



**THE OHIO STATE  
UNIVERSITY**

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# CSE 5525: Foundations of Speech and Language Processing

## Syntax I

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Many thanks to Prof. Greg Durrett @ UT Austin for sharing his slides.

Some slides adapted from Dan Klein, UC Berkeley

# This Lecture

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- ▶ Constituency formalism
- ▶ Context-free grammars and the CKY algorithm
- ▶ Refining grammars
- ▶ Discriminative parsers

# Syntax & Constituency

# Syntax

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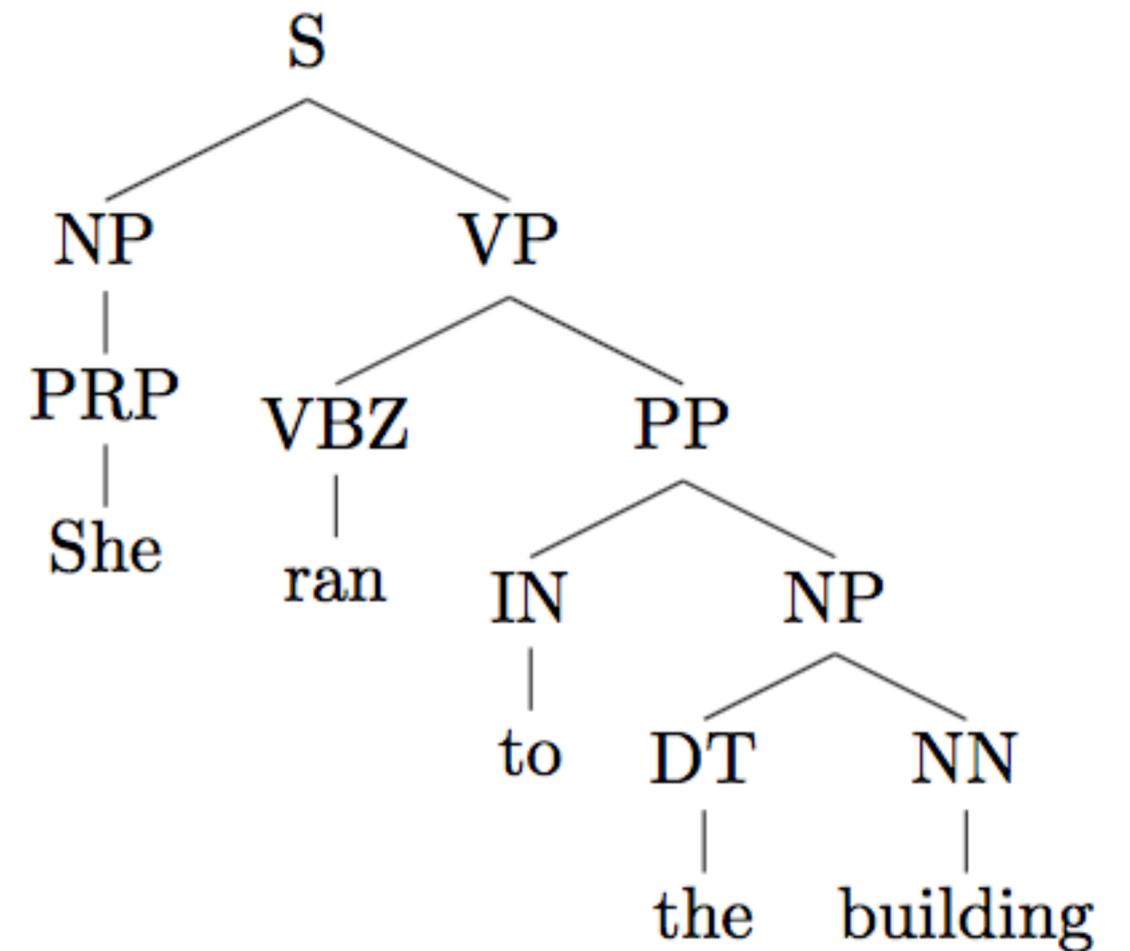
- ▶ Study of word order and how words form sentences
- ▶ Why do we care about syntax? (useful for many tasks like phrase extraction)
  - ▶ Multiple interpretations of words (noun or verb?)
  - ▶ Recognize verb-argument structures (who is doing what to whom?)
  - ▶ Higher level of abstraction beyond words: some languages are SVO, some are VSO, some are SOV, parsing can canonicalize

# Constituency Parsing

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- ▶ Tree-structured syntactic analyses of sentences

Constituency parsing is the task of breaking a text into sub-phrases, or constituents. Non-terminals in the parse tree are types of phrases, the terminals are the words in the sentence. —[AllenNLP demo](#)

