Myanmar NLP research and Usefulness of ALT data

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Natural Language Processing Lab in UCSY

မြန်မှန်တွတ်ရတ် တိုင်းမြန

- started in 2006 at University of Computer Studies, Yangon (UCSY) under Ministry of Science and Technology.
- Some of the works of the NLP lab are available online:
 - Network-based ASEAN Languages Translation Public Service (<u>http://www.aseanmt.org</u>)
 - English to Myanmar Statistical Machine Translation System (<u>http://www.nlpresearch-</u> <u>ucsy.edu.mm/NLP_UCSY/mtapplication.html</u>)
 - Myanmar-English-Myanmar bilingual dictionary (<u>http://www.nlpresearch-</u> ucsy.edu.mm/NLP_UCSY/dictionaryapplication.html)
 - Myanmar Word Segmentation (<u>http://www.nlpresearch-ucsy.edu.mm/NLP_UCSY/wsandpos.html</u>)



Research Collaboration

- NECTEC (Thailand National Electronics and Computer Technology Center)
- NICT (National Institute of Information and Communication Technology)
- For the purpose of
 - joint researches/projects,
 - researcher exchange,
 - publishing conference papers, journals and articles,
 - doing joint NLP workshops.









NLP Lab Members





NLP Research

Aim of Research

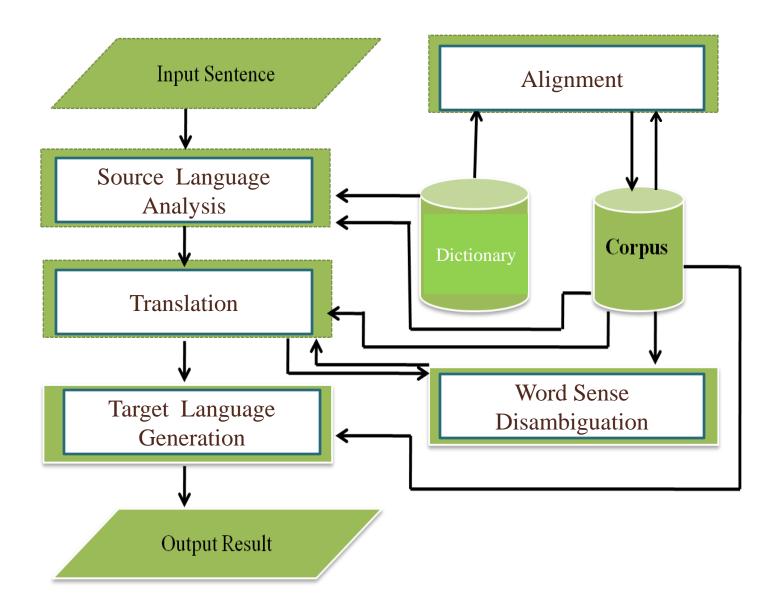
- to overcome language barrier
- to be applied conveniently in systems that are used by Myanmar

Domain of Research

- Myanmar-English-Myanmar Machine Translation
- Automatic Speech Recognition
- Text to Speech
- Myanmar Information Retrieval
- Myanmar Name Entity Recognition and Transliteration
- Myanmar Text Summarization
- Myanmar Text Categorization



