

# A Model based on Effective and Intelligent Sentiment Mining: A Review

Rajni Bhalla\* and Prateek Jain

School of Computer Application, Lovely Professional University, Phagwara - 144411, Punjab, India; rajni.b27@gmail.com, prateekjain2010@gmail.com

## Abstract

**Objectives:** Due to proliferation of internet spammers post fake audit to embrace or minimization items. Objective is to purpose a model that can extract spam reviews and implicit reviews. **Methods/Statistical Analysis:** Most research concentrated on extracting just explicit said highlights. Extraction of certain angle like implicit and spam gives more proficient result even in rating too. Pattern discovery method are puposed to known different behaviors to discover spam review. Detection metrics could be used to score every survey. **Findings:** Because of absence of dialect builds in the sentence implicit verifiable viewpoint extraction a mind boggling issue. Most research concentrated on extracting just explicit said highlights. The major weakness of the methods are lack of gold-standard dataset, unable to achieve better accuracy. The framework builds time arrangement of number of surveys for every brand and recognizes spam audits from genuine assessments subsequent to distinguishing suspicious intervals. **Novelty/Improvements:** Before obtaining anything, we need to know conclusion of others. By headway of social websites, opinion settles on potential choice for customer. Even manufacturers can enhance the nature of their item. This proposed model also has the capacity to cover dominant part of the elements which are the deciding factors for the effectiveness of aspect mining framework.

**Keywords:** Implicit Review, Spam Review, Sentiment Orientations

## 1. Introduction to ITARS

With the progression in technology, several individuals are giving survey online about each product. Even makers are constantly inspired by knowing individuals opinion, so they can enhance their products. Additionally, to build benefit also, marketers are quick to known individuals feeling. To collect numbers of reviews are not difficult. Numbers of reviews are available online in social, blogs, newsgroups, forums and product reviews etc. So, we need a method or technique that collect review that are noise free. There is a developing need of viably performing aspect based sentiment mining with respect to social network data.

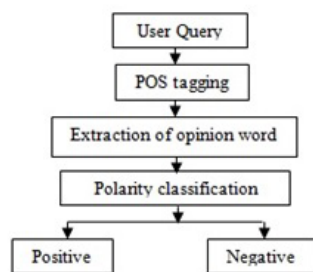
The fundamental motivation behind the sentiment mining is to recognize what people think around a specific item, administration, strategy or politics<sup>1</sup>. Before acquiring anything individuals need to know assessment of others about elements of product or sample: "This mobile phone has a decent battery however camera

quality is extremely poor". "Battery" is feature and "great" is feeling i.e. opinion which implies positive opinion. On the contrary, feature is "camera quality" and opinion is "extremely poor". Hence, aspect based sentiment mining is a rising science in the method for conclusion investigation and sentiment mining. A few investigates have as of now been done with respect to perspective based feeling mining; be that as it may, there is still need to enhance such models to make them mature. Aspect based sentiment mining is being utilized for the most part as a part of business knowledge. Ultimately; area and language adjustment implies that aspect based opinion mining calculation can be transferable to various spaces and languages.

Firstly, implicit features are those features which are unspoken in sentences. For example: "This phone is expensive". Here implicit feature is price, Secondly interrogative sentences in which we use positive or negative word, but that does not contain opinion. For example: "Is the photo nature of this camera is great".

\* Author for correspondence

Consequently client is keen on knowing photo nature, Thirdly, multi-aspects are those features in which opinion for one feature is positive but for other feature is negative. For instance: “The administration was great, yet nourishment was bad”. Fourthly factor is explicit opinion that are explicitly mention in sentences. For example: “This laptop battery is good”. Battery is feature that explicitly mention in a sentence. Fifth one is sarcastic sentences that contains opposite meaning what actually said. Actually remarks are made to hurt. For example: My ears are actually bleeding as a result of listening to your music. Ultimately; area and language adjustment implies that aspect based opinion mining calculation can be transferable to various spaces and languages. Sixth one is spam review that fraudsters activity to mislead potential customer and also to increase the profit of organization by giving fake review and preventing sentiment mining techniques from reaching accurate conclusion. There are diverse categorization levels of sentiment analysis like report level, sentence level and aspect level.



