

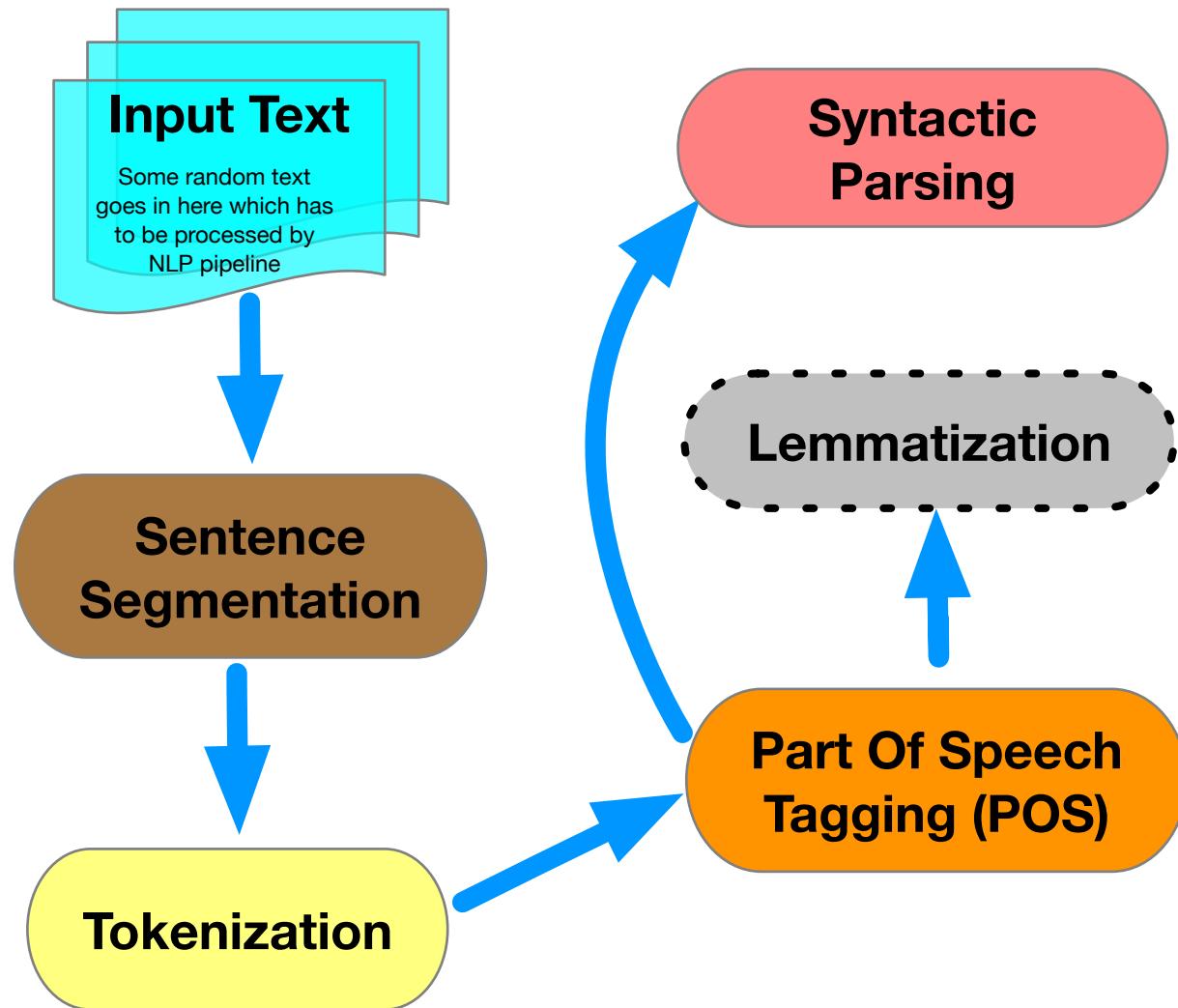
# Special Topics in Natural Language Processing

## CS6980

Ashutosh Modi  
CSE Department, IIT Kanpur



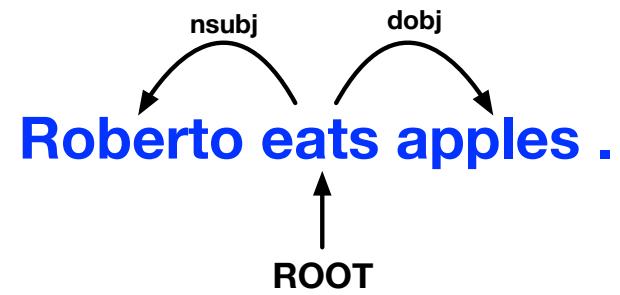
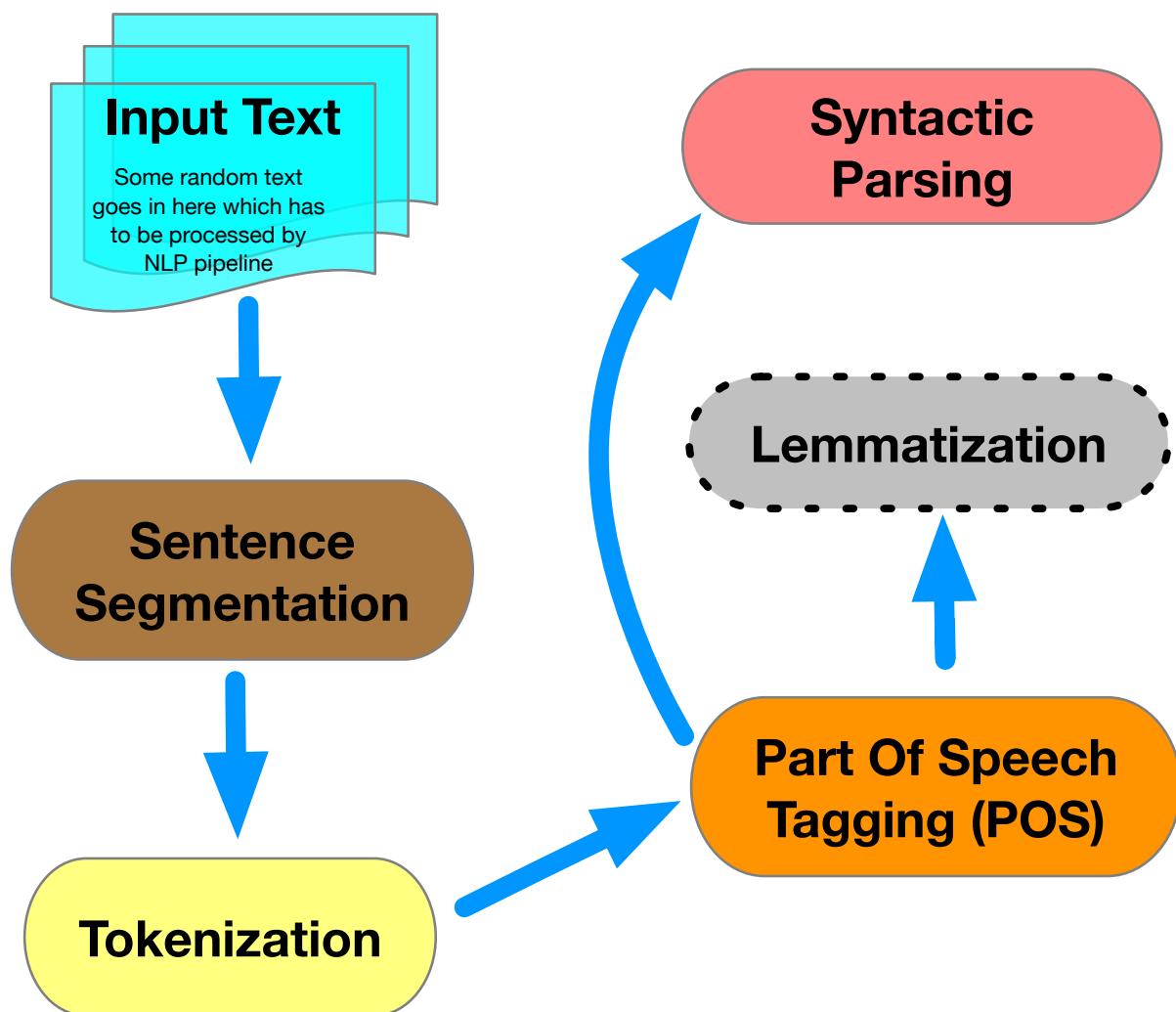
Lecture 19: Parsing 5  
Feb 24, 2020

