

Statistical Morphological Tagging and Parsing of Korean with an LTAG Grammar

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Overview

- Introduction to Supervised Statistical Parsing with LTAG
- LTAG grammar extracted from the Penn Korean Treebank
- Morphological Tagging: Motivation and Experiments
- Statistical parsing of Korean using a Morphological Tagger

Parsing as a machine learning problem

- S = a sentence
 T = a parse tree
A statistical parsing model defines $P(T | S)$
- Find best parse: $\arg \max_T P(T | S)$
- $P(T | S) = \frac{P(T, S)}{P(S)} = P(T, S)$
- Best parse: $\arg \max_T P(T, S)$
- e.g. for PCFGs: $P(T, S) = \prod_{i=1 \dots n} P(\text{RHS}_i | \text{LHS}_i)$

Parsing as a machine learning problem

- Training data for English: the Penn WSJ Treebank (Marcus et al. 1993)
- Convert Treebank into LTAG derivations using LexTract (Xia 2001)
- Train statistical LTAG parser from these events
- Evaluate accuracy on test data
- A standard evaluation:
Train on 40,000 sentences
Test on 2,300 sentences

Parsing as a machine learning problem

- Training data for Korean: the Penn Korean Treebank (Han et al. 2002)
- Train statistical morphological tagger and statistical LTAG parser
- Evaluate accuracy on test data
- Our evaluation:
Train on 4,653 sentences (49,473 words)
Test on 425 sentences (3,717 words)

Statistical Parsing with Tree Adjoining Grammars

- Substitution: $\sum_{\alpha} P_s(t, \eta \rightarrow \alpha) = 1$
- Adjunction: $P_a(t, \eta \rightarrow \text{NA}) + \sum_{\beta} P_a(t, \eta \rightarrow \beta) = 1$
- Multiple adjunctions at a node (Schabes and Shieber 1994):

$$P_{la}(\tau, \eta \rightarrow \text{NA}_l) + \sum_{\tau'} P_{la}(\tau, \eta \rightarrow \tau') = 1$$
$$P_{ra}(\tau, \eta \rightarrow \text{NA}_r) + \sum_{\tau'} P_{ra}(\tau, \eta \rightarrow \tau') = 1$$

Statistical Parsing with Tree Adjoining Grammars

- Start of a derivation: $\sum_{\alpha} P_i(\alpha) = 1$
- Probability of a derivation:

$$\begin{aligned} Pr(\mathcal{D}, w_0 \dots w_n) = & \\ & P_i(\alpha, w_i) \times \prod_p P_s(\tau, \eta, w \rightarrow \alpha, w') \times \\ & \prod_q P_a(\tau, \eta, w \rightarrow \beta, w') \times \prod_r P_a(\tau, \eta, w \rightarrow \text{NA}) \end{aligned}$$

