VOLUME 3, ISSUE 6, Jun.-2017

SENTIMENT ANALYSIS OF MARATHI LANGUAGE

SUJATA DESHMUKH

Department of Information Technology, FRCRCE, Affiliated to Mumbai University, Mumbai, India. sujata.p.deshmukh@gmail.com

NILEEMA PATIL

Dept. of Physics, LTCE, Affiliated to Mumbai University, Mumbai, India. patil. nileema@gmail.com

SURABHI ROTIWAR

Department of Information Technology, FRCRCE, Affiliated to Mumbai University, Mumbai, India. surabhi.rotiwar@gmail.com

IASON NUNES,

Department of Information Technology, FRCRCE, Affiliated to Mumbai University, Mumbai, India. jason.nunes@gmail.com

ABSTRACT:

Sentiment analysis offers many benefits and opportunities from business, government consumer perspective in this digital data explosive age. According to Google, in partnership with KPMG India report (April-2017), titled 'Indian Languages -Defining India's Internet', currently India today has 234 million Indian Language users who are online, compared to 175 million English web users and expected 536 million Indians to use regional languages while online by 2021. However Marathi users are expected to make significant contribution to define India Internet volume. By considering aspects of Marathi language and benefits of sentiment analysis, this paper presents a approach to overcome the barriers and difficulties being faced for analyzing text in Marathi language. The proposed system detects hidden sentiments in text of Marathi language. The system uses sentiment analysis methodology in order to achieve desired functionality. In this system, a corpus based approach is proposed, i.e the creation of a diverse up to date corpus of Marathi keywords, along with their individual polarities, with respect to the Word Net, which is consider as a corpus. The algorithm is used to calculate the cumulative polarity of the text and rank the sentence as positive, negative or neutral on a set scale standard.

KEYWORDS: Sentiment Analysis of Marathi Language, sentiment analysis, corpus based approach

I. INTRODUCTION:

Sentiment Analysis (SA) or Opinion Mining (OM) is the computational study of people's opinions, attitudes and emotions toward an entity. The entity can represent individuals, events or topics. Sentiment Analysis identifies the sentiment expressed in a text then analyzes it. Therefore, the target of SA is to find opinions, identify the sentiments they express, and then classify their polarity. It

is the field of study that analyzes people's opinions, sentiments, evaluations, appraisals, attitudes, and emotions towards entities such as products, services, organizations, individuals, issues, events, topics, and their attributes. It represents a large problem space [2, 3]. The analysis of sentiments may be document based, paragraph based or sentence based. In document approach the sentiment in the entire document is summarized as positive, negative or objective. In sentence based where individual sentences, bearing sentiments, in the text are classified. Sentiment Analysis task is considered a sentiment classification problem. Following are the general steps—

- To select text features i.e. obtaining a dataset.
- Identifying Parts of speech (POS) finding adjectives, as they are important indicators of opinions.
- Identifying Opinion words and phrases- these are words commonly used to express opinions including good or bad, like or hate.
- Negation handling-The appearance of negative words may change he opinion orientation like not good is equivalent to bad.

Calculate sentiment score and summarize as positive, negative or objective [2, 3].

II. LITERATURE REVIEW:

The following literature review provides, issues, challenges, approaches to understand the system. Walaa Medhat et al.(2014), highlighted that there is still a lack of resources and researches concerning languages other than English sentiment analysis [4]. Authors reviewed Fiftyfour of the recently published and cited articles, and were categorized and summarized. This paper provides basic understanding and helped to understand corpus and dictionary based approaches. Soha Ahmed et.al (2013), presented the challenges faced by researchers when conducting Social SA (SSA) on Arabic text [6]. Further

