

# A New Feature based Scoring Technique to Discover Sentiments Present in the Online Product



# R.Bhuvaneswari, S.Ravichandran

Abstract: The increasing usage of internet, online stores and social media has provided the users to express their opinion, attitude and views without any reluctance and fear on the World Wide Web. These opinions expressed by the users can be related to a product or service as well as any global issues. The colossal growth of the web technology has offered the consumers to know more about the products they intend to buy from the existing customer's reviews. This paper focuses on analyzing the opinion by splitting the positive and negative opinion and then guides the user about the ground truth regarding the performance and quality of the product. The important idea here is to categorize the important features of the product and then accordingly provide the feature wise computation instead of roughly promoting a product as good or bad.

Keywords: sentiment analysis, opinion mining, data mining, knowledge discovery.

#### I. INTRODUCTION

Sentiment analysis is computational process of discovering the individuals' opinions, sentiments, feelings, and dispositions. According to the author [Bing Liu, 2015] this entrancing issue is progressively critical in organizations and society. Sentiment analysis is a sort of Natural Language Processing (NLP) for following the state of mind of general society about a specific item or service. Sentiment analysis, additionally called opinion mining, is area of study related to opinions, sentiments, disposition, feelings, assessments, evaluations towards substances, for example, items, services, associations, people, issues, events, subjects, and their traits. According to the author [M.R. Saleh et al., 2011] enormous volume of data related to the client Opinions or views is very cumbersome to assess and need a computerized general approach to deal with the opinions gathered in the internet. Bounteous forums, journals, online business, social media, news reports and more importantly the web based resources aids the users to express their opinions, which can be utilized for understanding the opinions of the overall population and customers on online sites,

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social media, political activities and product sentiments. When a person wants to buy a product online he or she will typically start by searching for opinions written by other people on the various offerings. Hundreds of startups are developing sentiment analysis solutions and this area of research is gaining momentum of late. Sentiments and opinions are becoming more important mainly due to rapid growth of e-commerce sector, which provides copious data source for analyzing sentiments and views. Of late, most of the customers on e-commerce website depend heavily on the views and opinions posted by existing customers and users before taking the decision to buy the products. For example opinions given on e-commerce sites like Amazon, FlipKart can influence the customer's decision in buying products online.

#### II. DATA CHARACTERISTICS

Review related to products is an independent view or suggestion the users post in real time messages and have plethora of distinct characteristics, which throws new challenges in carrying out sentiment analysis. The following are the most important characteristics of a review message provided by the users,

# A. Data Length

The data length of the different user will be unique and the length of the message has no constraint and this makes the analysis quite cumbersome as the analysis has to be carried out on different longer size opinions.

# **B.** Writing Technique

The users usually deals with short form writing like ASAP, wrong spellings and slang related to cyber technology in their views. Mostly the user data are of different sizes, usage of acronym, and use emoticon and other special characters to convey special meanings to their views.

# **C.Emoticons**

Emoticons are tiny pictures of facial expression using punctuation and letters and the consumer use the emoticons to express their views and moods. Every emoticon has a special mood resemblance related to the product.

# D. Data Used

The data to be used in the proposed work in this paper is acquired from gsmarena.com, a portal specialized in reviewing mobile phones and comprises of huge volume of unstructured data in the form of views and remarks provided by the users.

#### III. OVERALL BLOCK DIAGRAM

The words present in the view are the terms user use to define their attitude and opinion (positive, negative or neutral). To discover the exact product opinion of the user, a new approach to extract the important features about the product and discover the opinions related to the features is proposed.

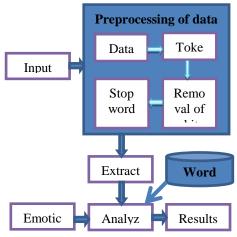


Figure 1: Overall Block Diagram

The preprocessing technique is illustrated here in this section, the first process is tokenization and this tokenization is nothing but separating words present in the view based on the common space separator. The tokenized words are stored separately with the opinion ID and the procedure to carry out the tokenization is shown in the figure 2.