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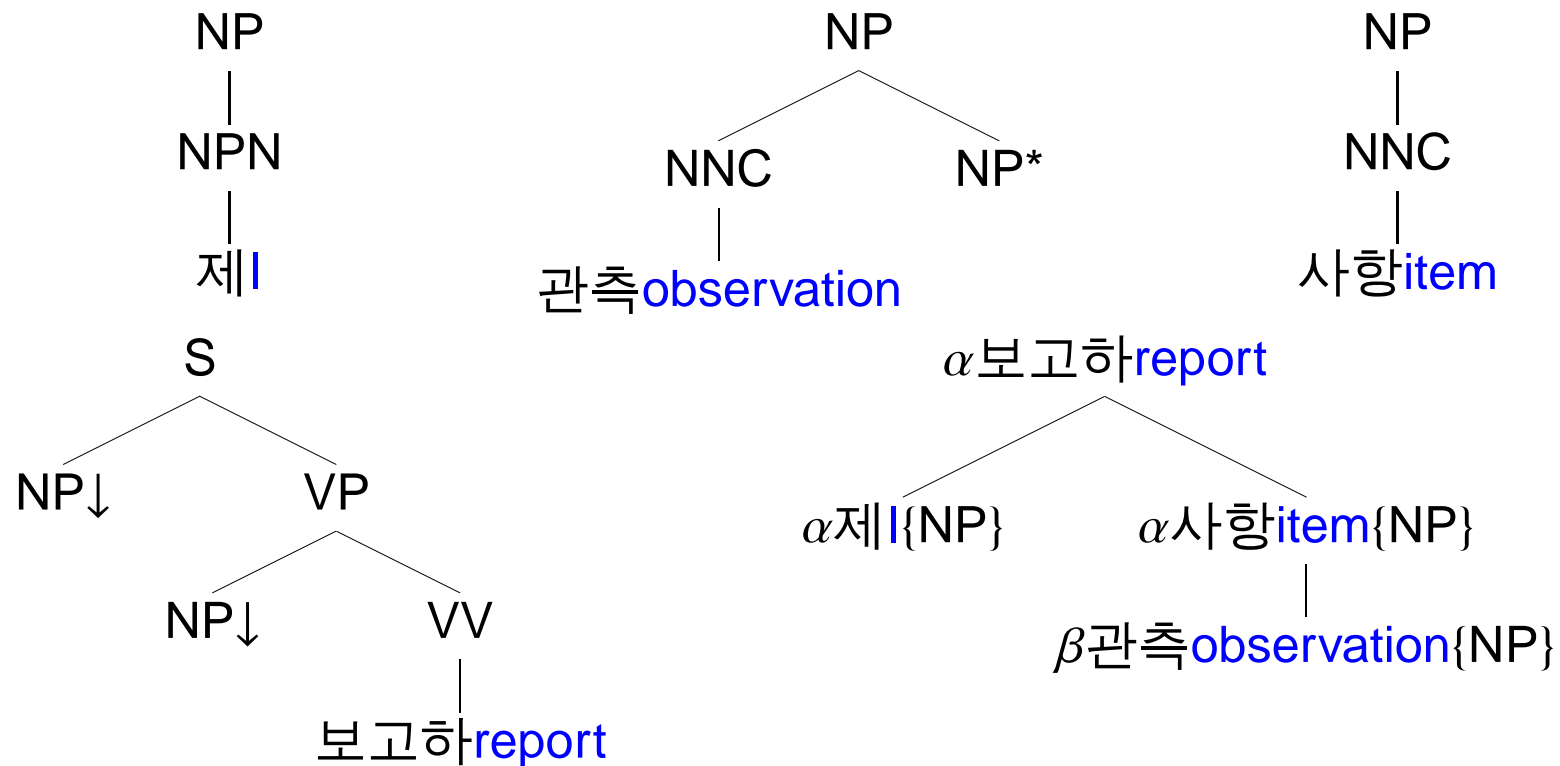
Korean Treebank

(S (NP-SBJ 제|I/NPN+가|nom/PCA)
(VP (NP-OBJ 관측observation/NNC
사항item/NNC+을acc/PCA)
보고하report/VV+었past/EPF+습니다decl/EFN)
./SFN)

→ I-Nom observation item-Acc report-Past-Decl

→ 'I reported the overvation items.'

LTAG Grammar and Derivation Tree using LexTract (Xia 2001)



