# Rule based POS Tagger for Sanskrit

Sharada Adinarayanan, J. Naren, P. Sriranjanie and Dr.G. Vithya

Abstract--- POS tagging is a process of attaching each word in a sentence with a suitable tag from the given set of tags. In the paper, rule based view of NLP is taken up for tagging the part of speech for Sanskrit words. The foundation for POS tagging is morphological analysis. The twelfth chapter of Bhagavad Gita is considered as input for POS tagging process. Annotated corpora will be developed and used for retrieving the grammatical category of the input text. Sanskrit is a language with very concrete grammar proposed by Panini (4000.B.C) and has layered grammatical structure. Thus rule based approach would fulfill the tagging process rather than stochastic or probabilistic approach (existing system).Therefore, the project aims to improve the accuracy by utilizing the efficient lookup strategies, searching and sorting techniques and finally rule formations(utilizing the richness of Sanskrit grammar) to quickly narrow down the assignment of grammatical category to words. The major challenge is the tokenization process of joined words. Since Sanskrit has many inflected noun and verb forms, identifying the correct grammatical category involves contextual meaning and semantics to be taken into view. Also, semantic analysis, derivative analysis and Sandhi analysis is done.

Keywords--- Annotated Corpora, Tokenization, Morphological Analysis.

# I. INTRODUCTION

Parts-of-Speech Tagging is a process of assigning appropriate tags to each word in a given unprocessed sentence. Parts -of -speech for a language according to the grammar can be categorized as noun, verb, adverb, pronoun, adjectives, conjunctions and their remaining categories. POS tagger performs all operations by referring dictionaries, rules and tag set.

The ambiguity is resolved by using probabilistic method. There are three techniques in POS tagging: Rule-based Model, Stochastic Model and Hybrid Model. Rule-based Model is one of the oldest approaches, which uses a set of Hand-coded rules to assign the appropriate tags for the given input. The accuracy remains high if more number of tagged words and hand coded rules are used. The Stochastic Model uses the concept of picking the most appropriate tag word in the unprocessed text. The various techniques used in stochastic model are Hidden Markov Model, Maximum Entropy Markov Model and Conditional Random Field Model. Hybrid model uses the combination of both Rule-based as well as stochastic model wherein the hand-coded rules are combined with techniques present in the Stochastic Model.

Sharada Adinarayanan, B.Tech Computer Science and Engineering, School of Computing, SASTRA Deemed University, Thanjavur, Tamil Nadu, India. E-mail: sharadha.adinarayanan@gmail.com

J. Naren, Assistant Professor, School of Computing, SASTRA Deemed University, Thanjavur, Tamil Nadu, India. E-mail: naren.jeeva3@gmail.com

P. Sriranjanie, B. Tech Computer Science and Engineering, School of Computing, SASTRA Deemed University, Thanjavur, Tamil Nadu, India. E-mail: nsriranjaniee@gmail.com

Dr.G. Vithya, Professor, School of Computing, KL University, Vijayawada, AP, India. E-mail: vithyamtech@gmail.com

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## **II.** LITERATURE SURVEY

The paper by A.J.P.M.P.Jayweera.et.al has proposed Hidden Markov Model based part of speech tagging and lexical semantics for Sinhala Language. Sinhala is the native language of ethnic group of Sri Lanka, spoken by more than seventeen million people. Sinhala is one of the morphologically rich languages of great complexity, and with various kinds of words in inflected form, for which grammatical tagging is very essential. The tag set comprises of 26 tags. Sinhala Text corpus containing 90551 words from 2754 sentences gathered from Sinhala newspapers. Beta version of corpus developed by UCSC is used. Virterbi algorithm is used for finding the best tag sequence based on the text corpus. Input to the algorithm is a string of words and tag set, the output is single best tag for each word. Supervised machine learning approach is used for training the tagger to find the transition probability of the given sequence of words. Statistical based approach is used and an accuracy of about 90.91% for known words is achieved. [1]

The paper by R Muni Prashanthi.et.al, have presented Implementation of Tree Tagger for Sanskrit. Annotation of text with POS and lemma information is done using Tree tagger tool. Tree Tagger involves two phases which are training and testing phases. Three files are used in training phases consisting of open class file, parameter file and lexicon. Two programs namely training and tagger programs are involved in the Tree Tagger. Tag set developed by J.N.U, New Delhi for Sanskrit is used. The training set was annotated using a tag set consisting of 134 tags. Errors were investigated and the reasons were analyzed. Tree tagger for Sanskrit yielded above average results though not better than the performance for English language. [2]