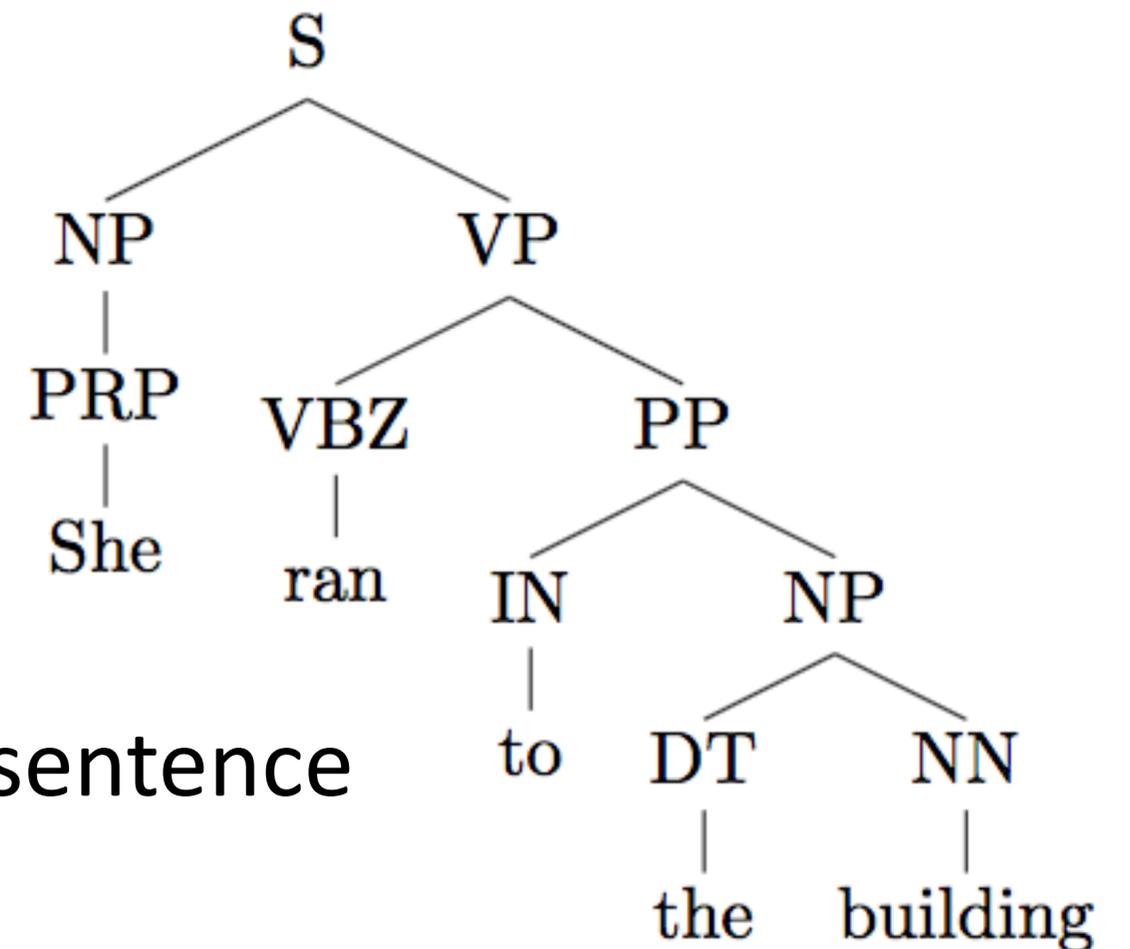


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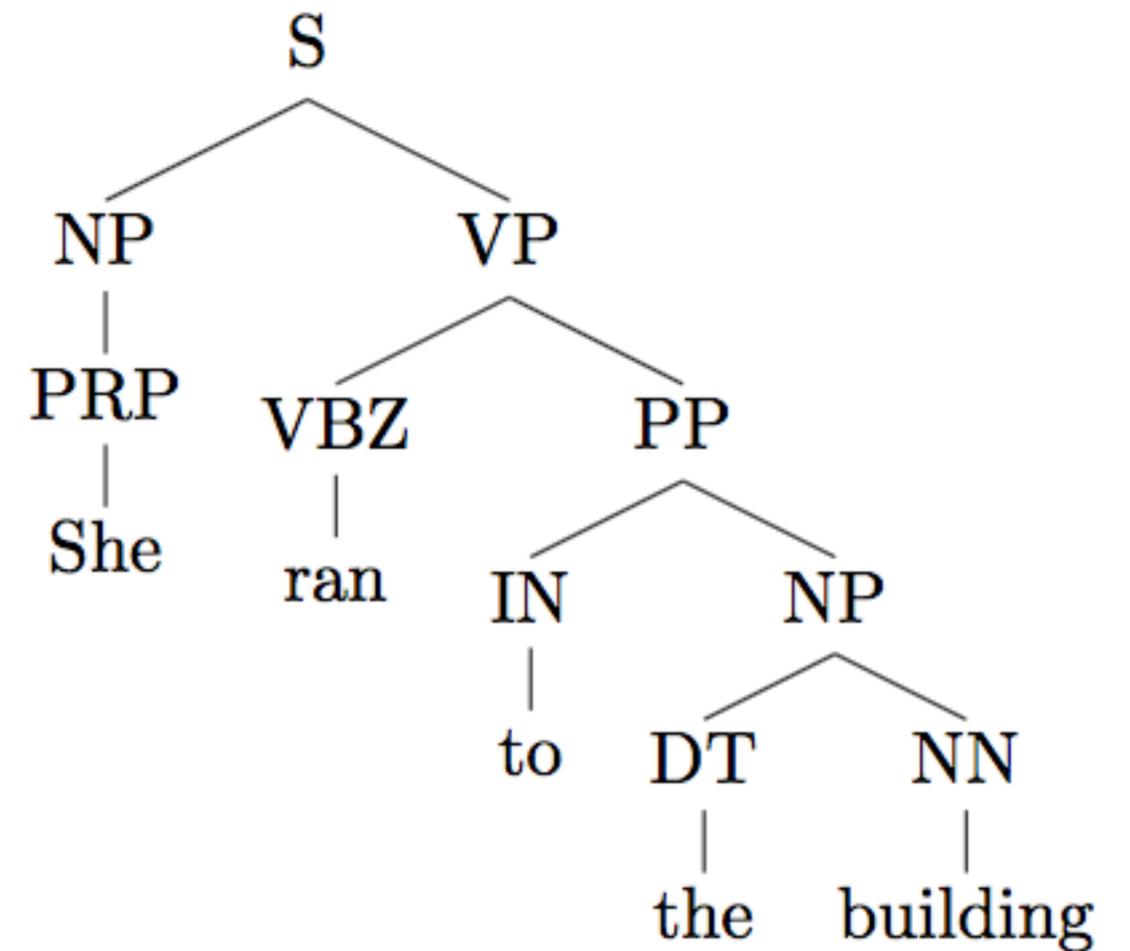


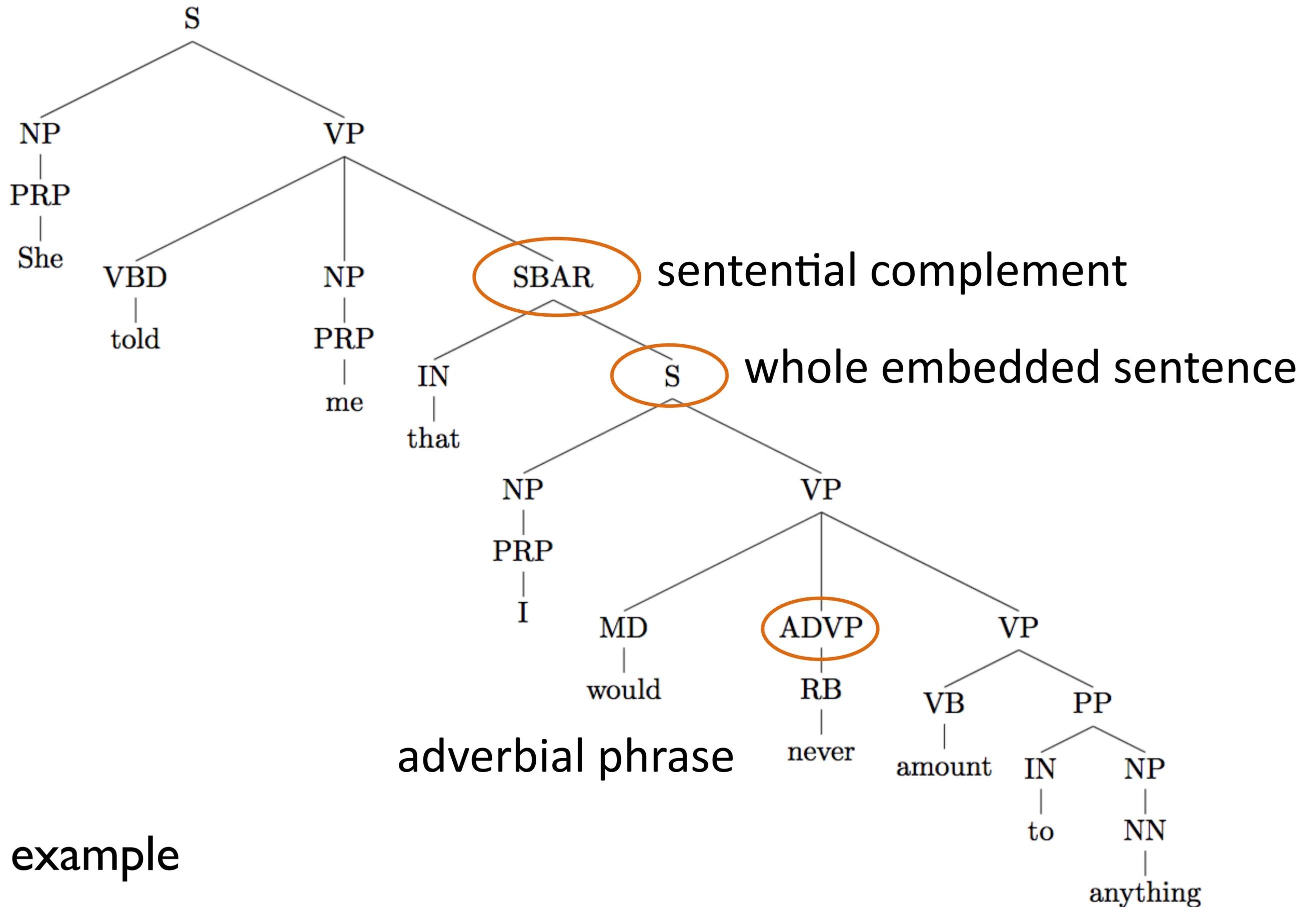
- ▶ Bottom layers are POS tags & words in the sentence
- ▶ Common things: noun phrases, verb phrases, prepositional phrases

