A State of the Art of Thai Language Resources and Thai Language Behavior Analysis and Modeling

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Abstract

As electronic communications is now increasing, the term Natural Language Processing should be considered in the broader aspect of Multi-Language processing system. Observation of the language behavior will provide a good basis for design of computational language model and also creating cost-effective solutions to the practical problems. In order to have a good language modeling, the language resources are necessary for the language behavior analysis.

This paper intended to express what we have and what we have done by the desire to make a bridge between the languages and to share and make maximal use of the existing lexica, corpus and the tools. Three main topics are, then, focussed: A State of the Art of Thai language Resources, Thai language behaviors and their computational models.

1. Introduction

As electronic communications are now increasing, the term Natural Language Processing should be considered in the broader aspect of Multi-Language Processing system. An important phase in the system development process is requirement engineering, which can define as the process of analyzing the problems in a certain language. An essential part of the requirementengineering phase is computational language modeling which is an abstract representation of the behavior of the language. In order to have a good language model for creating cost-effective solutions to the practical problems, the language resources are necessary for the language behavior analysis.

This paper intended to express what we have and what we have done by the desire to make a bridge between the languages and to share and make maximal use of the existing lexica, corpus and the tools. Three main topics are, then, focussed:

• A State of the Art of Thai language Resources that will give an overview of what we have in Corpus, Lexicon and tools for corpus processing and analysis. • Thai language behaviors (only in word and phrase level) analyzed from the varieties of corpus which consist of Lexicon growth, New word formation and Phrase/Sentence construction, and

• The computational models providing for those behaviors, which consist of Unknown Word Extraction and Name Entities identification, New word generation and Noun phrase recognition.

The remainder of the paper is organized as follows. In section 2, we give the gateway of Thai language resources. Thai Language behaviors are discussed in section 3. In section 4, then, provides Thai Language Computational Modeling as a basis for creating cost-effective solutions to those practical problems.

2. A State of the Art of Thai Language Resource

This section gives a survey of a state of the art of Thai Language Resources consisting of Corpus, Lexicon and Tools. Here, we will present only the resources that open for public access.

2.1 Corpus

The existing Thai corpus is divided into 2 types; speech and text corpus developed by many Thai Universities. Thai Language Audio Resource Center of Thammasart University (ThaiARC) (http:// thaiarc.ac.th) developed speech corpus aimed to provide digitized audio information for dissemination via Internet. The project pioneers the production and collection of various types of audio information and various styles of Thai speech, such as royal speeches, academic lectures, oral literature, etc.

For Text corpus, originally, the goal of the corpus collecting is used only inside the laboratory. Until 1996, National Electronics and Computer Technology Center (NECTEC) and Communications Research Laboratory (CRL) had a collaboration project with the purpose of preparing Thai language corpus from technical proceedings for language study and application research. It named ORCHID corpus (NECTEC, 1997). NAIST Corpus began in 1996 with the primary aim of collecting document from magazines for training and testing program in Written Production Assistance (Asanee, 1995). The existing corpus can be summarized as shown in Table 1.

List	Corpus	Туре	Amount	Status	
NECTEC	Orchid	POS-	2,560,000	Online	
	Corpus	Tagged Text	words		
Kasetsart	NAiST	Toyt	60,511,974	Onlina	
Univ.	Corpus	Text	words	Onnie	
Thammasart	Thai	Digitized	4000	onling	
Univ.	ARC	audio	words++	omme	

Table 1: The List of Thai Corpus

2.2 Lexicon

There are a number of Thai lexicons, which has been developed as shown in Table 2.