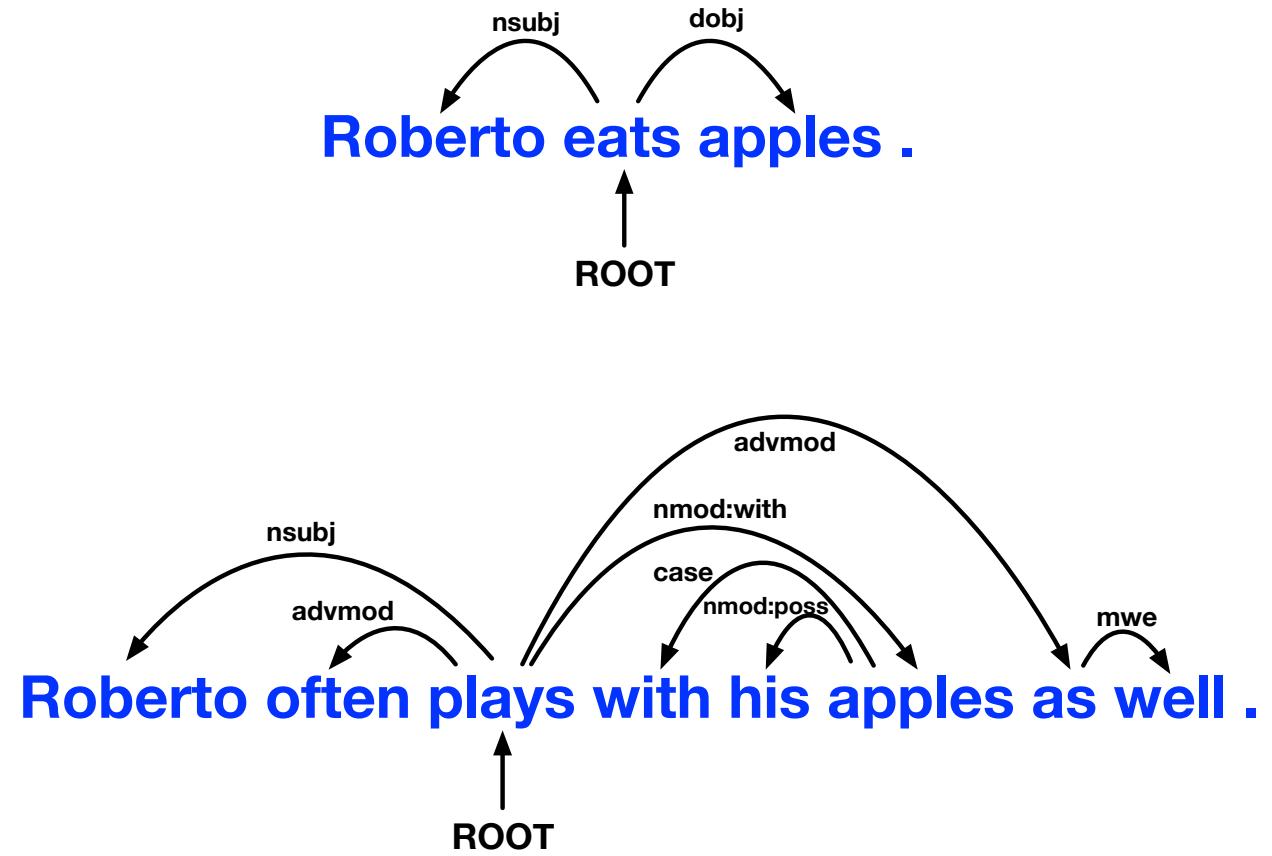
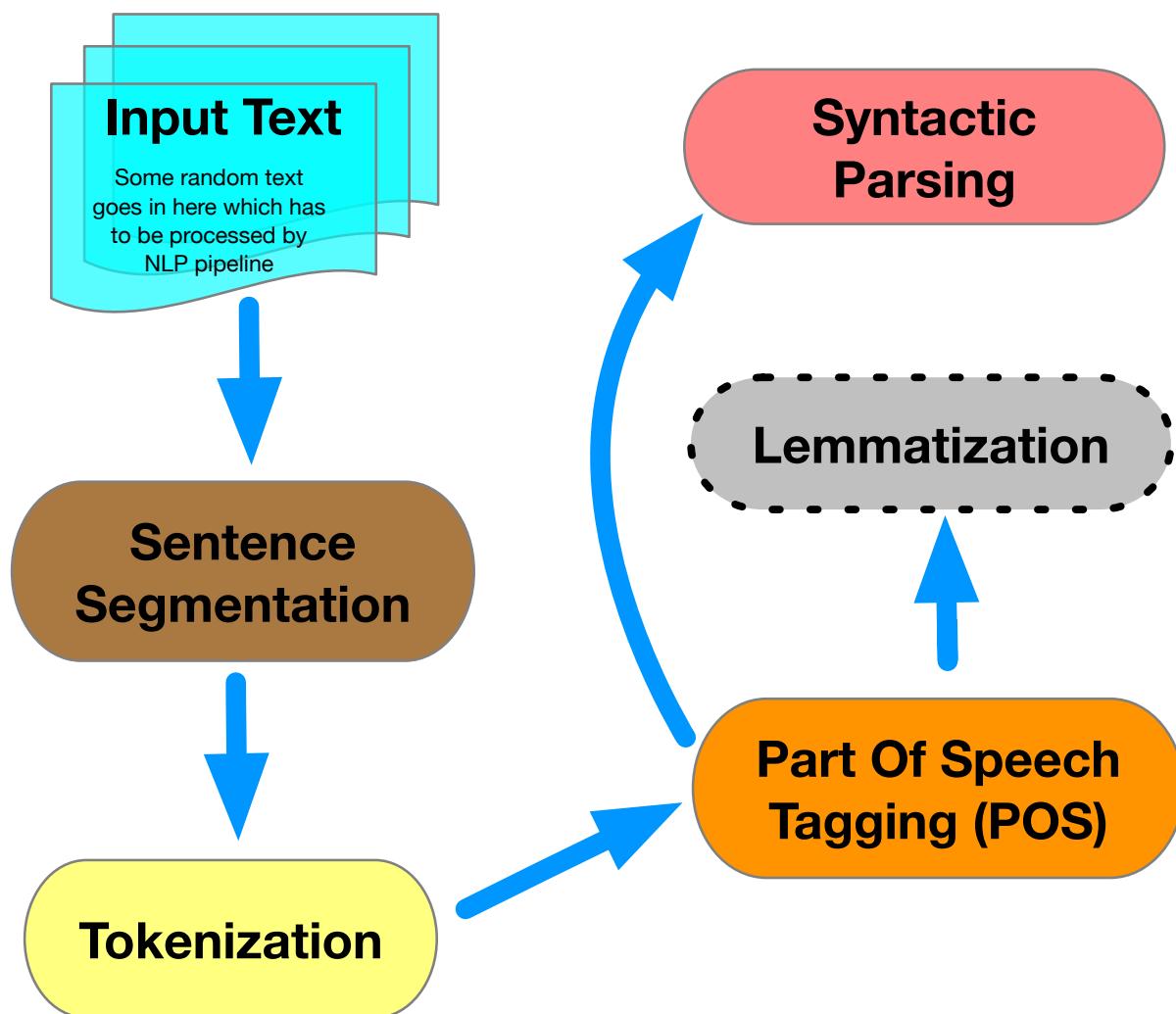


*Sentences in a language follow the grammar which is formally specified via the Syntax (rules of the language)*

*Parsing is the process for syntactically analyzing the sentences*

*One of the popular Parsing method is Dependency Parsing*





# Syntactic Analysis via Dependency Parsing

- Analyze language in terms of relations between words.
- Earlier versions of Dependency Grammar formulated by Panini in 4 B.C.E.
- Modern work done by French Linguist Tesnière (1959)
- Many different dependency grammar frameworks have been proposed: Prague School's Functional Grammar, Hudson Grammar, etc.

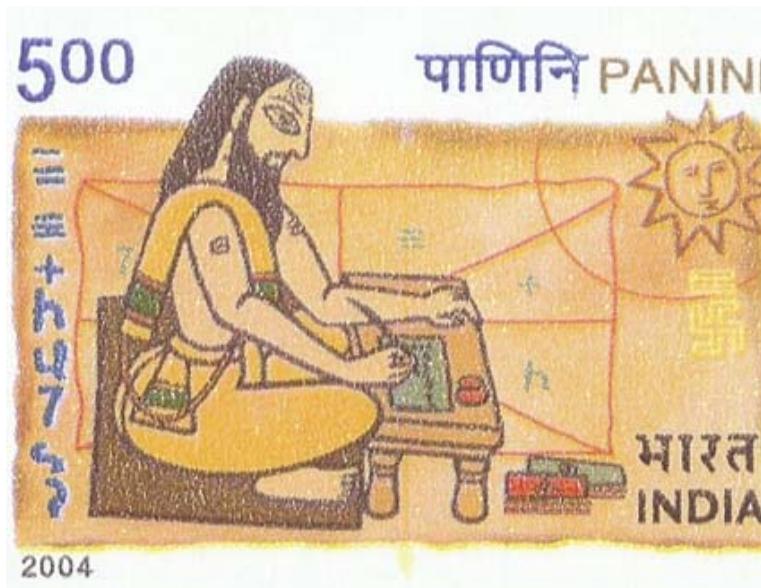


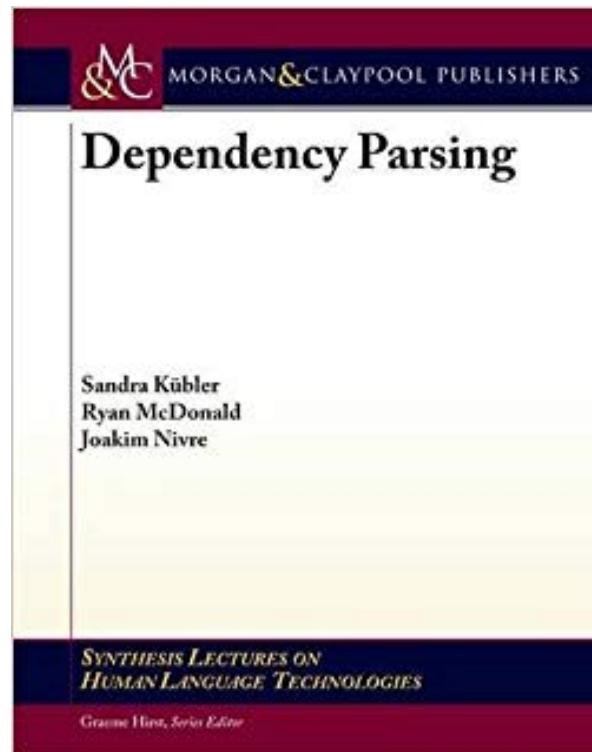
Image: <http://www-history.mcs.st-andrews.ac.uk/PictDisplay/Panini.html>



