



Amirkabir University of Technology  
(Tehran Polytechnic)

# Natural Language Processing

## Lecture 15: Dependency Parsing

Amirkabir University of Technology

Dr Momtazi

# Parsing

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- Finding structural relationship between words in a sentence

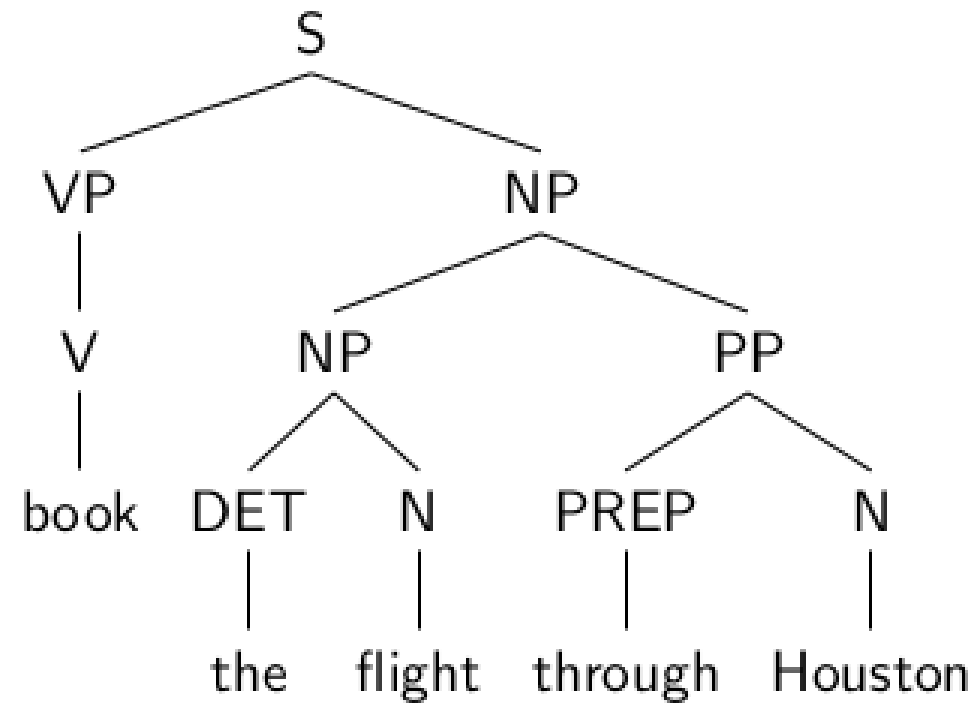
# Outline

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- **Constituency vs. Dependency**
- Dependency Relations and Formalisms
- Dependency Parsing
- Evaluation

# Constituency

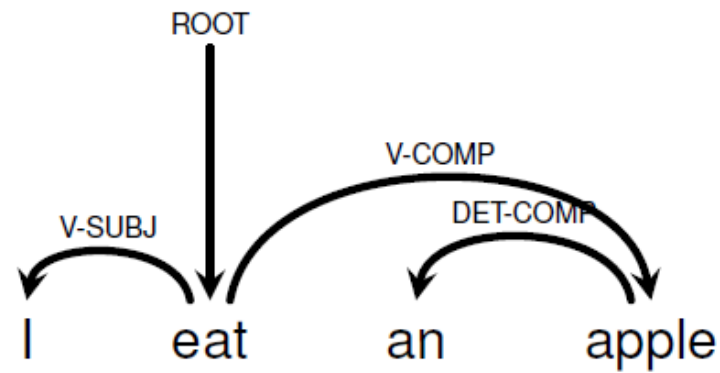
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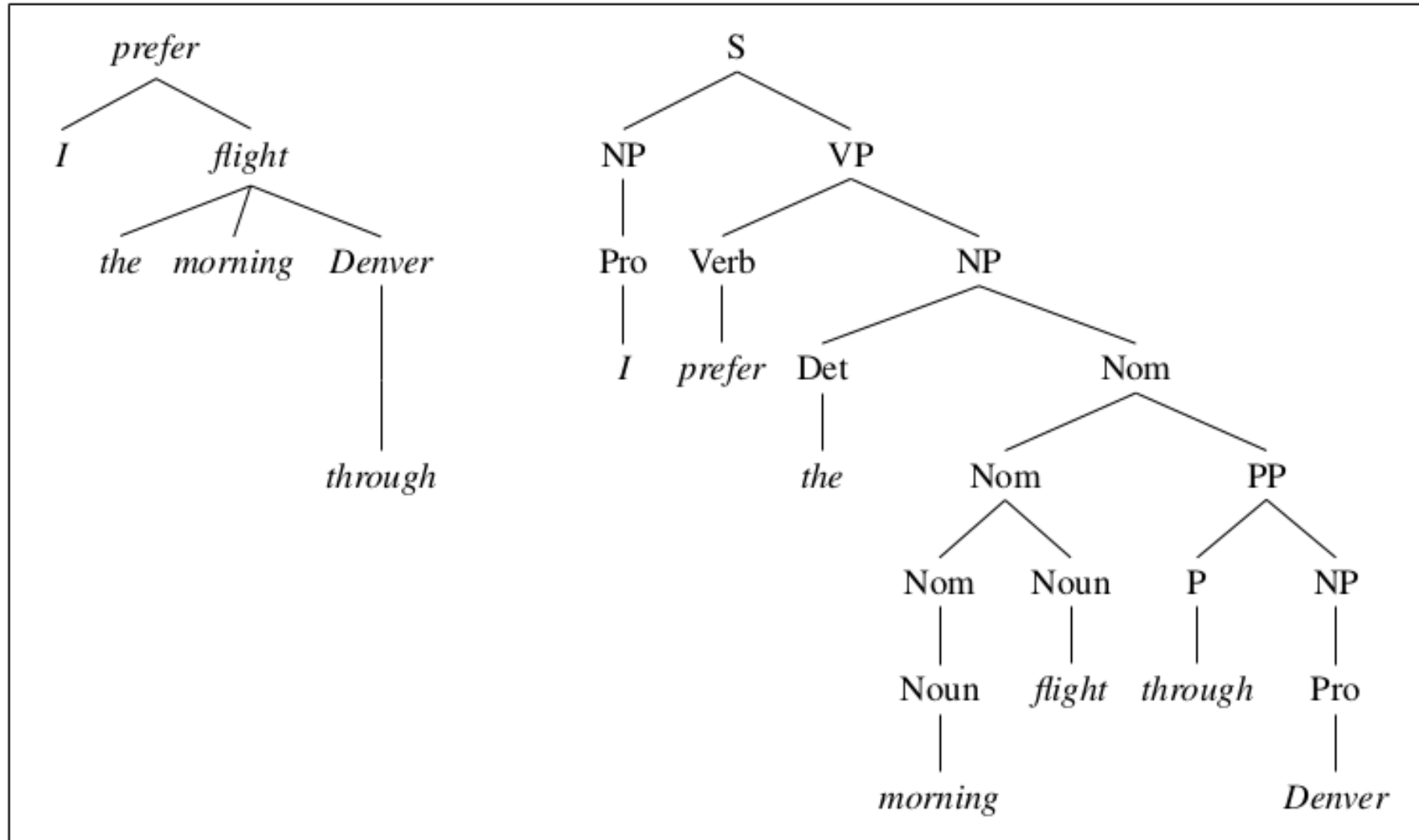
# Dependency

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- Identifying which words depend on (modify or arguments of) which other words