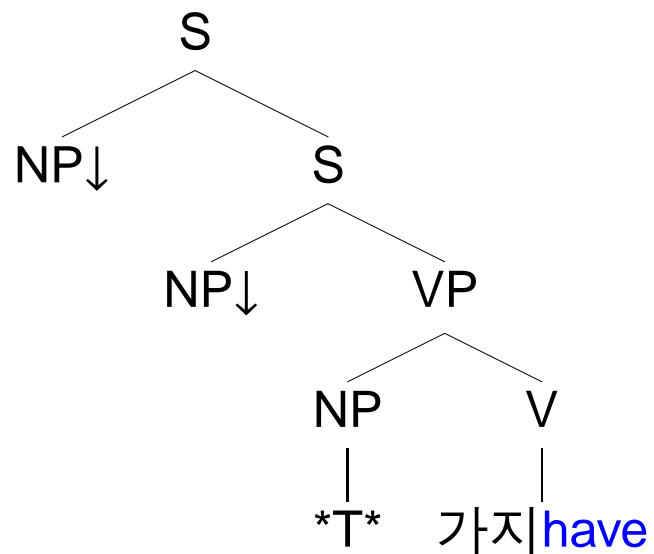
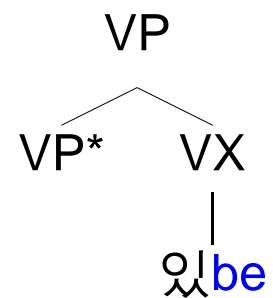
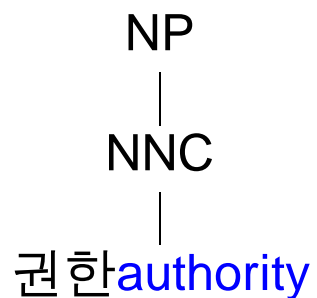
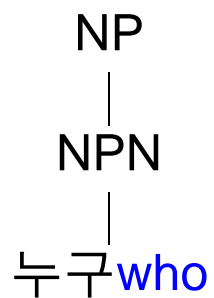
Korean Treebank

(S (NP-OBJ-1 권한authority/NNC+을acc/PCA)
(S (NP-SBJ 누구who/NPN+가|nom/PCA)
(VP (VP (NP-OBJ *T*-1)
가|지have/VV+고aux/EAU)
있be/VX+지|int/EFN))
?/SFN)

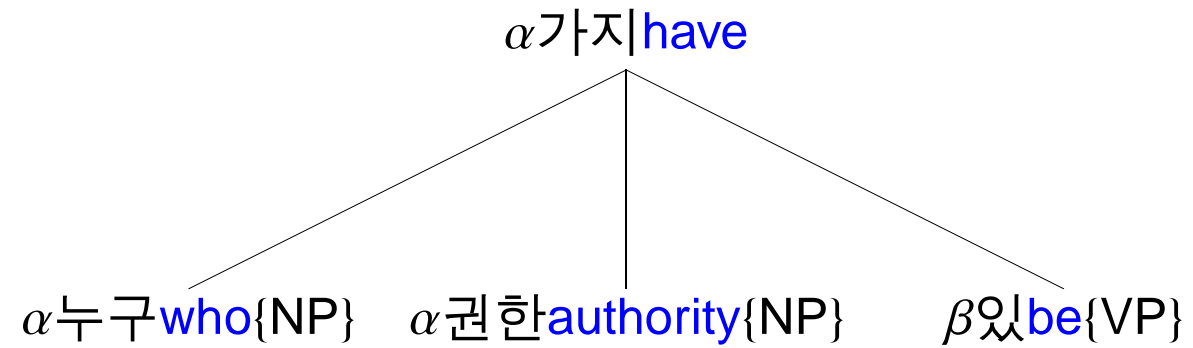
→ authority-Acc who-Nom have-AuxConnective be-Int

→ 'Who has the authority?'

LTAG Grammar for Korean using LexTract



LTAG Derivation Tree



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Motivation for Morphological Tagging

- Each substitution, adjunction is a relation between a pair of words
- Korean is an agglutinative language with a very productive inflectional system
- A fully inflected word seen in the training data will rarely occur in the unseen (test) data
- Sparse data problem is much worse than in English: the part-of-speech tags for inflected word forms are complex and can be novel in unseen data

Motivation for Morphological Tagging

- The morphological tagger provides lemma splitting plus part-of-speech tagging
- Instead of multiplying ambiguity in the parser, we choose to implement a statistical morphological tagger
(provides a single-best analysis of the input sentence)
- Both lemma splitting and tagging are trained using the Penn Korean Treebank (same training/test split as in the parser)
- Lexical stem and suffix information as well as part-of-speech information from the morphological tagger is used in the statistical parser

Example input and output from the morphological tagging phase

Input: 제가_관측_사항을_보고했습니다_.