ISSN: 2454-7875

VOLUME 3, ISSUE 6, Jun.-2017

authors suggested some solutions inspired from the literature, and discussed the difficulties faced when performing SSA on Arabic social media text. A preprocessing phase to sentiment analysis is proposed and showed the noticeably improvement to the results of sentiment extraction from Arabic social media data. This paper provided the improvements by pre-processing the text of SA. Hatem Ghorben and David Jacot (2012), showed that the combination of lexical, morpho-syntactic and semantic features achieves relatively good performance in classifying French movie reviews according to their sentiment polarity (positive, negative) [7]. In order to extract the semantic orientation of words from SentiWordNet, a standard word translation process is used. This paper provided to use SentiWordNet for corpus based approach for proposed system. Further it argued that dictionary-based approach could contribute better results. Yakshi Sharma et. Al (2015), proposed an approach Sentiment Analysis of Hindi Tweets using SentiWordNet with Subjective lexicon method [8]. Each entry in lexicon is categorized into Verb, Noun, Adjective, and Adverb. The results indicate that proposed algorithm. gave better accuracy than unigram presence method. In Hindi language, lexicon coverage can be increased as it has limited coverage now. Further authors suggested to do research on subjective lexicon which can be extended to machine learning, n-grams or both combined. This paper provides motivation to use SentiWordNet and lexicon method. Namrata Godbole et. Al (2012), showed a system that consists of a sentiment identification phase, which associates expressed opinions with each relevant entity, and a sentiment aggregation and scoring phase, which scores each entity relative to others in the same class [5]. Finally, authors evaluated the significance of scoring techniques over large corpus of news and blogs. The above literature summarized in following Table 1.

TABLE I. LITERATURE REVIEW TO IDENTIFY GAP

Title of paper	Conclusion	Relevance	Research
and authors	discussed		Gap
Sentiment	The various	Corpus based	Various other
analysis	sentiment analysis	approach was found to	techniques
algorithms and	algorithms were	be the best approach	are also
applications: A	noted and why one	for our project.	available and
survey	is better than the		a combination
Walaa Medhat,	other is found.		of techniques
Ahmed Hassan,			can also be
Hoda Korashy			favourable.
2014[4]	•		
Large-Scale	Authors concluded	Mapping should be	In large scale
Sentiment	that sentiment can	done to find the best	data it is
Analysis for	vary by	opinion while using	difficult to
News and Blogs	demographic	corpus based	map the
Namrata	group, news source	techniques	opinions and
Godbole,	or geographic		also due to
Manjunath	location.		the various
Srinivasiah,			combinations
Steven Skiena,			of the words.
2012 [5]			