# Constituency vs. Dependency



# Outline

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- Constituency vs. Dependency
- **Dependency Relations and Formalisms**
- Dependency Parsing
- Evaluation

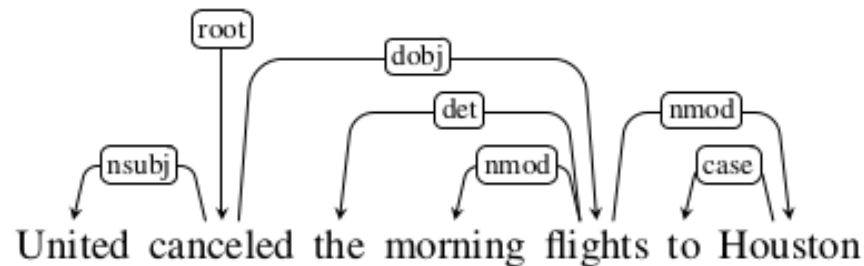
# Dependency Relations

- The traditional linguistic notion of grammatical relation provides the basis for the binary relations that comprise these dependency structures.
- The arguments to these relations consist of a head and a dependent.

Clausal Argument Relations	Description
NSUBJ	Nominal subject
DOBJ	Direct object
IOBJ	Indirect object
CCOMP	Clausal complement
XCOMP	Open clausal complement
Nominal Modifier Relations	Description
NMOD	Nominal modifier
AMOD	Adjectival modifier
NUMMOD	Numeric modifier
APPOS	Appositional modifier
DET	Determiner
CASE	Prepositions, postpositions and other case markers
Other Notable Relations	Description
CONJ	Conjunct
CC	Coordinating conjunction



# Dependency Relations



Relation	Examples with <i>head</i> and <b>dependent</b>
NSUBJ	<b>United</b> <i>canceled</i> the flight.
DOBJ	United <i>diverted</i> the <b>flight</b> to Reno. We <i>booked</i> her the first <b>flight</b> to Miami.
IOBJ	We <i>booked</i> <b>her</b> the flight to Miami.
NMOD	We took the <b>morning</b> <i>flight</i> .
AMOD	Book the <b>cheapest</b> <i>flight</i> .
NUMMOD	Before the storm JetBlue canceled <b>1000</b> <i>flights</i> .
APPOS	<i>United</i> , a <b>unit</b> of UAL, matched the fares.
DET	<b>The</b> <i>flight</i> was canceled. <b>Which</b> <i>flight</i> was delayed?
CONJ	We <i>flew</i> to Denver and <b>drove</b> to Steamboat.
CC	We flew to Denver <b>and</b> <i>drove</i> to Steamboat.
CASE	Book the flight <b>through</b> <i>Houston</i> .

# Dependency Formalisms

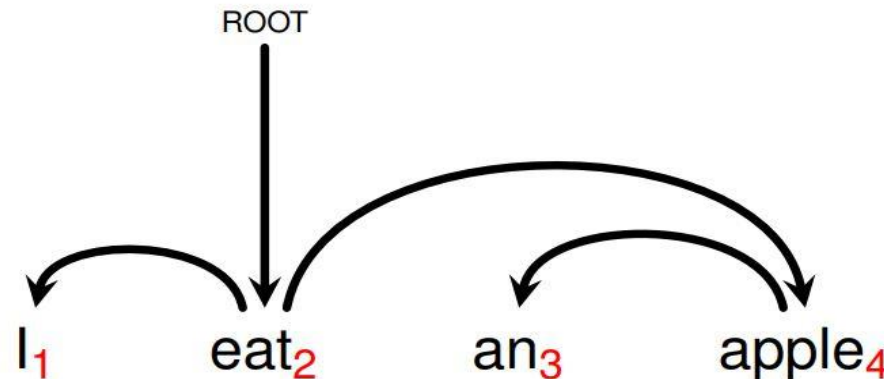
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- A dependency structure can be represented as a directed graph  $G=(V;A)$ , consisting of a set of vertices  $V$ , and a set of ordered pairs of vertices  $A$  (arcs).
  - The set of vertices,  $V$ , corresponds exactly to the set of words in a given sentence.
  - The set of arcs,  $A$ , captures the head dependent and grammatical function relationships between the elements in  $V$ .

# Unlabeled Dependency Parses

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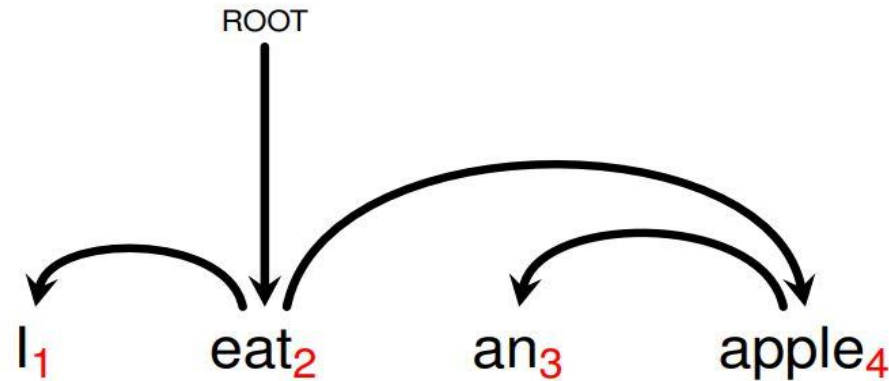
- A dependency tree is a directed graph that satisfies the following constraints:
  - There is a single designated root node that has no incoming arcs. It is used to point the head of the sentence
  - With the exception of the root node, each vertex has exactly one incoming arc. (Each dependency connect a head word  $h$  to a modifier  $m$ )
  - There is a unique path from the root node to each vertex in  $V$ .



# Unlabeled Dependency Parses

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- Dependencies in the following example are as follows:
  - (0,2)
  - (2,1)
  - (2,4)
  - (4,3)



