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## Design of Derivational Morphological Analyzer for Kannada Language

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### Abstract

We have developed a derivational morphological analyzer for Kannada Language. Derivational morphology deals with change of part of speech (POS) category from one basic category to another by the addition of derivational suffixes to the basic categories like noun, verbs and adjectives. Nouns can be derived from verbs and verbs can be derived from nouns and so on. There is no derivational morphological analyzer exists for Kannada. Existing systems have attempted only the inflectional morphology. The process of derivation is regular and productive in many instances for Kannada. But this is not true in all cases. Verbalizers are used in the process of derivation to verb. Nominalizers are used in derivation verb to noun. Finite state transducers are used for the implementation of the derivational analyzer. A set of verbs is

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Bhuvaneshwari C Melinamath, Ph.D. Scholar

Design of Derivational Morphological Analyzer for Kannada Language

used to derive verbs from adjective. Another set of pronoun suffixes are used to derive nouns form adjectives. The accuracy of derivation analyzer is around 90% in the case of nouns and around 85% in the case of verbs.

**Keywords**— Part of Speech (POS), Natural Language Processing (NLP), Finite State transducers (FST).

## **1. Introduction**

Natural Language Processing (NLP) is an area which is concerned with the computational aspects of the human language. The goal of the NLP is to analyze and understand natural languages used by humans and to encode linguistic knowledge into rules or other forms of representation. Statistical and machine learning algorithms have taken a lead over complex linguistic grammar. There has been a great progress in natural language processing, through the use of statistical methods trained on large corpora.

Morphology in the generative perspective started with the publication of Chomsky's remarks on nominalization (1970). Later on the foundation for generative morphology was laid by Mark Aronoff in his book 'Word formation in generative grammar'. According to Aronoff (1976) all regular word formation process are word based. A new word is formed or derived by applying a regular rule to a single already existing word. Both the new word and existing one are members of major lexical categories. Word formation rule operates on a single word, not on phrase or bound morpheme. Both input and output of WFR must be members of a major lexical category such as noun, verb, adjective, and adverb. Derivation on the other hand is a phenomenon which derives new words; derivation may be by adding an affix or by compounding. Words are formed from a combination of one or more free morphemes and zero or more bound morphemes. Free morpheme is units of meaning, which can stand on their own as words. Bound morphemes are also units of meaning; however, they cannot occur as words on their own: they can only occur in combination with free morphemes. From this definition, it follows that a word is either a single free morpheme, or a combination of a single free morpheme with other free and bound morphemes. The remaining part of the paper is organized as section2

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Bhuvaneshwari C Melinamath, Ph.D. Scholar

Design of Derivational Morphological Analyzer for Kannada Language

deals with literature survey, section 3 deals with morphology and its types. Section 4 deals with results and section 5 deals with conclusion.

## **2. Literature Survey**

In general there are several approaches attempted for developing morphological analyzer and Generator worldwide. Kimmo Koskenniemi (1983) has developed a two-level morphology formalism for Finnish language. Beesley (1996), has developed an Arabia finite state transducer for MA using Xerox finite state transducer (XFST). In case of Indian languages, (AU-KBC, 2010) the Research Centre of Anna University developed a finite state automata based morphological analyzer for Tamil language. Previous works show that very little work was done in computational aspect of Kannada: Prof. Kavi Narayan Murthy “Network and Process Model” handles only inflectional morphology; IIT Hyderabad and Department of CALTS have done some work in respect of Kannada Morphology, but the approach is based on paradigm, i.e. suffix list based. Performance of the system is limited by the size of the dictionaries and derivational morphology is not handled. E. Roche and Y. Schabes (1997) have shown that Finite state technique is a vital tool in the implementation of natural language morphology.

From The literature survey, it is clear that derivational morphology of Kannada is not exploited to the extent it is necessary to handle this aspect, which otherwise hinders the NLP applications. In this paper an attempt is made to cover both the inflectional and derivational morphology for Kannada nouns and verbs. Morphology aspects are discussed in the next section.