Key Issues in Conducting Sentiment Analysis on Arabic Social Media Text Soha Ahmed, Michael Pasquier, performing SSA on Ghassan Qadah, 2013 [6] approaches that have been developed when performing SSA on Arabic text in general and Arabic social media text in particular. Sentiment Analysis of French Movie Reviews Hatem Ghorben, David Jacot, 2012 [7] arabic text in Grom SentiwordNe, a standard words from SentiwordNe, and proach word words from SentiwordNe, and proach words from SentiwordNe, and proach w			• • •	EUME 3, 1330E 0,	juii. 2017
Arabic Social Media Text Soha Ahmed, Michael Pasquier, Ghassan Qadah, 2013 [6] general and Arabic social media text in particular. Sentiment Analysis of French Movie Reviews Hatem Ghorben, David Jacot, 2012 [7] standard words from Ghorben, David Jacot, 2012 [7] standard gaplication from Ghorben, David Jacot, 2012 [7] standard words from Ghorben, David Jacot, 2	Conduc Sentim	cting ent	highlights key	simple pre-processing steps and showed how	approach can be applied to
Michael Pasquier, Ghassan Qadah, 2013 [6] general and Arabic text in particular. Sentiment Analysis of French Movie Ghorben, David Jacot, 2012 [7] Sentiment Analysis of Hindi Tweets, Vakshi sharma, Veen mangat, Mandeep Kaur, 2015 [8] Using SentiWordNet for Multilingual sentiment Analysis Carstilled Cocument is classified according to its sentiment into one of the classes "positive" and "Negative". Sentiment An Unsupervised lexicon based approach for SA is discussed. A practical Approach to Sentiment Analysis of Hindi Tweets, Vakshi sharma, Veen mangat, Mandeep Kaur, 2015 [8] Document is classified according to its sentiment into one of the classes "positive" and "Negative". Sentiment analysis algorithms and applications: A survey Sentiment analysis algorithms and applications: A survey Walaa Medhat, Ahmed Hassan, Hoda Korashy developed when performing SSA on Arabic text in performing SSA on Arabic text in particular. Although translation wat does not necessarily of preserve the semantic or examination of unsage, in spite of all its side; thas been argued that dictionary-based approach could achieve better results. Lexicon "MaHl" can invery the sentence. \$6, these words are also considered in finding Polarity of text. Document is classified according to its sentiment into one of the classes "positive" and "Negative". Sentiment analysis algorithms were determined for these words. An interpretation of the classes "positive" and "Negative". Sentiment analysis algorithms were determined for these words. An interpretation of the country of the document polarity. Corpus based approach of the best approach a could achieve better results. Lexicon "MaHl" can invery the sentence. \$6, these words are also considered in finding Polarity of text. Sentiment analysis algorithms were determined for these words. An interpretation of the document polarity. Sentiment analysis algorithms were determined for these words. An interpretation of the document polarity. Sentiment analysis algorithms were determined for th	Arabic	Social	innovative		
Ghassan Qadah, 2013 [6]	Michae	1	developed when		
Analysis of French Movie orientation of Reviews Hatem Ghorben, David Jacot, 2012 [7] Sentiment Analysis of Manalysis of Ma	Ghassa	n Qadah,	Arabic text in general and Arabic social media text in		
French Movie Reviews Hatem Ghorben, David Jacot, 2012 [7] A Practical Approach to Sentiment Analysis of Hindi For Multilingual Sentiment Analysis Kerstin Denecke, 2008 [9] Sentiment Canalysis of Hindi For Multilingual Sentiment Analysis Canalysis algorithms and applications: A sentiment analysis algorithms and applications: A sentiment analysis algorithms and applications: A sentiment analysis algorithms and applications: A sentiment Ahmed Hassan, Hoda Korashy A Practical Approach to be used. An unsupervised lexicon based approach could achieve better results. An unsupervised lexicon based approach for SA is side it has been argued that dictionary-based approach could achieve better results. An unsupervised lexicon based approach for SA is discussed An unsupervised lexicon based approach for SA is discussed An unsupervised lexicon based approach for SA is discussed "Nahll" can invert the polarity of the sentence. \$60, these words are also considered in finding Polarity of text. By means of SentiWordNet, scores for positivity and negativity are determined for these words. An interpretation of the scores then leads to the document polarity. Sentiment analysis algorithms and applications: A survey Walaa Medhat, Ahmed