# Constituency Parsing

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- ▶ Tree-structured syntactic analyses of sentences
- ▶ Common things: noun phrases, verb phrases, prepositional phrases
- ▶ Bottom layers are POS tags & words in the sentence
- ▶ We will use English sentences as examples, but note that constituency makes sense for a lot of languages



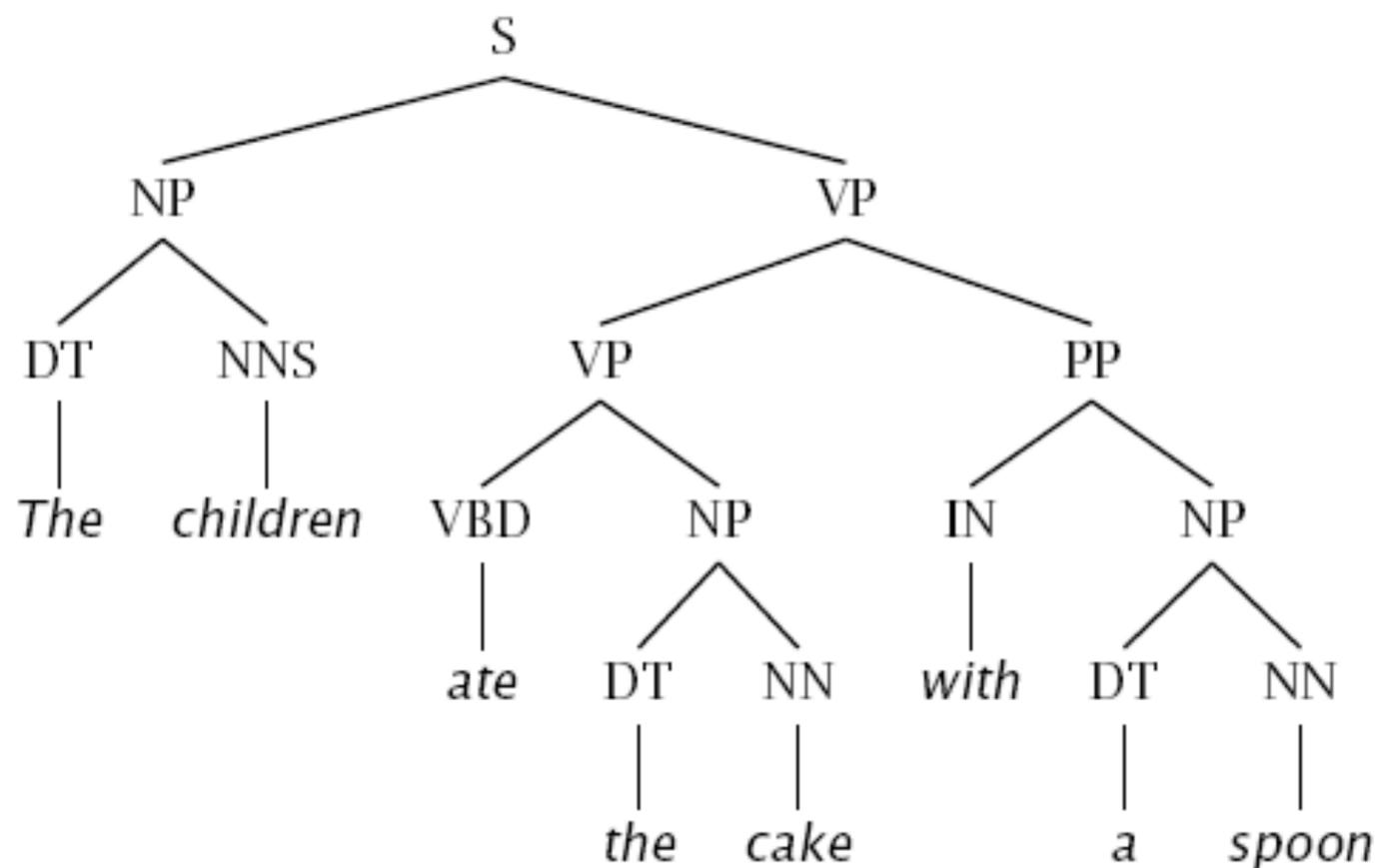