# Main Reference

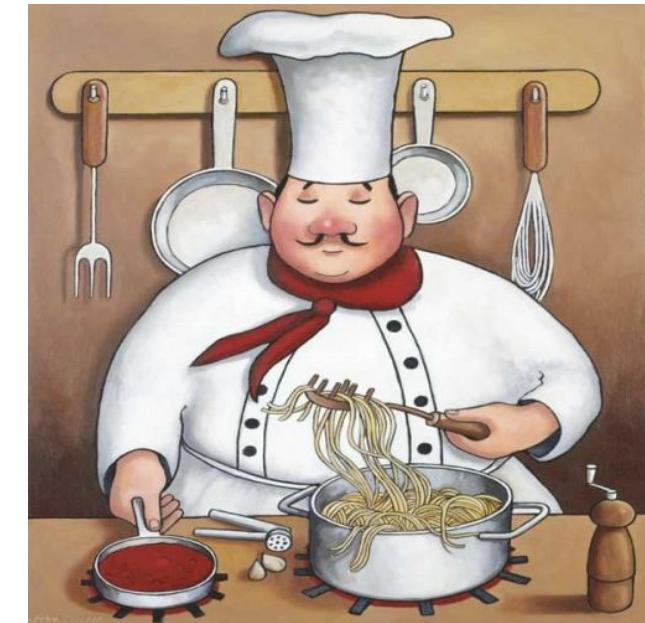
## Dependency Parsing

Sandra Kübler, Ryan McDonald, Joakim Nivre  
(Synthesis Lectures on Human Language Technologies,  
Morgan & Claypool Publishers)



# Dependency Grammar

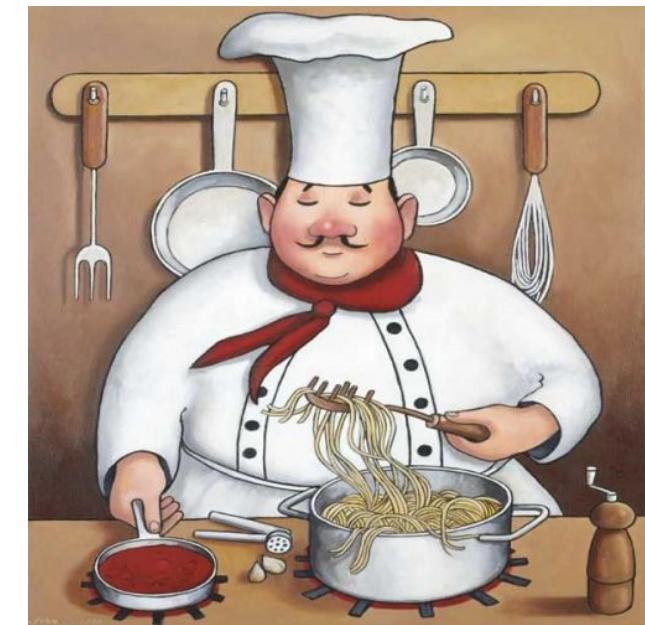
John cooked pasta.



# Dependency Grammar

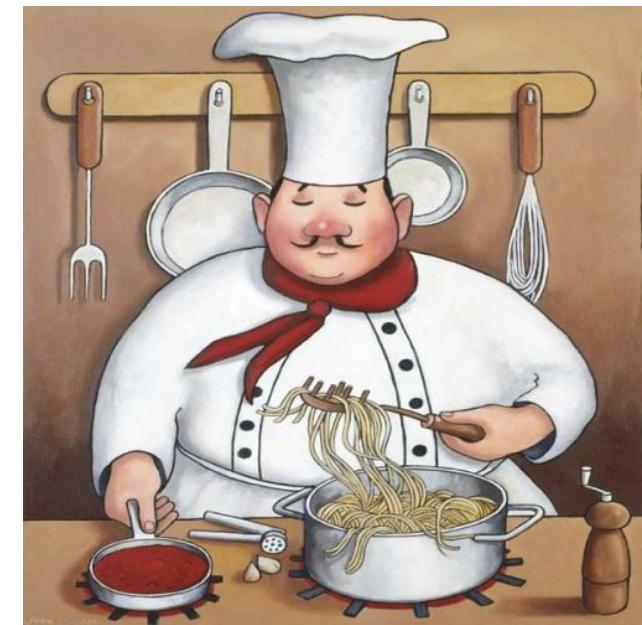
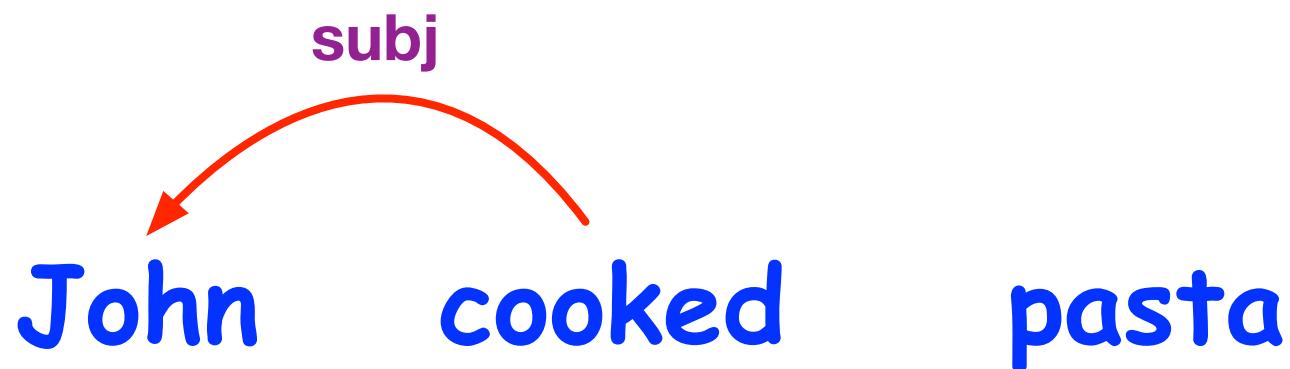
- Words in a sentence are related to each other via binary asymmetric relations called as dependency relations

John cooked pasta .



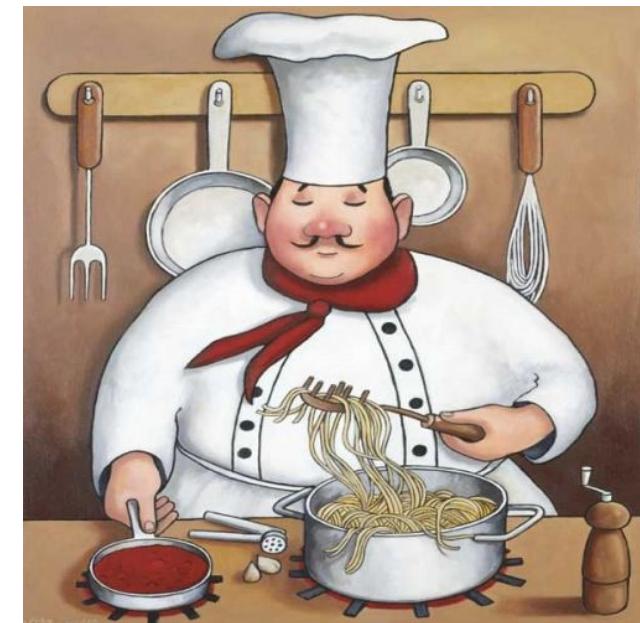
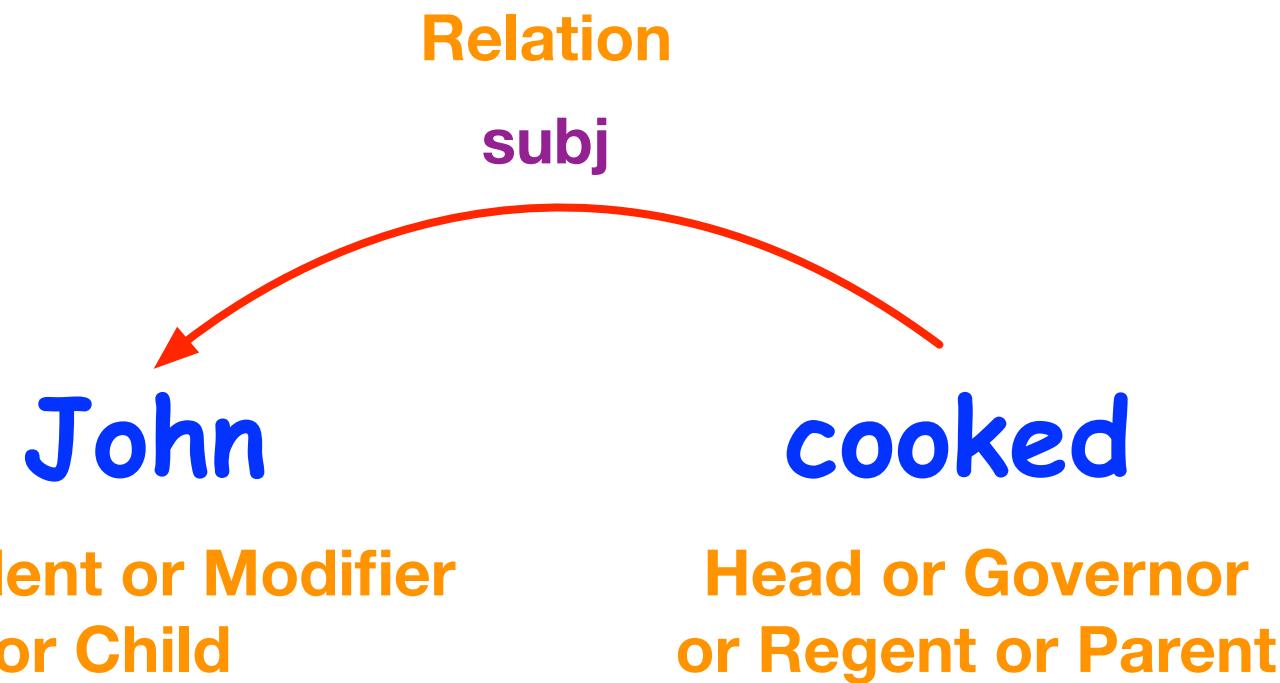
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- Words in a sentence are related to each other via binary asymmetric relations called as dependency relations



