

# Lecture 4 - Syntactic Structure and Dependency Parsing

## 1. Syntactic Structure : Consistency and Dependency

- **Consistency**
  - constituency = phrase
  - structure grammar = context-free grammars
  - **Phrase structure** organizes words into nested constituents
- **Dependency**
  - Dependency structure shows which words depend on (modify, attach to, or are arguments of) which other words.
- Ambiguity
  1. Phrase attachment ambiguity
  2. Coordination Scope ambiguity

## 2. Dependency Grammar and Treebanks

- Dependency syntax postulates that syntactic structure consists of relations between lexical items, normally binary asymmetric relations “arrows” called dependencies
  - An arrow connects a head (governor, superior, regent) with a dependent (modifier, inferior, subordinate)
  - Usually, dependencies form a tree (a connected, acyclic, single-root graph)
- Dependency Parsing

- A sentence is parsed by choosing for each word what other word it is a dependent of
- Projectivity
  - There are no crossing dependency arcs when the words are laid out in their linear order, with all arcs above the words
- The rise of annotated data : treebank's advantages
  - Reusability of the labor
  - Broad coverage
  - Frequencies and distributional information
  - A way to evaluate NLP system
- Dependency Conditioning Preferences
  1. Bilexical affiniteis
  2. Dependency distance
  3. Intervening material
  4. Valency of heads

### **3. Transition-based dependency parsing**

- Methods of Dependency Parsing
  1. Dynamic programming
  2. Graph algorithms
  3. Constraint Stisfaction
  4. "Transition-based parsing" or "deterministic dependency parsing"
    - Greedy transition-based parsing [Nivre 2003]

## Greedy transition-based parsing [Nivre 2003]



- A simple form of greedy discriminative dependency parser
- The parser does a sequence of bottom-up actions
  - Roughly like “shift” or “reduce” in a shift-reduce parser, but the “reduce” actions are specialized to create dependencies with head on left or right
- The parser has:
  - a stack  $\sigma$ , written with top to the right
    - which starts with the ROOT symbol
  - a buffer  $\beta$ , 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

Stanford

## Basic transition-based dependency parser

**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<sub>r</sub>  $\sigma | w_i | w_j, \beta, A \rightarrow \sigma | w_j, \beta, A \cup \{r(w_j, w_i)\}$

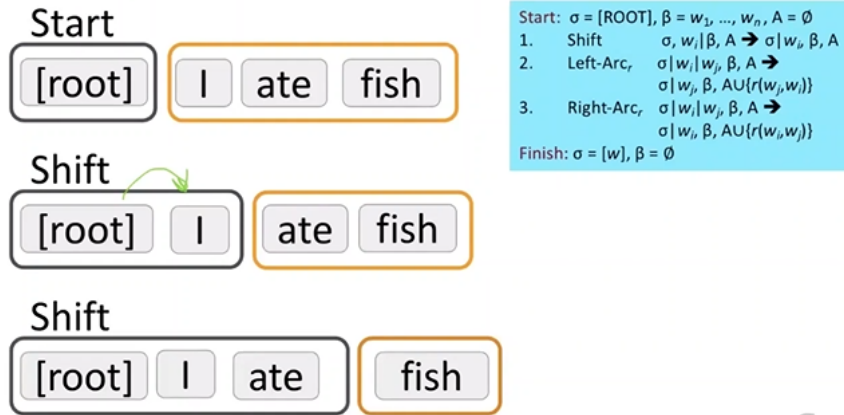
3. Right-Arc<sub>r</sub>  $\sigma | w_i | w_j, \beta, A \rightarrow \sigma | w_i, \beta, A \cup \{r(w_i, w_j)\}$

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

## Arc-standard transition-based parser

(there are other transition schemes ...)