Another example

# Challenges

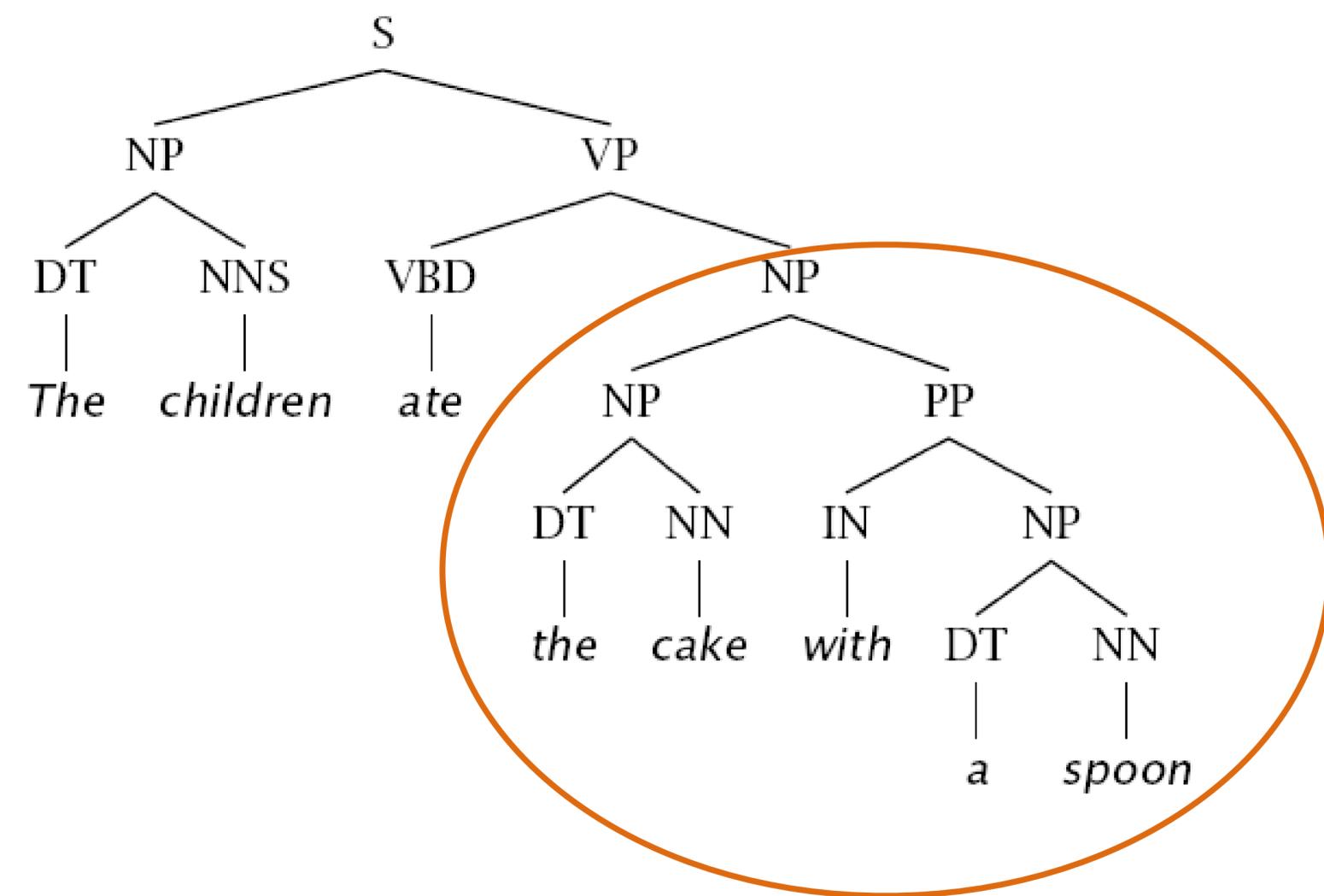
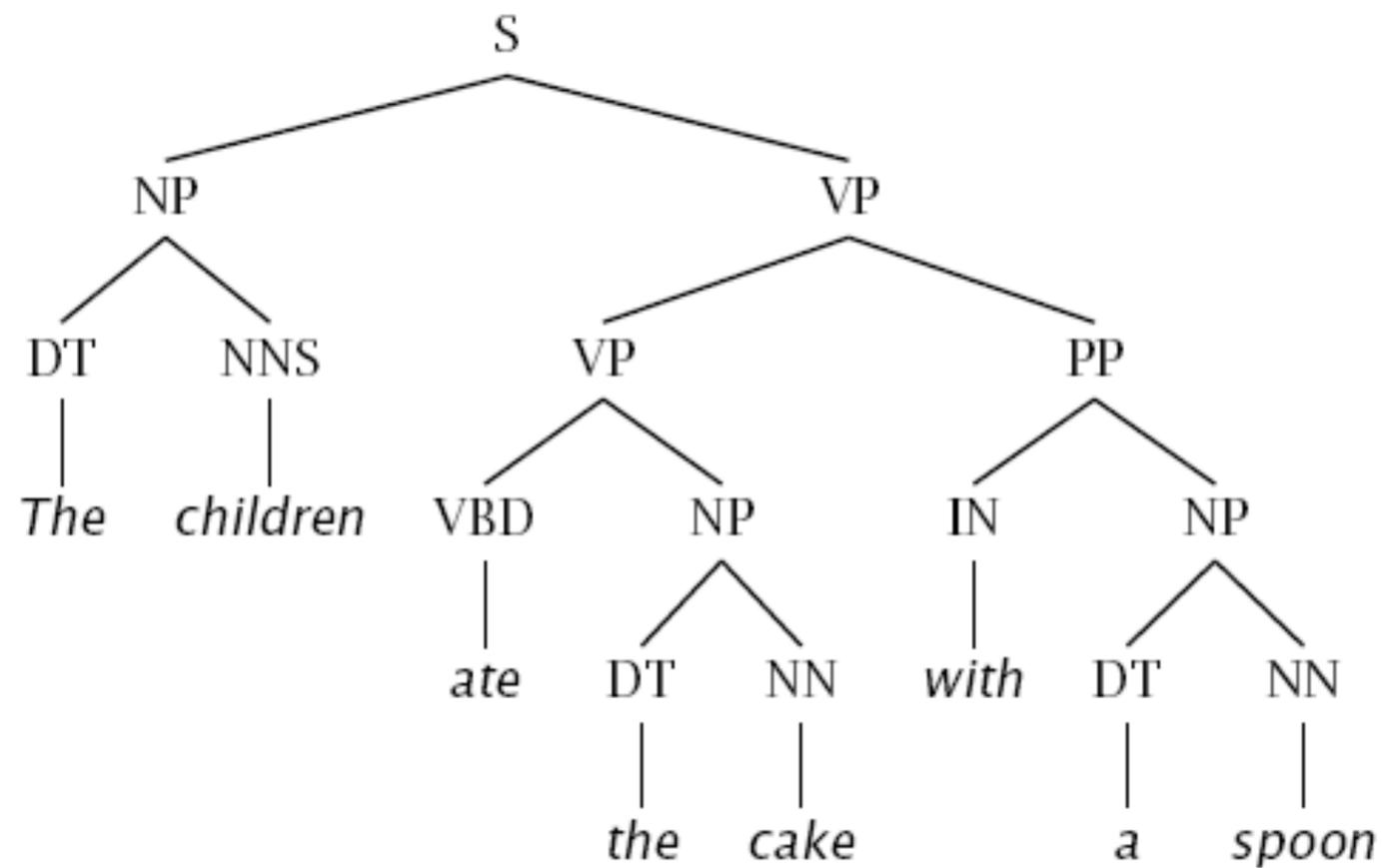
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- ▶ PP (Prepositional Phrase) attachment ambiguity



# Challenges

- ▶ PP (Prepositional Phrase) attachment ambiguity



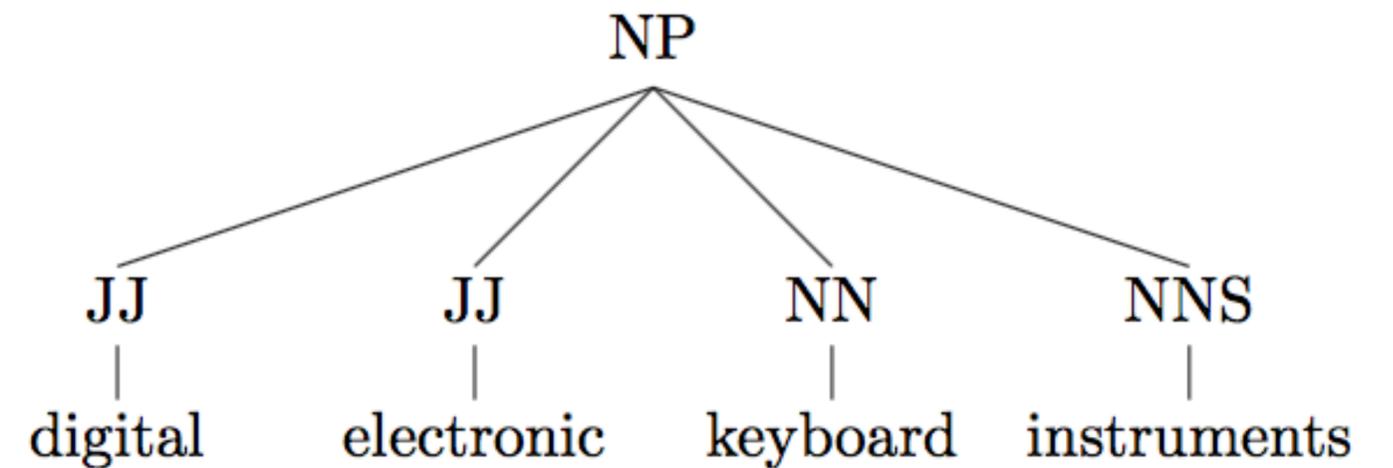
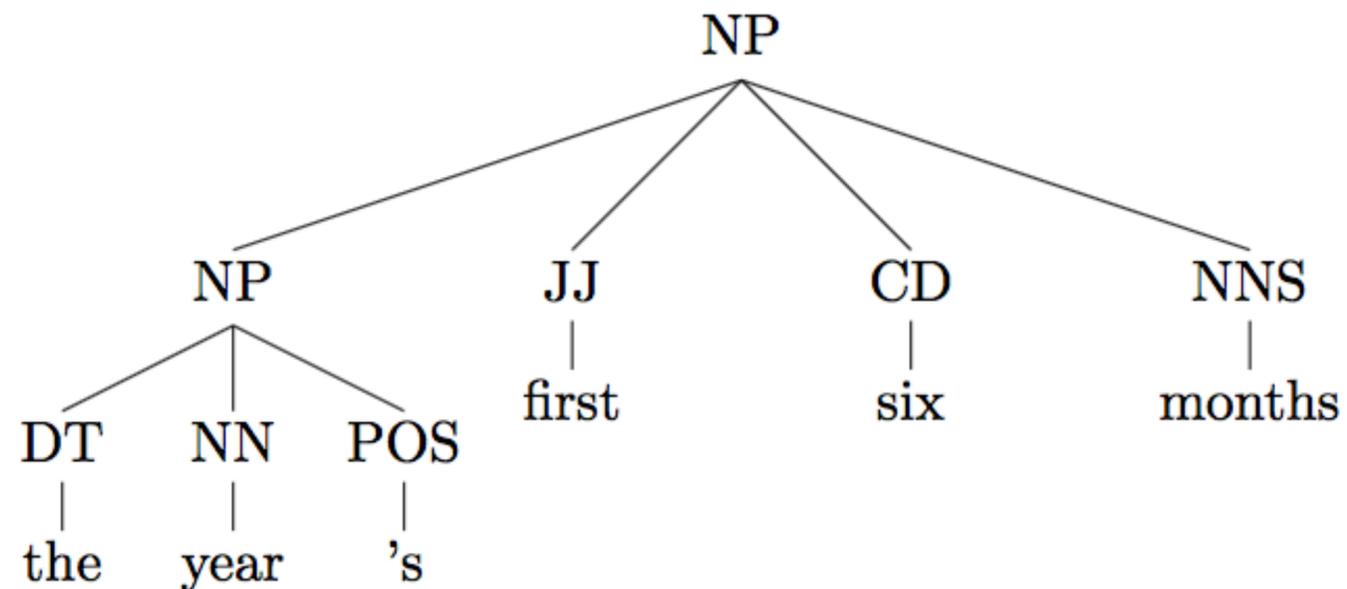
same parse as “the cake with some icing”