Table 2: The List of Thai Dictionaries

Dictionary	Туре	Size (word)	status	Web site
Royal Institute Dictionary	Mono	33,582	Online	http://rirs3.royin. go.th/riThdict/loo kup.html
Lexitron	Bi	50,000	Online	http://www.links. nectec.or.th/lexit/ lex_t.html
NaiST Lexibase	Mono	15,000	Online	http://beethoven. cpe.ku.ac.th/
So Sethaputra Dictionary	Bi	- 48,000 Eng words - 38,000 Thai words	Online	http://www.thais oftware.co.th/

Dictionary	Туре	Size (word)	status	Web site
Narin's Thailand homepage	Bi	-	Online	http://www.wiwi. uni- frankfurt.de/~sas cha/thailand/dicti onary/dictionary _index.html
Saikam online	Bi	133,524	Online	http://saikam.nii. ac.jp/.
Lao-Thai- English Dic.	Multi	5,000	Offline	-

From the table 2, Only Lexitron (from NECTEC) and NAiST Lexibase (from Kasetsart University) that were applied to NLP. NAiST Lexibase has been developed based on relational model for managing and maintaining easily in the future. It contains 15,000 words list with their syntax and the semantic concept information in the concept code.

2.3 Corpus and Language Analysis Tools

Corpus is not only the resource of Linguistic Knowledge but is used for training, improving and evaluating the NLP systems. The tools for corpus manipulation and knowledge acquisition become necessary.

NAIST Lab. has developed the toolkit for sharing via the Internet. It has been designed for corpus collecting, annotating, maintaining and analyzing. Additionally, it has been designed as the engine, which the end user could use with their data. (See a service on http://naist.cpe.ku.ac.th).

3. Thai Language Behavior Analysis

In order to have a good language model for creating cost-effective solutions to the practical problems in application development, language behavior must be observed. Next is Thai language behavior analysis based on NAiST corpus consisting of Lexicon growth, Thai word formation and Phrase construction.

3.1 Lexicon Growth

The lexicon growth is studied by using Word list Extraction tool to extract word lists from a large-scale corpus and mapping to the Royal Institute Dictionary (RID). It is noticeable that there are two types of lexicon: common and unknown words. The common word lists are some words in RID, which occur in almost every document, and use in daily life. They are primitive words but not being proper names or colloquial words. The unknown or new words occur much in the real document such as Proper names, Colloquial words, Abbreviations, and Foreign words.

The lexicon growth is observed from corpus size, 400,000, 2,154,700 and 60,511,974 words from Newspaper, Magazine and Agriculture text. We found that common word lists increased from 111,954 to 839,522 and 49,136,408 words according to the corpus size, while the unknown word lists increased from 288,046 to 1,315,178 and 11,375,566 words respectively as shown in table3.

Table 3 : Lexicon-growth

Size of Corpus/ words	Common words	Unknown
		words
400,000	111,954	288,046
2,154,700	839,522	1,315,178
60,511,974	49,136,408	11,375,566

Regarding to 60,511,974 words corpus in the table 3, it composes of 35,127,012 words from Newspaper, 18,359,724 words from Magazine and 7,025,238 words from Agricultural Text. Unknown words occur in each category as shown in table 4.

Table 4: The Categories of Unknown words according to the various corpus genres

0		1 0	
Types of	Newspaper	Magazine	Agricultural
unknown word	(words)	(words)	Text (words)
Proper name	4,809,160	1,272,747	1,170,076
Spoken words	58,335	8,787	0
Abbreviation	70,109	43,056	0
Foreign words	304,519	239,107	3,399,670
Total	5,242,123	1,563,697	4,569,746

According to table 3 and 4, we could observe that not only unknown words increase but common words also increase and the main categories of increasing unknown word are proper names and foreign words. Consequently, a computational model of unknown word extraction and name entity identification has been developed and also of new word construction.

3.2 New Word Formation and Core Noun

Regarding to the growth of common word shown in table 3, we studied how the new words come from.

3.2.1 Basic Information about Thai

Thai words are multi-syllabic words which stringing together could form a new word. Since Thai has no inflection and no word delimiters, Thai morphological processing is mainly to recognize word boundaries instead of recognizing a lexical form from a surface form as in English. Let C be a sequence of characters

Let U be a sequence of characters $C = c1c2c3...cn : n \ge 1$ Let W be a sequence of words $W = w1w2w3...wm : m \ge 1$

Where wi = $c1ic2i...ci r: i \ge 1, r \ge 2$

Since Thai sentences are formed with a sequence of words with a stream of characters, i.e., c1c2c3...cn mostly without explicit delimiters, the word boundary in "c1c2c3c4c5" pattern as shown below could have two ambiguous forms. One is "c1c2c" and "c3c4c5". The other one is "c1c2c3" and "c4c5" (Kawtrakul, 1997)



Figure 1: Word Boundary Ambiguity

From the figure 1, if characters were grouped differently, the meaning of words will be changed too. For example, "nonon" can be grouped to "non-on(fold one's arms across the chest)" and "no-non (a clump of flower)". From our corpus, we found that the sentence with 45 characters has 30 combinations of words sequence.

