tree algorithm is used for partitioning the training set into smaller subsets. These subsets further partitioned recursively as the tree is being built [18]. Decision tree algorithm uses information gain as its trait for partition. This selection method is based on Claude Shannon information theory. Suppose N represents the records of partition. The element having maximum information gain in this partition is chosen as the splitting attribute. This element reduces the information needed to categorize the records in the partition. This attribute gives the minimum randomness in the partition. This process of attribute selection and randomness minimization is repeated recursively. The expected information needed to classify a record D is given by:

$$Info(D) = -\sum p_i \log_2(p_i) \text{ Where } i = 1, 2, \dots, n$$

Where  $p_i$  is the probability with which record belongs to D with class  $C_i$ . Info (D) is the average amount of information required to recognize the class label of a record in D.

#### C. Artificial Neural Network

ANNs architecture is also called multi-layer perceptron (MLP) with back-propagation. Fig 3 [18] shows MLP feed forward Neural Network where Input layer used for providing the input. Input layer further provides data to the next layer i.e hidden layer. Hidden layer accepts and process the data received from the first layer (Input layer). There may many hidden layers and many neurons in each hidden layer for resolving the particular task. From hidden layer, records is given to the output layer. After this, data obtained from output layer and target values are matched. If there is a difference between these two, then the weights are attuned at each processing element. This process is reiterated until target and output values are matched or error reduces to the required limit [18].

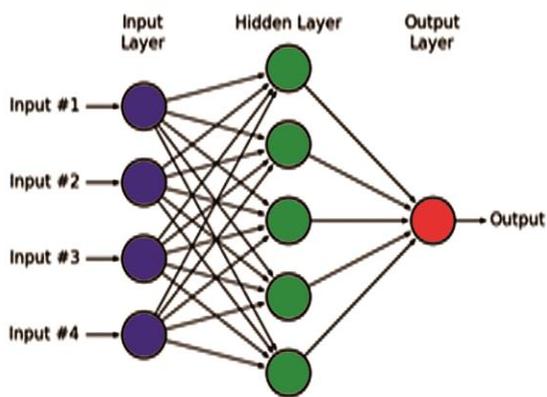


Fig. 3: Multi-Layer Perceptron

**D. Support vector machine**

Support Vector Machine (SVM) is a technique used for classification of data. A mapping technique is used to convert the data into a higher dimensions[18]. N-dimensional hyper plane used to separate the data into two categories by SVM. The optimal hyper plane separates the data in such a way that data with one class of the target variable is on one side of the plane and data with the other category is on the other side of the plane. The data points close to the hyper plane are called support vectors. The Fig 4 [18] gives the SVM process.

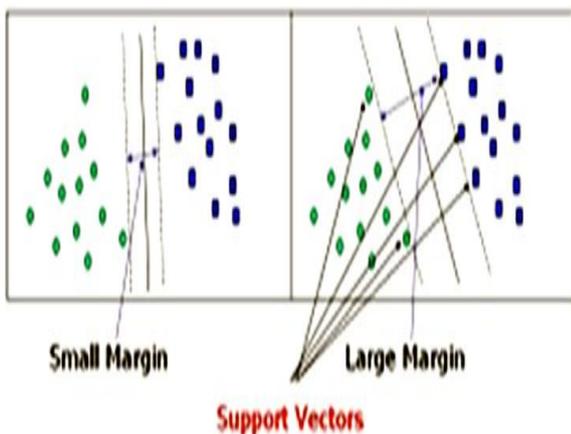


Fig. 4: SVM Topology

**E. Naïve Bayes method**

Naïve Bayes is used for classification. It shows the probabilistic relationship among a set of random variables and their conditional dependences. It also gives joint probability of the distribution. Fig 5 network illustrates the relationships among the random variable (season) of the given year (X1). If it is raining then its value is X2, if the sprinkler then its value is X3, if pavement is wet then its value is X4, and if the pavement is slippery its value is X5. There is no direct link between X1 and X5. It means that there is no direct influence of season on slipperiness. The influence is intervened by the wetness of the pavement.

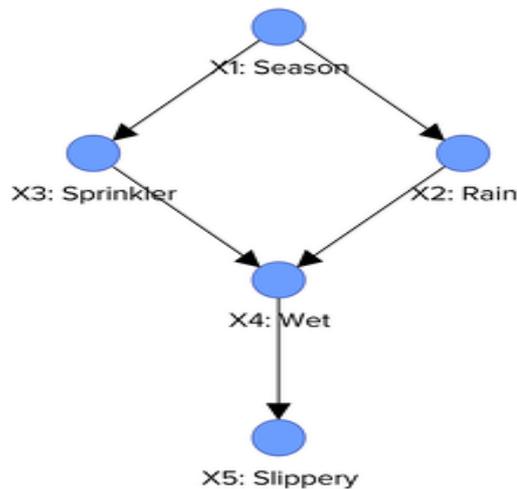


Fig. 5: Naïve bayes method

**IV. RESULTS AND ANALYSIS**

**A. Twitter Datasets**

The word "micro" in microblogging. This defines the limitation of the content of the view expressed on it. A Twitter user can make up their message up to 140 characters on a single tweet. Tweets is a combination of text data and metadata. It is not a simple text message. These attributes are the characteristics of tweets and used to express tweet content. The metadata can be used to discover the tweet domain. Tweet metadata is some entities and places. These entities contain mention of users, hashtags, URLs and multimedia users, user ID on Twitter. RT means retweets, "@" trailed by a user ID, signals the user, and "#" trailed by a word, shows a hashtag.

