

Overview

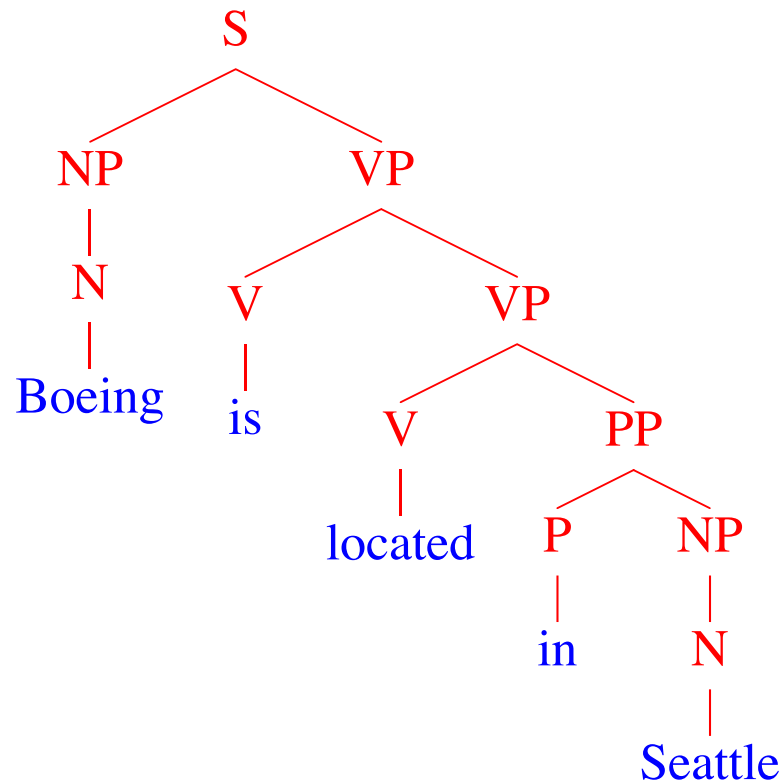
- An introduction to the parsing problem
- Context free grammars
- A brief(!) sketch of the syntax of English
- Examples of ambiguous structures
- PCFGs, their formal properties, and useful algorithms
- Weaknesses of PCFGs

Parsing (Syntactic Structure)

INPUT:

Boeing is located in Seattle.

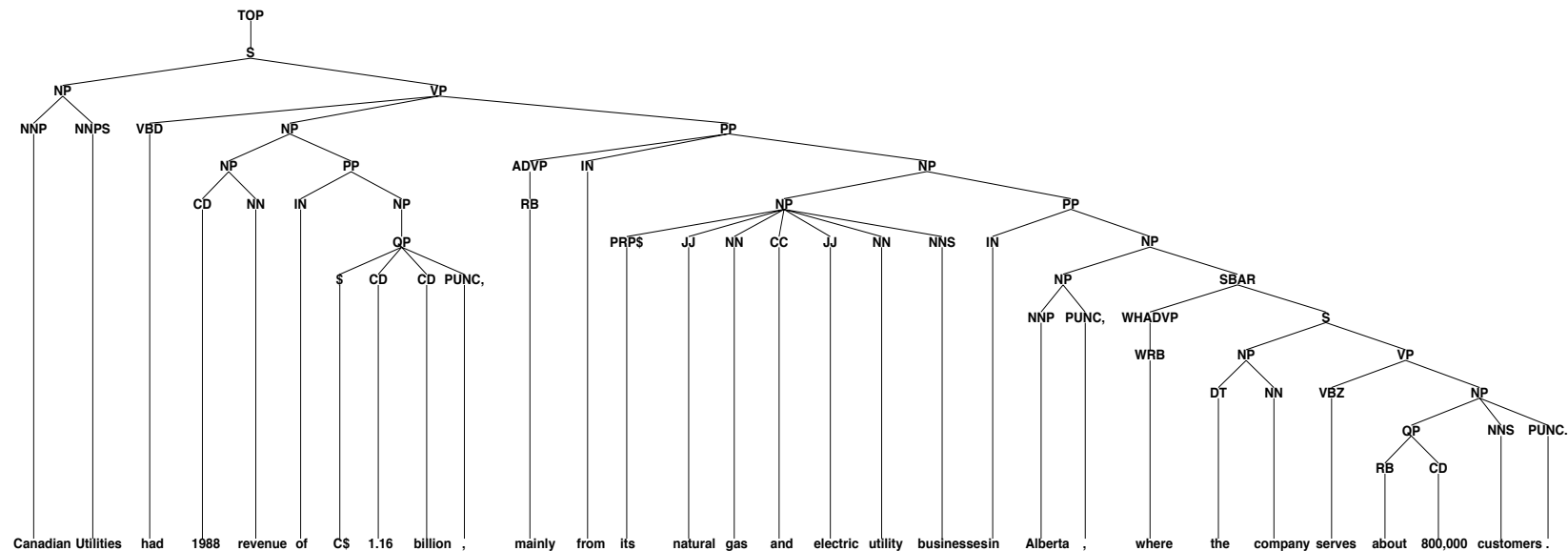
OUTPUT:



