

Lec 4. Syntactic Structure and Dependency Parsing

1. Syntactic Structure: Consistency and Dependency

-Two views of linguistic structure:

*Constituency=Phrase

structure grammar = context-free grammars(CFGs)

eg) VP → V, pp

S → NP VP

*Dependency structure

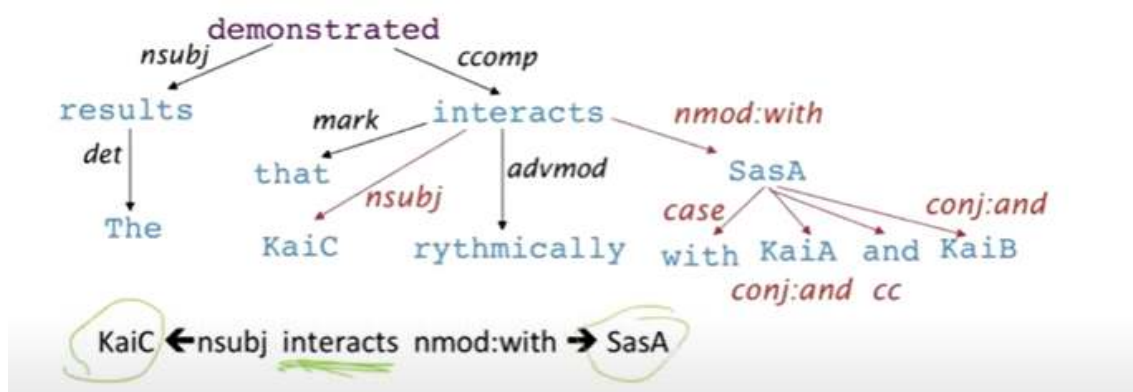
shows which words depend on (modify, attach to, or are arguments of) which other words → neural dependency parser



단어의 중의성, 모호성 등의 문제들 ...

뉴스제목은 특히 더 모호한 경향있음

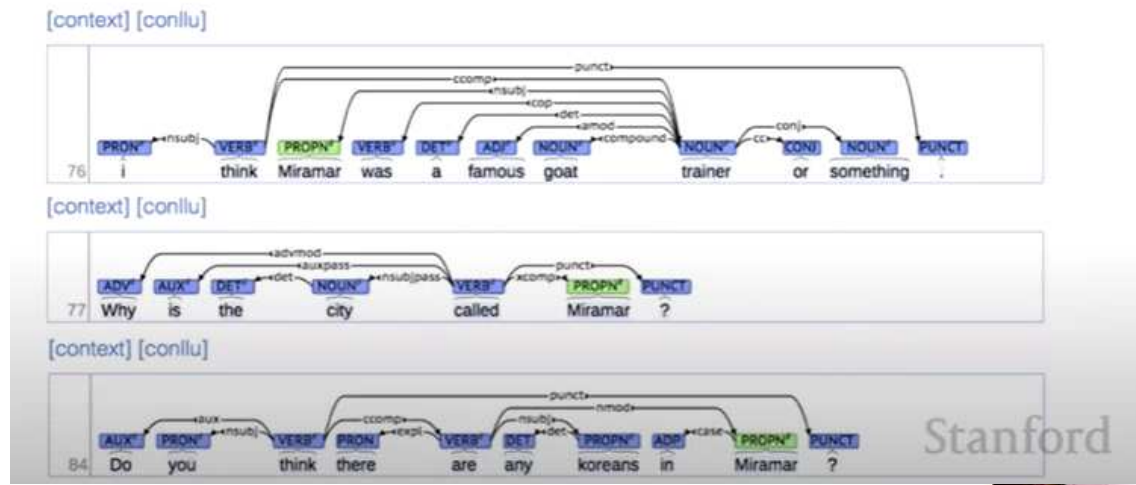
Modifier Ambiguity/VP attachment ambiguity



Dependency paths help extract semantic interpretation

*fake root

2. Dependency Grammar and Treebank



What are the sources of information for dependency parsing?

1. Bilexical affinities The dependency [discussion → issues] is plausible
2. Dependency distance Most dependencies are between nearby words
3. Intervening material Dependencies rarely span intervening verbs or punctuation
4. Valency of heads How many dependents on which side are usual for a head?

The rise of annotated data & Universal Dependencies tree

slower and less useful than writing a grammar

-> but, Reusability of the labor, broad coverage, not just a few intuitions, frequencies and distrivutional infrom, a way to evaluate NLP systems

3. Transition-based dependency parsing

-Dependency Parsing

A sentence is parsed by choosing for each word what other word(including ROOT) it is a dependent of Usually some constraints:

-> Only one word is a dependent of ROOT

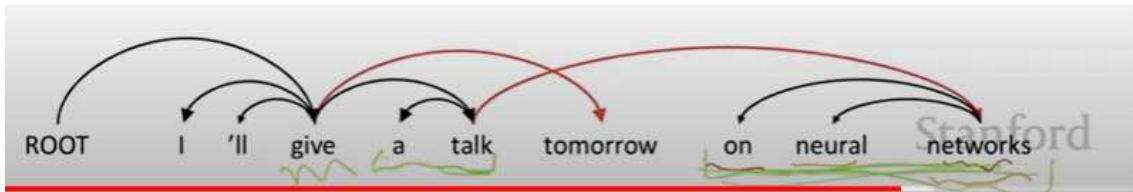
-> Don't want cycles A->B, B->A

This makes the dependencies a tree

Final issue is whether arrows can cross (be non-projective) or not

-Projectivity

<projective parse>: There are no crossing dependency arcs when the words are



laid out in their linear order, with all arcs above the words.