**Figure 1.** General sentiment classification flow.

Figure 1 shows the different levels of classification and also subtask of facet level sentiment analysis. From literature review, it would be clear that till now no aspect based opinion mining techniques fulfilled all factors.

The expansive sharing and use of client evaluation has raised a spam strikes subject on locales. Anyone can easily produce review and can freely and easily post on web without any constraints. It's also having positive as well as negative impact also. For example: a camera producer can hire number of persons to put fraudulent positive reviews, to increase profit of my product. Similarly the producer can hire number of persons to put a fraudulent negative review about competitor product. These types of reviews are given by those customers who have not been experience with that product. So this type of review will be called fake review or spam review. If spammers

take help of other to complete their objective that type of review will be called group spammers<sup>5</sup>.

In this paper, we have analyzed distinctive procedures and techniques for aspect extraction. Extraction from a large number of texts is challenging task in opinion mining, however then again we have talked about various methodologies proposed for the distinguishing proof of implicit aspects and spam detection. In this paper, we survey various algorithms and methods used for spam detection and implicit aspects.

### 1.1 Facet Extraction

This section deals with all the techniques that used for aspect based opinion mining along with strength and weakness of those techniques by analyzing. Exploring the impact of considering the assumptions score removed from client audits to rate restaurants by using rational choice theory<sup>6</sup>. The model introduced in<sup>3</sup> was utilized as a part of online gatherings, discussion groups and online journals. At the point when a client makes an online buy he/she as a rule checks alternate client's notions and feeling about that specific item/thing. The test with respect to this is the accessibility of the enormous information and clients frequently express negating and confounding sentiment around a specific item or service, moreover tradition question/answer system only answer factual questions for instance “In which day we celebrate republic day”. The present feeling QA frameworks have numerous issues one of them is that they just answer the inquiry in the event that they find significant substance about the inquiries in the archive. For example: they neglect to answer dominant part question, for example, “which is the best advanced camera?” the comparative inquiries, for example, “Is mobile X is superior to anything Y?”, the comprehensive inquiries, for example, “Is Canon X great?

To enhance the shortcoming of current technique, proposed model in<sup>3</sup> utilized five stages Question examination, Question development, astounding survey recovery, subject sentence extraction and answer gathering. The hugeness of this paper is that it conquers the shortcomings of current supposition QA framework by noting comparative, majority, and comprehensive questions. The fundamental drawback of this paper is that they have utilized little dataset: 2500 surveys.

Semi-supervised techniques have been used to extracts facets. Not only facets have been extracted but also cluster synonym word into one category. Two factual

models have been utilized which are: SAS model and ME-SAS model, SAS recognizes facets and gathering together related terms into one classification, whilst ME-SAS improves the productivity of SAS model to separate aspect and sentiment terms. They have taken inn survey dataset from [www.tripadvisor.com](http://www.tripadvisor.com) which comprises of 101,234 audits, which is fundamentally larger than different datasets. The principle impediment of this model is its inefficiency in distinguishing facets and after that gathering the related terms into a solitary classification, since, it requires client's assistance for starting seeds<sup>5</sup>.

In this paper<sup>2</sup> main aim of the author is to extract implicit aspects using several support vector machine classifier in Chinese product review. Moreover, they build explicit topic model by using LDA method. Since the primary target of this model is to separate implicit element on premise of explicit elements in the sentence, there is no system that distinguishes multi-angle features, comparative sentences and thorough sentences.

In this paper creator utilizes affiliation standard mining to discover verifiable angles as well. The significant distinction with others techniques that it produces its tenets from its co-occurrence framework of bipartite of assessment words and express view points<sup>10</sup>.

In this paper creator purpose to assess the extraction execution regarding precision, recall and F1 score on a fairly little dataset<sup>12</sup>.

In this paper creator purpose the model plays out the location with discovering single and multi-word angles, sifting by A-score metric and pruning. Experimental results show significant changes of the proposed model over ordinary strategies<sup>17</sup>.

In this paper creator reason an unsupervised model for recognizing angles in audit in light of heuristic guidelines and bootstrapping calculation<sup>18</sup>.