Overview of the System



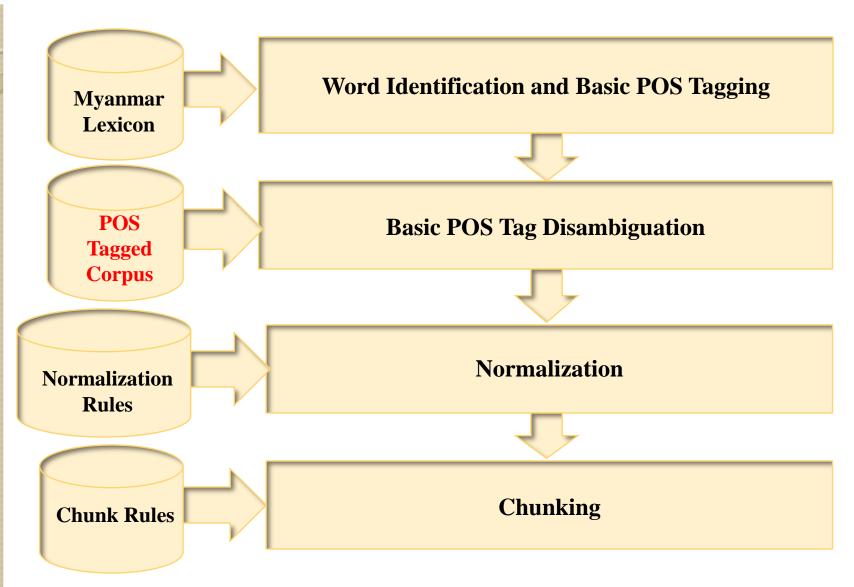


Source Language Analysis

- For Myanmar-English translation phase, it is the process of Myanmar Language Analyzer:
 - **Myanmar Part-of-Speech (POS) Tagging** and Chunking of Myanmar Language
 - Syntactic Analysis
 - Function Tagging and making Grammatical relation
- For English-Myanmar translation phase,
 - English POS and Chunking
 - Syntactic Analysis
 - Function Tagging and making Grammatical relation



Myanmar POS Tagging and Chunking





Pre-tagged Corpus Format :

Training Corpus

- Myanmar words are segmented and tagged with their respective basic POS tags and categories as follows ::
 - ✔ သူ/PRN.Person # ကျောင်း/NN.Building # သို့/PPM.Direction # သွား/VB.Common # သည်/SF.Declarative

 - ✓ ဤ/PRN.Distobj # ဢ/NN.Common # ကို/PPM.Obj # မည်သူ/PRN.Question # ရေး/VB.Common # ခွဲ /Part.Support # သနည်း/SF.Interrogative



Example : Tagging

Input Text

 ✓ သံလွင် မြစ် သည် မြန်မာပြည် တောင်ပိုင်း သို့ ဦးတည် စီးဆင်း သွား သည်။ (The river, Than Lwin, flows to south of Myanmar.)

□ Tagging with All Possible Tags on Each Word

- ✓ పఁ్గర్_#NNP.Location
- ✔ မြစ် _#NN.Location
- ✓ ාාරා _#SF.Declarative #PPM.Subj
- 🗸 မြန်မာပြည် _#NNP.Location
- ✓ တောင်ပိုင်း_#NN.Location
- 🗸 သို့ _#PPM.Direction
- ✓ ဦးတည်_#VB.Common
- ✓ သွား_#VB.Common#NN.Body#Part.Support



Disambiguation of Tags

- disambiguating all possible basic POS tags to produce the correct tag.
- training Myanmar pre-tagged Corpus with HMMs and LHMMs models.
- decoding using the Viterbi tagging algorithm to find out the best probable path (best tag sequence) for a given word sequence.



Example : Disambiguation

□ Disambiguation and Assigning with Correct Tag on Each Word

✓ పర్నర్_#NNP.Location	(Than Lwin)
🗸 မြစ် _#NN.Location	(The river)
🗸 သည် _#PPM.Subj	(null)
🗸 မြန်မာပြည် _#NNP.Location	(Myanmar)
🗸 တောင်ပိုင်း_#NN.Location	(south)
🗸 သို့ _#PPM.Direction	(to)
✓	(flows)
🗸 စီးဆင်း _#VB.Common	(flows)
🗸 သွား_#Part.Support	(flows)
🗸 သည် _#SF.Declarative	(null)



Example : Normalization

• forming more meaningful words and annotating with appropriate POS tags and categories.