Hassan, Hoda Korashy Sentiment and word-translation process due to the variation of language common usage, in spite of all its side it has been argued that dictionary-based approach could achieve better results. Lexicon Coverage can be increased should be increased should be increased for polarity of the sentiment of the sentiment on extranslation of the sentiment on the document is classified according to its sentiment into one of the classes "positive" and "Negative". Sentiment analysis algorithms and applications: A survey when the polarity of the sentence. \$60, these words are also considered in finding Polarity of text. Sentiment analysis approach was found to be the best approach as combination of techniques are also available and a combination of techniques are also for				· ·	
Jacot, 2012 [7] standard word-translation process was found to be used. language common usage, in spite of all its side it has been argued that dictionary-based approach could achieve better results.	French Review	Movie rs Hatem	orientation of words from	preserve the semantic orientation of words	SentiWordNet can be applied
A Practical Approach to Sentiment Analysis of Hindi Sentiment Sent		•	standard word-	language common	
A Practical Approach to Sentiment Analysis of Hindi Tweets, Yakshi sharma, Veenu mangat, Mandeep Kaur, 2015 [8] Using Sentiment Analysis Kerstin Denecke, 2008 [9] Sentiment analysis algorithms and applications: A survey Walaa Medhat, Ahmed Hassan, Hoda Korashy An unsupervised lexicon based approach for SA is depending on to to sentiment analysis algorithms were malased to the holar to the source of the favourable. An unsupervised lexicon based approach for SA is sentence. So, these words are also considered in finding Polarity of text. By means of SentiwordNet, scores for positivity and negativity are determined for these words. An interpretation of the scores then leads to the document polarity. Sentiment analysis algorithms and applications: A survey Walaa Medhat, Ahmed Hassan, Hoda Korashy An unsupervised lexicon based approach for SA is sentence. So, these words are also considered in finding Polarity of text. By means of SentiWordNet, scores for positivity and negativity are determined for these words. An interpretation of the scores then leads to the document polarity. Corpus based that present one which is limited. Corpus absed that present one which is limited. Corpus based approach was found to be the best approach for this system. Sentiment analysis algorithms were noted and why one is better than the other is found.				that dictionary-based	
Approach to Sentiment Analysis of Hindi Tweets, Yakshi sharma, Veenu mangat, Mandeep Kaur, 2015 [8] Using SentiWordNet for Multilingual Sentiment Analysis Kerstin Denecke, 2008 [9] Sentiment Analysis Sentiment Analysis Rerstin Denecke, 2008 [9] Sentiment Analysis Rerstin Denecke, and "Negative". Sentiment Analysis Rerstin Denecke, and "Negative" Sentiment Analysis Rerstin Denecke, and "Negative" Sentiment Analysis Sentiment Analysis Rerstin Denecke, and "Negative" Sentiment Analysis Anyois Anyo				achieve better results.	
Sentiment Analysis of Hindi Tweets, Yakshi sharma, Veenu mangat, Mandeep Kaur, 2015 [8] Using SentiWordNet for Multilingual Sentiment Analysis Kerstin Denecke, 2008 [9] Sentiment into one of the classes "positive" and "Negative". Sentiment analysis algorithms and applications: A survey Walaa Medhat, Ahmed Hassan, Hoda Korashy Sentiment Analysis of Hindi discussed spolarity of the sentence. So, these words are also considered in finding Polarity of text. By means of SentiWordNet, scores for positivity and negativity are determined for these words. An interpretation of the scores then leads to the document polarity. Corpus based approach was found to be the best approach for this system. Sentiment analysis algorithms were is better than the other is found. Sentiment Analysis Corpus based approach was found to be the best approach of or this system. Sentiment analysis algorithms were is better than the other is found.			-		