In this paper, this procedure positions the product agreeing to the client audits. After that, this strategy has been sent in search engine<sup>8</sup>. High adjective count and opinion ranking algorithm has been used for opinion ranking algorithm.

The primary shortcoming of this methodology is that the dataset has been utilized as a part of this paper is little in size (40 reviewers). This algorithm did not tackle sarcastic and multi-aspect review. Other weakness of this approach is that the review that are not explained properly like short reviews are not explained.

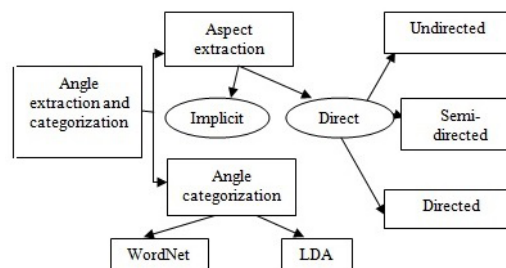


Figure 2. Facet mining approach.

## 1.2 Spam Review Discovery

Study of spam review must mandatory in both social and business point of review. Because it greatly effects the decision of potential customer, discovery of spam review happened for the first time<sup>26</sup>. In supervised learning techniques have been used to detect spam reviews. Behaviors that use to detect spam review was duplicated and nearby duplicate reviews are treated as spam or fake reviews<sup>27</sup>. Similarly in several behaviors are identified in review spammer to detect fake review. These behaviors support for discovering spammer reviewers<sup>28</sup>.

## 2. Methods

### 2.1 Methods for Implicit Aspects

More than 50 systems were abbreviated for the extraction of unequivocal perspectives. In any case, for unspoken aspect we have establish only 11 analyzes which listening carefully on the extraction of comprehended perspectives, at the same time as a couple thinks about focused on both unspoken and unequivocal points. Due to the limitless number of investigation papers for express points, we have disconnected the methodologies into three essential groupings i.e. undirected, semi-directed and directed, as illustrated in Figure 2.

The angle extraction and arrangement is partitioned into two classifications, accumulation of all features and after that order same sort of aspects into group. Features are further isolated into two sorts i.e. explicit facets and unspoken facets.

Dictionary (Word Net) or corpus based approaches used by some of the researchers for facet extraction<sup>13</sup>.

Regardless of of dictionary based approaches some of researcher used LDA (Latent Dirichlet Allocation) for group similar group of facets into one<sup>14</sup>.

In the following areas, we will quickly clarify diverse procedures utilized for the explicit and implicit facets. At that point the commitment of this study, an examination has been directed among diverse feature extraction procedures, explained the effectiveness in various areas and languages.

## 2.1 Method for Spam Detection

Spam detection also known as review spam, fraudulent review, opinion spam, fake review, fake sentiment, spam review. By using these terms we build up the search terms:

- We use these terms (define above to identify spam review) to extract relevant journal or conference papers for extraction of techniques and methods used for detection of spam review.
- Developed inquiry terms were utilized as a part of seeking various writing assets: sciencedirect.com, googlescholar.com, IEEEexplore.com and the ACM digital library.
- The third footstep was superiority examination: the amassed documents were bankrupt down in regards to whether they were related to our investigation. Those identifying with our investigation were picked, and the rest were slighted.

Next, we surveyed references of the consequent papers to gain extra appropriate papers. Finally, we reviewed in a general sense all assembled papers

## 3. Aspect Extraction

With the end goal of accommodation, we have orchestrated

the frameworks of unequivocal angle extraction into three huge sorts i.e. undirected, semi-directed and directed

### 3.1 Undirected

Undirected learning calculation tries to discover concealed example from unlabeled information on the premise of closeness measure. We have info information with no named reactions. The most well-known calculation utilized as a part of undirected learning is grouping. Group examination utilizes a plan of comparability measure to discover concealed information. Comparability measure is characterized upon measurements, for example, Euclidean separation or probabilistic separation. By a long shot the greater part of the methods of insight has focused customer review dataset which was at initially utilized by Hu and Liu<sup>15</sup>. These datasets were prepared for the thing audit zone yet distinctive spaces were also explored. Table 1 demonstrates number of procedures in undirected strategy where implicit viewpoints are recognized.