The tokenized words are then processed to remove the stop words or common words which are no longer useful in determining the orientation of the opinion and words like "and", "the", "so", "that", "what" are removed from the list of words to reduce the size of the words to be processed.

o reduce the size of the words to be processed			
Procedure TokenizeWords			
( Opinion Data D )			
INPUTS: Opinion Data File D			
OUTPUT: Tokenized words			
<ol> <li>Load Opinion Data D</li> </ol>			
2. While [D ≠ EOF] do begin			
3. Word=♦			
4. For each character C in D do			
begin			
5. IF [ C ≠ Space ]			
6. Word = Word + $C$			
7. Else			
8. Store Word in Array			
9. Word = $\phi$			
10. End IF			
11. End For			
12. End While			
13. Return Array			
END procedure			

Figure 2: Pseudo code to tokenize the words

Procedure RemoveStopWord ( Tokenized Word Array TA, Stop word reference file SF )

**INPUTS:** Tokenized Word Array TA,

Stop words reference file SF

**OUTPUT:** Tokenized words without stop words

- 1. Load Tokenized Word Array TA
- 2. For each word W in TA do begin
- 3. IF [W present in SF]
- 4. Remove W
- 5. End IF
- 6. End For
- 7. Return Array TA

END procedure

Figure 3: Pseudo code for stop word removal

The figure 3 shows the pseudo code to remove the stop words that are no longer has any effect to the overall opinion or expression of the user. Since the young and new users use many emoticons to represent their emotions regarding the product, the emoticons are extracted and compared with the emoticon reference file to compute the emoticon value as shown in figure 4.

Figure 4: Pseudo code to find emoticon value

# Procedure EmoticonValue ( Tokenized Word Array TA, Emoticon reference file EF )

**INPUTS:** Tokenized Word Array TA, Emoticon reference file SF

**OUTPUT:** Emoticon value with OpinionID

- 1. Load Tokenized Word Array TA
- 2. For each word W in TA do begin
- 3. IF [ W not present in Dictionary]
- 4. IF [ W present in EF]
- 5. Compute the Emoticon value →E-value with ID
- 6. End IF
- 7. End IF
- 8. Return E-value with ID

END procedure

**Table 1: Emoticon reference file** 

Emoticon Reference file			
Emoticon	Meaning	Value	
;):)	Нарру	0.5	
=):-)			
XD	Laughing	1	
:D	Big smile	1	
BD	Big Smile	1	
	with glass		
`:*	kiss	0.5	
: :\	undecided	0	
:(	Sad	-0.5	
B(	Sad with	-0.5	
	goggles		
:'(	Crying	-1	
X-(	Angry	-1	

The feature extraction is carried out to extract three basic attributes from the opinions and they are Appearance,



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Affordability and Essential. Here the appearance is the look and the style, sleekness, weight, colors, design, and sturdiness. The affordability deals with the price tag, tariff, and the rate of the product. The essentials are basically the hardware and the software used like battery, screen size, screen resolution, speakers, camera, connectivity and processor. The tokenized document is further segregated according to the three features explained earlier and the tokens are stored in the respective feature heads. A special feature reference document is used to separate the review tokens according to the basic three features. The pseudo code to extract the basic features is shown in the figure 5.

Word TA, Feature reference doc FD)  INPUTS: Tokenized Word TA, Feature reference doc FD OUTPUT: Emoticon value with OpinionID  1. Load Tokenized Word Array TA 2. For each word W in TA do begin 3. IF [ W present in AppearanceSection of FD] 4. Fetch the complete phrase tokens 5. Store in their Appearance array 6. ElseIF [ W present in AffordabilitySection of FD] 7. Fetch the complete phrase tokens 8. Store in their Affordability array 9. Else [ W present in EssentialSection of FD] 10. Fetch the complete phrase tokens 11. Store in their Essential array	Procedure ExtractFeature ( Tokenized			
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<ul><li>10. Fetch the complete phrase tokens</li><li>11. Store in their Essential array</li></ul>				
11. Store in their Essential array				
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12 End IE	12. End IF			
13. End For				

Figure 5: Pseudo code of basic feature extraction.