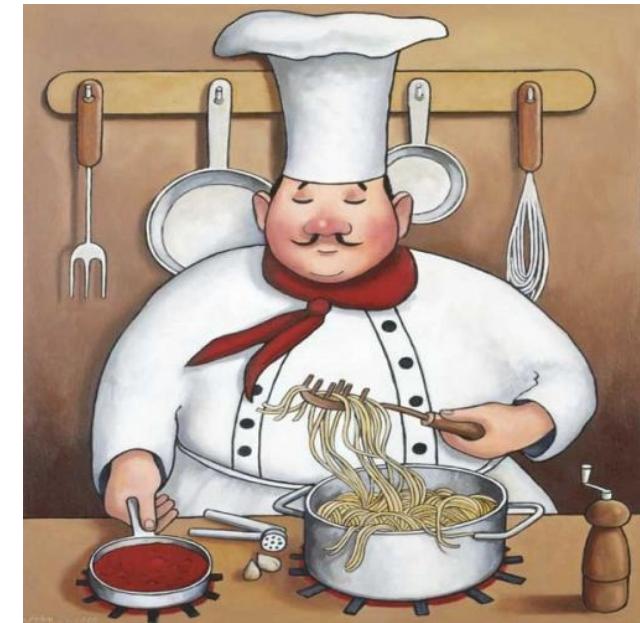
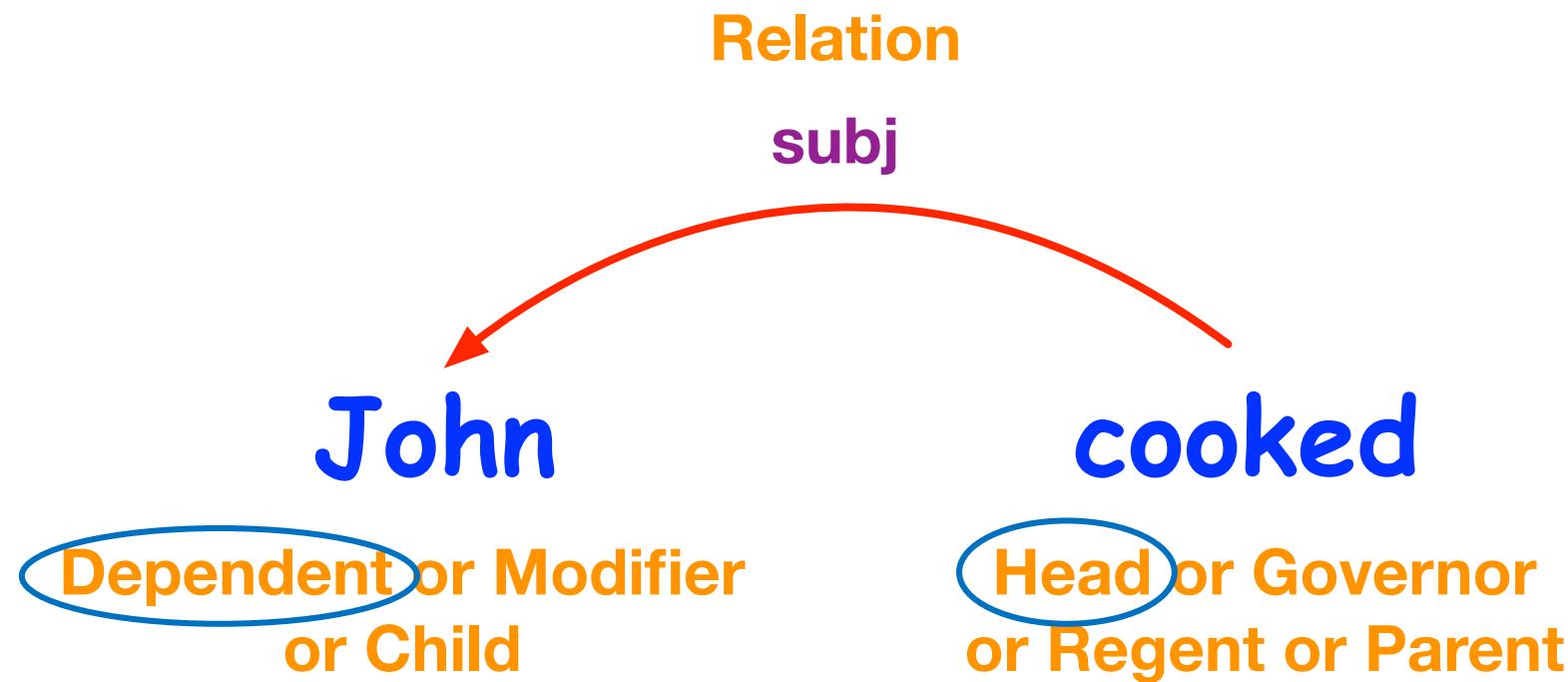
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- Words in a sentence are related to each other via binary asymmetric relations called as dependency relations