## **3. Morphology**

In linguistics, morphology is the study of the internal structure and the transformational processes of words. The internal structure of words is its morphemes. Each morpheme is an individual unit of meaning. Morphology is classified as inflection and derivational morphology.

### **3.1 Inflectional Morphology**

This is the most significant part of morphology from the syntax point of view.

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Bhuvaneshwari C Melinamath, Ph.D. Scholar

Design of Derivational Morphological Analyzer for Kannada Language

Inflectional morphology deals with words which are syntactically significant word forms projected towards morpho-syntactic functions, often overtly marked by corresponding bound morphemes. Inflection involves formation of a distinct word form of a given word, words are formed by the base or root words by the addition of certain affixes, to express certain grammatical relationships and functions. The changes are with respect to gender, number, case markers, personal suffixes in case of noun and pronoun, tense, aspect, and modal categories in case of verb. Inflectional categories vary with respect to language. For example, English does not have gender as an inflectional category, since its subject verb grammatical agreement rule does not require the gender information of the subject noun, but only person and number of the subject. In contrast, Kannada requires gender information along with person and number to mark the verb form to show the agreement with the subject. Inflectional words are words which are used in syntax to compose sentences, when different forms of the same lexeme are generated, and there is no change in category.

### **3.2 Derivational Morphology**

Derivation is the process of formation of new words or inflect able stem from another word or stem. It typically occurs by the addition of an affix. The derived word is often of a different word class from the original. It may thus take on the inflectional affixes of the new word class. Derivation has the following characteristics:

- Typically produces a greater change of meaning from the original form.
- Is more likely to result in a form which has a somewhat idiosyncratic meaning.
- Often changes the grammatical category of a root
- Kindness is derived from kind.

Here are some kinds of derivational operations:

- Operations that change the grammatical category of a root. Verbs and adjectives can be turned into nouns: Consider the examples like amaze → amazement, speak → speaker, perform → performance, soft → softness, warm → warmth
- Operations that change the valence (transitivity) of a root.

### 3.3 Kannada Derivational Morphology

Derivation morphology deals with change of category from one to another. Noun is derived from verb or adjective, as we have shown a sample of the addition of suffix. VERB + v saMdhi.

**Nominalizers:** nominalizers are used to derive nouns from verbs. Consider an example:

- Use of derivational suffix “vike”  
maaDu ‘do’ (verb) + vike → maaDuvike (noun) “process of doing”.
- Causation Kannada suffix “isu”  
prasne ‘question’ + isu → prasnisu is verb which means questioning.

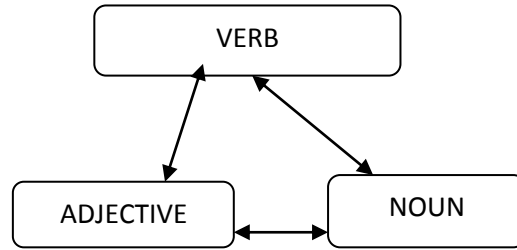


Figure 1: Derivation Morphology

**Verbalizes** A set of verbs used to derive verbs from nouns and adjectives. There is a set of verbs like ‘aagu’, biDu, iru, goLLu, aaDu, hoDe, maaDu which are used in deriving verbs from nouns.

Noun+verb=Verb. Consider the following examples.

gaaDi+hoDe= gaaDihoDe

nidde+maaDu=niddemaaDu

- Use of verb (aagu) ‘become’ in verbalization.

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Bhuvaneshwari C Melinamath, Ph.D. Scholar

Design of Derivational Morphological Analyzer for Kannada Language

Most of the nouns in Kannada are derived as verbs by adding verb aagu to basic noun roots. The following examples illustrate the derivation Process.

- Noun to verb derivation using (aagu) verbalizer.

a) peTTu (Injury) 'noun' + aagu 'become/happen' → peTTaagu

maLe (rain) 'noun' + aagu → maLeyaagu (raining)

beLe (crop) 'noun' + aagu → beLeyaagu (cropping)

paasu (pass) 'noun' + aagu → paasaagu (paas)

ishTa (like) 'noun' + aagu → ishTavaagu (like)

- Adjective to Verb Derivation Process

b) haaLu 'adjective' + aagu → haaLaagu (get ruined)

suMdara 'beautiful' + aagu → suMdaravaagu (become beautiful)

- Use of Verb aagu in Complex verb Derivation

In Kannada we have a set of verbs which do not have aspect and tense inflections. Like beeku (want), beeDa (donot want), kuuDadu (no), bahudu etc. By adding verb 'aagu' to these verbs aspect and tense inflections can be added.