END procedure

The sample special feature reference document file is showcased in the figure 6 and this file is used to segregate the important features present in the mobile phones and the sentiment analysis is carried out based on these three features extracted. Since the opinions will have a mixed responses like, a great camera capabilities, good screen resolution but will be having poor battery, high rate and very less signal strength. Here the proposed approach provides a new method by delivering the opinion values separately according to the features and this will definitely offer a better insight related to a product and the consumer will experience the real ground truth about a product with break up details and has an opportunity to know about the strengths and the weakness of the product.

Table 2: Basic Special features reference file

Special Feature Reference file			
Tokens	Meaning		
Price, Rate, affordable			
Tariff, Cost, \$			
Expensive, Cheap	Affordability Feature		
Amount, INR, Economy			
Screen, Size, build, inch			
Design, Sleek, Sturdy			
Weight, Grams, pixels, resolution	Appearance Feature		

Lavish, Classy, Elegant	
Battery, body, Camera	
MP, FPS, Video, Face	
OS, Dual, Glass, SIM	
Hybrid, Octo, Quad,	Essential Features
USB, OTG, Wifi, Charger	
NFC, Chip, RAM, External,	
Memory	

# IV. CALCULATING THE OPINION SCORE

The main portion of the proposed approach consists of calculating the verb score and adverb score of the tokens present in the three feature arrays extracted from the opinions of the users. Despite the fact that, one can figure out the exact meaning of a specific writings depending on the semantic introduction of the descriptive words, yet including adverb modifiers is very important. This is essentially on the grounds that there are a few adverbs in semantics, (like, "not") which is extremely fundamental to be mulled over as they would totally change the entire meaning of the opinions (i.e.) "not good" is a negative view and not a positive view. For example,

**User 1:** "The battery life is great"

**User 2:** "The battery life is not great"

Here by just seeing great, the opinion should not be placed in the positive side as the first user's opinion is positive and the second user's opinion is on the negative side. The next question is if both the views are positive or negative, what is the actual score given for the phrase as every user will have an unique way in expressing their views. For example,

User 1: "The battery life is great"

**User 2:** "The battery life is extremely great"

The adverb present in the opinion helps a lot in deciding the level of positivity or negativity and here the User1's opinion is positive, whereas the User2's opinion is strongly positive and a score is provided to find this level as shown in the table

VERB and ADVERB SCORE CHART			
Tokens	Score	Type	
Like, Love,	1		
Adore, Enjoy, prefer	0.9		
Excite, Attract,	0.8		
recommend		Verb	
Hate, dislike	-1		
Suck, regret	-0.9		
Suffer, reject, hurt	08		
Most, Complete, Best	1		
Too, extremely, very	0.9		
Much, More, pretty	0.8	Adverb	
Never, Hardly	-1		



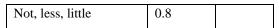


Table 3: Common Verb and adverb scores in the opinion

The table 3 showcases most of the frequently employed verbs and adverbs usually in the opinions and their values are ranged from +1 to -1. The Word net is used here to recognize the verbs and adverbs. The token that cannot be identified by the Word net is pruned off as they may not be usable words.

Usually to computerize the user's opinion, plethora of approaches is used to predict the user's sentiments which include Natural Language Processing (NLP) and Machine Learning (ML) algorithms.

The opinion score is calculated after finding the scores of the adjectives and verbs. The adjectives and verbs are combined together to form adjective set and similarly the adverbs and verbs are combined to form verb set. The products of adjective scores, adverb scores and verb scores are initially computed and then the feature based opinion score is calculated by the following equation,

$$\underline{S(F_{app})} = \frac{1}{N} \sum_{i=1}^{N} S(Adj) + S(Vb) + (En * S(E))$$

$$\underline{S(F_{aff})} = \frac{1}{N} \sum_{i=1}^{N} S(Adj) + S(Vb) + (En * S(E))$$

$$\underline{S(F_{ess})} = \bigoplus_{N} \sum_{i=1}^{N} S(Adj) + S(Vb) + (En * S(E))$$

Where, S(Fapp) = Score of the feature Appearance, S(Faff) = score of the feature affordability, S(Fess) = Score of the feature essentials, N = number of tokens, S(Adj) = adjective set, S(Vb) = verb set, En = number of emoticons, S(E) = Emoticon score.