### 3.2 Semi-Directed

Semi-directed learning is a class of supervised learning assignments and procedures that additionally create utilization of undirected information intended for preparing - regularly a little measure of labeled information with a lot of unlabeled information. As we have seen approaches followed by unsupervised techniques on product reviews. Likewise unsupervised techniques, semi-supervised also focus on product reviews.

In this paper author learning product features and

**Table 1.** Rundown of undirected systems where implicit aspects are considered<sup>9,16,19,11,2,24</sup>

| Reference | Year | Model                   | Algorithm                    | Domain  | Language |
|-----------|------|-------------------------|------------------------------|---------|----------|
| [9]       | 2006 | PMI                     | PMI                          | Product | English  |
| [16]      | 2009 | FB2*                    | Recurrence based             | Item    | Chinese  |
| [19][24]  | 2012 | Association rule mining | Feature based opinion mining | Product | English  |
| [11]      | 2014 | Rule based*             | Rule based                   | Product | English  |
| [2]       | 2015 | Semi-supervised         | LDA                          | Product | Chinese  |

**Table 2.** Rundown of semi-directed techniques where implicit aspects are considered<sup>20-23</sup>

| Ref  | Year | Model | Algorithm                  | Domain  | Language |
|------|------|-------|----------------------------|---------|----------|
| [20] | 2008 | BST2* | Bootstrapping              | Product | Chinese  |
| [21] | 2011 | DP    | Twofold engendering        | Item    | English  |
| [22] | 2010 | DP1*  | Twofold engendering        | Item    | English  |
| [23] | 2015 | EXPRS | Page rank dictionary based | Item    | Chinese  |

sentiment simultaneously is an important concept. Context dependency parser techniques have been used for teaching feature and opinion together.

Table 2 demonstrates that utilized semi-directed procedures that cover implicit angles.

### 3.3 Directed

Directed learning is a machine learning calculation. The primary point is to fabricate a model that can make expectation for new information. Model in directed taking in calculation work from known info values and known reactions to that information values. After we construct model, it's anything but difficult to make expectation in view of past estimation. Understood angles have not been considered.

## 4. Proposed Idea

### 4.1 Proposed Idea Model for Implicit Aspects

Figure 3 shows the model that gives an idea to find implicit aspect from large reviews. To find implicit aspect, we can divide into number of phases. Each phase is depends on output of previous phase. One phase is related with other phase.

Firstly we gather dataset, it could be in various dialects. As we are gathering dataset from various sites. So dialect could be contrast. To interpret into basic dialect, dialect recognizable proof instrument is required. Information is pre-prepared in such an arrangement, to the point that

adequate to the information pre-handling calculations. Assessment mining motor is utilized to recognize grammatical form pos tag,noun and adjectives.

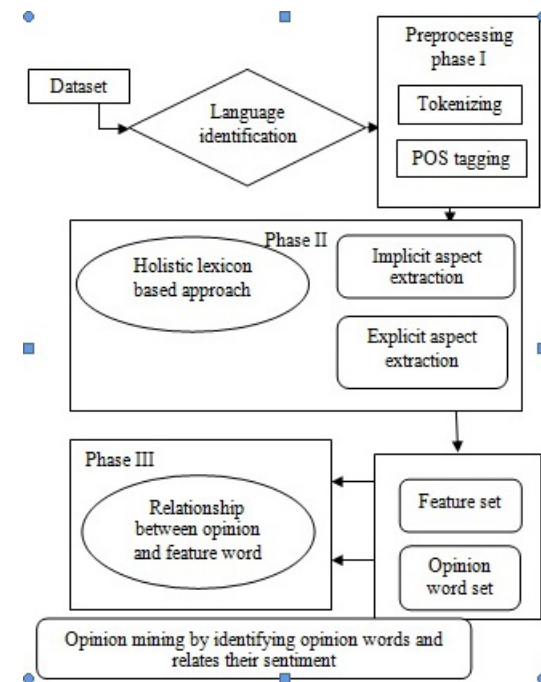


Figure 3. Implicit aspect extractions.

