

DEPARTMENT OF INFORMATION TECHNOLOGY PILLAIS INSTITUTE OF INFORMATION TECHNOLOGY ENGINEERING, MEDIA STUDIES AND RESEARCH NEW PANVEL - 410 206 UNIVERSITY OF MUMBAI Academic Year 2015-16

Α

Stage One Dissertation Seminar On

'Sentiment Analysis of transliterated hindi and marathi script'

Ву

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Under the Guidance of

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Introduction

Motivation

Literature Survey

Proposed Approach

Application

Conclusion

Introduction

- Presence of translited text in Literature
- ► Transliteration as vital source of sentiments
- Current state of affairs on the subject
- Why it matters now?

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Motivation

- Reviews
- Discourse Analysis
- ► Feedback Analysis
- Other areas

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Paper	Author	Approach	Result	Limitation
Hindi sub-	Bakliwal,	Usage of	Resource	Doesn't
jective lexi-	Akshat,	sense based	for hindi	work on
con: A lexi-	Piyush	sentiment		translitera-
cal resource	Arora ad	analysis and		tion of hindi
for hindi po-	Vasudev	devlopment		text
liarty classi-	Varma	of HSWN		
fication [1]				

Paper	Author	Approach	Result	Limitation
A frame-	Pooja	Devnagiri	Usage of	Doesn't
work for	Pandey ,	sentiment	negati-	work with
sentiment	Sharvari	analy-	on and	hindi trans-
analysis in	Govilkar	sis using	discourse	literation
hindi using		HWSN	analyse	
HSWN [4]		and use of	improves	
		negation	the result	
		discourse	of polarity	
		analysis	detection	

Paper	Author	Approach	Result	Limitation
Automatic	Kundu,	Statistical	Accuracy	There is a
Detection	Bijoy and	method	upto nearly	great scope
of English	Chandra,	developed	72 percent.	of increa-
words in	Swarup	which is		sing the
Benglish		language		accuracy by
text: A		indepen-		improving
statistical		dent and		methodolo-
approach		can be used		gy.
[2]		to detect		
		any foreign		
		language.		

Paper	Author	Approach	Result	Limitation
Sentiment	Namita	Negation	Accuracy	Transl - ite-
Analysis of	Mittal,	and dis-	upto nearly	ration not
Hindi Re-	Basant	course	81 percent	covered.
view based	Agwarwal,	relation		Scope for
on nega-	Garvit Cho-	were iden-		accuracy
tion and	uhan and	tified and		improve-
discourse	Nitin Bania	HSWN im-		ment.
relation [3]		provement		
		carried out		

Paper	Author	Approach	Result	Limitation
Text Nor-	PYKL Sri-	Language	Accuracy	Negation
malization	niva and	identifica-	upto 85	and dis-
of Code	Shashank	tion and	precent	course
Mix and	Sharma	translite-		analysis
Sentiment		ration to		not being
Analysis [5]		devnagiri		performed.
		script and		
		then using		
		HSWN for		
		sentiment		
		analysis.		

Introduction

Motivation

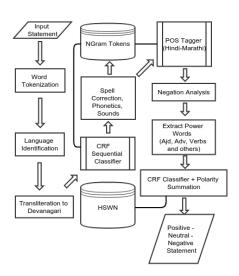
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Proposed Approach: Flow diagram



Proposed Approach

- 1. Text Normalization
- 2. NLP and Sentiment Analysis

Proposed Approach - Text Normalization

- 1. Normalization Process for hindi
 - 1.1 Spelling correction
 - 1.2 Ambiguous words
 - 1.3 Sounds
 - 1.4 Phontic words
 - 1.5 Transliteration
- 2. Dictionary method for marathi using bilingual lexicon

Proposed Approach - NLP and Sentiment Analysis

- 1. POS Tagging to identify nouns, adjectives and adverbs
 - 1.1 Use machine learning to figure out POS for hindi
 - 1.2 In case of marathi, use bilingual dictionary and then tag the statement
- 2. Adjective and Adverb extraction
 - 2.1 Easier to do with POS tagged statements.
 - 2.2 If POS doesn't work, then use lexicon lookup
 - 2.3 Use senti wordnet or HSWN for looking up polarity count
 - 2.4 If lookup fails, use extended wordnet api for getting sense
- 3. Negation and Classification using Classifier
 - 3.1 Figureout negation in statements and then invert the POS tagged adjectives and adverbs
 - 3.2 Semi supervised naive bayes/svm classifier to classify the polarity or -
 - 3.3 Simple summation of polarity will give a result to. Compare the two results

Proposed Approach: Algorithm for language identification

```
Arguments
w: word to identify language for
sentence: sentence to which word belongs
if Leh is None:
         Leh = Qeh(Lhi) for i in Lh
if I em is None:
          Lem = Qem(Lmi) for i in Lm
if not model.
         Model = CRF(FNGram(Leh), FNGram(Lem),
                                                            FN-
Gram(Le))
         if not w in D:
                  w = stem word(w)
                           if w not in D.
                                    w = find most similar word(D,
w)
I = arg max(Model, sentence, w)
return l
```

Algorithm for pos tagged with negetation

Algorithm steps

if max(tags in sentence) is english:

TaggedSentence = POSTaggerEnglish(sentence)

else max(tags in sentence) is hindi:

TaggedSentence = POSTaggerHindi(sentence)

return replace negetive phrases with antonyms(TaggedSentence)

Algorithm for polarity identification

```
Algorithm steps
languageTaggedSentence = (LanguageIdentifier(sentence, word) for
word in sentence).join('')
posTaggedSentence = POSTag(languageTaggedSentence)
polarWords = extractAdjectivesAdverbs(posTaggedSentence)
wordPolarity = dict()
for word in polarWords:
        if word tagged as english:
                 wordPolarity[word] = sentiwordnetPolarity(word)
        elif word tagged as hindi:
                 wordPolarity[word] = hindiSentiwordnetPolari-
ty(word)
        else:
                 hindiWword = hindiMarathiBilingualDictiona-
ry(word)
                 wordPolarity[hindiWord] = hindiSentiwordnetPo-
larity(hindiWord)
                                        return wordPolarity
```

Algorithm to classify polarity

Algorithm steps

Arguments

sentence: sentence to which word belongs

Variables and methods

NaiveBayesClassifier -¿ Trained to return polarity of the entire sentence given tokens with polarity values

LinearCalculation -¿ Simple Summation based polarity classifier

word Dictionary = Polarity Indentification (sentence)

$$\label{lem:condition} \begin{split} \text{return} & \text{NaiveBayesClassifier(wordDictionary)}, & \text{LinearCalculation(wordDictionary)} \end{split}$$

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Conclusion

- 1. Accuracy of 95 percent on target
- 2. Imporvement of transliteration sentiment analysis in general
- 3. Usage of pre trained models over on the fly dynamic models for peformance gains

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