The paper by Pallavi Bagul.et.al, Archana Mishra, deals with POS Tagging of Marathi language text. The word in a given sentence is tagged by using rule based approach. Tokenization is used to split the given string of inputs into tokens, and the result of this process i.e. tokens, are compared with WordNet to assign the correct POS Tags for each token. Ambiguity in Marathi is resolved by using Marathi grammar rules. WordNet, corpus, tagset has been used as a database for providing correct POS tags to every word in a given sentence. Accuracy can be increased by increasing the number of tagged words in WordNet. [3]. The paper by Kanak Mohnot .et.al has presented A Hybrid Based Part of Speech Tagger for Hindi. Corpus of over 80,000 words having 7 standard different part of speech tags is evaluated by the proposed system for Hindi Language. Also accuracy of the proposed tagger is determined. The input words are searched for in the database and are tagged if present, else HMM model is used. The basic dataflow of the proposed system involves:-accepting user text, tokenizing the text, converting them to singular form, applying rules if word is found in database else assign category using Hindi NER then apply rules by predictive analysis else if multiple Tag condition is encountered, then Hindi NER and HMM are used. The average accuracy the system has attained for POS Tagging is 89.9%. [4]

#### **III. PROBLEM ASSESSMENT**

The project deals with using Rule-Based technique of NLP (Natural language Processing) to perform POS (Parts-Of- Speech) Tagging for Sanskrit Text by assigning appropriate tags according to the correct grammatical categories to the words in unprocessed sentences. Unavailability of considerable amount of annotated corpora of sound quality for South-Asian languages like Sanskrit and tokenization of joined words are the major challenges.

# **IV. PROPOSED SYSTEM**

The proposed system uses Rule-based model for POS tagging. The objective of the study is to analyze and improve accuracy of POS Tagging process for Sanskrit words in an unprocessed sentence by using Rule-Based technique of NLP wherein hand-coded rules are used for assigning the tags according to Sanskrit Grammar. [5] Since rule-based approach is proposed, if specific rules are unavailable for tagging a specific category of words, tagging process does not take place as compared to the stochastic model (existing system) where probabilistic frequency is taken into account even if rules are unavailable and tags are assigned. The assumed rule may or may not be correct although in most cases it is correct. Rule-based Approach removes the probabilistic nature of stochastic model and assigns rules only if available thus improving the correctness of the tagged words in an unannotated Sanskrit Text. [8][9]

# V. SYSTEM DESIGN

The design of the system is based on the following Data- Flow diagrams. The proposed work is divided into 4 individual models assisting POST for Sanskrit. The first module is POS Tagger Module, the second is Semantic Analyzer Module, the third is Derivative Analyzer and the Fourth is Sandhi Analyzer. The first illustrates data-flow diagram for POS Tagger.



Figure 1.5.1: Data Flow Diagram for POS Tagging

The second illustrates the data-flow diagram for Semantic Analyzer.



Figure 1.5.2: Data Flow Diagram for Semantic Analysis The third illustrates the data-flow diagram for Derivative Analyzer







Figure 1.5.4: Data Flow Diagram for Sandhi Generator

## **VI. SYSTEM IMPLEMENTATION**

A Rule-based technique for POS Tagging for Sanskrit Text is proposed. A specific domain, Sanskrit Slokas from 12<sup>th</sup> chapter of Bhagavad Gita is chosen. Hand-coded rules are used for the tagging purpose with No specific Algorithm is usage. Derivative analysis, Semantic analysis and Sandhi analysis is also done.

# POS Tagger

A Rule-Based POS tagger for Sanskrit is proposed. Sanskrit language has concrete grammar rules. Unannotated Sanskrit text is taken as input, tokenized into words and processed to produce tagged Sanskrit words as output. The 12<sup>th</sup> chapter Slokas from Bhagavad Gita is taken as input text for Part-Of-Speech Tagging. The Slokas are categories according to their Sanskrit Grammar and tags are assigned. [1][6][7]

The steps followed for POS Tagging are as follows:

Step1: Stemming – Manual Stemming is done. The stemmed words are populated in database. Step2: Suffix-Stripping-The end string or simply the suffix of each word is stripped. The hand- coded rules are checked with the suffixes.