In this document, we recommend a holistic lexicon based proposition to manage dealing with the issue by abusing external affirmations in addition, etymological customs of normal dialect expressions. This approach

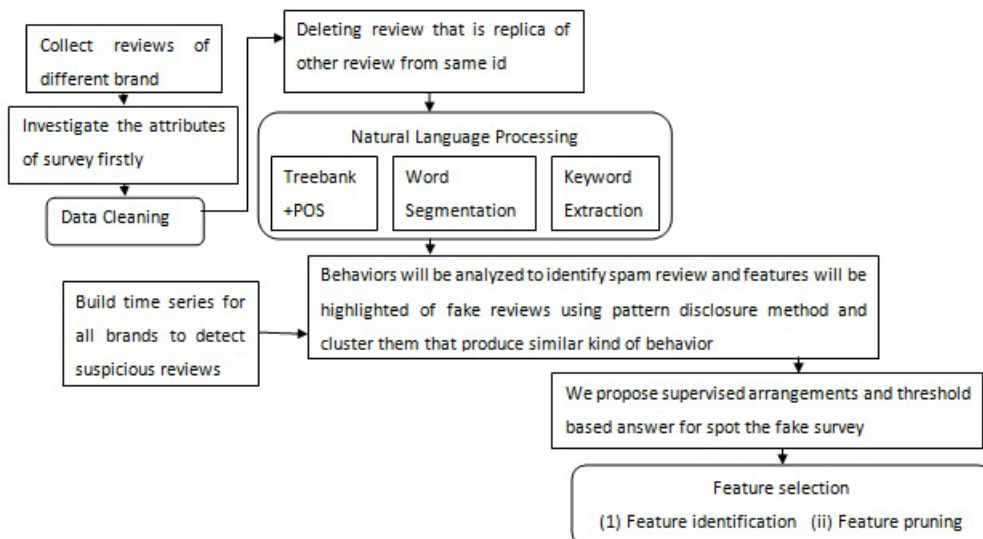


Figure 4. Purposed model for spam detection.

permits the framework to handle sentiment words that are context dependent, which bring about significant troubles for existing algorithms.

## 4.2 Proposed Model for Spam Detection

Instead of considering all reviews, firstly narrow down reviews on which having doubt. Question arise what kind of review to consider. We can use pattern discovery method to known different behaviors to discover spam review. In this manner, detection metrics are then could be used to score every survey. At last, audits with spam scores more noteworthy than the characterized edge are set apart as spam.

In this paper, we expect to propose a structure which is talented to perceive spam overviews effectively. The framework builds time arrangement of number of surveys for every brand and recognizes spam audits from genuine assessments subsequent to distinguishing suspicious intervals.

### 4.2.1 Data Cleaning

Spam reviews are of different categories. Reviews that contain no opinion considered as spam reviews. Reviews that mislead the reviewer whether for the purpose of promoting their product for profit or intentionally writing spiteful about competitor product. It is mandatory to remove these kinds of reviews so that it won't be able to mislead opinion readers. It is compulsory to delete duplicate or nearby duplicate reviews. We can use clustering algorithm that group similar kind of review and after we can delete those reviews.

### 4.2.2 Word Segmentation and POS

Diverse clients have distinctive written work styles; they utilize diverse words to express same importance, so to discover syntactic relations amongst highlight and supposition words are not dense, particularly if there should arise an occurrence of extensive information sets. Thus, we have to utilize some other way to deal with concentrate relationship amongst aspect and sentiment words from substantial information sets. We can use Treebank for this purpose. A treebank is a parsed content corpus that comments on syntactic or semantic sentence structure. A characteristic dialect parser is a task that works out the semantic structure of sentences, for event, which gatherings of words go together (as "expressions") and which words are the subjects or challenge of a verb.

To parse a sentence, we need to know the rules for the syntax of language. We can understand grammatically also.

Example: - John ate a frog. "John" is a grammatical subject "ate a frog" is a predicate "a frog" is a direct object "ate" the Verb.

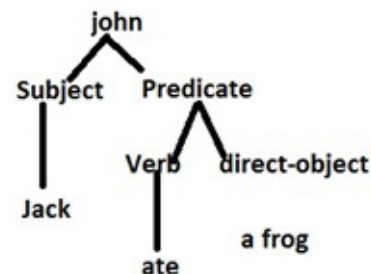


Figure 5. Tree notations for a diagrammed sentence.