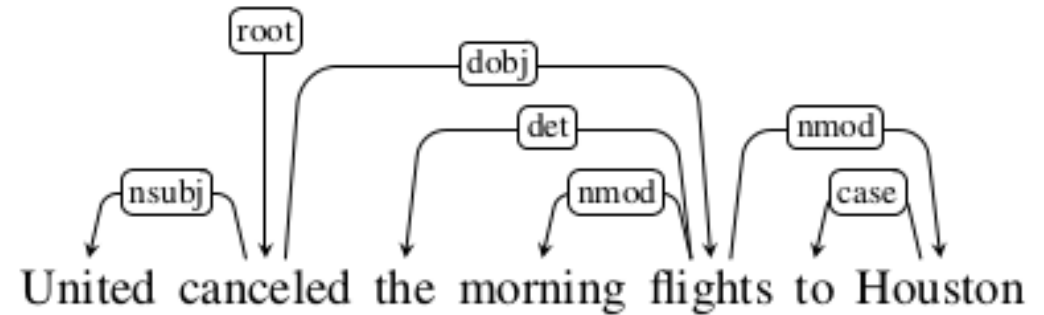
# Projectivity

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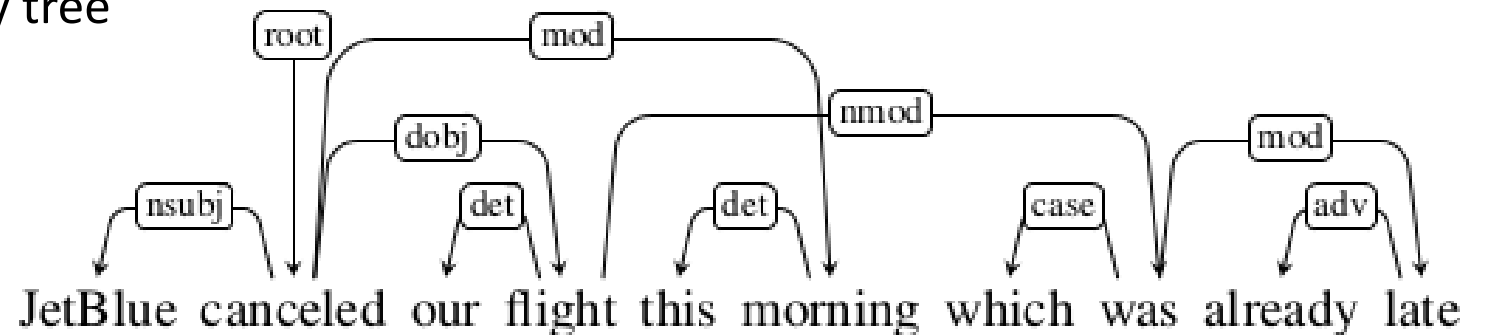
- The notion of projectivity imposes an additional constraint that is derived from the order of the words in the input.
- An arc from a head to a dependent is said to be projective if there is a path from the head to every word that lies between the head and the dependent in the sentence.
- A dependency tree is then said to be projective if all the arcs that make it up are projective (There is no crossing dependencies).

# Projectivity

- Example
  - Projective dependency tree



- Non-projective dependency tree



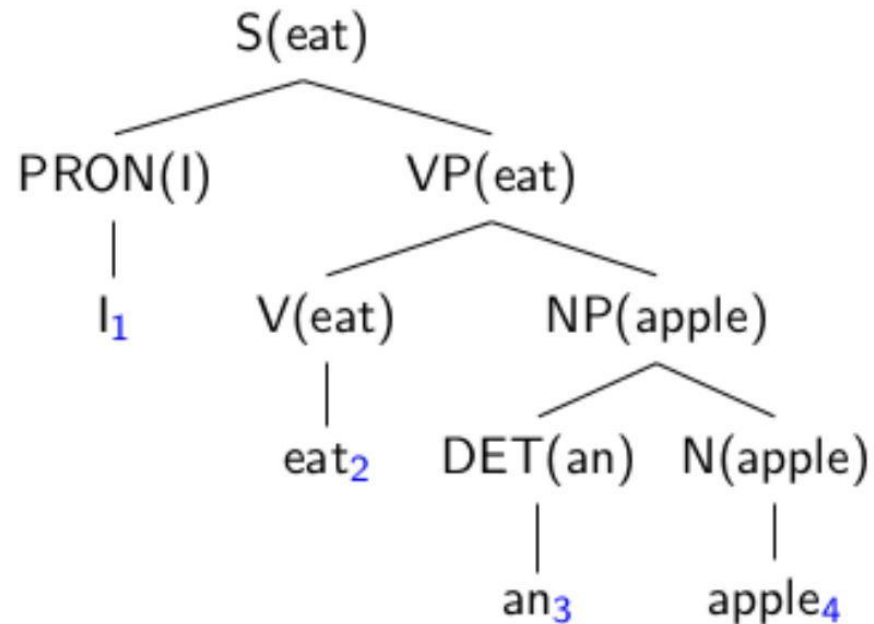
# Dependency Treebanks

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- For some languages, there exist “Dependency banks”; e.g., Czech
- If no dependency bank is available for a language, it is possible to extract it from a constituency-based treebank:
  - Mark the head child of each node in a phrase structure, using the appropriate head rules.
  - In the dependency structure, make the head of each non-head child depend on the head of the head-child.

# Dependency Treebanks

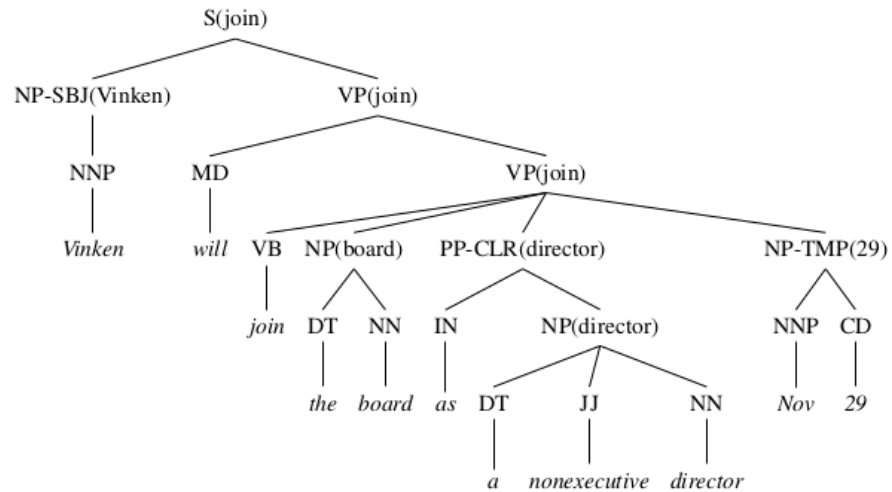
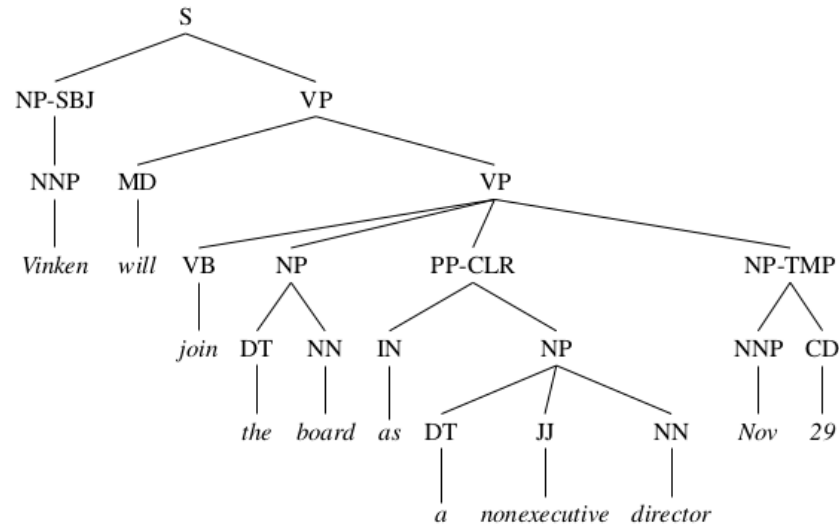
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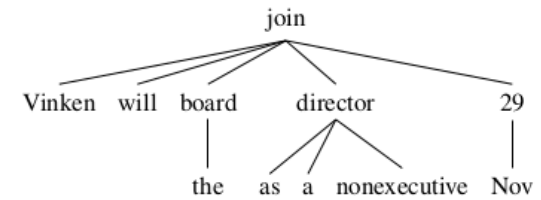
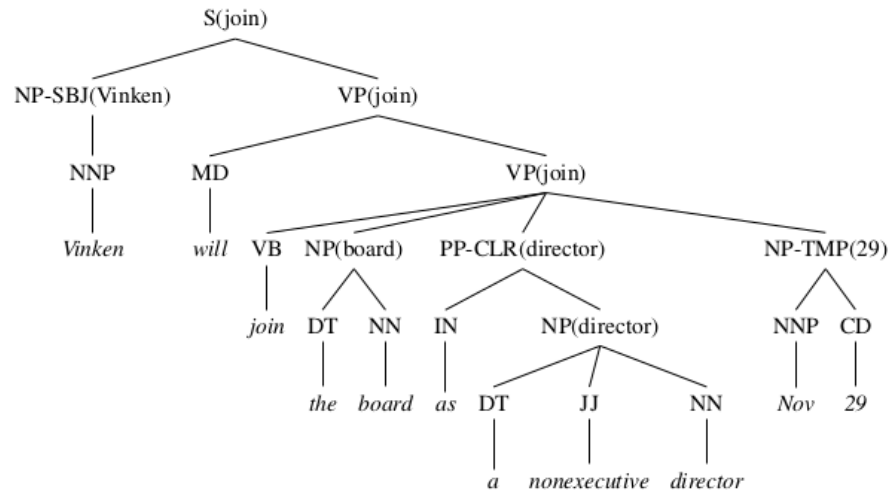
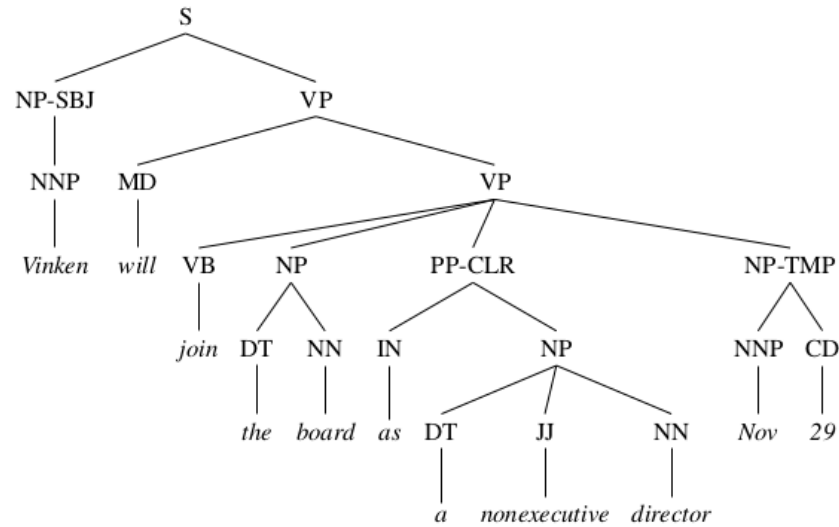
- Unlabeled Dependencies:
  - (0,2) root → eat
  - (2,1) eat → I
  - (2,4) eat → apple
  - (4,3) apple → an