Step3: Tag Assignment- The appropriate tags are checked for in the database and in the program to assign the correct tags to the words in sentence.

not =	Adverb •	Prefix + pary	Noun + satatayukta	Verb ·	ye	Pronouns + aham	sarvaani
- 1901 140F	- North 1	1.1					
	api	ati	bhakta	358	tesaam	mayi	etat
	ty	abhi	aksara	Kara	yen	maam	idam
	niveshaya	prati	avyakta	kru	sah	tvaam	yasmaat
	sarvatra	pra	yogavittama	hrusha	yuh	rama	maam
	evani	upa	mina	shoce	me	rama	rama
	hi	athi	nityayukta	64	sarva	rama	rama
	atah	94	shraddhayaa	aaveshya	0e.	rama	rama
	atva	pira	upeta	rataah	ke	rama	rama
	nə	pra	yuktatama	aasakta	mad	rama	rama
	uurdhvam	vi	mata	avaapyate	rama	rama	rama
	atha	ei	anirdeshya	shyaayanta	rama.	rama	rama
	samaadhaatun	anu	sarvatraga	V85	rama	rama	rama
	sthiram	ava.	kuutastha	samshayah	rama	rama	rama
	tatah	pra	achala	shaknosi	rama	rama	rama
	aaptum	ei .	drova	iccha	rama	rama	rama
	sannyasiya	upa	endriyagraamu	ase	rama	rama	rama
	madartham	54	samabuddhi	tur	rama	rama	rama
	kartum	adhi.	sarvabhuutahi	NID	rama	rama	rama
	anantaram	ati	kleshah	arpi .	rama	rama	rama

Figure 1.6.1: The POS Database

#### Semantic Analyzer

Semantics refers to finding the meanings of words in a sentence. Sanskrit words with their meanings are fed in database table. Words from the Sanskrit Slokas of the 12<sup>th</sup> chapter of Bhagavad Gita are taken as input for which semantic analysis is done. [2]

The steps followed for Semantic Analysis is as follows: Step 1: Get a word as input. Step 2: Check if the word is present in the database. Step 3: Print its meaning

Step 4: If word is not present, check if the root is present in the database.

Step 5: Strip the suffix if root is present in database and perform Semantic analysis

meaning1	\$ root1 +	means +	
meaning1 : Table	dhava	jump	
	yaacha	beg	
	bhaja	chant	
	aas	desire	
	aap	get	
	bhava	to be	
	evam	thus	
	ye	who	
	tu	Indeed	
	dharmaamrta	immortal dhar	
	idam	this	
	yathayukta	as declared	
	paryupaasate	follow	
	shraddadhaan	endued with s	
	matparamaah	regarding me a	
	bhaktaah	devotees	
	te	they	
	atiiva	exceedingly	
	me	to me	
	priyaah	dear	
	yah	who	
	na	not	
	hrsyati	rejoices	
	Record: 14 1 of 1	37 F H HI W. No Filler	Search

Figure 1.6.2: The Semantic Analyzer Database

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#### Derivative Analyzer

Derivatives are also known as Kridantas in Sanskrit. The root words and the corresponding derivatives are fed in the database. The derivatives for the parasmaipadi and aatmanepadi forms are displayed as output. [4]

The steps followed for Derivative Analysis is as follows: Step 1: Get a word as input. Step 2: Check if the word is present in database.

Step 3: If the word is present, the derived words of the input root word are displayed.

derivation	11	Root	Pressrit. =	Add New Field
derivation : Table		gaaya	ti	
derivation1	A	hrsyati	rama	
derivation1 Table		hrusha	rama	
derivation2	8	techa	rama	
derivation2 : Table		Jaya	tah	
derheatlen 3	8	Kaarriksa	1.01111.0	
derivations : Table		kara	rama	
derivationd	8	khada	anti	
derivation4 : Table	- 22	keitda	51	
derivation3		kru	rama	
	2	kur	rama	
derivation5 : Table		likha	thah	
derivation6	*	nama	tha	
derivations : Table		nanda	aami	
		patha	aavah	
		rakalta	aarnah	
		rataah	rama	
		samsbayab	rama	
		shaknosi	FAITTA	
		shoca	rama	
		smara	rama	
		sech	rama	
		ud	rama	
		Record: H = 1 of	545 P PL PUT	S. Hu Filler Search

Figure 1.6.3: The Derivative Database

#### Sandhi Analyzer

Sandhis are syllables that help in joining two distinct meaningful words to produce another meaning word. Sandhis mainly occur in many Indian Languages. Sanskrit consists of complex Sandhi rules. There are three main categories of Sandhi namely: Swara Sandhi, Vyanjana Sandhi and Visaraga Sandhi. The Sandhi module deals with Basic Swara Sandhi Generator. There are 8 types of Swara Sandhi. [3]

The steps followed for Sandhi generator is as follows: Step 1: Get two input strings. Step 2: Apply Basic Swara Sandhi Rules.

Step 3: Display the output string after Sandhi processing.

## Graphical User Interface

Java Jframes are used to design Graphical User Interface which is easy to use and User-friendly. The Graphical User Interface is designed for all four modules. The different parts of the GUI for different modules are shown in the screenshots below.