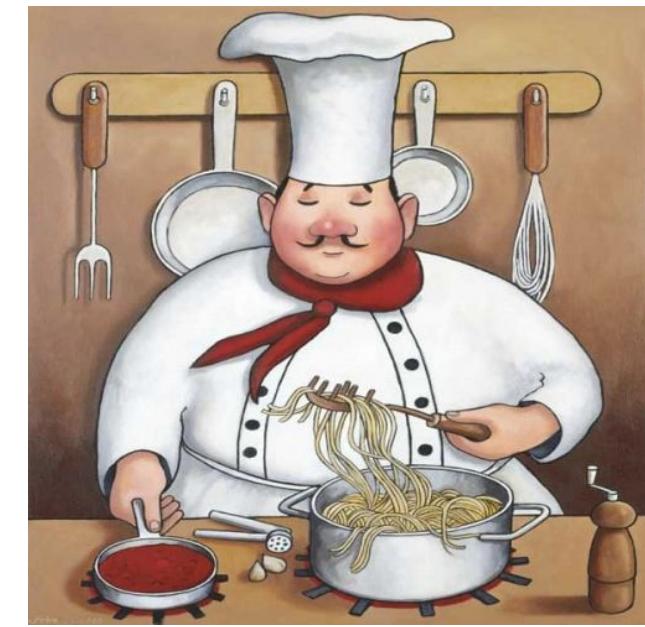
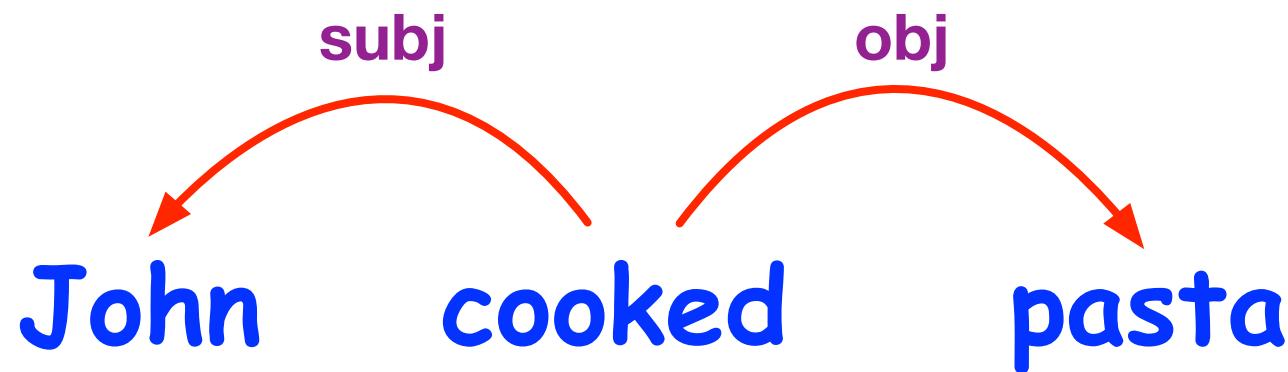
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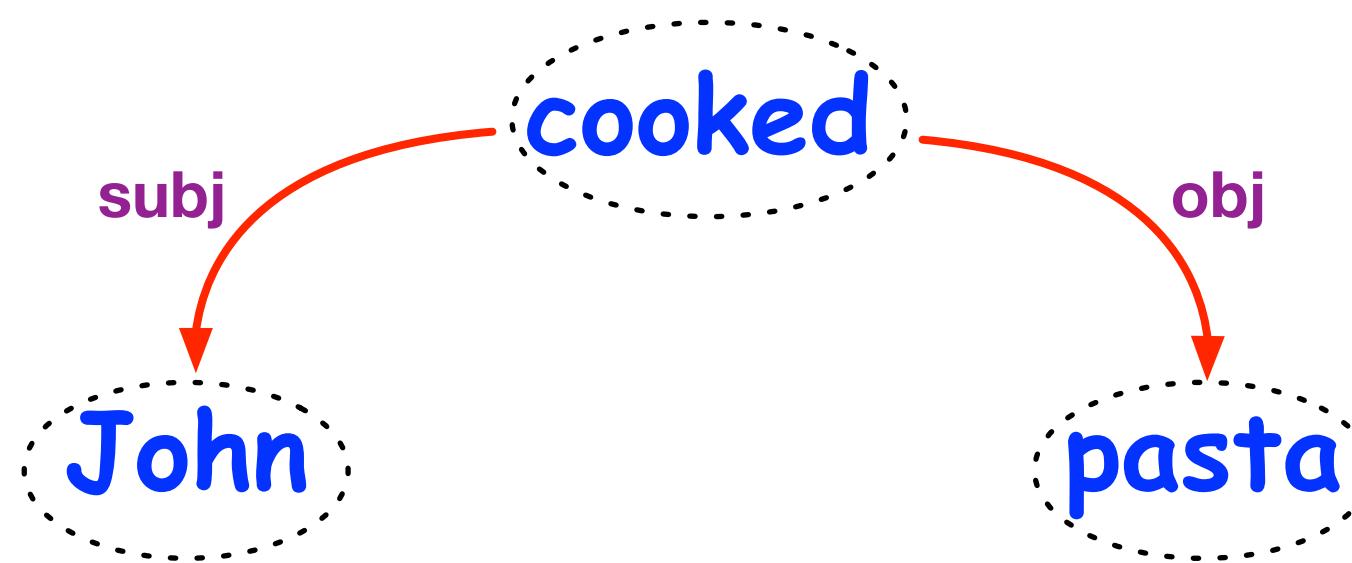
# Dependency Grammar

- Words in a sentence are related to each other via binary asymmetric relations called as dependency relations



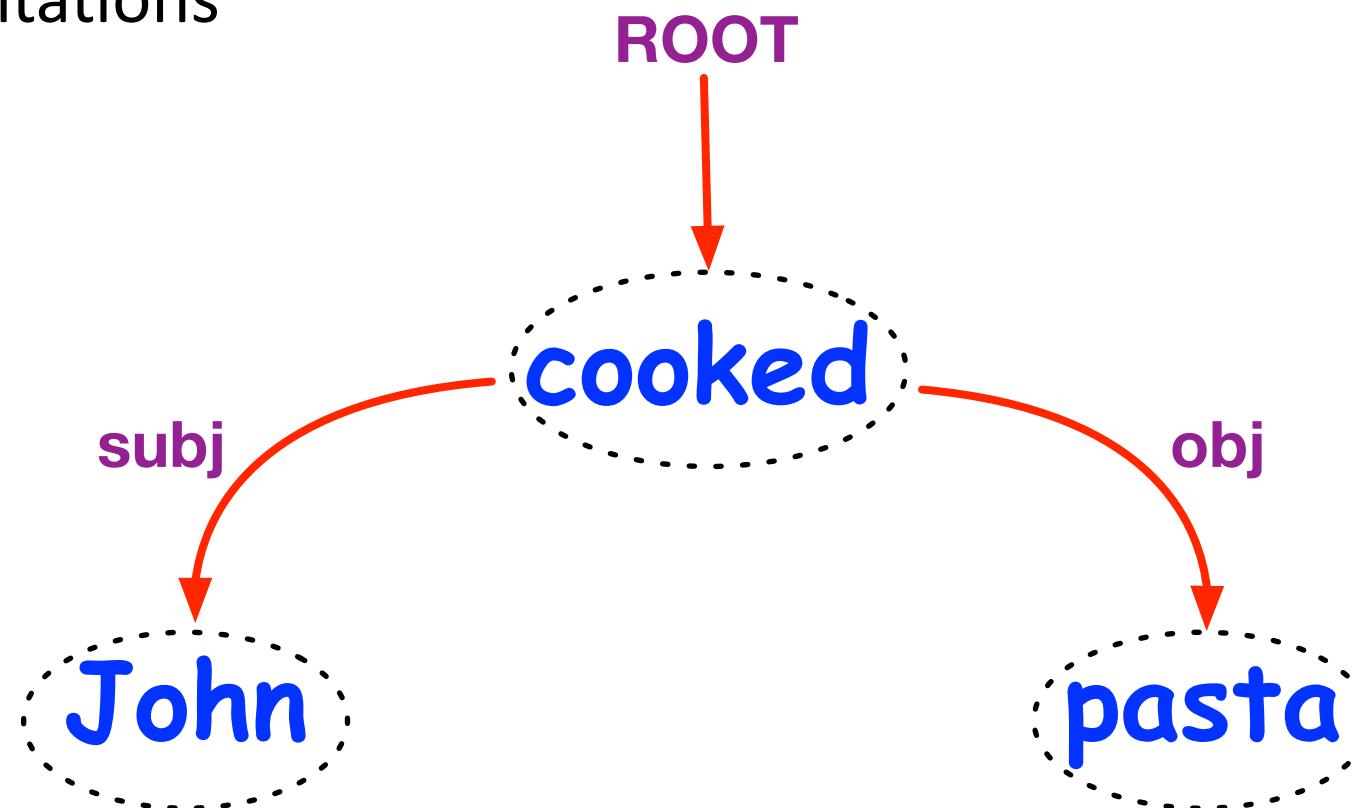
# Dependency Graph

- Labeled Directed Graph
- Nodes → Words
- Labelled Edges → Dependency Relations



# Dependency Graph

- Every dependency graph has a **ROOT**
- It is for technical convenience which helps in computational implementations

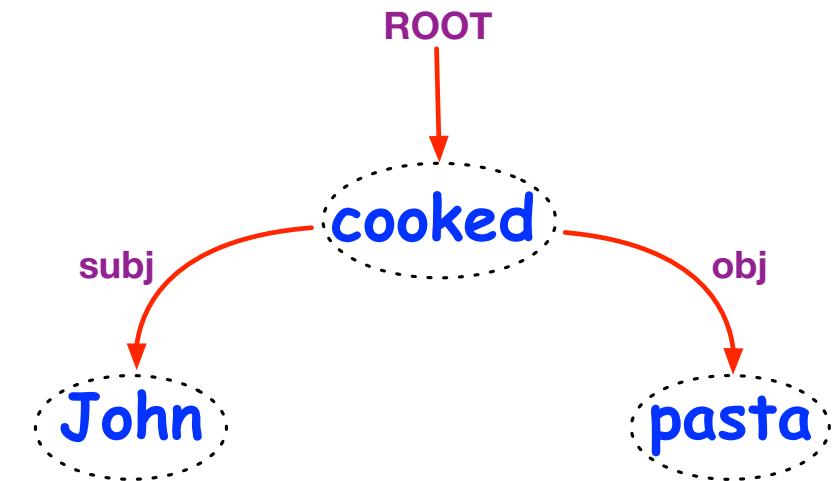


# Dependency Parsing (DP)

- The task of automatically analyzing dependency structure of a given input sentence.

John **cooked** pasta.