3.2.2 New word construction

Almost all-Thai new words are formed by means of compounding and nominalization, by using a set of prefixes.

3.2.2.1 Nominalization

Nominalization is a process by which a word can be formed as a noun by using prefixes added. Noun words formed by using prefixes "nii(ka:n)" and "nii(k^hwa:m)"are nouns which signal state or action. Words formed by using prefixes "ğ(p^hu:)" "ini(tç^ha:w)" and "ŭn(nak)"are nouns which signal human or profession.

Prefix " החז (ka:n)" האור (ka:n)" היה (ka:m)" are used in the process of forming a noun from verb or verb

phrase and sometimes from noun (Nominalization). MIG(ka:n) that co-occur with noun, represents the meaning about duty or function of noun it relates to. MIG(ka:n) that co-occur with verbs, always occur with action verbs.

3.2.2.2 Compounding

Thai new words can, also, be combined to form compound nouns and are invented almost daily. They normally have at least two parts. The first part represents a pointed object or person such as คน(man), หมือ(pot), หาง(tail), พืช(plant). The second part identifies what kind of object or person it is, or what its purpose is like บับรถ(drive a car), หุงขัวว(cook rice), เสือ(tiger), น้ำ(water). Table 5 shows the examples of compound noun in Thai.

Table 5: The Examples of That Compound Noul					
What or who	What type / what purpose				
คน(man)	ขับรถ(drive a car),				
หม้อ(pot)	หุงข้าว(cook rice)				
หาง(tail)	เสือ(tiger)				
พืช(plant)	น้ำ(water)				

Table 6 shows the patterns of compound noun. **Table 6:** Compound noun pattern

Compound	Examples	Meaning
noun	-	
structure		
	หาง(tail)เสือ(tiger)	Rudder
$\Pi + \Pi$	พืช(plant)น้ำ(water)	Water Plant
	ห้อง(room)นอน(sleep)	Bedroom
n + v	เก้าอี้(chair)โยก(rock)	rocking chair
	คน(man)ขับ(drive)รถ (car)	Driver
n + v + n	หม้อ(pot)หุง(cook)ง้าว(rice)	A Pot For Cooking Rice
	เด็ก(child)ผม(hair)ยาว(long)	A Long Hair Child
n + n + v	คน(human)บา(leg)เป้(lame)	A Lame Man
	บ้าน(home)ทรง(shape)ไทย	Thai Style House
n + n + n	(Thai)	
	ข้าว(rice)ขา(leg)หมู(pig)	A kind of dishes
	ໃນ(leaf) vັນ(drive) ຈື່(ride)	Driving License
n + v + v	ห้อง(room)นั่ง(sit)เล่น(play)	Living Room

From Table 6, it has shown that some compound nouns maintain some parts of those primitive word meaning but some changed to a new meaning. In this paper, we are only interested in compound noun grouping from primitive words which were changed the meaning to more abstract but still maintain some parts of those primitive word meanings, e.g. "คนรถ(driver) คนครัว(cooker) etc." The word "nu" maintains its meaning which has a concept of human, but when it was compounded with "sn(car)" and "nšo(kitchen)", their meanings have changed to the occupation by the word relation in the equivalent level. In case of compound noun that change a whole meaning such as "ลูกเสือ (a boy scout)", it will be kept in the lexicon.