Dependencies corresponding to a CFG tree must be projective.

1. Dynamic programming
2. Graph algorithms
3. Constraint Satisfaction
4. “Transition-based parsing” or “deterministic dependency parsing”

-Greedy transition-based parsing: a simple form of greedy discriminative dependency parser. bottom-up actions

- The parser has:
 - a stack σ , written with top to the right
 - which starts with the ROOT symbol
 - a buffer β , written with top to the left
 - which starts with the input sentence
 - a set of dependency arcs A
 - which starts off empty
 - a set of actions

Start: $\sigma = [\text{ROOT}]$, $\beta = w_1, \dots, w_n$, $A = \emptyset$

1. Shift $\sigma, w_i | \beta, A \rightarrow \sigma | w_i, \beta, A$

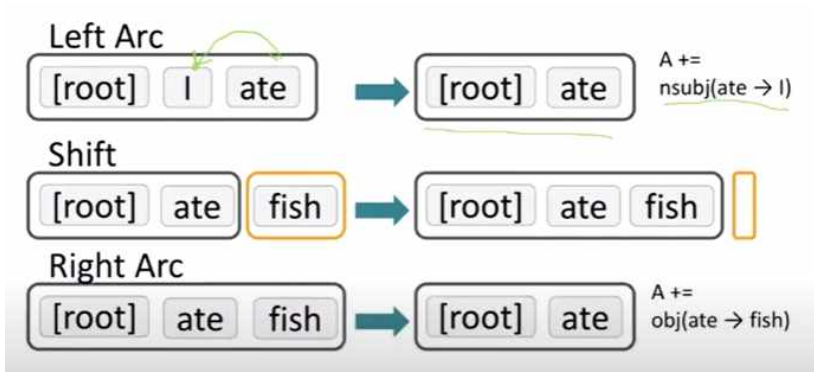
2. Left-Arc_r $\sigma | w_i | w_j, \beta, A \rightarrow \sigma | w_j, \beta, A \cup \{r(w_j, w_i)\}$

3. Right-Arc_r $\sigma | w_i | w_j, \beta, A \rightarrow \sigma | w_i, \beta, A \cup \{r(w_i, w_j)\}$

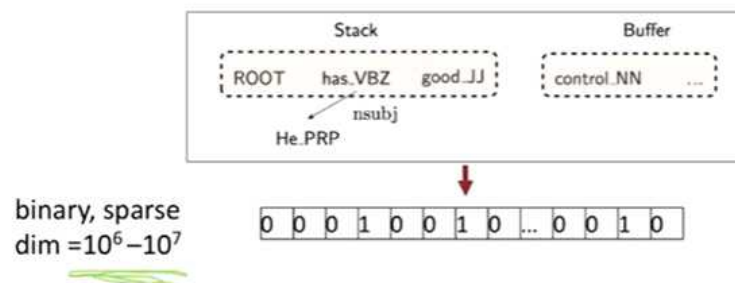
Finish: $\sigma = [w]$, $\beta = \emptyset$

-MaltParser -> 다음 action 어떻게 정할 것인지?

-> Stand back, machine learning, each action is predicted by a discriminative Classifier (eg. softmax classifier), no search (in the simple form), model's accuracy -> fractionally below the state of the art in dependency parsing but very fast linear time parsing with high accuracy- good for web!



Conventional Feature Representation



Indicator features

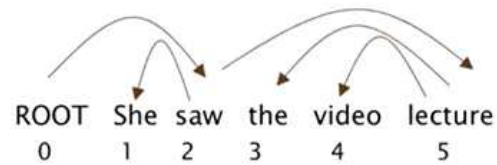
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$s1.w = \text{good} \wedge s1.t = \text{JJ}$
 $s2.w = \text{has} \wedge s2.t = \text{VBZ} \wedge s1.w = \text{good}$
 $lc(s2).t = \text{PRP} \wedge s2.t = \text{VBZ} \wedge s1.t = \text{JJ}$
 $lc(s2).w = \text{He} \wedge lc(s2).l = \text{nsubj} \wedge s2.w = \text{has}$

Stanford

dependency accuracy

Evaluation of Dependency Parsing: (labeled) dependency accuracy



$$\text{Acc} = \frac{\text{\# correct deps}}{\text{\# of deps}}$$

$$\text{UAS} = 4 / 5 = 80\%$$

Gold			
1	2	She	nsubj
2	0	saw	root
3	5	the	det
4	5	video	nn
5	2	lecture	obj

Parsed			
1	2	She	nsubj
2	0	saw	root
3	4	the	det
4	5	video	nsubj
5	2	lecture	ccomp

Stanford