# Challenges

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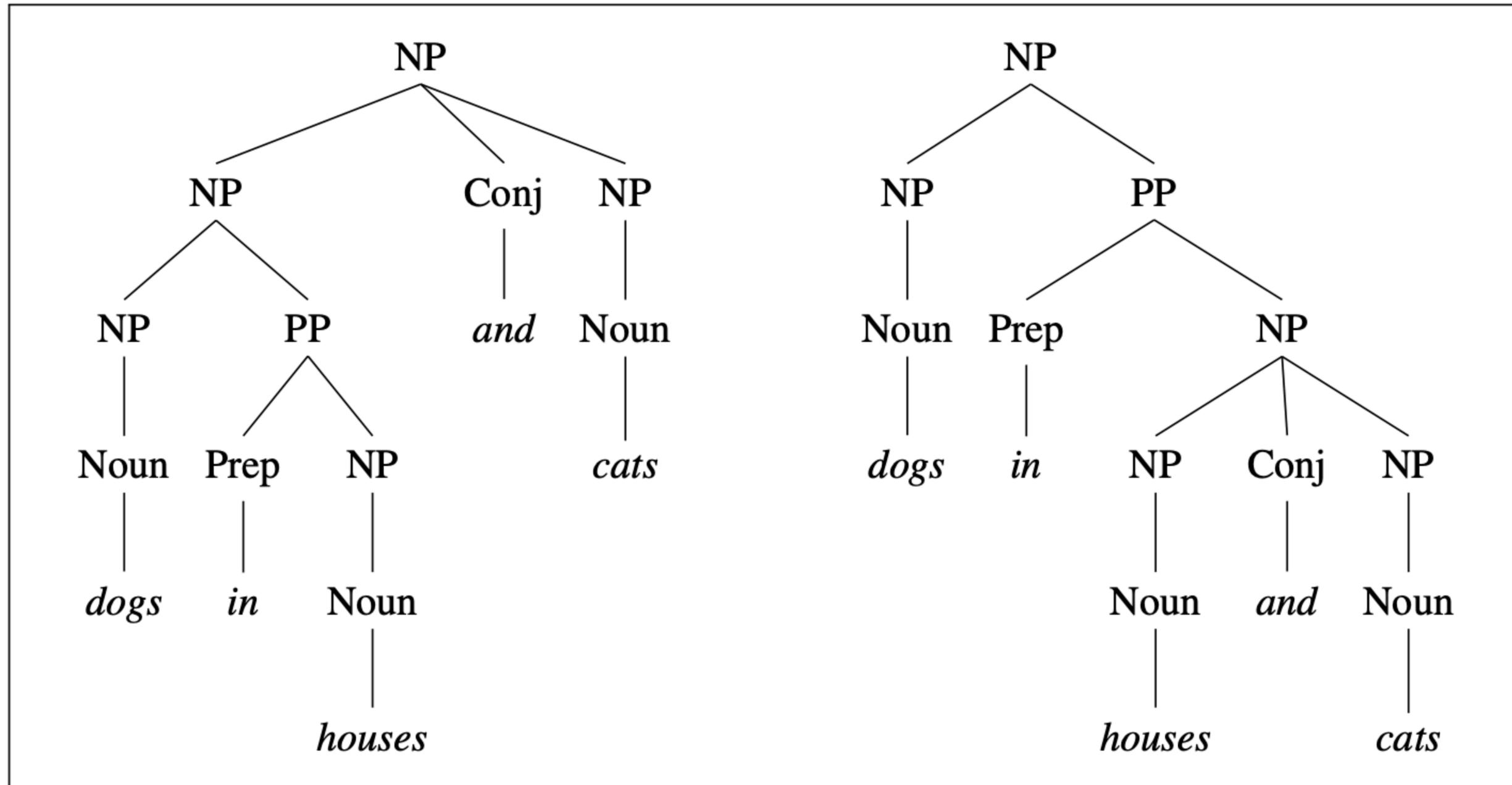
- ▶ NP internal structure: tags + depth of analysis