- Usage of aagu with Modal auxiliaries

c) beeku (want) 'verb' + aagu → beekaagu (wanted). Now beekaagu is a compound verb

- We had another set of conflicts like whether maataaDu 'talking'(maatu + aaDu is noun+verb) is compound should be stored in dictionary, or generated by morph system by adding verbalizes like aagu iru, aaDu, paDu, goLLu etc. finally We decided

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Bhuvaneshwari C Melinamath, Ph.D. Scholar

Design of Derivational Morphological Analyzer for Kannada Language

to handle it through morph system.

Kannada language has complex morphology, many suffixes get added to form the complex inflected root. Consider the formation of verbal noun in example shown below:

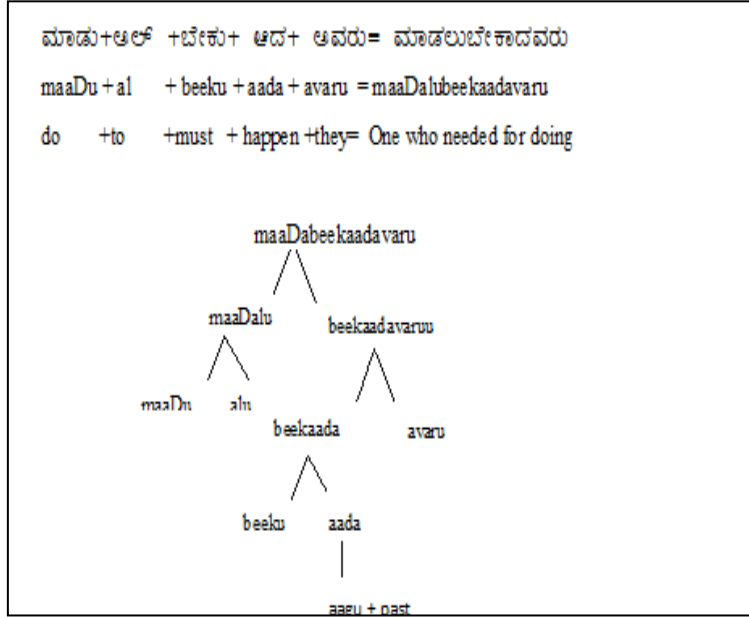


Figure 2. Sample Derivation Tree of Verbal Noun (Gerund)

### 3.4 Finite-state Transducers

Finite State Automaton (FSA) is a model of computation consisting of a finite set of states, a start state, an input alphabet, and a transition function that maps input symbol and current state to next state. A state transition usually has some rules associated with it that dictate when the transition may occur. The result given by a FSA is limited: either the string is accepted, or it is rejected. The transducer is defined as  $T = (Q, L, \delta, qI, F)$  where  $Q$  is a finite set of states,  $L$  a set of transition labels,  $qI \in Q$  the initial state,  $F \subseteq Q$  the set of final states, and  $\delta: Q \times L \rightarrow 2^Q$  the transition function (where  $2^Q$  represents the set of all finite sets of states). The set of transition labels is  $L = (\Sigma \cup \{\epsilon\} \times (\Gamma \cup \{\epsilon\}))$  where  $\Sigma$  is the alphabet of input symbols,  $\Gamma$  the alphabet of output symbols, and  $\epsilon$  represents the empty symbol. On next input symbol, it can move to state 2 without consuming any input symbols, and thus there is an ambiguity: is the system in state 1 or state 2, before consuming the letter a. Because of this ambiguity, it is more

convenient to talk of the set of possible states the system may be in. Thus, before consuming letter a, the NFA-epsilon may be in any one of the states out of the set {1, 2}. Equivalently, one may imagine that the NFA is in state 1 and 2 'at the same time': and this gives an informal hint of the power set construction  $2^Q$ . The following table shows the transition table for Kannada derivational Morphology:

v-verb, n-noun, adj-adjective:

