# A DEEP SYNTACTIC PARSER FOR THE TAMIL LANGUAGE

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Degree of Doctor of Philosophy

Department of Computer Science & Engineering

University of Moratuwa Sri Lanka

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Thesis submitted in partial fulfillment of the requirements for the degree of Doctor of Philosophy

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### **DECLARATION**

I declare that this is my own work and this dissertation does not incorporate without acknowledgement any material previously submitted for a Degree or Diploma in any other University or institute of higher learning and to the best of my knowledge and belief it does not contain any material previously published or written by another person except where the acknowledgement is made in the text.

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The above candidate has carried out research for the PhD thesis under our supervision.

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#### ABSTRACT

#### A Deep Syntactic Parser for the Tamil Language

Natural Language Processing (NLP) applications have become integral to human life. A syntactic parser is a vital linguistic tool that shows syntactic relations between the words in a sentence. These may then be mapped to a tree, a graph, or a formal structure. Syntactic parsers are helpful for building other NLP applications. In addition, they help linguists to understand a language better and perform cross-lingual linguistic analysis. A syntactic parser that performs a deeper analysis and captures argumentative, attributive and coordinative relations between the words of a given sentence is called a deep syntactic parser. Tamil is considered a low-resourced language in terms of tools, applications, and resources available for others to use and build NLP applications or carry out linguistic analyses. Not many resources, such as treebanks and annotated corpora, or linguistic analysis tools such as POS taggers or morphological analysers, are publicly available for Tamil. Available off-the-shelf language-agnostic syntactic parsers show comparatively low performance because of the rich morphosyntactic properties of Tamil. This study elaborates on how I developed the first grammar-driven parser for Tamil, which uses the Lexical-Functional Grammar formalism, and a state-of-the-art data-driven parser using the Universal Dependencies framework. I have also proposed an approach to evaluate a syntactic parser's syntactical coverage, experimented with transition-based and graph-based approaches, and for the first time, tried multi-lingual training to develop a data-driven parser for Tamil. A part of speech tagger, a morphological analyser cum generator, pre-processing tools, and treebanks are the other tools and resources I have developed to facilitate the development of the parsers. While all these tools give the current best score for their respective tasks, these resources are also available online for others to build upon. Moreover, the study also documents my contributions toward understanding different linguistic aspects of the Tamil language.

**Keywords**: Deep Syntactic Parser; Grammar-driven parser; Data-driven parser; Part of Speech tagger; Morphological Analyser

## **DEDICATION**

அப்பா - அம்மா

 $app\bar{a}$  -  $amm\bar{a}$ 

'Father - Mother'

for their unconditional love, support, and being the reason of who I am today.

பெரியப்பா - பெரியம்மா

 $periyapp\bar{a}$  -  $periyamm\bar{a}$ 

'Uncle - Aunt'

for being my guardians when crossing the most important part of my life.

இயற்கை

 $iya\underline{r}kai$ 

'the great Nature'

(the god)

for always putting together and aligning things I required for the progressions of this study.

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காலத்தி னாற்செய்த நன்றி சிறிதெனினும் ஞாலத்தின் மாணப் பெரிது. - திருக்குறள் (102) kālatti nārceyta nanri ciriteninum ñālattin māṇap peritu. - tirukkuraļ (102) "A favour conferred in the time of need, though it be small (in itself), is (in value) much larger than the world."

Thank you!

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## LIST OF ABBREVIATIONS

#### Abbreviation Description

1S First person – Singular

1PLE First person - Plural - Epicene
3SN Third person - Singular - Neuter
3SM Third person - Singular - Masculine
3SF Third person - Singular - Feminine

ABL Ablative
ACC Accusative
ADJ Adjective
ADJL Adjectivalizer

ADV Adverb

ADVL Adverbializer AGR Agreement

AP Adjectival Participle

ARG Argument

ASCII American Standard Code for Information Interchange

ASS Associative AUX Auxiliary

AVM Attribute-Value-Matrix

BEN Benefactive

BILSTM Bidirectional Long Short-Term Memory

BLEX Bi-LEXical dependency score

CAR Cardinal

CAUS Cause Marker
COMP Complementiser
COND Conditional
CONJ Conjunction

COP Copula

CRF Conditional Random Field

DAT Dative

DEC Declarative

DIST Distal

DOM Differential Object Marking

DUR Durative
EMPH Emphatic
EPI Epicene
F Feminine
FEATS Features

FST Finite-State Transducer

FUT Future Tense
GEN Genitive
GEND Gender

HDT Hamburg Dependency Treebank

HON Honorific

HON Double Honorific

HORT Hortative
IMP Imperative
INCL Inclusive

INESS Infrastructure for the Exploration of Syntax and Semantics

INF Infinitive
INS Instrumental
IRRAT Irrational
INTJ Interjection
IOBJ Indirect Object

LOC Locative

LAS Labelled Attachment Score LFG Lexical Functional Gramma

LV Light Verb

M Masculine Analyzer
MA Morphological Azad

MLAS Morphology-aware Labeled Attachment Score

MWTT Modern Written Tamil Treebank

N Neuter NEG Negative

NER Named Entity Recognizer

NLP Natural Language Processing

NLTK Natural Language ToolKit

NMLZ Nominaliser

NOM Nominative NPNoun Phrase NTYPE Noun Type NUM Number OBJ Object Oblique OBLORD Ordinal Particle PART PASS Passive Permissive PERM PERS Person PLPlural

POS Parts of Speech POSS Possessive

PP Postposition Phrase

PRED Predicate
PRES Present tense
PROG Progressive
PRON Pronoun

PRS Present Tense
PSP Postposition
PST Past Tense
QUOT Quotative
RAT Rational
REL Relativiser

RNN Rrecurrent Neural Network

SAN Sandhi SEM Semantic

SER Singular - Epicene - Rational

SG Singular SUBJ Subject

SVC Serial Verb Construction

SYM Symbol
TB Treebank
TNS-ASP Tense-Aspect
TTB Tamil TreeBank

UAS Unlabelled Attachment Score
UD The Universal Dependencies
UPOS Universal Part of Speech

VP Verbal Pharse

VPART Adverbial Participle

VTYPE Verb type

XCOMP Non-finite clause argument XLE Xerox Linguistic Engine

XPOS Language-specific Part of Speech

## TRANSLITERATION SCHEMA

Vow	els	Consonants		
<u> </u> அ	a	க்	k	
<b>ஆ</b>	ā	ங்	'n	
₹ ₩ ₩ ₩	i	· · · · · · · · · · · · · · · · · · ·	n c ñ	
FTF	ī	ஞ்	ñ	
೨_	u		ţ	
<u>ഉണ</u>	ū	ண்	ņ t	
ส	е	த்		
ஏ	ē	ந்	n	
ඉ ස ඉ ඉ	ai	· · · · · · · · · · · · · · · · · · ·	p	
ஒ	О	ம்	m	
ஓ	ō	ய்	y r	
ஒள	au	ij	r	
		ல்	1	
		வ்	V	
		් ම ම ම ම	<u>1</u> 1	
		ள்		
		ற்	<u>r</u>	
		ன்	${f n}$	

Note: Composite characters are formed by adding consonants and vowels together. For instance, Tamil letter  $\mathfrak s$  is transliterated as ka as  $\mathfrak s=\dot{\mathfrak s}$  (k)+  $\mathfrak A$ (a) In this way there are 216 composite Tamil letters are formed by composing 18 consonants with 12 vowels, and the composite letters will be transliterated accordingly.