There are a variety of components an NP can contain



Review tags, if needed: <https://cs.nyu.edu/grishman/jet/guide/PennPOS.html>

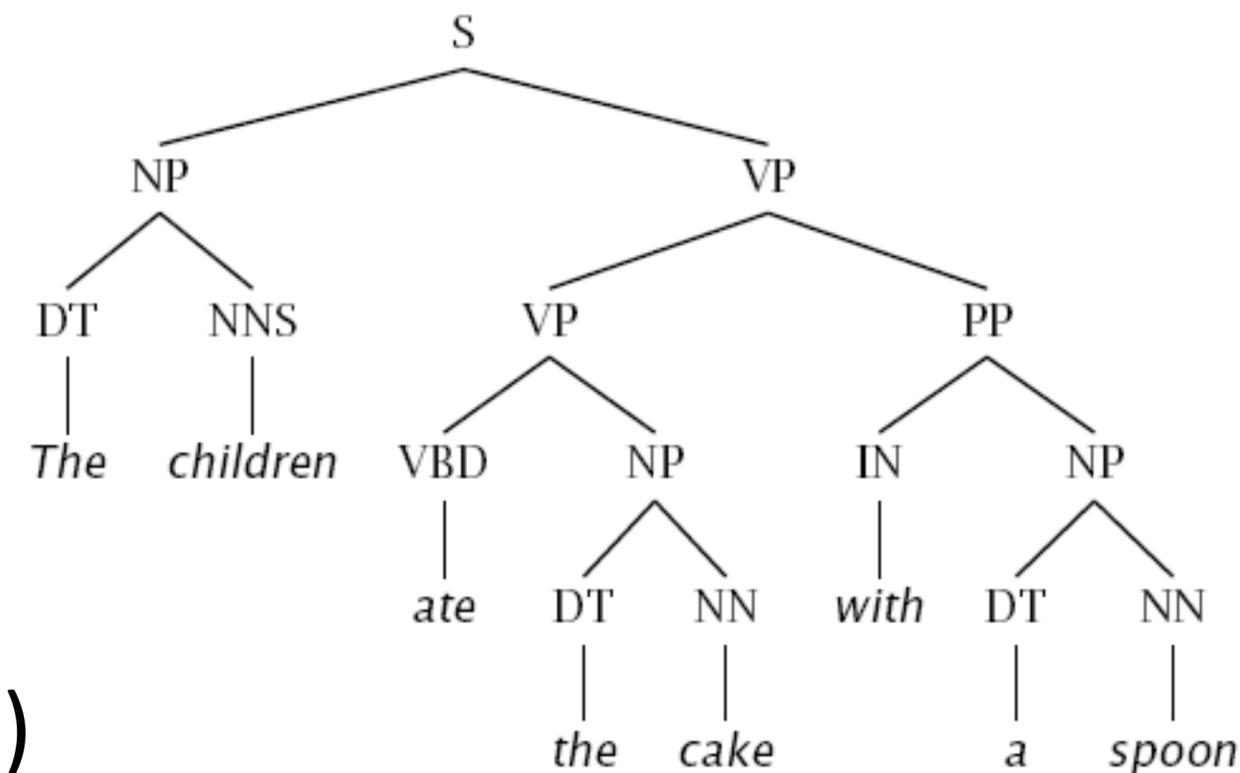
# Coordination ambiguity



**Figure 14.7** An instance of coordination ambiguity. Although the left structure is intuitively the correct one, a PCFG will assign them identical probabilities since both structures use exactly the same set of rules. After [Collins \(1999\)](#).

# Constituency

- ▶ How do we humans know what the constituents are?
- ▶ “Constituency tests”:
  - ▶ Substitution by *pro-form* (e.g., pronoun)
  - ▶ Clefting (*with a spoon is how the children ...*)
  - ▶ Answer ellipsis (What did they eat? *the cake*)  
(How? *with a spoon*)



# Context-Free Grammars

The most widely used formal system for modeling constituent structure in English and other natural languages

Also called Phrase-Structure Grammars

Chapter 12.2 in textbook JM: <https://web.stanford.edu/~jurafsky/slp3/12.pdf>

# CFGs

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## Rules (or, productions)

ROOT  $\rightarrow$  S

S  $\rightarrow$  NP VP

NP  $\rightarrow$  DT NN

NP  $\rightarrow$  NN NNS

NP  $\rightarrow$  NP PP

VP  $\rightarrow$  VBP NP

VP  $\rightarrow$  VBP NP PP

PP  $\rightarrow$  IN NP

## Lexicon

NN  $\rightarrow$  interest

NNS  $\rightarrow$  raises

VBP  $\rightarrow$  interest

VBZ  $\rightarrow$  raises

- ▶ Rules/productions: symbols which rewrite as one or more symbols
  - ▶ each rule expresses the ways that symbols can be grouped and ordered together
- ▶ Lexicon consists of “preterminals” (POS tags) rewriting as terminals (words)

# CFGs

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The following rules express that an **NP** (or **noun phrase**) can be composed of either a *ProperNoun* or a determiner (*Det*) followed by a *Nominal*; a *Nominal* in turn can consist of one or more *Nouns*.

*NP*  $\rightarrow$  *Det Nominal*

*NP*  $\rightarrow$  *ProperNoun*

*Nominal*  $\rightarrow$  *Noun* | *Nominal Noun*

Context-free rules can be hierarchically embedded, so we can combine the previous rules with others, like the following, that express facts about the lexicon:

*Det*  $\rightarrow$  *a*

*Det*  $\rightarrow$  *the*

*Noun*  $\rightarrow$  *flight*

## Example

# CFGs

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## Rules/productions

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- ▶ Rules/productions: symbols which rewrite as one or more symbols
- ▶ Lexicon consists of “preterminals” (POS tags) rewriting as terminals (words)
- ▶ CFG is a tuple (N, T, S, R): N = nonterminals, T = terminals, S = start symbol (generally a special ROOT symbol), R = rules
- ▶ Terminals: words in the language. Non-terminals: symbols that express abstractions over these terminals. What is the item to the left of the arrow?

# CFGs

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## Rules/productions

ROOT  $\rightarrow$  S

S  $\rightarrow$  NP VP

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- ▶ Probabilistic CFG (PCFG): conditional probabilities associated with rules

# CFGs

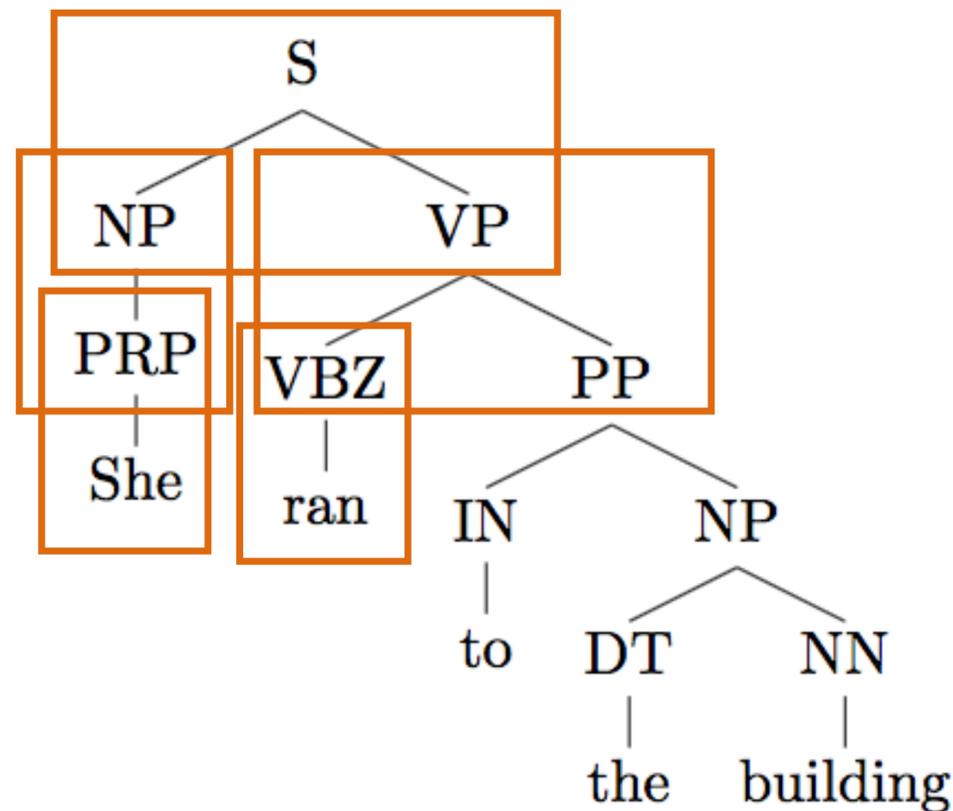
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Rules/productions				Lexicon	
ROOT → S	1.0	NP → NP PP	0.3	NN → interest	1.0
S → NP VP	1.0	VP → VBP NP	0.7	NNS → raises	1.0
NP → DT NN	0.2	VP → VBP NP PP	0.3	VBP → interest	1.0
NP → NN NNS	0.5	PP → IN NP	1.0	VBZ → raises	1.0

- ▶ Rules/productions: symbols which rewrite as one or more symbols
- ▶ Lexicon consists of “preterminals” (POS tags) rewriting as terminals (words)
- ▶ CFG is a tuple (N, T, S, R): N = nonterminals, T = terminals, S = start symbol (generally a special ROOT symbol), R = rules
- ▶ **PCFG**: conditional probabilities associated with rules (num. are made up)

# Estimating PCFGs

- ▶ Tree  $T$  is a series of rule applications  $r$ . (Each rule expands a non-terminal node.)