# Dependency Treebanks

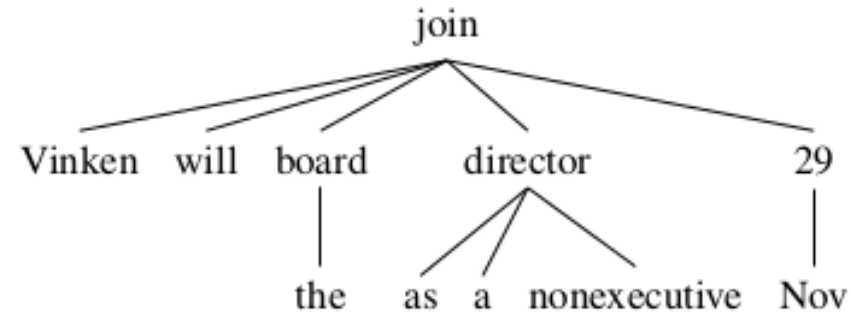


# Dependency Treebanks



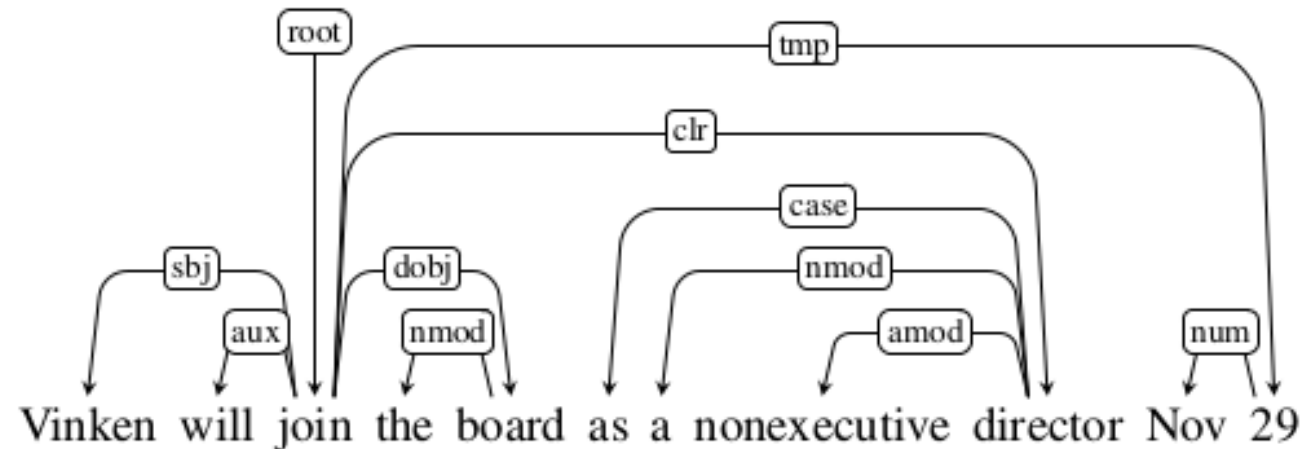
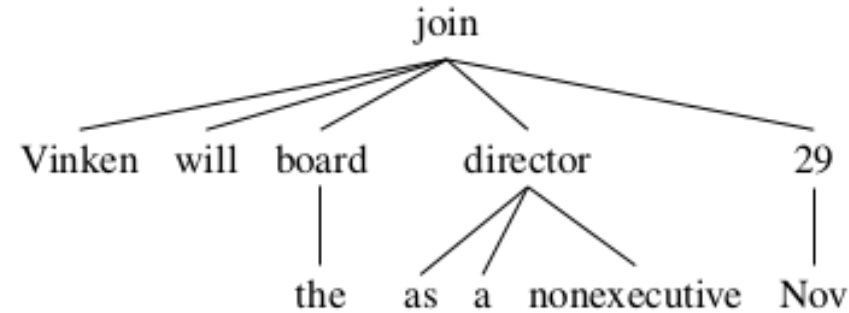
# Dependency Treebanks