□ Before normalization,

"**ကျန်းမာ/VB.Common # ඛිර්:/Part.Common #** သည် /PPM.Subj # လာဘ်/NN.Common # တစ်/NN.Cardinal # ပါး/Part.Type # ဖြစ်/VB.Common # သည် /SF.Declarative"

□ After normalization,

"**ကျန်းမာခြင်း/NN.VBConvert** # သည် / PPM.Subj # လာဘ် / NN.Common # တစ် / NN.Cardinal # ပါး / Part.Type # ဖြစ်/ VB.Common # သည် / SF.Declarative "



Example : Chunking

• assemble the POS tagged words and identify chunk tag.

Before chunking,

သူတို့/NNR.Person # သည်/PPM.Subj # အတန်း/NN.Common # ထဲတွင်/PPM.Extract # အတော်ဆုံး/JJS.Common # ကျောင်းသားများ/NNR.Person# ဖြစ်/VB.Common # ကြ/Part.Support # သည်/SF.Declarative

□ After chunking,

NC [သူတို့/NNR.Person] # PPC [သည်/PPM.Subj] # NC [အတန်း/NN.Common] # PPC [ထဲတွင်/PPM.Extract] # NC [အတော်ဆုံး/JJS.Common # ကျောင်းသားများ/NNR.Person] # VC [ဖြစ်/VB.Common # ကြ/Part.Support] # SFC [သည်/SF.Declarative]





- Identifying word correspondence that are translations of each other based on information found on parallel text.
- Developing a Myanmar-English bilingual corpus:
 - Dictionary lookup approach
 - Corpus-based approach



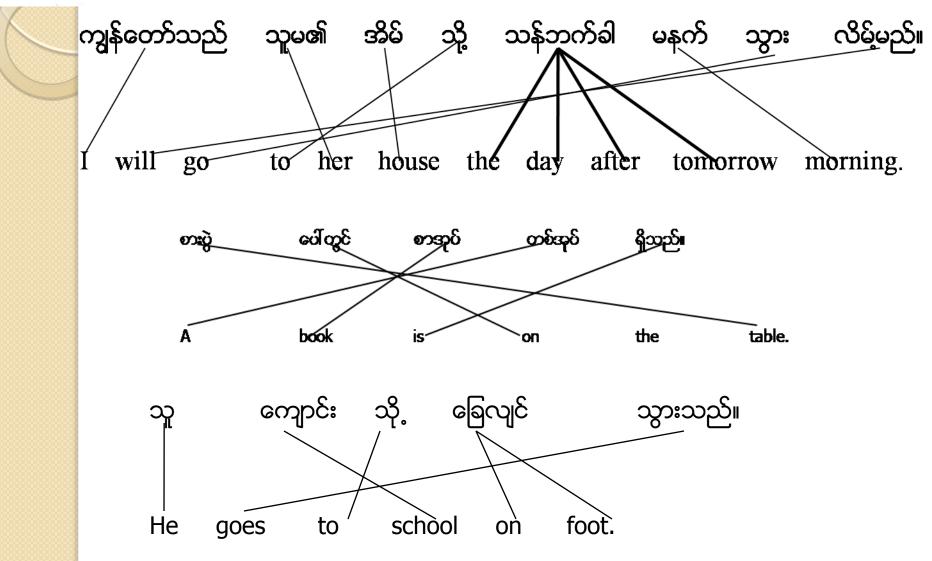
Word Alignment Algorithm

Step 1: Accept pair of Myanmar and English sentences.

- Step 2: Tag English sentence with Part-Of-speech (POS) Tagger and it will produce tagged output also with root word.
- Step 3: Segment Myanmar sentence into words.Removes the stop words.Make morphological analysis of the noun and verb affixes using trigram method.
- Step 4: Align the output English and Myanmar words from step 2 and 3 based on the first three IBM models and EM algorithm using parallel corpus.
- Step 5: Align the remaining words (i.e unaligned) using Myanmar-English bilingual dictionary.



Example Alignment





Problems in Alignment

Scarce Resource

- No publicly available POS-tagged corpus for Myanmar and English.
- □ The constructed POS-tagged corpus has a limited number in size.

Linguistic Problem

- □ Parallel sentence pairs might not be equal size.
- □ Myanmar and English word order could be significantly different.
- Myanmar language is a morphologically rich and verb final language. English is a verb-second language.