Transition State	Derivation Category	Derivation Suffix
25-26	v>n	oosuga
35-4	v>n	udu
17-3	v>n	vike
# Gerund = tense + udu, negative are like tense too:		
35 4	v>n	udu gerund
# Nominalizers:		
17 3	v>n	vike
# Relative Participles:		
# Adjective has no morphology as such. Derived adjectives behave		
# like nouns, morphologically.		
20 50	v>adj	(a) rp
26 4	v>n	takkaddu→ compulsive_gerund
26 50	v>adj	takk(a)→ rp_compulsive
26 4	v>n	aaraddu gerundn_capabilitative
26 50	v>adj	baarad(a)→ rp_prohibitive
30 50	v>adj	illada→ rp_negative
30 4	v>n	illaddu→ gerund_negative
35 4	v>n	baaraddu→ baaradu_gerund
35 4	v>n	lolladdu olla_gerund
35 50	v>adj	loll(a) rp_olla
35 50	v>adj	lollada rp_ollada
50 94	adj>adv	aaga particle_aaga
#50 94	adj>adv	aagalee particle_aagalee
50 94	adj>adv	aagaloo particle_aagaloo
50 94	adj>adv	aagina particle_aagina
50 94	adj>adv	meele particle_meele
50 94	adj>adv	naMtara particle_naMtara
50 94	adj>adv	baLika particle_baLika
50 94	adj>adv	oDane particle_oDane
50 94	adj>adv	atta particle_atta
50 94	adj>adv	haage particle_haage
50 94	adj>adv	aMt(e) particle_aMte



Transition State	Derivation	Derivation Suffix
50 94	adj>v	aMtilla existential_negative
50 94	adj>v	haagilla existential_negative
50 4	adj>n	ashTu particle_ashTu
50 3	adj>n	alli particle_alli
50 0	adj>n	eDe particle_eDe
52 94	adj>adv	aaga particle_aaga
#52 94	adj>adv	aagalee particle_aagalee
52 94	adj>adv	aagaloo particle_aagaloo
52 94	adj>adv	aagina particle_aagina
52 94	adj>adv	varege particle_varege
52 94	adj>adv	tanaka particle_tanaka
52 94	adj>adv	maTTige particle_maTTige
52 94	adj>adv	atta particle_atta
52 94	adj>adv	haage particle_haage
52 94	adj>adv	aMt(e) particle_aMte
52 100	adj>adv	vareginaparticle_varege
52 94	adj>v	aMtilla prohibitive
52 94	adj>v	haagilla prohibitive
52 4	adj>n	ashTu particle_ashTu
52 3	adj>n	alli particle_alli
52 0	adj>n	eDe particle_eDe
# Adj+Pron=Pron		
50 100	adj	aMtaha similaritive
50 3	adj>n	aMtahadu similaritive_pr_adu
50 3	adj>n	aMtahavu similaritive_pr_avu
53 3	adj>n	avanu
53 3	adj>n	avanobba(nu)
53 3	adj>n	avanaata
53 3	adj>n	avaniita(nu)
53 3	adj>n	aata(nu)
53 3	adj>n	aake(yu)
53 3	adj>n	avaLu
53 3	>n	avaLobbaLu
53 3	>n	avaLaake
53 3	adj>n	avaLiike(yu)
53 100	>v	avaLillade
53 3	adj>n	avaru
53 3	adj>n	avarobbaru
53 3	adj>n	avarellaru
53 3	adj>n	avaraaru
53 100	adj>v	avarillade
53 3	adj>n	avugaLu pr_avu
53 3	adj>n	ellavugaLu pr_avu
100 0	n>v	ennu +v_saMdhi
100 0	v	ennu +v_saMdhi
100 0	n>v	uMTu +v_saMdhi
100 0	n>v	aagu +v saMdhi

Box 1. Derivation Morphology Continued.