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# Dependency Treebanks

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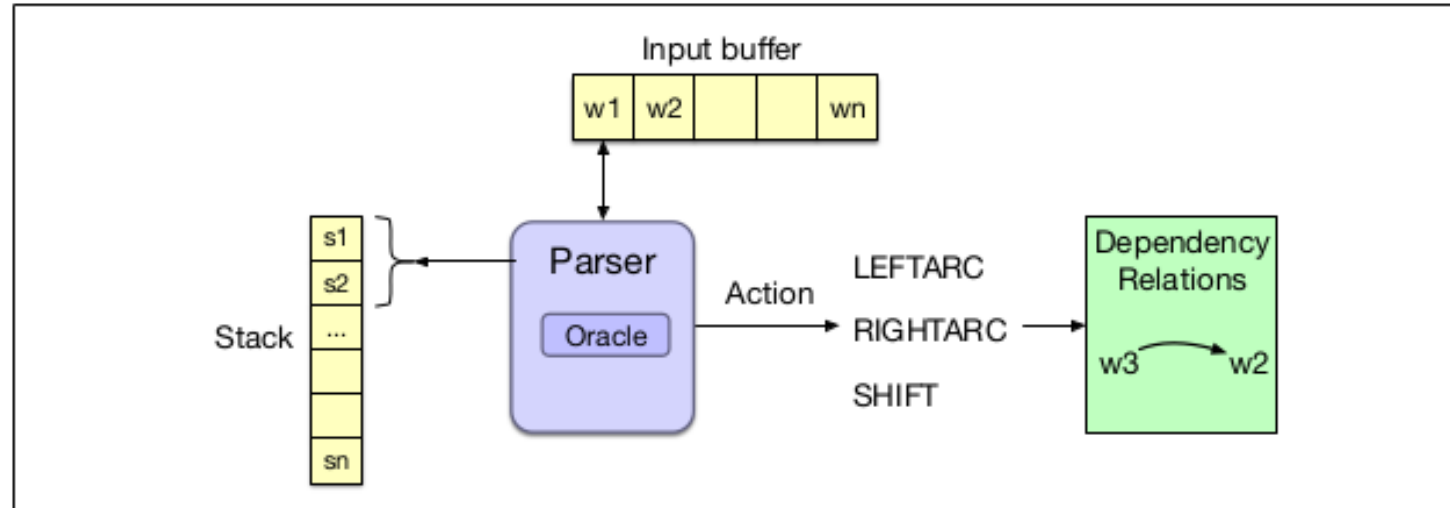
# Outline

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- Constituency vs. Dependency
- Dependency Relations and Formalisms
- **Dependency Parsing**
  - Transition-based Dependency Parsing
  - Sequence Modeling
- Evaluation

# Transition-based Dependency Parsing

- Using on shift-reduce parsing, a paradigm originally developed for compilers
- Main elements:
  - A stack on which we build the parse
  - A buffer of tokens to be parsed
  - A parser which takes actions on the parse via a predictor called an oracle



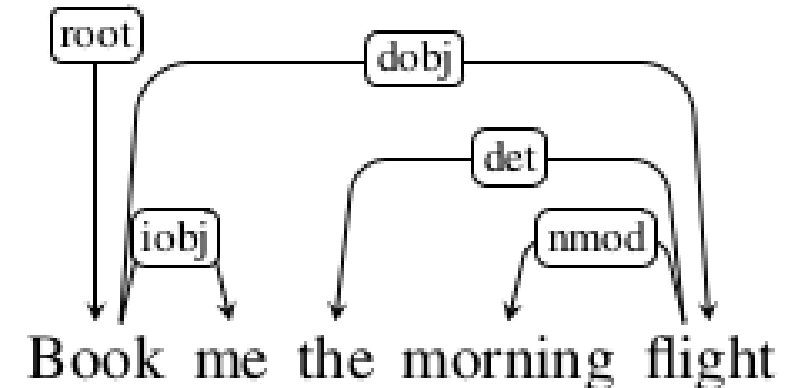
# Transition-based Dependency Parsing

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- The parser walks through the sentence left-to-right, successively shifting items from the buffer onto the stack.
- At each point we examine the top two elements on the stack, and the oracle makes a decision about what transition to apply to build the parse.
- The possible transitions:
  - LEFTARC: Assigning the current word as the head of some previously seen word
  - RIGHTARC: Assigning some previously seen word as the head of the current word
  - SHIFT: Postponing to deal with the current word, storing it for later processing
- LEFTARC and RIGHTARC are also known as reduce operations, based on a metaphor from shift-reduce parsing

# Transition-based Dependency Parsing

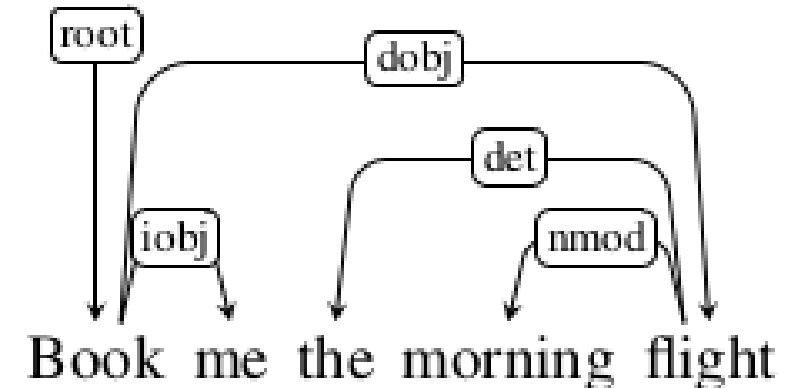
Step	Stack	Word List	Action	Relation Added
0	[root]	[book, me, the, morning, flight]	SHIFT	





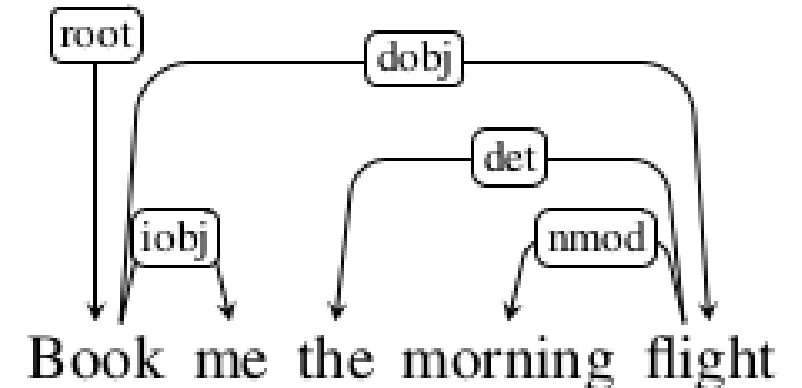
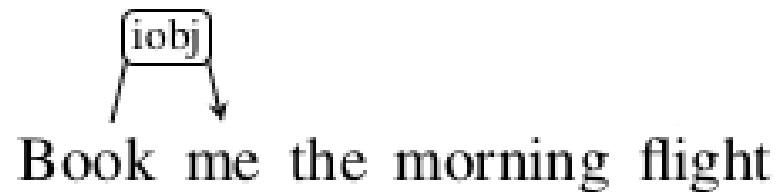
# Transition-based Dependency Parsing

Step	Stack	Word List	Action	Relation Added
0	[root]	[book, me, the, morning, flight]	SHIFT	
1	[root, book]	[me, the, morning, flight]	SHIFT	



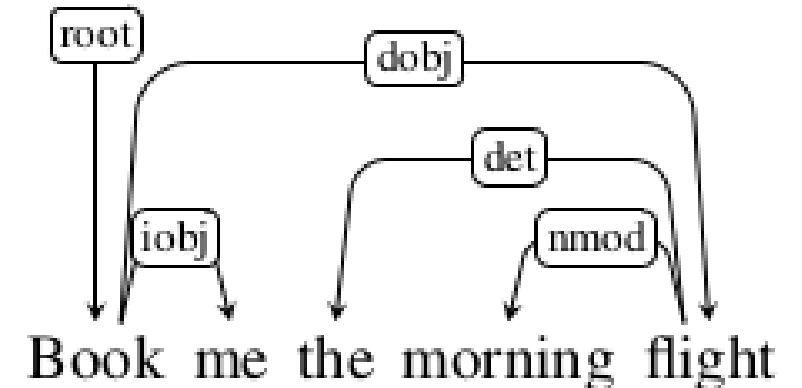
# Transition-based Dependency Parsing

Step	Stack	Word List	Action	Relation Added
0	[root]	[book, me, the, morning, flight]	SHIFT	
1	[root, book]	[me, the, morning, flight]	SHIFT	
2	[root, book, me]	[the, morning, flight]	RIGHTARC	(book → me)