3.2.2.3 Compound noun extraction problems

There are three non-trivial problems

- Compound Noun VS Sentence Distinction
- Compound Noun Boundary Ambiguity
- Core noun Detection

Compound Noun VS Sentence

Several NP structures have the same pattern as sentences. Since Thai language is flexible and has no word derivation, including to preposition in compound noun can be omitted, etc. This causes a compound noun having the same pattern as sentence. Thus, Thai NP analysis in IR system is more difficult than English. (See Figure 2)

Sentence: u	กกินผลไม้ (birds)	eat fruit)	
In Thai:	นก	กิน	ผลไม้
	Birds	eat	fruit
Syntactic	cn	tv	cn
Category			
C	™ ⇔ต่อ ข	(<u>1</u>	
Compound	Noun: โต๊ะกินข้าว	(a dining	table)
Compound In Thai:	Noun: โต๊ะกินข้าว โต๊ะ(ถำหรับ)	(a dining กิน	table) ข้าว
Compound In Thai:	Noun: โต๊ะกินข้าว โต๊ะ(<mark>สำหรับ)</mark> table	(a dining กิน eat	table) ข้าว rice
Compound In Thai: Syntactic	Noun: โด๊ะกินข้าว โด๊ะ(สำหรับ) table cn	(a dining กิน eat tv	table) ข้าว rice cn

Figure 2: The comparison of noun phrase and sentence structure

In figure 2, compound noun "โต๊ะกินข้าว" (a dining table) actually omit the preposition "สำหรับ (for)", which is a relation that point to the purpose of the first noun "โต๊ะ(table)".

The Compound Noun Boundary Ambiguity

After we have extracted noun phrase aiming for enhancing the IR system, we have to segment

that noun phrase into sub noun phrase or compound noun in order to specify the core noun as index and its modifier as sub-index. For example, compound noun with "noun + noun + verb" structure: idn(child/N)MU(hair/N)MU (long/V) etc. In this case, the second noun and verb have to be grouped firstly since it behaves similarly to a modifier by omitting the relative pronoun that represents its purpose, i.e., "who has".

Another case of Compound Noun Boundary Ambiguity is word combination. Consider the sequence of words as the example of NP that composes of four words as follows:

$$NP = N_1 N_2 N_3 N_4$$

There are 8 word combinations of compound noun as shown in figure 3.





In figure 3, word string has to be grouped correctly for the correct meaning.

The ambiguity of noun phrase boundary has also directly effected the efficiency of text retrieval.

Core Noun detection

Due to the Information Retrieval, a head or core of noun phrase detection is necessary. In this paper, core noun refers to the most important and specific word that the information retrieval and extraction can directly retrieve or extract without over generating candidate words. However, by the observation, the core of noun phrase needs not to be the initial words. Some of them are at the final position and some have word relation in the equivalent level (As shown in Table 7).

	_	CD1		1 0			•	A TT
Table	7:	The	examp	les of	core	noun	ın	NF

Noun phrase(NP)	Core noun
W1 W2 โครงสร้าง + ประโยค	be W1 located at the initial position
structure + sentence	
W1 W2	be W2 located at the final position
รอย + วงปี	
stain annual ring	
W1 W2 W3	be W2 located at the second
ผล + มะละกอ + คิบ	position
fruit papaya green	

As mentioned above, the models of New Word Generation and Noun Phrase Recognition become one of the interesting works in Thai processing.

3.3 Phrase and Sentence Construction

Next, we will indicate the main problems that influence to MT, IE and IR system. These are constituent movement, zero anaphora and iterative relative clause.

3.3.1 Constituent Movement

Constituent is the relationship between lexicon units, which are parts of a larger unit. Constituency is usually shown by a tree diagram or by square brackets:

Ex. [[การประชุมคณะกรรมการ] [อย่างราบรื่น]]

[[meeting committee] [very smoothly]].

Constituent acts as a chunk that can be moved together and it often occurs in Thai language (see Fig. 4). The constituents can be moved to the front, the middle or the end of the sentence.



Figure 4 The movements of constituent ตอนเช้า ชาวประมง ออกเรือ หาปลา

- In the morning, the fisherman goes to catch the fish ชาวประมง ออกเรือ หาปลา ตอนเช้า
- The fisherman goes to catch the fish in the morning. ชาวประมง ออกเรือ ตอนเช้า หาปลา

The fisherman goes to in the morning, catch the fish ตอนเช้า หาปลา ชาวประมง ออกเรือ

In the morning, catch the fish, the fisherman goes to.