Tweets, Yakshi sharma, Veenu mangat, Mandeep Kaur, 2015 [8] Using SentiWordNet for Multilingual Sentiment Analysis Kerstin Denecke, 2008 [9] Sentiment into one of the classes "positive" and "Negative". Sentiment analysis algorithms and applications: A survey Walaa Medhat, Ahmed Hassan, Hoda Korashy Mandeep Kaur, 2015 [8] Document is considered in finding Polarity of text. By means of SentiWordNet, scores for positivity and negativity are determined for these words. An interpretation of the scores then leads to the document polarity. Corpus based approach was found to be the best approach for this system. Sentiment analysis algorithms were is better than the other is found. Words are also considered in finding Polarity of text. Performance of system is depending on translation software. Corpus based approach was found to be the best approach for this system. Sentiment for these words. Corpus based approach was found to be the best approach of its system. Sentiwent is depending on translation software. Various other techniques are also available and a combination of techniques can also be favourable.	Sentim	ent	approach for SA is	polarity of the	be increased
sharma, Veenu mangat, Mandeep Kaur, 2015 [8] Using SentiWordNet for Multilingual Sentiment Analysis Kerstin Denecke, 2008 [9] Sentiment into one of the classes "positive" and "Negative". Sentiment analysis algorithms and applications: A survey Walaa Medhat, Ahmed Hassan, Hoda Korashy Sentiment mangat, Considered in finding Polarity of text. By means of SentiWordNet, scores for positivity and negativity are determined for these words. An interpretation of the scores then leads to the document polarity. Corpus based approach was found to be the best approach for this system. Corpus based approach was found to be the best approach for this system. Sentiment of various other techniques are also a combination of techniques can also be favourable.			discussed		
Using SentiWordNet translated into English using Sentiment Analysis Kerstin Denecke, 2008 [9] Translation software. Then, the translated document is classified according to its sentiment into one of the classes "positive" and "Negative". Sentiment analysis algorithms and applications: A survey is better than the Walaa Medhat, Ahmed Hassan, Hoda Korashy Document is SentiWordNet, scores for positivity and negativity are determined for these words. An interpretation of the scores then leads to the document polarity. Corpus based approach was found to be the best approach for this system. Various other techniques are also available and a combination of techniques can also be favourable.	sharma	, Veenu		considered in finding	
SentiWordNet for Multilingual Sentiment Analysis Translation Software. Then, the translated document is classified according to its sentiment into one of the classes "positive" and "Negative". Sentiment analysis algorithms and applications: A survey walaa Medhat, Ahmed Hassan, Hoda Korashy Translated into English using standard negativity are determined for these words. An interpretation of the scores then leads to the document polarity. Corpus based approach was found to be the best approach for this system. SentiwordNet, scores for positivity and negativity are determined for these words. An interpretation of the scores then leads to the document polarity. Corpus based approach was found to be the best approach for this system. SentiwordNet, scores of system is depending on translation software. Corpus based approach was found to be the best approach for this system. SentiwordNet, scores of system is depending on translation software. Corpus based approach was found to be the best approach for this system. SentiwordNet, scores of prositivity and negativity are determined for these words. An interpretation of the scores then leads to the document polarity. Corpus based approach was found to be the best approach for this system. Sentiment into one of the classes "positive" and interpretation of the scores then leads to the document polarity.		-			
for Multilingual Sentiment standard Translation Sentiment Analysis Translation Software. Then, the translated document is classified according to its sentiment into one of the classes "positive" and "Negative". Sentiment analysis algorithms and applications: A survey is better than the Walaa Medhat, Ahmed Hassan, Hoda Korashy English using for positivity and negativity are determined for these words. An interpretation of the scores then leads to the document polarity. Corpus based approach was found to be the best approach for this system. Corpus based approach was found to be the best approach for this system. Sentiment analysis algorithms were applications: A survey is better than the other is found.		ordNet		-	