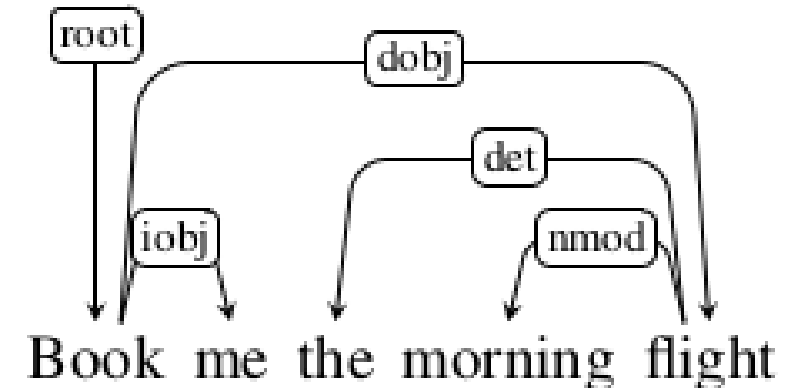
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Step	Stack	Word List	Action	Relation Added
0	[root]	[book, me, the, morning, flight]	SHIFT	(book → me)
1	[root, book]	[me, the, morning, flight]	SHIFT	
2	[root, book, me]	[the, morning, flight]	RIGHTARC	
3	[root, book]	[the, morning, flight]	SHIFT	



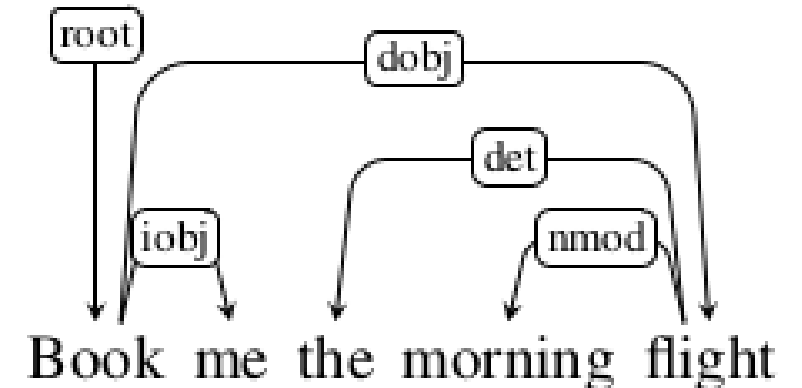
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1	[root, book]	[me, the, morning, flight]	SHIFT	
2	[root, book, me]	[the, morning, flight]	RIGHTARC	
3	[root, book]	[the, morning, flight]	SHIFT	
4	[root, book, the]	[morning, flight]	SHIFT	



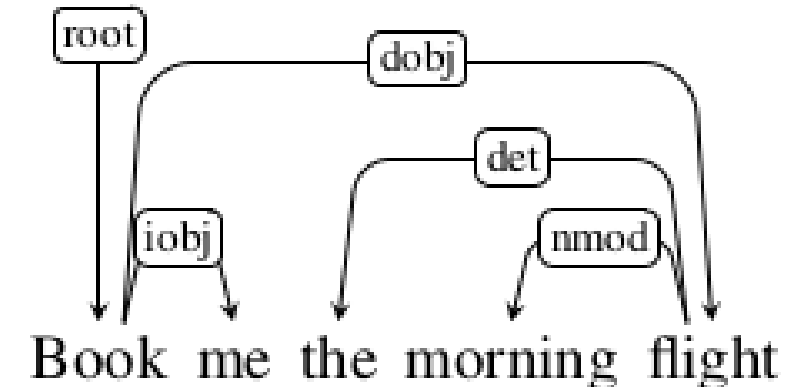
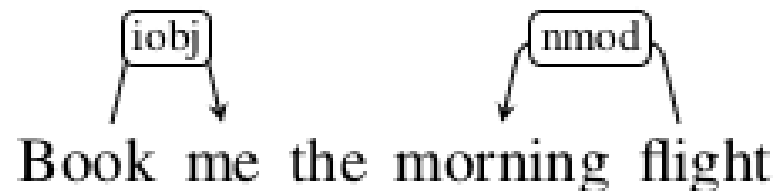
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Step	Stack	Word List	Action	Relation Added
0	[root]	[book, me, the, morning, flight]	SHIFT	(book → me)
1	[root, book]	[me, the, morning, flight]	SHIFT	
2	[root, book, me]	[the, morning, flight]	RIGHTARC	
3	[root, book]	[the, morning, flight]	SHIFT	
4	[root, book, the]	[morning, flight]	SHIFT	
5	[root, book, the, morning]	[flight]	SHIFT	



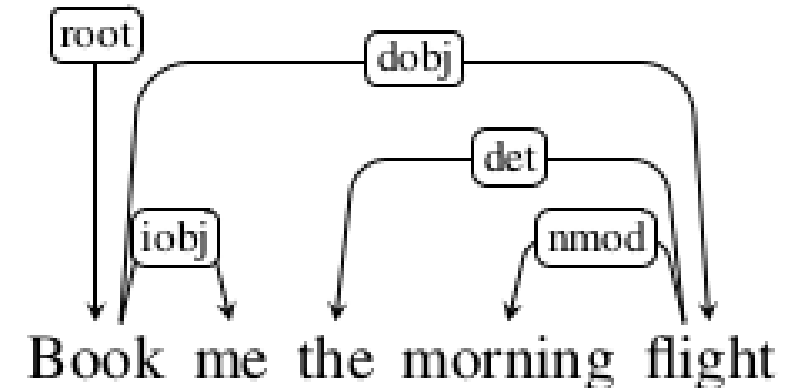
# Transition-based Dependency Parsing

Step	Stack	Word List	Action	Relation Added
0	[root]	[book, me, the, morning, flight]	SHIFT	(book → me)
1	[root, book]	[me, the, morning, flight]	SHIFT	
2	[root, book, me]	[the, morning, flight]	RIGHTARC	
3	[root, book]	[the, morning, flight]	SHIFT	
4	[root, book, the]	[morning, flight]	SHIFT	
5	[root, book, the, morning]	[flight]	SHIFT	(morning ← flight)
6	[root, book, the, morning, flight]	[]	LEFTARC	



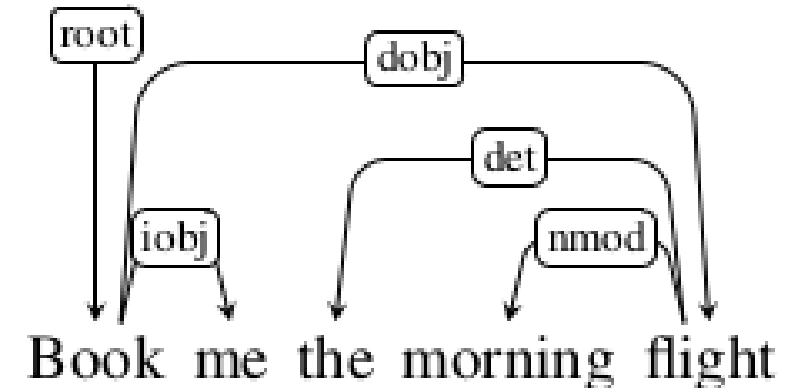
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Step	Stack	Word List	Action	Relation Added
0	[root]	[book, me, the, morning, flight]	SHIFT	(book → me)
1	[root, book]	[me, the, morning, flight]	SHIFT	
2	[root, book, me]	[the, morning, flight]	RIGHTARC	
3	[root, book]	[the, morning, flight]	SHIFT	
4	[root, book, the]	[morning, flight]	SHIFT	
5	[root, book, the, morning]	[flight]	SHIFT	(morning ← flight)
6	[root, book, the, morning, flight]	[]	LEFTARC	
7	[root, book, the, flight]	[]	LEFTARC	