Sanstarit side meat invol	n satatavaida ve tihaidaas tvaam parvupiaasate	
	n zastopusta je struktas, takan parjupakče osnecilos Nadel You entered string - evam satatajvilta ye bhaltasa-tusam parjupaasale Nam Koven Settapolta sambodan vibhašti, singular, masodine satatajvilta e osnori, mazodine ye pratonajkajstani vibhašti, dvaljalogular, masodine financie-dual-potama satatajvita mastasa Noue-baltasa vasta Pretoun - s. tuam dollja vibhašti, singular, masodinejneutral avjatta dollja vibhašti, singular, financie samusa gadi, present, 3rd pencen, singular ana	

Figure 1.6.5: POS Tagger

and SERVICENTEC INF	and WSPE		In provide the state of the state
			incorrection and styram contents
And and a second	You entered sking of	také Nehathan ma develéné Antérés an ésté départenté Martin anter a ser ésté de la partenté antre anter a ser ésté de la partente de la parte de la partente de la partente de la partente de la partente	And a second sec

Figure 1.6.6: Semantic Analyzer

apart hours	destrotive analysis	A CONTRACTOR OF
	enclosed analysis	exclusion
For Acture territor of partmental parts Incoderputs Incoderputs Incoderputs Incoderputs Incoderputs Incoderputs Incoderputs Incoderputs Incoder In		

Figure 1.6.7: Derivative Analyzer

second string: upamam	
Connection Made: You entered stong2 : atrma This is Deerga Sandhi: parama You entered stong2 : atrma You entered stong1 : prakuthya You entered stong1 : prakuthya You entered stong1 : exacuthyarva Na Na You entered stong1 : at You entered stong2 : aeta This is You Sandhi. Upaetat You entered stong2 : aeta You entered stong2 : amruta You entered stong2 : upamam	
	Connection Made Connection Made Connection Made Connection Made You entered String2 attrma This is Deerga Sandhi: prakuthya You entered string3 : prakuthya You entered string3 : prakuthya This is Viruthi Sandhi: prakuthyawa You entered string2 : attai You entered string2 : attai You entered string3 : attai You entered string3 : upamati Amouta Sandhi: prakuthyawa You entered string3 : attai You entered string3 : attai You entered string3 : attai You entered string3 : attai You entered string3 : amouta You entered string3 : amouta You entered string3 : attai You entered string3 : attai Partial : att

Figure 1.6.8: Sandhi Analyzer

## A. Output for POS Tagger is as given below

You entered string: evam satatayuktaah ye bhaktaas tvam pary upaasate Evam-Adverb Satatayuktaah-Noun sambodan vibhakti, singular, masculine satatayukta-root Ye-pronoun, masculine Bhaktaas-NounPratama vibhakti, singular, masculine Tvaam-Pronoun, same (gender) Pary-verb Upaasate-aatmane padi, present, 3rd person, singular asa-root.

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#### B. Output for Semantic Analyzer is as given below

You entered string: mad bhaktah-my devotee You entered string: bhaktimaan-full of devotion You entered string: shraddadhaana-endued with shraddha You entered string: dharmaamrta- immortal dharma **Output for Derivative Analyzer is as given below:** You entered string: shaknosi.

The derivative is For present tense for parasmai padi Shaknositi Shaknositah Shaknosianti Shaknosisi Shaknosithah Shaknosiaami Shaknosiaavah Shaknosiaamah Similarly, for past, present future of parasmai and aatmane padi is done.

## C. Output for Sandhi Analyzer is as given below

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You entered string1: parama You entered string2: athma This is Deerga Sandhi: paramaathma

You entered string1: prakuthya You entered string2: eva This is Viruthi Sandhi: prakuthyaiva

#### Test Cases:

TEST CASES	INPUT	OUTPUT	RESULT
POS Tagging	mayi manah ye maam	mayi pronoun - same mayi manah noun manah pratama vibhakti, singular, masculine mana ye pronoun-masculine ye pratma(saptami vibhakti, dual(singular, masculine)feminine-dual-pratama maam pronoun-same maam dutiya vibhakti, singular, masculine)neuter dutiya vibhakti, singular, feminine	PASS
Semantic Analysis	Pata	Read	PASS
Derivative Analysis	Pata	All DERIVATIVES OF Pata in Present Tense patati patatah patanti patasi patathah patatha pataami pat aavah pataaamah	PASS
Sandhi Generator	Ganesha astakam	Ganeshaashtakam	PASS

Table 1.6.1: Test Cases for POS Tagging

## Accuracy Calculation: POS Tagging:



Figure 1.6.7.1: Pie Chart for POS Tagging

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Figure 1.6.7.3: Pie Chart for Derivative Analysis



Figure 1.6.7.4: Pie Chart for Sandhi Generator

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