Analysis Kerstin Denecke, 2008 [9] Translation software. Then, the translated document is classified according to its sentiment into one of the classes "positive" and "Negative". Sentiment analysis algorithms and applications: A survey Walaa Medhat, Ahmed Hassan, Hoda Korashy Translation software. These words. An interpretation of the scores then leads to the document polarity. Corpus based approach was found to be the best approach for this system. of this system. of techniques can also be favourable.	for Mul	tilingual	English using	for positivity and	depending on
Kerstin Denecke, 2008 [9] translated document is classified according to its sentiment into one of the classes "positive" and "Negative". Sentiment analysis algorithms and applications: A survey is better than the Walaa Medhat, Ahmed Hassan, Hoda Korashy Kerstin Denecke, software. Then, the translated document into one of the scores then leads to the document polarity. Corpus based approach was found to be the best approach for this system. of this system. Sentiment analysis algorithms were noted and why one is better than the other is found.				· ·	
document is classified according to its sentiment into one of the classes "positive" and "Negative". Sentiment analysis algorithms and applications: A survey is better than the Walaa Medhat, Ahmed Hassan, Hoda Korashy document is scores then leads to the document polarity. Corpus based approach was found to be the best approach for this system. of techniques are also available and a combination of techniques can also be favourable.	Kerstin	Denecke,	· ·	words. An	
classified according to its sentiment into one of the classes "positive" and "Negative". Sentiment analysis algorithms and applications: A survey is better than the Walaa Medhat, Ahmed Hassan, Hoda Korashy classified according to its sentiment and the document polarity. Corpus based approach was found to be the best approach for this system. Corpus based approach was found to be the best approach for this system. are also available and a combination of techniques can also be favourable.	2008 [9	9]		-	
into one of the classes "positive" and "Negative". Sentiment analysis algorithms and applications: A survey walaa Medhat, Ahmed Hassan, Hoda Korashy into one of the classes "positive" and "Negative". Corpus based approach was found to be the best approach for this system. for this system. Corpus based approach was found to be the best approach for this system. are also available and a combination of techniques can also be favourable.			classified according		
classes "positive" and "Negative". Sentiment analysis algorithms and applications: A survey walaa Medhat, Ahmed Hassan, Hoda Korashy classes "positive" and "various based approach was found to be the best approach for this system. Corpus based approach was found to be the best approach for this system. are also available and a combination of techniques can also be favourable.					
"Negative". Sentiment analysis sentiment analysis algorithms and applications: A survey is better than the Walaa Medhat, Ahmed Hassan, Hoda Korashy "Negative". Corpus based approach was found to be the best approach for this system. is better than the other is found. Corpus based approach was found to be the best approach for this system. is better than the other is found. Corpus based approach was found to be the best approach for this system. is better than the other is found. Corpus based approach was found to be the best approach are also available and a combination of techniques can also be favourable.					
Sentiment analysis sentiment analysis algorithms and applications: A noted and why one survey is better than the Walaa Medhat, Ahmed Hassan, Hoda Korashy The various Corpus based approach was found to be the best approach for this system. The various corpus based approach was found to be the best approach are also available and a combination of techniques can also be favourable.					
algorithms and applications: A noted and why one survey is better than the Walaa Medhat, Ahmed Hassan, Hoda Korashy algorithms were noted and why one is better than the other is found. be the best approach for this system. are also available and a combination of techniques can also be favourable.	Sentim	ent		Corpus based	Various other
applications: A noted and why one survey is better than the Walaa Medhat, Ahmed Hassan, Hoda Korashy is noted and why one for this system. available and a combination of techniques can also be favourable.			-	* *	-
survey is better than the Walaa Medhat, Ahmed Hassan, Hoda Korashy is better than the other is found. a combination of techniques can also be favourable.				1.1	
Ahmed Hassan, Hoda Korashy can also be favourable.	survey		is better than the	,	a combination
Hoda Korashy favourable.	1		other is found.		
	Hoda K	orashy			

