



Handling Noun-Noun Coreference in Tamil

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Outline

- Objective
- Co-reference chain
- Our Approach
- Experiment and Results
- Error Analysis
- Intrinsic Errors



Objective

- Co-reference chains bring coherence
- Reference markers which bring cohesiveness
 - Pronominal, Reflexives, Reciprocals, Distributives, One-anaphors, Noun–noun reference
- Focus on resolution of noun-noun anaphors in Tamil
- Challenges in resolving it in Tamil



Co-reference Chains

- Coreference chains are formed by grouping various anaphoric expressions referring to the same entity.
- Early work in Co-reference resolution using ML
 - Soon et al (2000)
- Different ML Approaches
 - Decision Tree
 - First order probabilistic model
 - Multiple sieve based approach
 - Deep neural network based approach



Characteristics of Tamil

- South Dravidian family of language
- Relatively free word order language
- Verb final language and allows scrambling
- Nominative-accusative language
- Has Person, Number and Gender (PNG) agreement
- Clausal constructions are introduced by non-finite verbs.
- Copula drop, Accusative drop, Genitive drop, and PRO drop (Subject drop)



Our Approach

- Noun-Noun Anaphors
 - task of identifying the referent of the noun which has occurred earlier in the document.
 - Noun phrase may be repeated as a full noun phrase, partial noun phrase, acronym, or semantically close concepts such as synonyms or superordinates.
 - Named entities, Acronyms, Demonstrative noun phrases
Definite descriptions

Our Approach(Contd...)

- Machine Learning Technique
 - Conditional Random Fields
- Data preparation:
 - Training data:
 - Positive and negative pairs of NPs (NP_i and NP_j)
 - Testing data:
 - pairs of NPs (NP_i and NP_j)
- Pre-processing of data:
 - Processed with morphological analyser, Part of Speech tagger, Chunker, Clause boundary identifier and Named Entity Recognizer.

Our Approach(Contd...)

- Features Used: Individual Features
 - Single Word:
 - Is NP_i a single word; Is NP_j a single word
 - Multiple Words:
 - Number of Words in NP_i; Number of Words in NP_j
 - PoS Tags:
 - PoS tags of both NP_i and NP_j.
 - Case Marker:
 - Case marker of both NP_i and NP_j.
 - NE Category :
 - Named Entity tags of both NP_i and NP_j.
 - Presence of Demonstrative Pronoun:
 - Check for presence of Demonstrative pronoun in NP_i and NP_j.

Our Approach(Contd...)

Comparison Features

- Full String Match:
 - Check the root words of both the noun phrase NP_i and NP_j are same.
- Partial String Match:
 - In multi world NPs, calculate the percentage of commonality between the root words of NP_i and NP_j .
- First Word Match:
 - Check for the root word of the first word of both the NP_i and NP_j are same.
- Last Word Match:
 - Check for the root word of last word of both the NP_i and NP_j are same.
- Last Word Match with first Word is a demonstrator:
 - If the root word of the last word is same and if there is a demonstrative pronoun as the first word.
- Acronym of Other:
 - Check NP_i is an acronym of NP_j and vice-versa.

Experiment and Evaluation

- Collected 1,000 News articles from Tamil News dailies online
- Preprocessed and Noun-Noun anaphoric relations are tagged using PALinkA tool
- Statistics of Corpus

| | | |
|---|----------------------------------|---------|
| 1 | Number of Web Articles annotated | 1,000 |
| 2 | Number of Sentences | 22,382 |
| 3 | Number of Tokens | 272,415 |
| 4 | Number of Words | 227,615 |

Result and Analysis

| S. No. | Task | Precision (%) | Recall (%) | F-Measure (%) |
|--------|-------------------------------|---------------|------------|---------------|
| 1 | Noun-Noun Anaphora Resolution | 86.14 | 66.67 | 75.16 |

Errors

- Intrinsic Errors of the Noun-Noun resolution Engine

| S. No | Intrinsic Errors (%) |
|-------|----------------------|
| 1 | 17.48 |

Errors due to Preprocessing modules

Considering the total errors as 100%

| Percentage of error contributed by Each Preprocessing module | | | |
|--|----------------|-------------|-----------------------------|
| Morphological Analyser (%) | PoS Tagger (%) | Chunker (%) | Named Entity Recogniser (%) |
| 11.56 | 18.78 | 36.44 | 33.22 |

Ex.a

aruN vijay kapilukku pathilaaka *theervu_ceyyappattuLLar.*

Arun(N) vijay(N) Kapli(N)+dat instead select(V+past)

(Instead of Kapil, Arun Vijay is selected)

Ex.b

vijay muthalil kalam iRangkuvaar.

Vijay(N) first(N)+loc groud(N) enter(V)+future+3sh

(Vijay will be the opener.)

System output: vijay kapilukku ,vijay

Intrinsic Errors

- Fails to handle definite NPs,
 - no definiteness marker, these NPs occur as common noun.

Ex.a

maaNavarkaL pooRattam katarkaraiyil nataththinar.

Student(N)+PI demonstration(N) beach(N)+Loc do(V)+past+3pc

(The students did demonstartions in the beach.)

Ex.b

kavalarkaL maaNavarkaLai kalainthu_cella ceythanar.

Police(N)+PI students(N) disperse(V)+INF do(V)+past+3pc

(The police made the students to disperse.)

Here in both the sentences 'maaNavarkaL' (students) has occurred referring to the same entity.

Intrinsic Errors

Challenge in noun-noun anaphora resolution

- Popular names and nicknames
 - 'Gandhi' -> 'Mahatma', 'Baapuji'; 'Subhas Chandra Bose' -> 'Netaji';
 - 'Vallabhbhai Patel' -> 'Iron man of India'.
- Shortening of names
 - 'thanjaavur' (Thanjavur)-> 'thanjai' (Tanjai),
 - 'nagarkovil' (Nagarkovil) -> 'nellai' (Nellai),
- Usage of anglicized words
 - 'thiruccirappalli' (Thirucharapalli) -> 'Tirchy', 'thiruvananthapuram' (Thiruvananthapuram) -> 'trivandrum', 'uthakamandalam' -> 'ooty'.

Intrinsic Errors

Challenge in noun-noun anaphora resolution

- Spell variations
 - 'raaja' (Raja) -> 'raaca'.
 - Person names are usually written in different spelling
- Named Entities mentioned with description referring to the Named Entities

mumbai, inthiyaavin varththaka thalainakaram

Mumbai, India's Economic Capital

kaaci, punitha nakaram

Kasi, the holy city

Intrinsic Errors

Challenge in noun-noun anaphora resolution

Errors in identifying synonymous NP entities

Ex.a

makkaL muuththa kaavalthuRaiyinarootu *muRaiyittanar*.

People(N) senior(Adj) police(N)+soc argue(V)+past+3p

(People argued with the senior police people.)

Ex.b

antha athikaarikaLiyin pathiLai eeRRu cenRanar.

That(Det) officer(N)+PL+gen answer(N) accept(V)+vbp go(V)+past+3p

(Accepting the officer's answer they left.)

KaavalthuRaiyinarootu, athikaarikaL refers to the same entity.



Thank you