T includes:

$S \rightarrow NP VP$

$NP \rightarrow PRP$

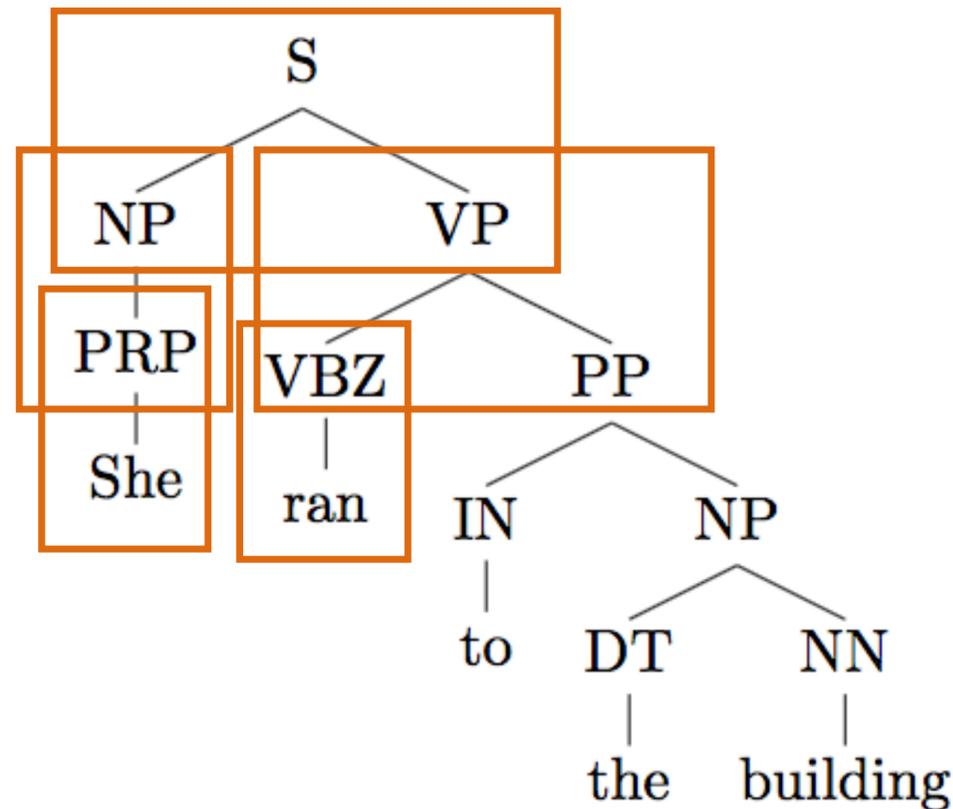
$NP \rightarrow DT NN$

...

# Estimating PCFGs

- ▶ Tree  $T$  is a series of rule applications  $r$ .  $P(T) = \prod_{r \in T} P(r | \text{parent}(r))$

T includes:



$S \rightarrow NP VP$

$NP \rightarrow PRP$

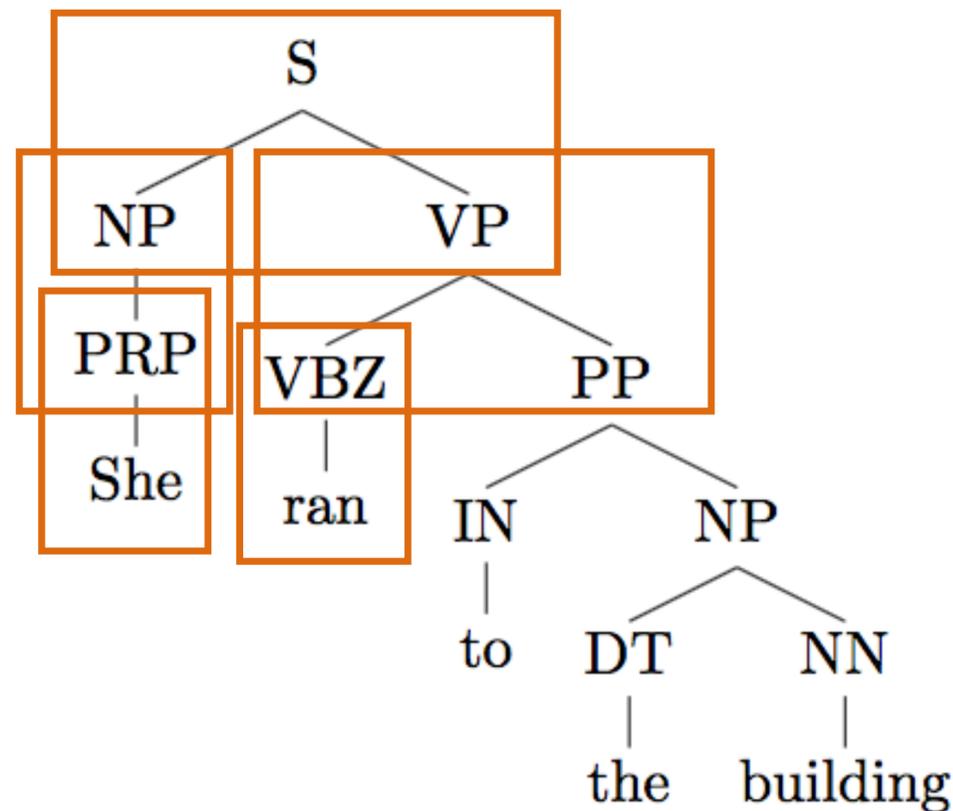
$NP \rightarrow DT NN$

...

The probability of a particular parse tree  $T$  is defined as the product of the probabilities of all the rules used to expand each of the non-terminal nodes in the parse tree  $T$