Noun, adverb, and prepositional phrase are often move while verb phrases are.

3.3.2 Zero Anaphora

Ex.:

To make the cohesion in the discourse, the anaphora is used as a reference to "point back" to some entities called referent or antecedent, given in the preceding discourse. Halliday, M.A.K. and Hasan, Rugaiya (1976) divided cohesion in English into 5 categories as shown in Table 8:

Table 8: Categories of anaphora

Reference	- Personal Reference, Demonstrative
	Reference, Comparative Reference
Substitution	- Nominal Substitution, Verbal Substitution,
	Causal Substitution
Ellipsis	- Nominal Ellipsis, Verbal Ellipsis, Causal
	Ellipsis
Conjunction	- Additive, Adversative, Casual, Temporal
Lexical Cohesion	- Reiteration(Repetition, Synonym or Near
	Synonym, Super ordinate, General word)
	- Collocation

Observing from the corpus in: news, magazine and agricultural text, there are 4 types of anaphora. Ellipsis or zero anaphora was found most frequently in Thai documents and other anaphora happened as show in table 9.

¥1			
Type of Anaphora	Magazine	news	agriculture
Zero anaphora	49.88%	52.38%	50.04%
repetition	32.04%	27.78%	34.49%
personal reference	12.18%	12.70%	1.87%
nominal substitution	5.90%	6.08%	13.60%

Table	9 :	Types	of reference
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Zero anaphora is the use of a gap, in a phrase or clause that has an anaphoric function similar to a pro-form. It is often described as "referring back" to an expression that supplies the information necessary for interpreting the gap

The following is a sentence that illustrates zero anaphora:

มีถนนสองสาขที่ต้องไป ตรงแต่แกบ และกว้างแต่กดเกี้ยว

• There are two roads to eternity, straight but narrow, and broad but crooked.

In this sentence, the gaps in *straight but narrow* [*gap*], *and broad but crooked* [*gap*] have a zero anaphoric relationship to *two roads to eternity*.

Table 10 also shows the occurrence of zero anaphora in various parts of a sentence.

	A D	• . •	0	C	•	
Tahla T	n. p	ocition	ot re	terence	1n	centencec
I ADIC I	U . I	USILIOII	UT IC.	lononee	ш	sentences

Position	Frequency
Subject	49.88%
Object	32.04%
Possessive Pronoun	12.18%
Following a Preposition	5.90%

It is noticeable that zero anaphora in the position of the subject occurs with high frequency (49.88%). It shows that in Thai language, the position of subject is the most commonly replaced.

3.3.3 Iterative Relative Clause

Thai relative pronouns " \vec{n}' " (thi) " \vec{n}_3 (sung)" and " $\tilde{\mathfrak{Du}}(un)$ " relate to group of nouns or other pronouns (The student " \vec{n}' " (thi) studies hardest usually does the best.). The word " \vec{n}' " (thi) connects or relates the subject, *student*, to the verb within the dependent clause (*studies*). Generally, we use " \vec{n}' " (thi) and " \vec{n}_3 (sung)" to introduce clauses that are parenthetical in nature (i.e., that can be removed from the sentence without changing the essential meaning of the sentence. The pronoun " \vec{n}' " (thi) and " \vec{n}_3 (sung)" refers to things and people and " \vec{n}_4 (un)" usually refers to things, but it can also refer to event in general.

The relative pronoun is sometimes omitted because it makes the sentence more efficient and elegant.

- ์หนังสือ ที่/ซึ่ง คุณ สั่งซื้อ จาก ร้านนั้น มาถึงแล้วเมื่อ 2 วันก่อน
- The book that you ordered from that shop arrived two days later.

Sometimes relative pronoun refers to an event that takes place repeatedly in a phrase.



Figure 5 The structure of relative clause

Ex. [พ่อครัว]N [(ที่) ชนะการแข่งขันทำอาหาร]Rel Cl.