# V. EXPERIMENTAL RESULT AND ANALYSIS

To explain the score calculation, a sample data is considered and the scores are calculated to find the sentiment involved in the data.

"This mobile is too good:-) and I love the overall design as it is very sleek and light weight. The battery life is great:-):-). Both the front and the rear camera are the best in the market and the photos are excellent and I enjoy it XD.. But the rate of the phone is very costly:(:(. "

After cleaning the input data, the stop words are removed from it. The basic three features are initially extracted and they are grouped separately to find the score separately. The adjective set and the verb sets are extracted from the separated features data along with the emoticons if present.

Appearance Feature = This mobile is too good :-) and I love the overall design as it is very sleek and light weight.

# Adjectives extracted

Too good, very sleek, light weight

Verb extracted

Love

**Emoticons extracted** 

:-)

**Affordability feature** = But the rate of the phone is very costly:(:(

**Essential feature** = The battery life is great :-) :-). Both the front and the rear camera are the best in market and the photos are excellent and I enjoy XD

The adjective set and the verb set along with the emoticons are extracted from these features and then the scoring is computed.

(i) 
$$S(F_{app}) = (1/6) * ((0.9 x 1) + (0.9 x 1) + (0.7 x 0.5) + (1 x 0.5))$$
  
= 0.441

(ii) 
$$S(F_{aff}) = (1/3) * ((0.9 x - 1) + (-0.5) + (2 x - 0.5))$$
  
= -0.133

(iii) 
$$S(F_{ess}) = (1/7) * ((0.9 x 1) + (0.9 x 1) + (0.9 x 1) + (1 x 1))$$
  
= 0.525

This mobile phone user feedback is classified into three categories and the user has provided a positive opinion about its appearance, positive opinion about its physical nature, but the user has given a negative opinion about its affordability. Overall the opinion may fall on the

positive side but the negativity should also be considered and here the rate or the price tag proved to be its negative face. The sample data are shown in the table 4 and its corresponding scores related to appearance, affordability, essentials and the overall rating of the opinion is calculated showcased.

Table 4: Sample data and its score calculations

Table 4: Sample data and its score calculations				
Sample data	S	S	S	S
	(App)	(Aff)	(Ess)	<b>(R)</b>
Waste mobile internal	-	-	-0.32	-ve
memory only 8GB can't			5	
install more app			3	
very very worst X-(				
I am using this phone last	0.327	0.21	0.107	+ve
3 month phone is good		2		
:D, best battery life it's		2		
looking and working				
good but				
sound quality is not good.				
Very Cheap phone :-)				
I think it beats all phone	0.356	0.21	0.425	+ve
in market		1		
and it is a best phone to		1		
buy at this low price 6				
inch screen is very large				
and good I Love this				
phone. :D :D				





EXCELLENT PHONE	0.367	0.22	0.512	+ve
:-), I recently		1		
purchased the phone in		1		
Dubai cheaply whilst on				
holiday and finally I can				
say the features are super				
it is light and classy. My				
interest in a phone is				
battery life				
obviously includes				
charging speed and				
clarity of both image and				
sound. The phone				
charges to the max in				
under 30 mins and				
gives me a lot more :D I				
LOVE it!. :-)				

# VI. CONCLUSION

The evolution of online sites selling many products to the consumers offers an exceptional prospect to design and employ plethora of new technologies to search, extract and mine the user's emotions and opinions related to the products that are being sold on the online sites. The work presented in this paper specifies a novel approach for opinion mining on online product data sold. To unearth the sentiments hidden underneath the user fata, the opinion words (a combination of the adjectives along with the verbs and adverbs) are extracted along with the emoticons present in the user sample data. The corpus-based method is employed to discover the adjective set and the dictionary-based approach is employed to discover the verb set. After finding the individual scores of the three important features present, the overall opinion was then found.

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