Data for Parsing Experiments

- Penn WSJ Treebank = 50,000 sentences with associated trees
- Usual set-up: 40,000 training sentences, 2400 test sentences

An example tree:

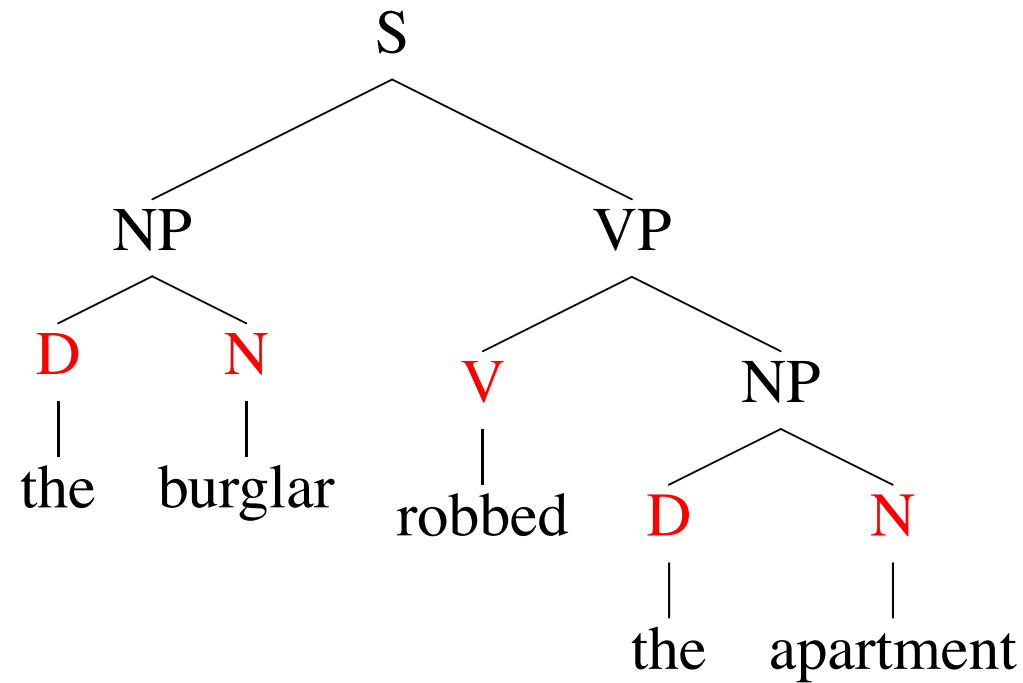


Canadian Utilities had 1988 revenue of C\$ 1.16 billion , mainly from its natural gas and electric utility businesses in Alberta , where the company serves about 800,000 customers .