Output: 제/NPN+가/PCA_관측/NNC_사항/NNC+을/PCA_
보고하/VV+있/EPF+습니다/EFN _./SFN

The part-of-speech tags for inflected word forms are complex and can be novel in unseen data

Evaluation of the Morphological Analyzer/Tagger

	unseen test data (3,717 words) precision/recall (%)
Treebank trained	95.78/95.39
Off-the-Shelf	29.42/31.25

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Morphological Analysis Incorporated into the Statistical Model

In each probability model used in the parser where inflected word forms are used we incorporate the output of the morph tagger as a backoff level

For example, take the probability model for adjunction:

$$P_a(t, \eta \rightarrow t') = Pr(t', p', w' \mid \eta, t, w, p) \quad (1)$$

$$= Pr(t' \mid \eta, t, w, p) \times \quad (2)$$

$$Pr(p' \mid t', \eta, t, w, p) \times$$

$$Pr(w' \mid p', t', \eta, t, w, p)$$

Morphological Analysis Incorporated into the Statistical Model

- e_1 = lexicalized model using stems;
 e_2 = part-of-speech tags from the morphological tagger:

$$Pr_{e_1} = Pr(t' \mid \eta, t, w, p)$$

$$Pr_{e_2} = Pr(t' \mid \eta, t, p)$$

- The backoff model is computed as follows:

$$\lambda(c) \times Pr_{e_1} + (1 - \lambda(c)) \times Pr_{e_2}$$

Parsing Experiment: Training and Test Data

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Example derivation reported by the statistical parser

Index	Word	Gloss	POS tag (morph)	Elem Tree	Node Address	Subst/ Adjoin
0	모든	every	DAN	$\beta NP^*=1$	root	2
1	호출	call	NNC	$\beta NP^*=1$	root	2
2	대호+는	sign-topic	NNC+PAU	$\alpha NP=0$	0	6
3	매일	everyday	ADV	$\beta VP^*=25$	1	6
4	24		NNU	$\beta NP^*=1$	0	5
5	시+에	hour-at	NNX+PAD	$\beta VP^*=17$	1	6
6	바뀌+게	switch-aux	VV+ECS	$\alpha S-NPs=23$	-	TOP
7	되+지요	be-decl	VX+EFN	$\beta VP^*=13$	1	6
8	.		SFN	-	-	-

Parser evaluation results

	On training data	On unseen test data (425 sents)
Current Work	97.58	75.7
(Yoon et al. 1997)	–	52.29/51.95 P/R

Summary

- First LTAG-based parsing system for Korean.
- LTAG-based statistical parsing is feasible for a language with free word order and complex morphology.
- Our system has been successfully incorporated into a Korean/English machine translation system as source language analysis component.

Summary

- The tagger/analyzer obtained the correctly disambiguated morphological analysis with 95.78/95.39%
- The statistical parser obtained a dependency accuracy of 75.7%
- These performance results are better than an existing off-the-shelf Korean morphological analyzer and parser run on the same data.

Grazie . . .

Experiments with and without the Morphological Tagger

- Even the part-of-speech tags are often unseen in the test data
- When we lexicalize trees we use words from the training data and for unknown words the output of a part-of-speech tagger
- Without a morphological tagger the lexicalization step becomes infeasible
(We can annotate the Treebank with a new smaller tagset, but the number of trees for unknown words explodes)
- Thus, we could not easily compare parsing with and without a morphological tagger