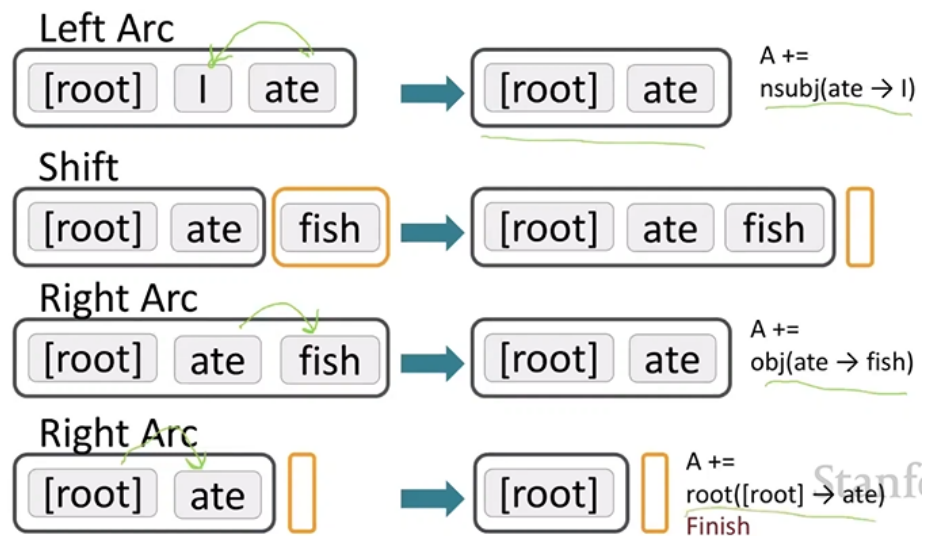
Analysis of "I ate fish"



Stanf

## Arc-standard transition-based parser

Analysis of "I ate fish"



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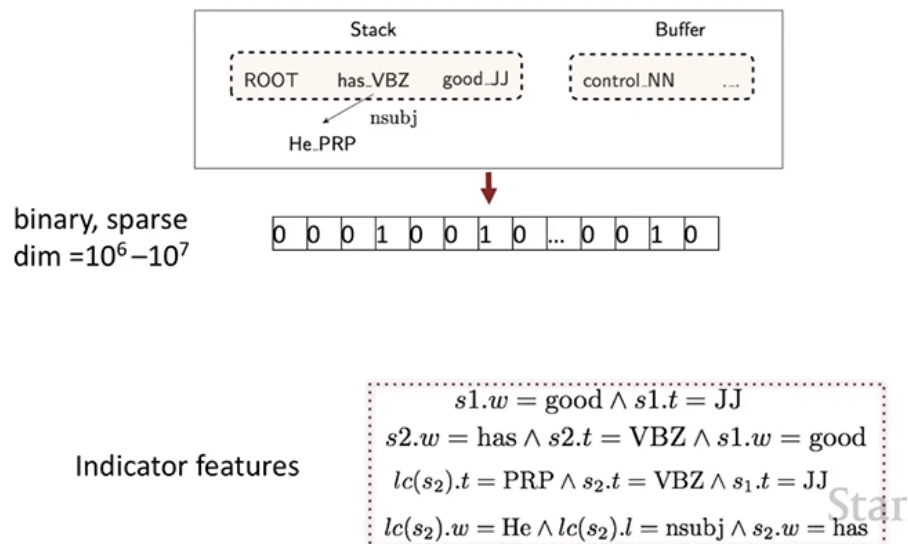
- MaltParser [Nivre and Hall 2005]

## MaltParser [Nivre and Hall 2005]



- We have left to explain how we choose the next action 🤖
  - Answer: Stand back, I know machine learning!
- Each action is predicted by a discriminative classifier (e.g., softmax classifier) over each legal move
  - Max of 3 untyped choices; max of  $|R| \times 2 + 1$  when typed
  - Features: top of stack word, POS; first in buffer word, POS; etc.
- There is NO search (in the simplest form)
  - But you can profitably do a beam search if you wish (slower but better): You keep  $k$  good parse prefixes at each time step
- The model's accuracy is *fractionally* below the state of the art in dependency parsing, but
- It provides **very fast linear time parsing**, with high accuracy – great for parsing the web

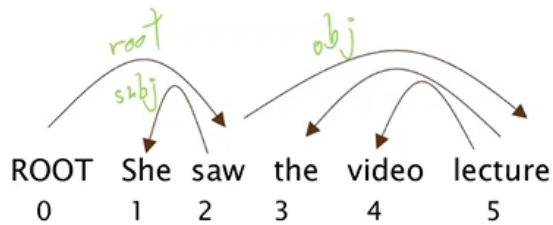
## Conventional Feature Representation



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- Evaluation of Dependency Parsing

## Evaluation of Dependency Parsing: (labeled) dependency accuracy



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

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

$$\text{LAS} = 2 / 5 = 40\%$$

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