100	0	n>v	iru	+v_saMdhi
#100	0	adj>v	aagu	+v_saMdhi
#100	0	adj>v	iru	+v_saMdhi
100	0	adv>v	aagu	+v_saMdhi
100	0	adv>v	iru	+v_saMdhi
# External saMdhi with iru forms:				
98	100	any	ide	+v_saMdhi_ide
98	100	any	ive	+v_saMdhi_ive
98	100	any	iddu	+v_saMdhi_iddu
# External saMdhi with sub_conj etc:				
#98	100	any	aadudariMda	+sub_conj

### Box 1. Derivation Morphology Continued

### 3.5 Compounding

Compounding is a word formation process, whereby two or more words are combined to produce a new word. The meaning of the resulting word cannot be understood as the combination of meaning of combined words. Compounds yield new meaning, even though they are made of existing words, for example **greenhouse** is 'a building for plants', but not a house which is green in colour. In linguistics, word formation is the creation of a new word. Word formation is sometimes contrasted with semantic change, which is a change in a single word's meaning. The line between word formation and semantic change is sometimes a bit blurry; what one person views as a new use of an old word, another person might view as a new word derived from an old one and identical to it in form. Word formation can also be contrasted with the formation of idiomatic expressions, though sometimes words can form from multiword phrases.

Compounding involves two or more words rather than affixes. Syntactic rules are not involved in the internal structure of compound words. In syntax, morphologically complex words are not different from simple words. Internal structure of compounds is inaccessible to syntax.

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Bhuvaneshwari C Melinamath, Ph.D. Scholar

Design of Derivational Morphological Analyzer for Kannada Language

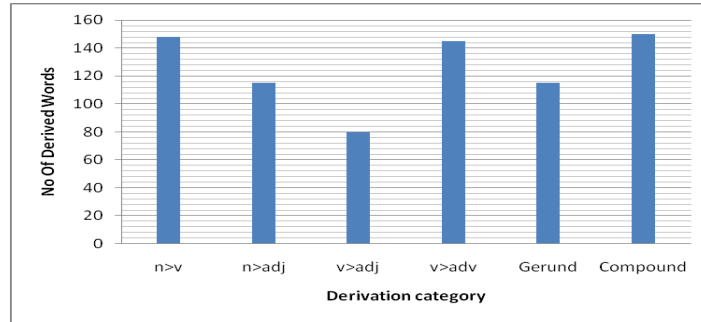
Compound agrees with derivation in not allowing inflection inside. Hence compounds are syntactic atoms like other words. Processing of meaning of compounds from the constituents is not always easy: noun+noun compounding like redfort “ keMpukoTe”.

#### 4. Experiments and Results

We have considered DoE CILL corpus for experimentation, a sample file Bank1.aci.out is considered, the size of the file is 1084 words. We observe the following derivation types in the corpus. More than 50 % words in the corpus are derivational types. This shows the importance of handling the derivational morphology.

Derivation Type	No of words
n>v	148
n>adj	115
Gerund	115
Compound	150
v>adj	80
v>adv	145

**Box 1: Derivation Types in Bank1.aci.out**



**Figure 3. Derivation Category in Bank1.aci.out File**

$$Recall = Tp / (Tp + fn) \dots\dots\dots(1)$$

$$Precision = TP / (TP + FP) \dots\dots\dots(2)$$

$$F - Measure = 2 * Precision * Recall / Precision + Recall \dots\dots(3)$$

**Table 1. Showing Recognition Efficiency**

<b>Total Derivation Words</b>	<b>842</b>
Correct Recognized	700
Wrong Recognized	53
Not Recognized	89
Recognized	753
Precision	95%
Recall	92%
Fmeasure	93%

## 5. Conclusion

We have attempted to develop a derivational morphological analyzer for Kannada, this tool is important in morphological process. Morphological analyzers and generators are useful in many NLP applications, like machine translation, parsing systems etc. Kannada language, having a complex morphology, designing analyzers or generator is a challenging job, due to agglutinative nature of Kannada.

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Design of Derivational Morphological Analyzer for Kannada Language

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