# Transition-based Dependency Parsing

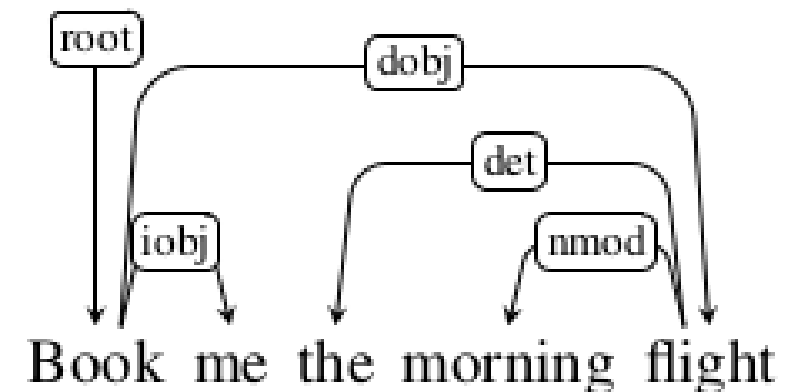
Step	Stack	Word List	Action	Relation Added
0	[root]	[book, me, the, morning, flight]	SHIFT	(book → me)
1	[root, book]	[me, the, morning, flight]	SHIFT	
2	[root, book, me]	[the, morning, flight]	RIGHTARC	
3	[root, book]	[the, morning, flight]	SHIFT	
4	[root, book, the]	[morning, flight]	SHIFT	
5	[root, book, the, morning]	[flight]	SHIFT	(morning ← flight)
6	[root, book, the, morning, flight]	[]	LEFTARC	
7	[root, book, the, flight]	[]	LEFTARC	
8	[root, book, flight]	[]	RIGHTARC	(book → flight)





# Transition-based Dependency Parsing

Step	Stack	Word List	Action	Relation Added
0	[root]	[book, me, the, morning, flight]	SHIFT	(book → me)
1	[root, book]	[me, the, morning, flight]	SHIFT	
2	[root, book, me]	[the, morning, flight]	RIGHTARC	
3	[root, book]	[the, morning, flight]	SHIFT	
4	[root, book, the]	[morning, flight]	SHIFT	
5	[root, book, the, morning]	[flight]	SHIFT	(morning ← flight)
6	[root, book, the, morning, flight]	[]	LEFTARC	
7	[root, book, the, flight]	[]	LEFTARC	
8	[root, book, flight]	[]	RIGHTARC	
9	[root, book]	[]	RIGHTARC	
10	[root]	[]	Done	(root → book)



# Parsing Process

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- Generating training data
  - Simulating the operation of the parser by running the algorithm and relying on a new training oracle to give us correct transition operators for each successive configuration
- Training a classifier on the generated training data
  - Classic feature-based algorithm
  - Neural classifier using embedding features

# Feature-based Classifier

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- Using classical ML algorithms; e.g., logistic regression, SVM
- Feature-based classifiers generally use different features, such as
  - Word forms
  - Lemmas
  - Parts of speech
  - The head
  - The dependency relation to the head
- Other features may be relevant for some languages, such as
  - Morphosyntactic features like case marking on subjects or objects

# Feature-based Classifier

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- The features are extracted from the training configurations, which consist of the stack, the buffer and the current set of relations.
- Most useful are features are extracted from:
  - The top levels of the stack
  - The words near the front of the buffer
  - The dependency relations already associated with any of those elements

- Feature template

$\langle s_1.w, op \rangle$        $\langle s_2.w, op \rangle$

$\langle s_1.t, op \rangle$        $\langle s_2.t, op \rangle$

$\langle b_1.w, op \rangle$        $\langle b_1.t, op \rangle$

$\langle s_1.wt, op \rangle$

# Feature-based Classifier

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- Example

Stack	Word buffer	Relations
[root, book, flight]	[through, Houston]	(the $\leftarrow$ flight)

$\langle s_1.w = flights, op = shift \rangle$

$\langle s_2.w = canceled, op = shift \rangle$

$\langle s_1.t = NNS, op = shift \rangle$

$\langle s_2.t = VBD, op = shift \rangle$

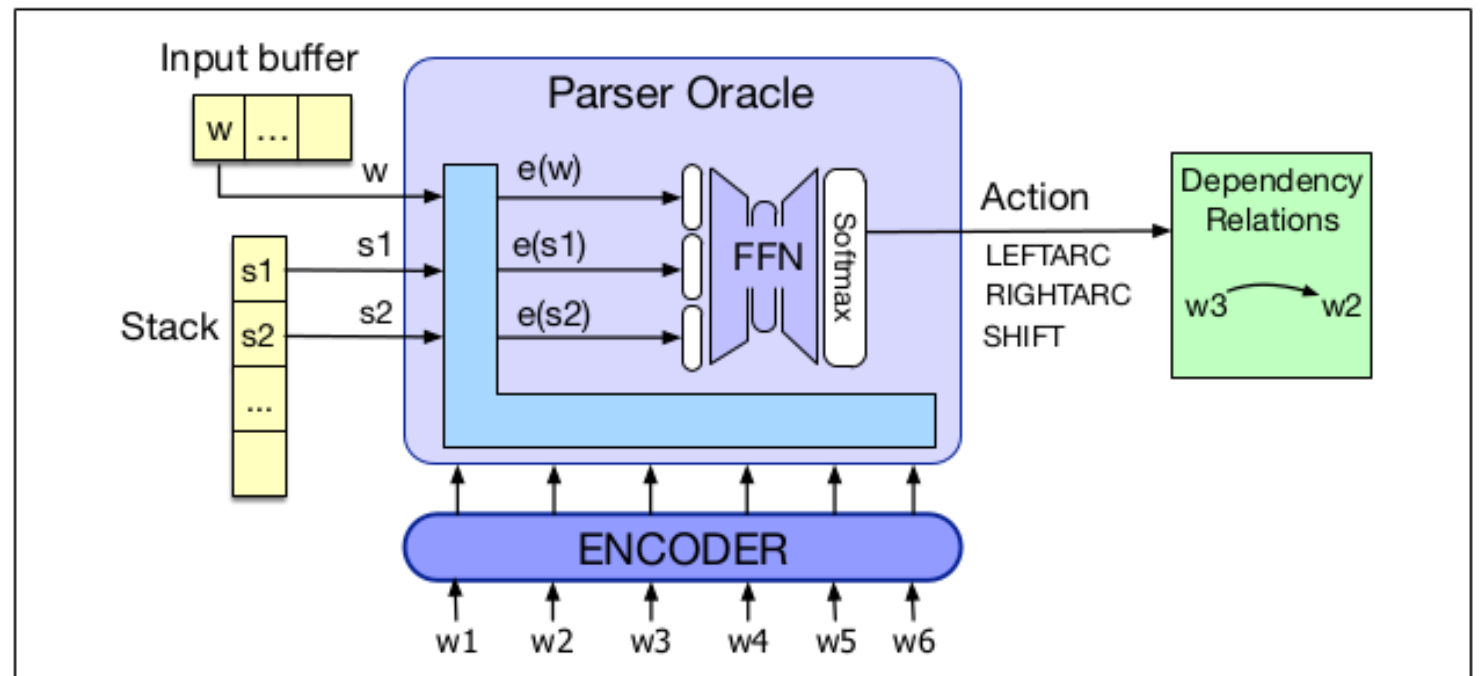
$\langle b_1.w = to, op = shift \rangle$