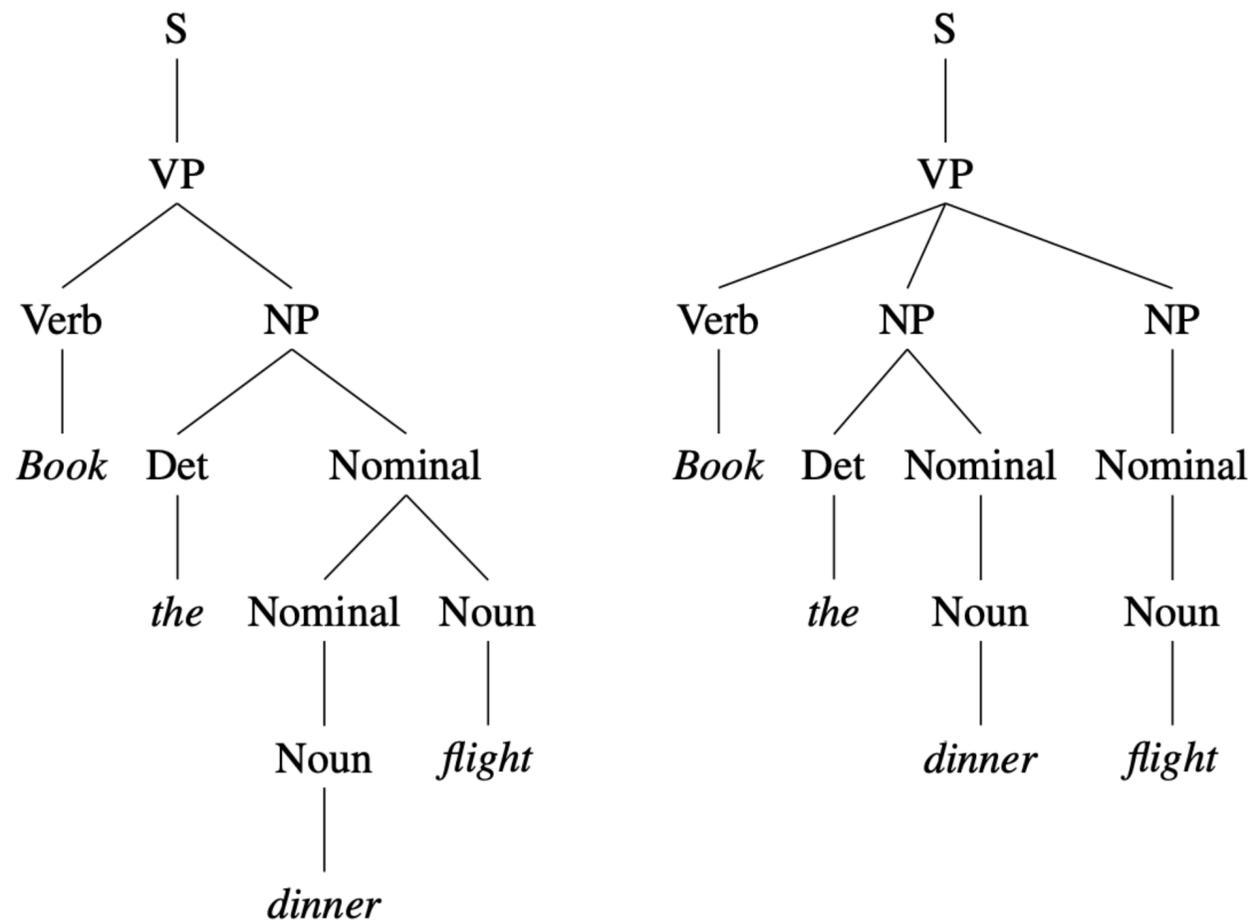
# Estimating PCFGs

- ▶ Tree  $T$  is a series of rule applications  $r$ .  $P(T) = \prod_{r \in T} P(r | \text{parent}(r))$



$S \rightarrow NP VP$       1.0  
 $NP \rightarrow PRP$       0.5  
 $NP \rightarrow DT NN$       0.5  
...

- ▶ Maximum likelihood PCFG: count and normalize! Same as HMMs / Naive Bayes



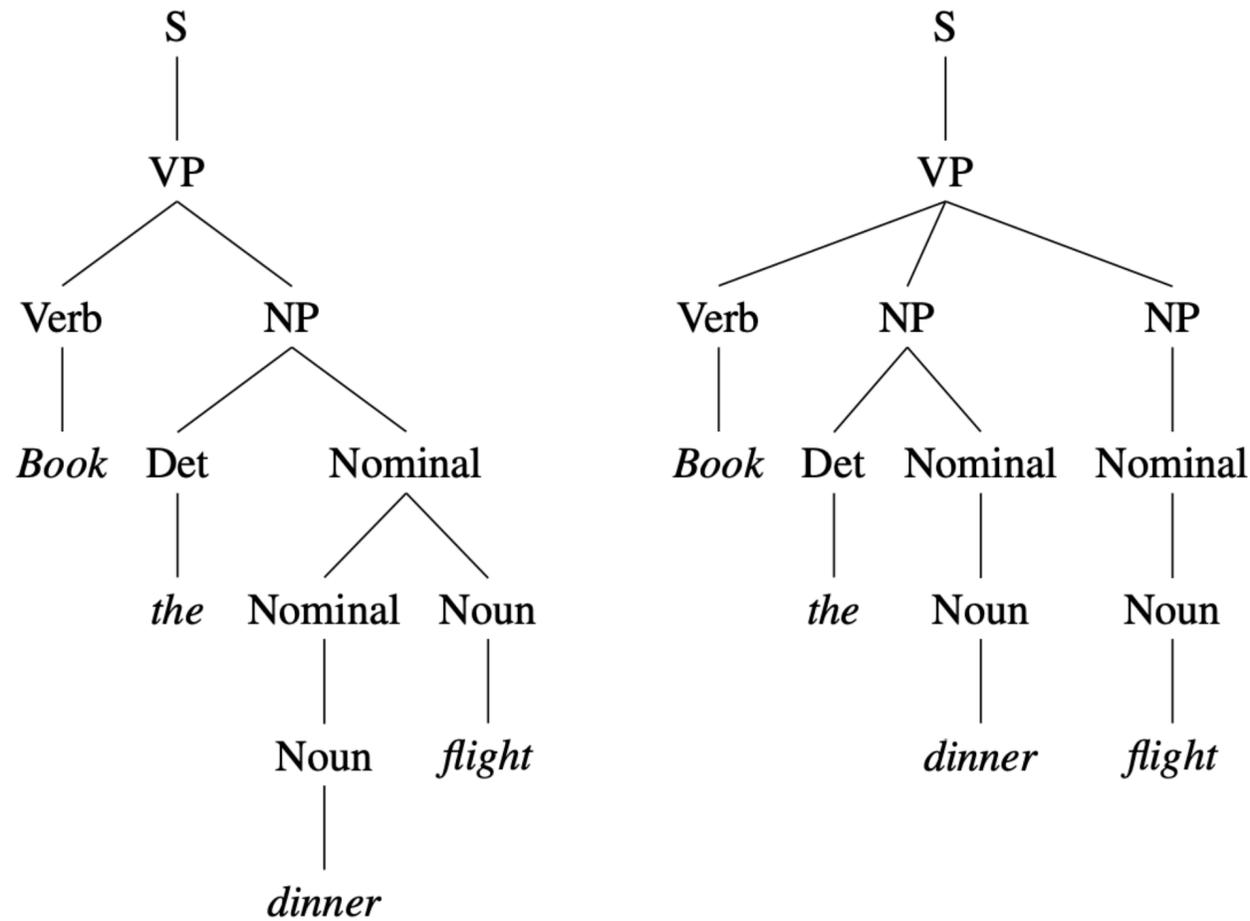
	Rules	P		Rules	P
S	→ VP	.05	S	→ VP	.05
VP	→ Verb NP	.20	VP	→ Verb NP NP	.10
NP	→ Det Nominal	.20	NP	→ Det Nominal	.20
Nominal	→ Nominal Noun	.20	NP	→ Nominal	.15
Nominal	→ Noun	.75	Nominal	→ Noun	.75
			Nominal	→ Noun	.75
Verb	→ book	.30	Verb	→ book	.30
Det	→ the	.60	Det	→ the	.60
Noun	→ dinner	.10	Noun	→ dinner	.10
Noun	→ flight	.40	Noun	→ flight	.40

**Figure 14.2** Two parse trees for an ambiguous sentence. The parse on the left corresponds to the sensible meaning “Book a flight that serves dinner”, while the parse on the right corresponds to the nonsensical meaning “Book a flight on behalf of ‘the dinner’ ”.

$$P(T_{left}) = .05 * .20 * .20 * .20 * .75 * .30 * .60 * .10 * .40 = 2.2 \times 10^{-6}$$

$$P(T_{right}) = .05 * .10 * .20 * .15 * .75 * .75 * .30 * .60 * .10 * .40 = 6.1 \times 10^{-7}$$

## How to produce the most-likely parse T?



Rules	P	Rules	P
S → VP	.05	S → VP	.05
VP → Verb NP	.20	VP → Verb NP NP	.10
NP → Det Nominal	.20	NP → Det Nominal	.20
Nominal → Nominal Noun	.20	NP → Nominal	.15
Nominal → Noun	.75	Nominal → Noun	.75
		Nominal → Noun	.75
Verb → book	.30	Verb → book	.30
Det → the	.60	Det → the	.60
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