Dependency  
Parsing



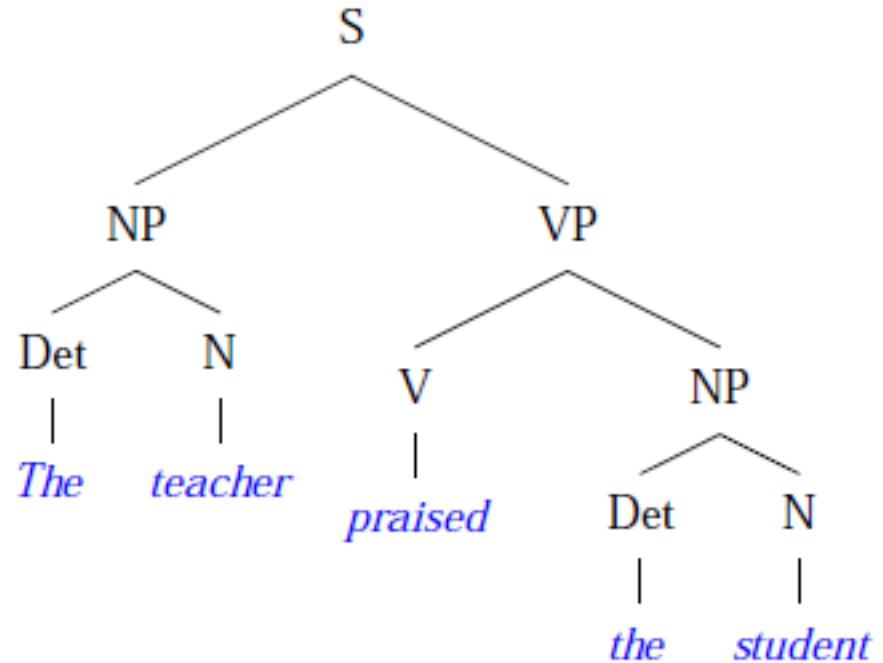
# Why Dependency Parsing (DP)?

- Helpful in resolving ambiguities.
- Useful for extracting information due to relationship between words
- Provides input to many NLP tasks, like Semantic Role Labeling, Question Answering, etc.
- Useful for analyzing languages with free word order



# Constituent Parsing (CFG Parsing)

- Sentences divided into constituent phrases
- Relationship between words is not binary as in Dependency parsing
- Defined by context free grammar rules



$S \rightarrow NP\ VP$

$NP \rightarrow N\ Det$

$N \rightarrow \text{dog, teacher}$

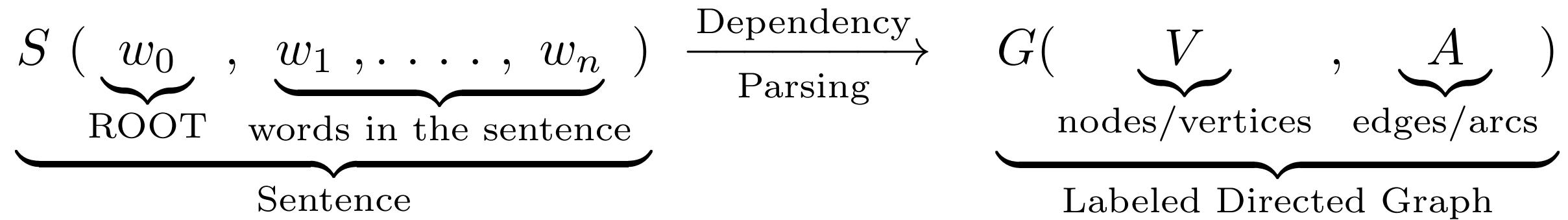
$VP \rightarrow V\ NP$

# Dependency Parsing (DP)

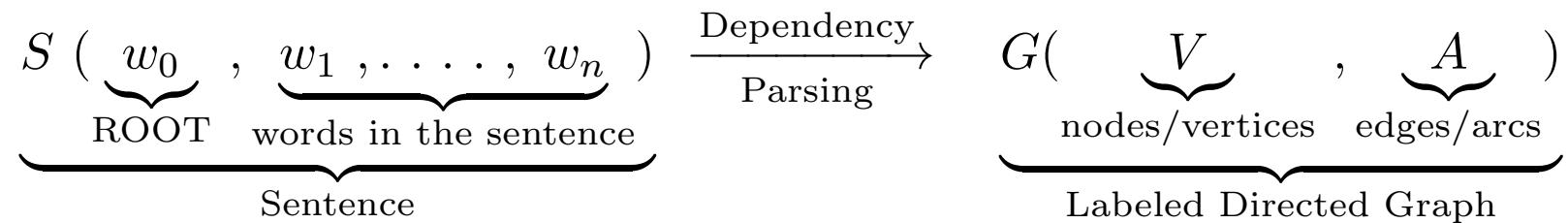
$$S(w_0, w_1, \dots, w_n) \xrightarrow[\text{Parsing}]{\text{Dependency}} G(V, A)$$



# Dependency Parsing (DP)



# Terminology



Dependency Relations

$$R = \{r_1, r_2, \dots, r_m\}$$

Nodes

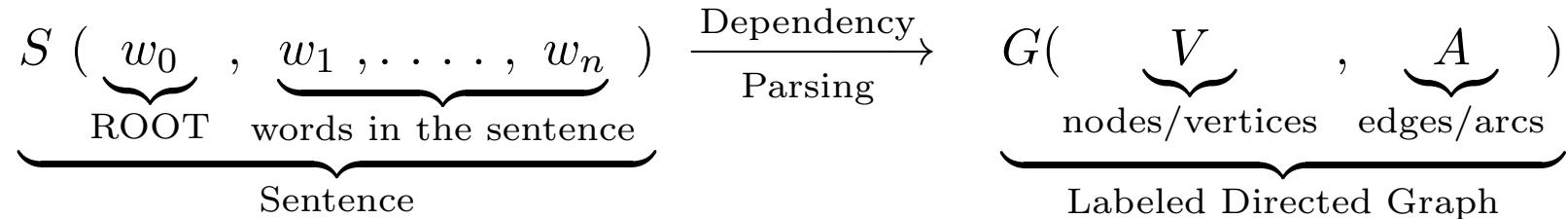
$$V \subseteq \{w_0, w_1, \dots, w_n\}$$

Arcs

$$A \subseteq V \times R \times V$$



# Terminology



Dependency Relations

$$R = \{r_1, r_2, \dots, r_m\}$$

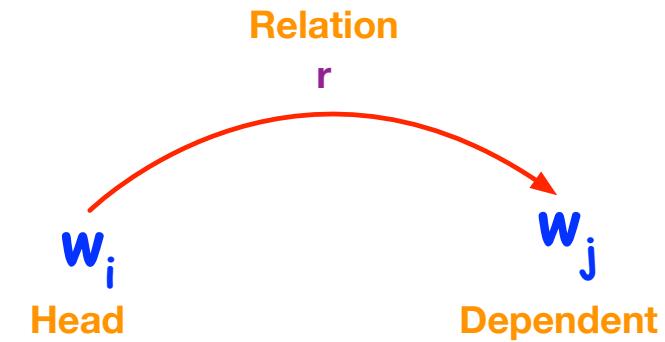
Nodes

$$V \subseteq \{w_0, w_1, \dots, w_n\}$$

Arcs

$$A \subseteq V \times R \times V$$

$$(w_i, r, w_j)$$



# Constraints



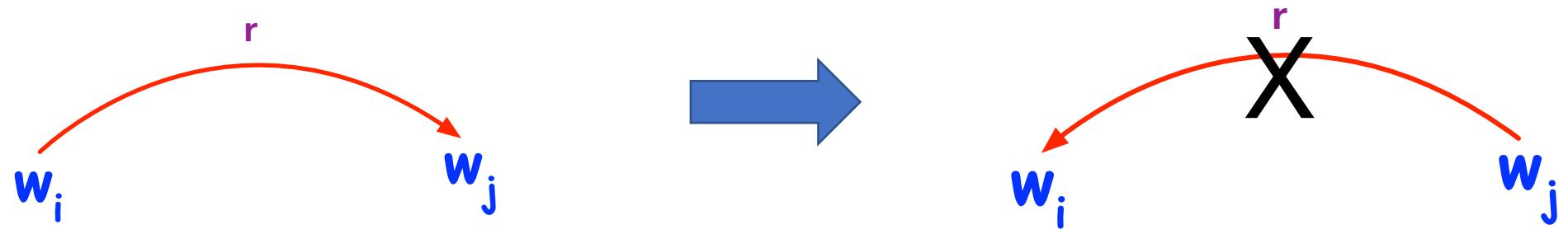
# Constraints

*if  $(w_i, r, w_j) \in A$  then  $(w_j, r, w_i) \notin A$*



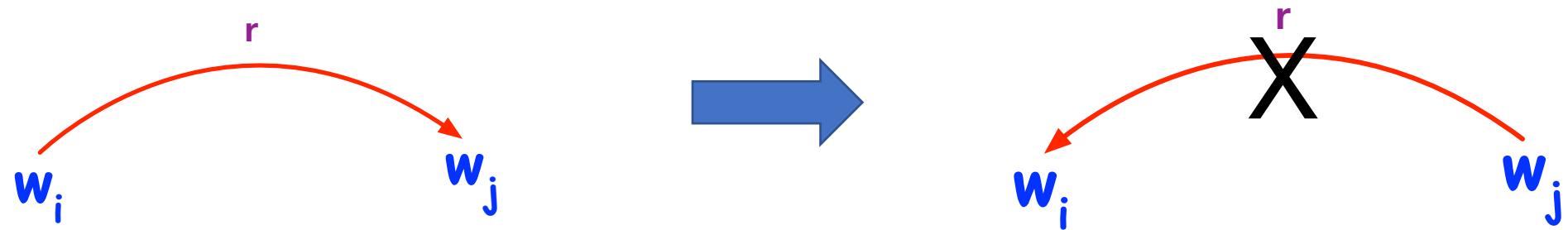
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# Constraints