III. PROBLEM STATEMENT:

The aim of this system is to address the overall problem of sentiment analysis being faced in the Marathi language. The idea behind the development of this system is twofold; Firstly, to create a diverse up to date corpus consisting of Marathi keywords like adverbs, adjectives etc. along with their individual polarities. This in keeping with other sentiment analysis literature can be called Marathi Word Net. Secondly, aim of this system is to carry out advanced sentiment analysis using optimal algorithm to obtain the cumulative polarity of text. Thus, this twofold problem statement is converted into following objectives:

- To create an system that that uses a corpus based approach and can carry out advanced Sentiment Analysis for a vernacular language i.e. Marathi that detects hidden sentiments in text and analyzes the content accordingly.
- To create an system that that uses a corpus based approach and can carry out advanced Sentiment Analysis for a vernacular language i.e. Marathi that detects hidden sentiments in text and analyzes the content accordingly.

IV. PROPOSED SYSTEM:

The proposed sentiment analysis approach is a Corpus based approach. It is twofold and involves two modules. First module provides foundation to the proposed system by creating of feasible corpus for Marathi language from English SentiwordNet and second module presents mapping of keywords to analysis sentiments. The main tasks of creation of feasible corpus from English SentiwordNet are Categorization of keywords with respect to parts of speech (POS) and allocating polarity to keywords. The main steps for Mapping of keywords to analysis of sentiments are accept Marathi content as input for analysis and use the devised algorithm to find polarity and degree of given input.

A. CORPUS GENERATION FROM ENGLISH SENTI WORD NET:

This system proposed the creation of a Marathi corpus. In order to generate the corpus, steps are as follows:

- 1. Collection of data online: Find text files which contained lists of various Marathi keywords and their meanings etc.
- 2. Segregation and lemmatization: It pre-processed the data to categorize words as PoS he parts of speech, ie Adjective etc and another corresponding lemma file.
- 3. Conversion to csv format: After pre-processing, it converted the text files into 11 csv format in order to be able to index the keywords properly. Every keyword was attributed a unique identification number. Then Data File is created, which has the same corresponding UIDs and the meaning of the words, followed by its synonyms to facilitate mapping.
- 4. Attributing polarity to each keyword: polarity of each word is calculated for word using English sentiwordnet. Then, for the purpose of easy indexing, the text files are converted to csv format.

B. ALGORITHM

The algorithm first find individual polarity of each word in the sentence and then find the cumulative polarity to determine if the sentiment is positive, negative or neutral and to what degree. Sentence in Marathi is a input for the algorithm. The steps are follows:

• Elimination of stop words. Any words that do not attribute any specific polarity to the sentence are found to

be redundant while calculating the overall polarity of the sentence. Hence they are known as stop words.

- Obtaining the polarity of relevant keywords from the corpus. Here, it is +0.75 and +0.5.
- Calculating cumulative polarity.

The cumulative polarity P is calculated as P=(a+b+c+....)/n where; a,b,c are the individual polarities mapped from the corpus, n=No of words.

V. RESULTS AND DISCUSSION:

This system is implemented in Java. Following screen shot shows the output of the system.



Figure 1: Highly positive output

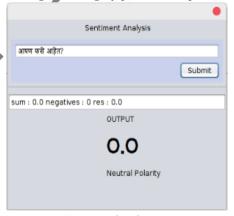


Figure 2 Neutral polarity output



Figure 3: Negative polarity output

VOLUME 3, ISSUE 6, Jun.-2017

Following Table 2 provides different test cases its polarity.

Table 2: Test cases with Result

Input	Output	
मला वाटते ती, अतिशय सुंदर आहे	0.84, Highly Positive	
किती सुंदर आहे ती!	0.75 Highly Positive	
ती आहे म्हणून दु: खी?	0.0 Neutral Polarity	
मी खूप आनंदी आहे	0.14 Positive Polarity	
मी आनंदी नाही	-0.12 Negative Polarity	
मी नाही धन्य किंवा सुखी	-0.125, Negative polarity	
तो नाही आहे की, मी आनंदी नाही	0.125, Positive Polarity	
"तुम्ही कसे आहात?"	0.0 Neutral Polarity	
""मी दु: खी आहे"	0.0 Neutral Polarity	

After careful analysis of the test cases, the few discrepancies in the obtained results were mainly due to the following 3 factors:

- 1) Corpus word limit and Translation accuracy: If words are not present in corpus built, then online Yandex translator is used and sentiment is analyzed with English SentiWordNet. The freely available Yandex translator that system is using provides an accuracy in the range of 60-70%. Thus, especially for Asian, vernacular languages, the translation is only partially accurate. Thus, these inaccurate translations affect the overall polarity of the sentences which can result in minor discrepencies in the overall output. However, for smaller and simple sentences, the translation accuracy is comparatively higher than that of complex compound sentences. It was an observation that the framing of the sentences makes a difference in the sentiment analysis.
- 2) Limited scope of English SentiWordNet: The English SentiwordNet 3.0 which system is using to obtain the polarities of the respective words, also needs a lot of optimization. Many words which are actually of neutral polarity, are misclassified as adjectives having higher or lower polarity and hence either provide wrong polarities or tamper with the algorithm.
- 3) Non acceptance of special characters: This system perfectly analyzes and parses basic punctuation like commas, question marks, exclamation marks etc., but a special character like "" etc. is parsed along with the word, the system fails to recognize and process it and doesn't give any output. Hence there is a need to include a separate exception class to eliminate these special characters.