[The chef] [who won the cooking competition] [(ซึ่ง) จัดขึ้นที่ประเทศฝรั่งเศส] Rel Cl. [(ที่) ฉันจ้างมา] Rel Cl. [which compete at France] [that I employ]

Although a sentence, which has several clauses inside, will be grammatical, but it is not a good style in writing and always causes a problem for parser and noun phrase recognition.

4. The Computational Model

The computational models in word and phrase level are developed according to the phenomena mentioned in section 3.

4.1 Unknown Word Extraction

Unknown word extraction model composes of 2 sub-modules: unknown word recognition and name entity identification.

4.1.1 Unknown word recognition

The hybrid model approach has been used for unknown word recognition. The approach is the combination of a statistical model and a set of context based rules. A statistical model is used to identify unknown word's boundary. The set of context based rules, then, will be used to extract the unknown word's semantic concept. If the unknown word has no context, a set of unknown word information, which has defined through corpus analysis, will be generated and the best one will be selected, as its semantic concept, by using the semantic tagging model. Unknown word recognition process is shown in figure 6.



Figure 6: Unknown word recognition process

4.1.2 Name Entity Identification

After unknown words have been extracted, Named Entity (NE) Identification will define the category of unknown word. The model based on heuristic rules and mutual information. Mutual information or statistical analysis of word collocation is used to solve boundary ambiguity when names were composed with known and unknown words. We use Knowledge based such as list of known name (such as country names), clue word list (such as person's title) to support the heuristic rules. Using clue word or common noun that precedes the name can specify NE categorization. Based on the case grammar, NE categories can also defined. Moreover, the lists of the names from predefined NE Ontology can be used for predicting category too. The overview of our system is shown in figure 7. More detail sees (Chanlekha, H. et al, 2002)



Figure 7 : Named Entity Recognition System

4.2 New Word Generation

Word formation is proposed to reduce the lexicon size by constructing new words or compound noun from the existing words. Based on word formation rules and common dictionary, the shallow parser will extract a set of candidate compound nouns. Then probabilistic approach based on syntactic structure and statistical data is used to solve the problem of over- and undergeneration of new word construction and prune the erroneous of compound noun from the candidate set. The process of new word construction is shown in figure 8. See more detail in (Pengphon, N. et al, 2002)



Figure 8 : New Word Construction process

4.3 Noun Phrase Recognition

Entities or concepts are usually described by noun phrases. This indicates that text chunks like noun phrases play an important role in human language processing. In order to analyze NP, both statistical and linguistic data are used. The model of NP analysis system is shown in figure 9. More detail sees (Pengphon, N. et al, 2002)



Figure 9 The architecture of system

The first step is morphological analysis for word segmentation and POS tagging. At the second step, the compound word is grouped into one word by using word formation module (see 4.2). The third step, statistical-based technique is used to identify phrase boundary. This step was provided for identifying the phrase boundary by using NP rules. Next step is Noun Phrase Segmentation. The condition of noun phrase segmentation is shown in figure 10.

\neg w ₂ \in clue word set	; w_2 =start sub np			
$\{w_1, w_2, R_{12}\} - R_{12} < threshold$; w_2 = start sub np			
L else	; $w_1 w_2$ is one unit			

Figure 10 Noun phrase Segmentation

After noun phrase is correctly detected, the relation in noun phrases will be extracted. There are 2 types of relation: head-head noun phrase and head-modifier noun phrase. The process is based on statistical techniques by considering the frequency (f_i) of each word (w_i) in the document (See figure 11).



Figure 11: Noun phrase relation

5. Conclusion

The computational language models for Thai in word and phrase level, consisting of Unknown Word Extraction and Name Entities identification, New word generation and Noun phrase recognition, are studied on the basis of their behavior analysis from the varieties of corpus. We expected that it could create costeffective solutions to the practical problems in the application developments especially in Thai Information Retrieval and Information extraction system. We also give the gateway to access Thai language resources with hoping that it could be the bridge of the international collaboration for developing Multi-Language Processing applications.

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