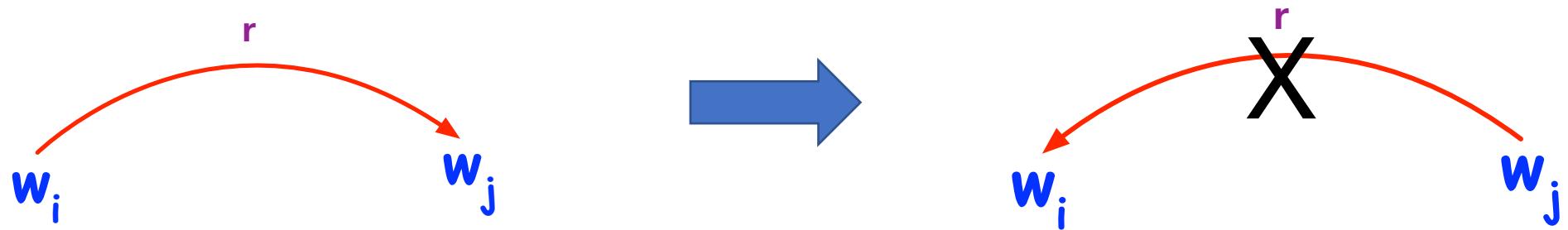
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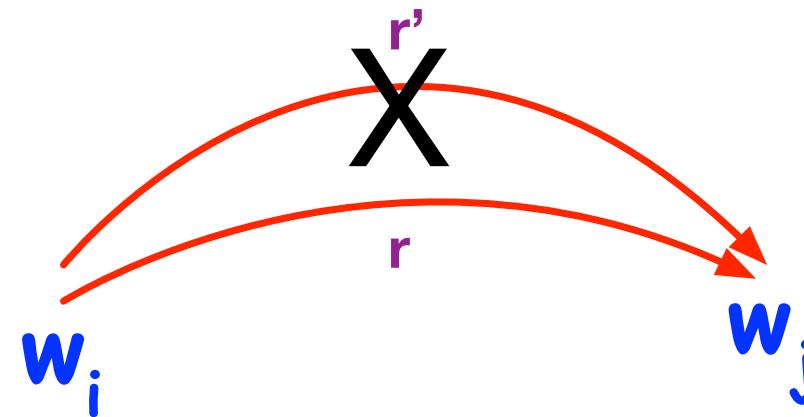
*if  $(w_i, r, w_j) \in A$  then  $(w_i, r', w_j) \notin A \forall r' \neq r$*

# Constraints

*if  $(w_i, r, w_j) \in A$  then  $(w_j, r, w_i) \notin A$*

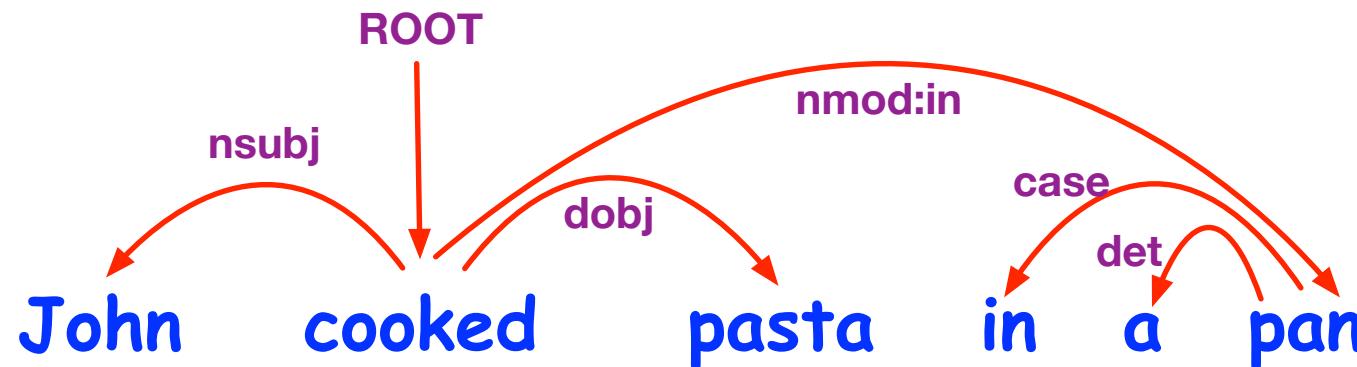


*if  $(w_i, r, w_j) \in A$  then  $(w_i, r', w_j) \notin A \forall r' \neq r$*



# Dependency Tree

**Spanning Node Set:** Node set having all the words of the sentence ( $V_S$ )



$$G = (V, A)$$

$$V = V_S = \{ROOT, John, cooked, pasta, in, a, pan\}$$

$$\begin{aligned} A = \{ & (ROOT, PRED, cooked), (cooked, nsubj, John) \\ & (cooked, dobj, pasta), (cooked, nmod : in, pan) \\ & (pan, det, a), (pan, case, in) \} \end{aligned}$$

# Dependency Tree

Well-Formed Dependency Graph  $G(V,A)$  for an input sentence  $S$  and dependency relation set  $R$  is any dependency graph that is a directed tree originating out of node  $w_0(\text{ROOT})$  and having spanning node set  $V = V_S$

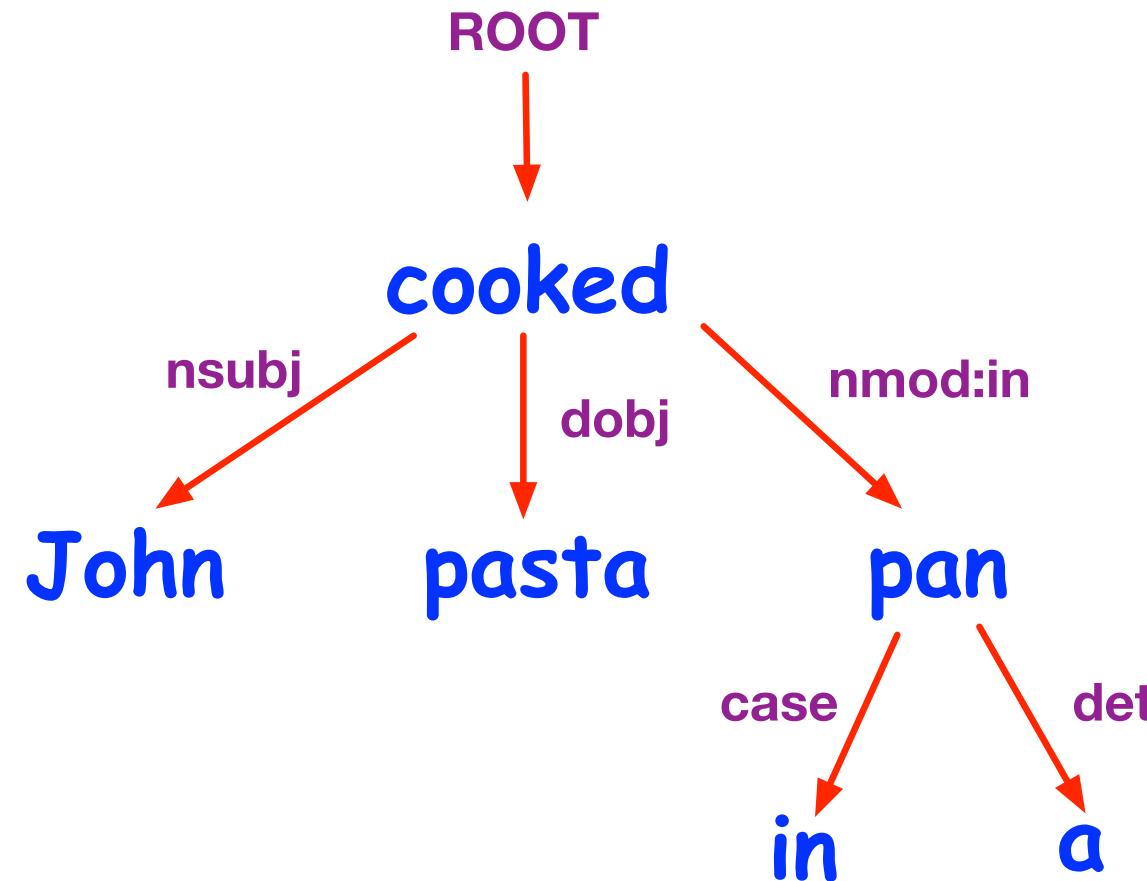
Well formed dependency graphs are called as *Dependency Trees*

In this lecture, we consider only those parsing systems where dependency graphs are dependency trees