$\langle b_1.t = TO, op = shift \rangle$

$\langle s_1.wt = flightsNNS, op = shift \rangle$

# Neural Classifier

- Classification scenario:
  - Passing the sentence through an encoder
  - Taking the presentation of the top 2 words on the stack and the first word of the buffer
  - Concatenating the representations and presenting to a feedforward network that predicts the transition to take



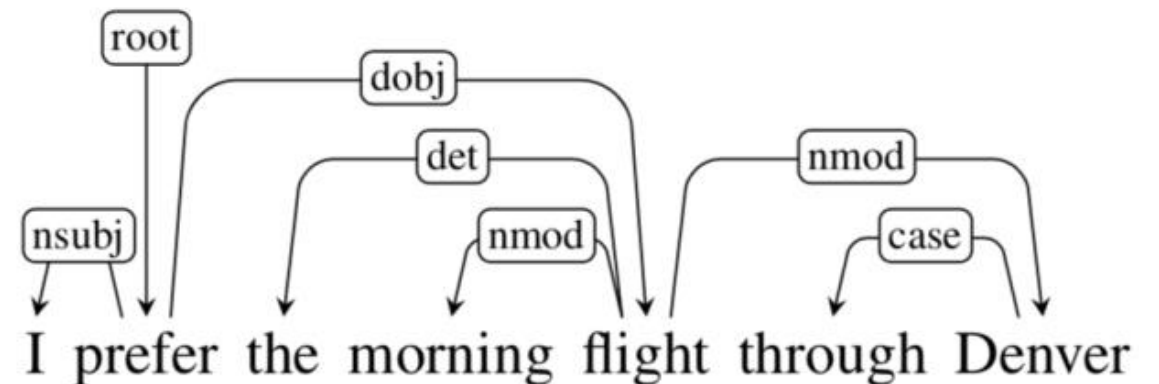
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# Sequence Modeling for Dependency Parsing

- Bi-directional LSTM and Label learning
- Predicting a label for each word
  - The head word of the target word
    - the: flight
    - I: prefer
    - flight: prefer
  - The path to the head word
    - the → 2 R
    - I → 1 R
    - flight → 3 L



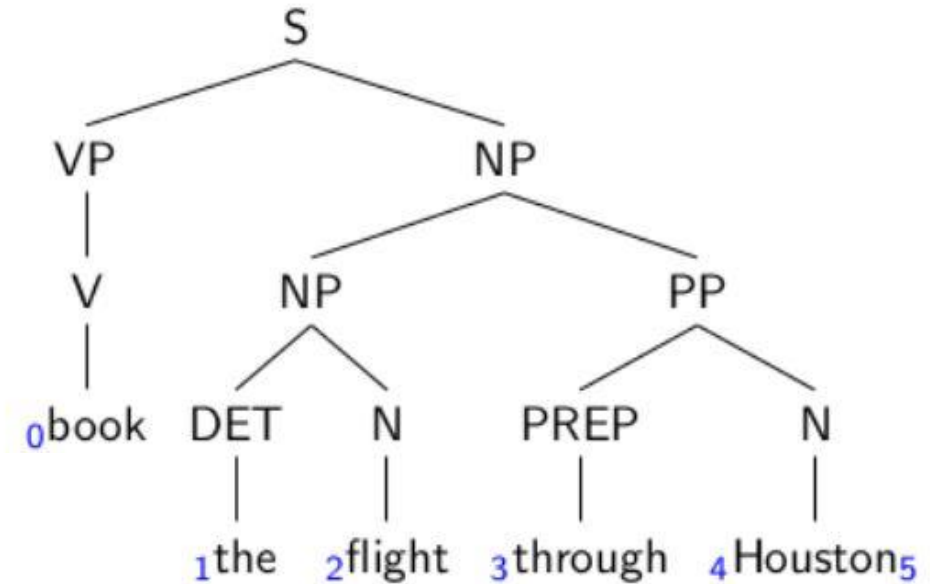


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# Dependencies in Parse Tree



Head	Word	Rule
Root	book <sub>1</sub>	ROOT
book <sub>1</sub>	flight <sub>3</sub>	$S \rightarrow_1 VP \ NP$
flight <sub>3</sub>	through <sub>4</sub>	$NP \rightarrow_1 NP \ PP$
flight <sub>3</sub>	the <sub>2</sub>	$NP \rightarrow_2 DET \ N$
through <sub>4</sub>	Houston <sub>5</sub>	$PP \rightarrow_1 PREP \ N$

# Dependency Accuracy

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- The number of dependencies in both gold and guess parse trees are equal to the number of words
- Dependency accuracy:  
The number of dependencies matches in both trees

# Dependency Accuracy

Head	Word	Rule
Root	book <sub>1</sub>	ROOT
book <sub>1</sub>	flight <sub>3</sub>	$S \rightarrow_1 VP\ NP$
flight <sub>3</sub>	through <sub>4</sub>	$NP \rightarrow_1 NP\ PP$
flight <sub>3</sub>	the <sub>2</sub>	$NP \rightarrow_2 DET\ N$
through <sub>4</sub>	Houston <sub>5</sub>	$PP \rightarrow_1 PREP\ N$

Head	Word	Rule
Root	book <sub>1</sub>	ROOT
book <sub>1</sub>	through <sub>4</sub>	$S \rightarrow_1 VP\ PP$
book <sub>1</sub>	flight <sub>3</sub>	$VP \rightarrow_1 V\ NP$
flight <sub>3</sub>	the <sub>2</sub>	$NP \rightarrow_2 DET\ N$
through <sub>4</sub>	Houston <sub>5</sub>	$PP \rightarrow_1 PREP\ N$

$$\text{Dependency Accuracy} = \frac{3}{5} = 0.6$$

# Results

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- Comparing LPCFG with Dependency model
- Available results on Penn treebank (Wall Street Journal)
  - LPCFG (Collins 1997): 91.4% Dependency Accuracy
  - Dependency parsing (McDonald 2005): 90.7%
    - using dynamic programming
- Efficiency
  - Dependency parsing is much more faster than constituency parsing

# Behind Dependency Accuracy

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- Precision and recall can still be used, but for a particular dependency type (e.g.,  $\text{NP} \rightarrow_1 \text{NP PP}$ )

# Behind Dependency Accuracy

---

- Precision and recall can still be used, but for a particular dependency type (e.g.,  $\text{NP} \rightarrow_1 \text{NP PP}$ )

⇒ Can be used for error analysis

- subject-verb: above 95% recall and precision
- object-verb: above 92% recall and precision
- PP attachments  $\approx$  82% recall and precision
- Coordination  $\approx$  61% recall and precision

# Further Reading

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- Speech and Language Processing (3<sup>rd</sup> ed. draft)
  - Chapters 18