**B. Pre-processing of the datasets**

This step executes the mechanism of text preprocessing. Text is divided into tokens and then it removes the stopwords like ., ?, # etc. and later on it utilizes stemming process for the reduction of noise and every word is defined as a root word.

**C. Performance Measures and Results**

Confusion matrix is used to explain classification results. In this paper confusion matrix is shown below in Table 1. The upper left corner shows the quantity of data points classified as sensitive as they called True Positive ( TP), and the lower right cell shows the quantity of data points classified as non-sensitive as they called True Negative (TN). The remaining columns show the quantity of misclassified data points.

Table 1: Confusion Matrix

	Classified as sensitive	Classified as not sensitive
Actually sensitive	TP	FN
Actual not sensitive	FP	TN

Below formulae were used to calculate accuracy, precision and recall [18]:

$$\text{Sensitivity} = TP / (TP + FN)$$

$$\text{Specificity} = TN / (TN + FP)$$

$$\text{Accuracy} = (TP + TN) / (TP + FP + TN + FN)$$

$$\text{Precision} = TP / (TP + FP)$$

$$\text{Recall} = TP / (TP + FN)$$

A measure that is used to combine precision and recall is called F Measure and it is the Harmonic mean of precision and recall.

$$\text{F Measure} = 2(\text{Precision} * \text{Recall}) / (\text{Precision} + \text{Recall})$$

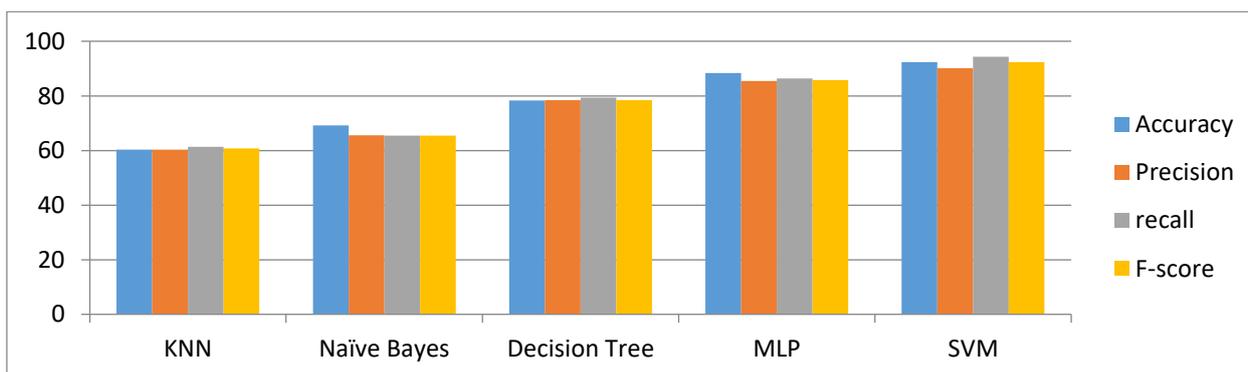
In Table 2 five machine learning techniques (K Nearest Neighbors (KNN), Decision Tree, Artificial Neural Networks (ANNs), Naïve Bayes and Support Vector Machine (SVM)) are evaluated on basis of Accuracy, Precision, Recall and F-score. As results shown in Fig 6, Accuracy of SVM is 4% to 32% higher than others. Consequently models have good predictive capabilities then the others. SVM has attained precision value 94.34% which is 5% to 30% higher among all

techniques. SVM has attained recall value 92.34% which is 8% to 33% higher among all techniques which implies there is a greater degree of agreement between actual and predicted class and model is stable.

F Measure values of Decision Tree, MLP and SVM are 78.45%, 85.78% and 92.34% respectively. F Measure for SVM nearly equal to 100%. It shows that there is excellent relation between actual and predicted class values.

**Table.2 Classification methods based on Accuracy, Precision, Recall, and F-Measure**

Classification Method	Accuracy %	Precision %	Recall %	F-score %
KNN	60.34	60.25	61.34	60.75
Naïve Bayes	69.16	65.56	65.45	65.45
Decision Tree	78.34	78.45	79.34	78.45
MLP	88.34	85.45	86.34	85.78
SVM	92.34	90.12	94.34	92.34



**Fig. 6: Comparison of various performance measures**

### V. CONCLUSION

In this paper five machine learning techniques, K-Nearest Neighbors (KNN), Decision Tree, Artificial neural networks (ANNs), Naïve bayes and Support Vector Machine (SVM) are used for Opinion Mining. These Techniques are analyzed on Twitter dataset. Performances of these methods are compared through various performance metrics such as accuracy, precision, recall and F measure. Results show that SVM outperforms others. Consequently models have good predictive capabilities then the others. All the metrics Accuracy, Precision, Recall and F-score are high in case of SVM. Our Analysis shows that SVM

technique comes out to be most outstanding classifier for Opinion Mining.

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## AUTHORS PROFILE



**Dr. Sanjeev Kumar**, did his M.Tech and Ph.D in Comp. Sc. & Engg from Guru Jambheshwar University of Science and Technology Hisar(Haryana). He is working as Associate Professor in CSE deptt. GJUS&T, Hisar. He had more than 50 papers in his account. His area of interest are machine learning, computer networks and cloud computing.



**Nirmal Godara**, is doing her Ph.D D in Comp. Sc. & Engg from Guru Jambheshwar University of Science and Technology Hisar(Haryana). Her area of interest are machine learning and cloud computing.