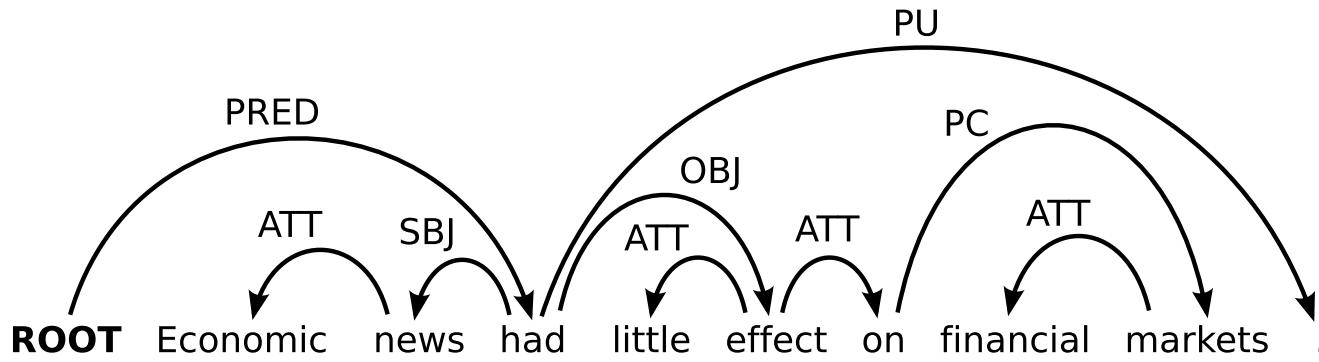


# Dependency Tree

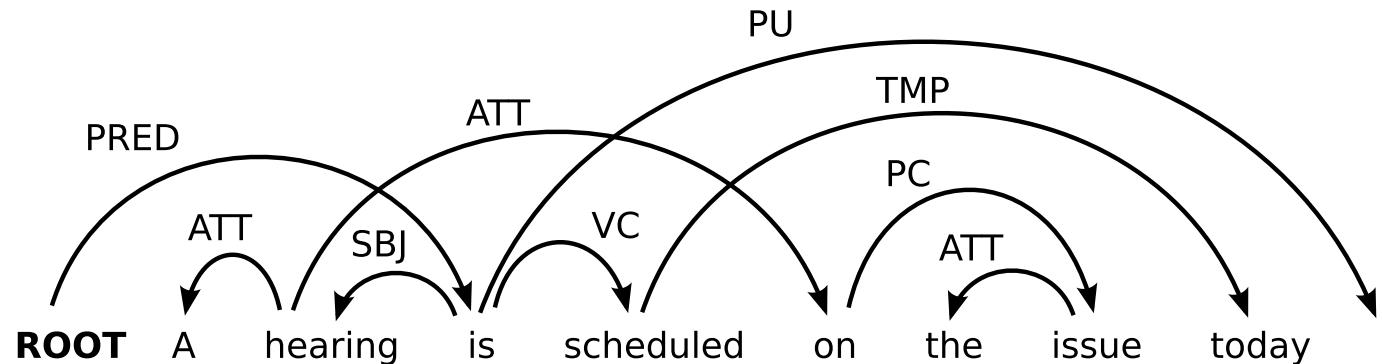
In this lecture, we consider only those parsing systems where dependency graphs are dependency trees



# Projective vs Non-Projective Dependency Parse



Projective Dependency Tree



Non-Projective Dependency Tree



# DP Approaches



# DP Approaches

**Grammar**  
(Formal Grammars)

**Data Driven**  
(Machine Learning)

**Hybrid**  
(ML + Grammars)



# DP Approaches

**Grammar**  
(Formal Grammars)

**Data Driven**  
(Machine Learning)

**Hybrid**  
(ML + Grammars)

Context-Free



# DP Approaches

## Grammar

(Formal Grammars)

### Context-Free

- Exploits a mapping from dependency structures to context-free phrase structure representations.
- Reuses parsing algorithms originally developed for CFG
- E.g. Chart parser, shift reduce parser, etc.

## Data Driven

(Machine Learning)

## Hybrid

(ML + Grammars)



# DP Approaches

## Grammar

(Formal Grammars)

### Context-Free

### Constraints-based

- Exploits a mapping from dependency structures to context-free phrase structure representations.
- Reuses parsing algorithms originally developed for CFG
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## Data Driven

(Machine Learning)

## Hybrid

(ML + Grammars)

- Views parsing as constraint satisfaction problem
- Grammar is set of constraints on well-formed Dep. Graphs
- Some approaches allow soft weighted constraints and score Dep. Graphs by combination of weights of constraints violated by the graph.



# DP Approaches

**Grammar**  
(Formal Grammars)

Context-Free

Constraints-based

**Data Driven**  
(Machine Learning)

Unsupervised

Supervised

**Hybrid**  
(ML + Grammars)



# DP Approaches

## Grammar

(Formal Grammars)

### Context-Free

### Constraints-based

- Defines a transition system (or state machine) for mapping sentence to its Dep. Graph
- Induces model for predicting the next state transition, given the transition history
- The induced model is used to construct optimal transition sequence for the input sentence
- Based on Shift-Reduce parser

## Data Driven

(Machine Learning)

### Unsupervised

### Supervised

## Transition Based

## Hybrid

(ML + Grammars)



# DP Approaches

- Defines a space of candidate Dep. Graphs for a sentence
- Induces a model for scoring candidate Dep. Graphs for a given sentence
- The learned model is used find the highest scoring Dep. Graph
- Maximum Spanning Tree parsing

**Context-Free      Constraints-based**

- Defines a transition system (or state machine) for mapping sentence to its Dep. Graph
- Induces model for predicting the next state transition, given the transition history
- The induced model is used to construct optimal transition sequence for the input sentence
- Based on Shift-Reduce parser for CFG

**Data Driven**  
(Machine Learning)

**Hybrid**  
(ML + Grammars)

**Unsupervised**

**Supervised**

**Transition  
Based**

**Graph  
Based**



# Dependency Parsing Model

**Dependency Parsing  
Model**

$$\mathcal{M} = (\Gamma, \lambda, h)$$



# Dependency Parsing Model

**Dependency Parsing  
Model**

$$\mathcal{M} = (\Gamma, \lambda, h)$$



**Set of constraints**  
**Defines set of**  
**permissible Dep. Trees**



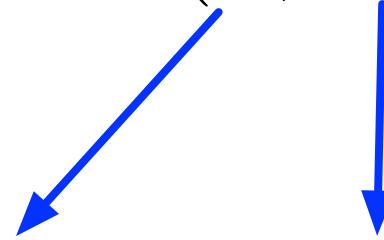
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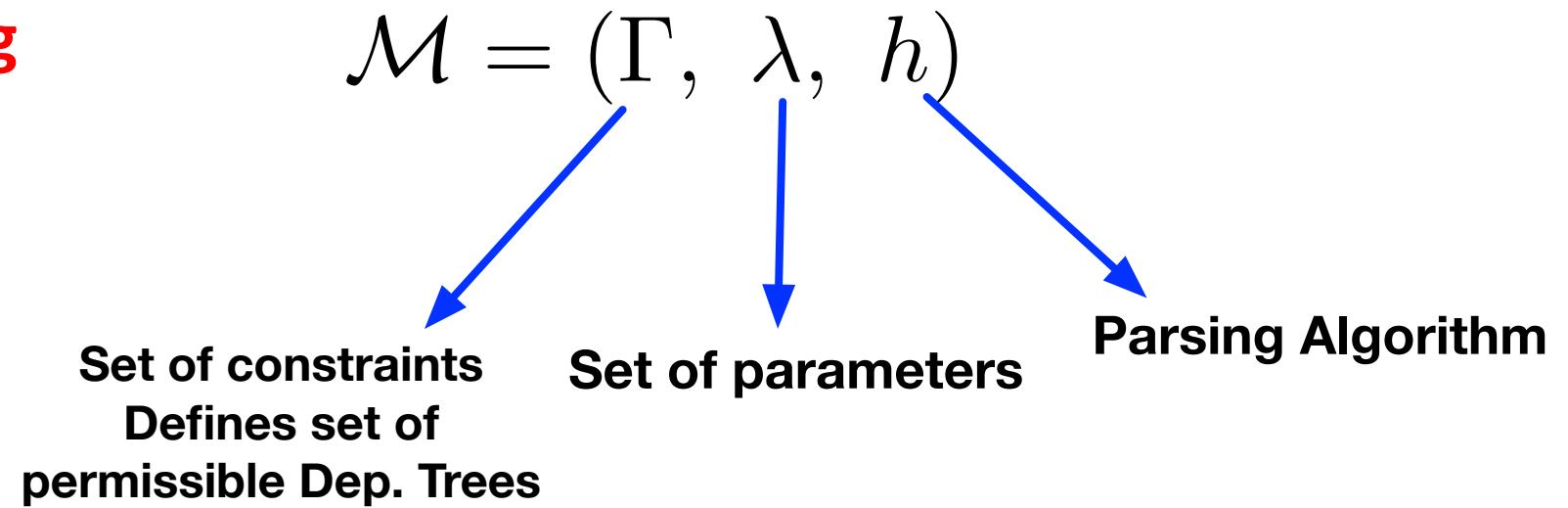
**Set of constraints**  
Defines set of  
permissible Dep. Trees

**Set of parameters**



# Dependency Parsing Model

## Dependency Parsing Model



# Supervised Dependency Parsing

- Given a corpus of annotated sentences:  $\mathcal{D} = \{(S_d, G_d)\}_{d=0}^{|\mathcal{D}|}$
- Learn a model to automatically generate the dependency tree for a given sentence



# Supervised Dependency Parsing

- Two main computational problems:
  - **Learning:** Given a training set of annotated sentences, induce a parsing model that can be used to parse a new sentence



# Supervised Dependency Parsing

- Two main computational problems:
  - **Learning:** Given a training set of annotated sentences, induce a parsing model that can be used to parse a new sentence
  - **Parsing:** Given a parsing model and a sentence derive the optimal dependency tree