Translation

- Phrase/word Translation pairs Extraction
- Morphological Analysis
- Word Sense Disambiguation



Phrase/word Extraction

- For each phrase we identified by its start position, end positions phrase length and target phrase to ensure that there are no gaps and no overlap.
- Applying N-gram methods using Corpus,

Source phrase	Start position	End position	Phrase Length	Target phrase	Translation probability
ငှက်	1	1	1	Bird	1.0
ငှက်များ	1	2	2	Birds	1.0
ပျံ	4	4	1	Fly	1.0
ပျံကြသည်	4	6	3	Fly	1.0

Translation

ငှက်များ - birds ပျံကြသည် - fly



Example : Morphological Analysis of verbs

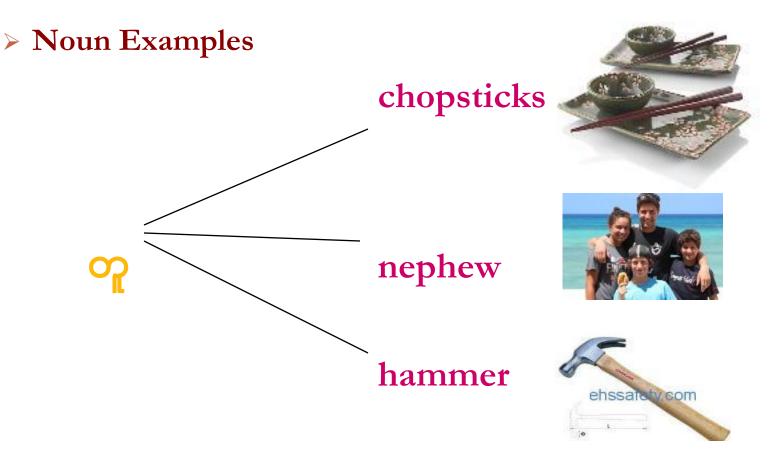
- Myanmar unknown verb: ကြည့်ခဲ့ပါသည်
- Verb suffiex: ခဲ့ပါသည်
- Tense particle: ఫై
- Translation of main verb (using corpus): look
- ကြည့်ခဲ့ပါသည်/looked



Word Sense Disambiguation for Myanmar Language

- Purpose:
 - to solve the ambiguity of Myanmar words for Myanmar-English machine translation





- > သူသည်တူဖြင့်ခေါက်ဆွဲစားသည်။ He eats the noodle with chopsticks.
- > သူ့မှာတူသုံးယောက်ရှိသည်။ He has three nephews.
- > လက်သမားသည်တူကိုသုံးသည်။ Carpenter uses the hammer.



WSD Algorithm for Myanmar Word

Step1:Preprocessing

-Segment input sentence

-Remove stop words from input sentence and create ambiguous vector **Step2:Multi-sense Look-up**

-Retrieve all possible sense meanings of ambiguous word from corpus

-Collect training data concerning with these sense from corpus Step3:Build context vectors for each sense based on collected

training data

-For all context vectors do

-Remove stop words

-Remove redundant words

-End For

Step4:Calculate the cosines between ambiguous vector and each of the context vectors

 $\cos \theta = \frac{A.B}{\|A\|.\|B\|} = \frac{\sum_{i=1}^{n} A_i.B_i}{\sqrt{\sum_{i=1}^{n} A_i^2} \cdot \sqrt{\sum_{i=1}^{n} B_i^2}}$ where A represents each word in ambiguous vector B represents each word in each context vector Step5:Choose correct sense of the target word s' = argmax score(s_i)



Conclusion

- The data sparseness is most important in many research regarding NLP because of the followings:
 - The rules only can not be solved for all problems for many languages.
 - So, the researches are coming based on the statistical model.
 - The more availability of data in developing the system/tools, the more accuracy we can get.
- So, ALT data is very useful not only for Myanmar language but also for all languages to be applied in various kinds of NLP researches.

Thank you!