<http://nlp.stanford.edu:8080/corenlp/>. We can build tree notation using syntactic categories that tells the structure of items themselves.

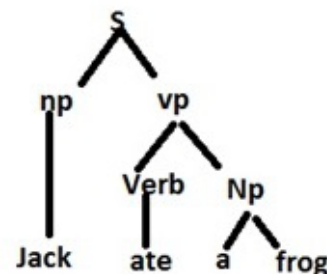


Figure 6. Syntactic categories.

We note that "jack" is a proper noun. We can put this information in the tree too. We will call such structures "a parse trees" or "phrase markers".

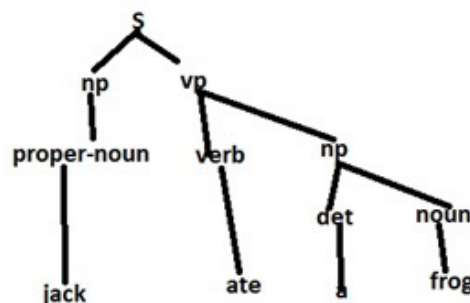


Figure 7. Parse tree.

**Table 3.** Description of spam review approaches along with strength and weakness<sup>29-31</sup>

| Ref | Author              | Dataset                                                                    | Technique                                                                 | Strength                                                                                                                        | Weakness                                                                                                        |
|-----|---------------------|----------------------------------------------------------------------------|---------------------------------------------------------------------------|---------------------------------------------------------------------------------------------------------------------------------|-----------------------------------------------------------------------------------------------------------------|
| 29  | Heydari (2016)      | Product review from amazon.com                                             | 1)Pattern recognition technique                                           | Reducing heavy computation by for investigation purposes.                                                                       | Lack of gold-standard dataset                                                                                   |
| 30  | Yuming Lin (2014)   | http://www.cs.uic.edu/liubIFBS/senti-ment-analysis.html from amazon        | 1)Supervised solution 2)Threshold                                         | 1)Observation shows it work good with high precision and recall<br>2)Without training samples can make up a good effect         | Work with small data set                                                                                        |
| 31  | Dongzhi Wang (2015) | Various sources such as e-commerce stores, search engines, blogs and webs. | 1)NLP<br>2) Machine Learning algorithms<br>3)Python program-ming language | 1) Extract reviews smartly, valuable uses are acquainted with assist particular control-lers and customers to make exact choice | Small data set, Unable to achieve better accuracy, Other appli-cation must introduce to support decision making |

### 4.2.3 Keyword Extraction

For each item property watchword, we recognized its supposition words and related degree verb modifiers with the SVM classifier.

After this step, behaviors are investigated of analyst. At sudden, we are seeing quantities of audits are negative or positive. This is called blasted assault either to advance or downgrade the notoriety of product. So we can break down conduct by utilizing pattern disclosure method. After we can bunch into one gathering using clustering method.

We generally offer score to every audit to compare with limited value. On the off chance that that esteem is more noteworthy than limit esteem then that will be called spam survey.

Table 3 Gives description of spam review approaches. In this table all strength and weakness of all techniques have mentioned. Till now, less work done on spam survey. Spam survey is a key subject in supposition mining. This is purposed framework. In future, we try to implement this framework. Till now, very less work done on spam reviews.

## 5. Conclusion

With the advancement in web, large amounts of information is produced every day, which made it troublesome for the clients, makers or notwithstanding for interpersonal organization users to get precise and right data. This drove presentation of viewpoint based conclusion mining. We highlighted the variables which

are for the most part essential for a compelling and intelligent aspect mining framework. Notwithstanding this, we proposed applied model for successful feeling mining framework. The explanation behind purposing new strategies that prior rating deviation based separating framework is effectively conned by savvy spammers. In this paper holistic lexicon based approach framework is purposed to distinguish implicit review and hearty survey spam discovery framework is purposed. To distinguish burst examples of spam audit, we are appraising every review. Clustering method is used to bundle reviews into one group to detect spam review. This proposed model also has the capacity to cover dominant part of the elements which are the deciding factors for the effectiveness of aspect mining framework. Future work lies towards the implementation of the proposed model display and to approve the model utilizing genuine datasets.

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