VI. CONCLUSION:

Sentiment Analysis has been quite popular and has lead to building of better products, understanding user's opinion, executing and managing of business decisions. With rapidly increasing technology, the early approach of word-of-mouth has been shifted towards the mass opinion what the people like and appreciate in majority. The rise in user-generated content for Marathi language across various genres- news, culture, arts, sports etc has opened the data to be explored and mined effectively, to provide

better services and facilities in terms of sentiment analysis. The scarcity of resources is one of the biggest challenges while dealing with sentiment analysis for Marathi language. This system focused on resource creation which includes building of an up to date corpus for Marathi language. The algorithm being proposed is used to give cumulative polarity using the Wordnet. This sentiment analysis model proposes a novel and effective approach to achieve desired functionality for Marathi language. This system mainly focused on creation of a vast and diverse up-to-date corpus, efficient mapping of data and generation of accurate sentiments for the data to erase the language barriers faced in the field of Sentiment analysis for Marathi.

VII. FUTURE SCOPE:

The scope of this system is limited for sentence level which can further be increased so as to analyze the sentiment for paragraphs and even larger text documents. This can be done, by firstly finding the individual polarity of the sentences using our suggested algorithm and then finding the net cumulative polarity of all sentences to obtain the overall polarity of the paragraph or document. There is a need for an optimized and accurate algorithm to do the same. Further limited size of corpus can be increased by not only considering adjectives but also verbs and nouns which are already present in the English Senti Word Net. This while evaluating the sentiment, better accuracy and results can be obtained. Also, increase the dictionary size by regularly updating it with new words. Further real time update of dictionary can be future research direction in the field of sentiment analysis of Marathi language.

REFERENCES:

- 1) Report 'Indian Languages Defining India's Internet', a study by KPMG in India and Google April 2017.
- 2) Sentiment analysis, https:// en.wikipedia.org/ wiki/ Sentiment_ analysis, accessed in Aug. 2016.
- Alessia D'Andrea, Fernando Ferri, Patrizia Grifoni, "Approaches, Tools and Applications for Sentiment Analysis Implementation", International Journal of Computer Applications (0975 – 8887) Volume 125 – No.3, September 2015
- 4) H. K. Walaa Medhat, Ahmed Hassan, "Sentiment analysis algorithms and applications: A survey," Ain Shams Engineering Journal (2014) 5, 1093–111, 2014.
- 5) S. S. Namrata Godbole, Manjunath Srinivasaiah, "Large-scale sentiment analysis for news and blogs," ICWSM'2007, Boulder, Colorado, USA, 2007.
- 6) G. Q. Soha Ahmed, Michael Pasquier, "Key issues in conducting sentiment analysis on Arabic social media text," IIT`13, 2013.

- 7) D. J. Hatem Ghorben, "Sentiment analysis of french movie reviews," Proceedings of the 7th Atlantic Web Intelligence Conference, AWIC 2011, pg. no. 19-28, 2011.
- 8) M. K. Yakshi sharma, Veenu mangat, "A practical approach to sentiment analysis of Hindi tweets," 1st International Conference on Next Generation Computing Technologies (NGCT), pg. no. 677-680, 2015.
- 9) Denecke, Kerstin. "*Using sentiwordnet for multilingual sentiment analysis.*" Data Engineering Workshop, 2008. ICDEW 2008. IEEE 24th International Conference on. IEEE 2008