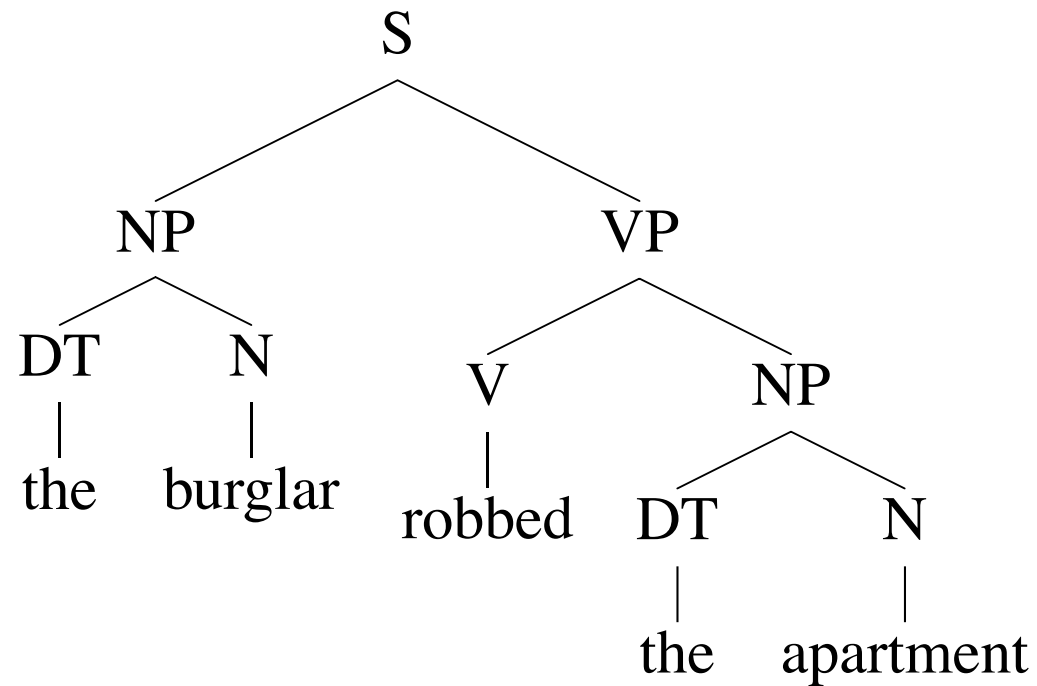
The Information Conveyed by Parse Trees

1) Part of speech for each word

(N = noun, V = verb, D = determiner)



2) Phrases

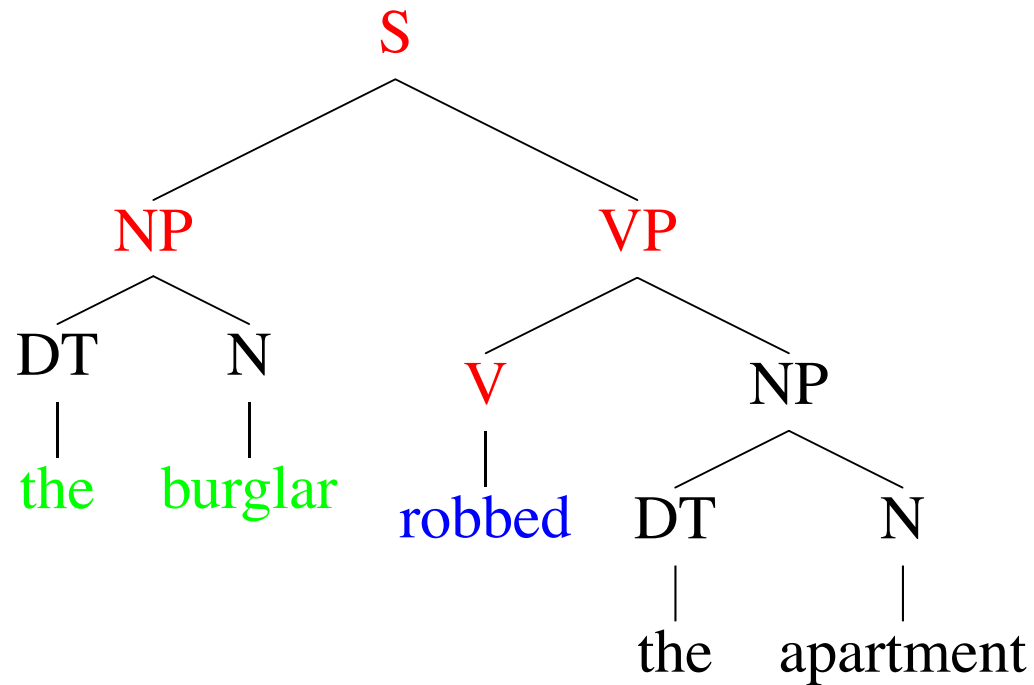
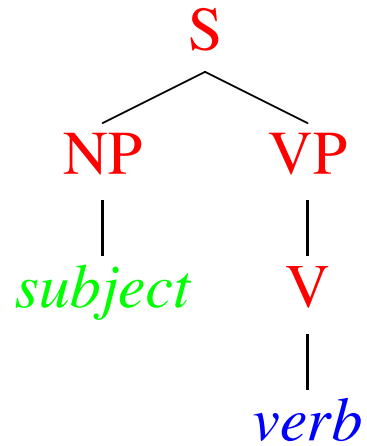


Noun Phrases (NP): “the burglar”, “the apartment”

Verb Phrases (VP): “robbed the apartment”

Sentences (S): “the burglar robbed the apartment”

3) Useful Relationships



⇒ “the burglar” is the subject of “robbed”

An Example Application: Machine Translation

- English word order is *subject – verb – object*
- Japanese word order is *subject – object – verb*

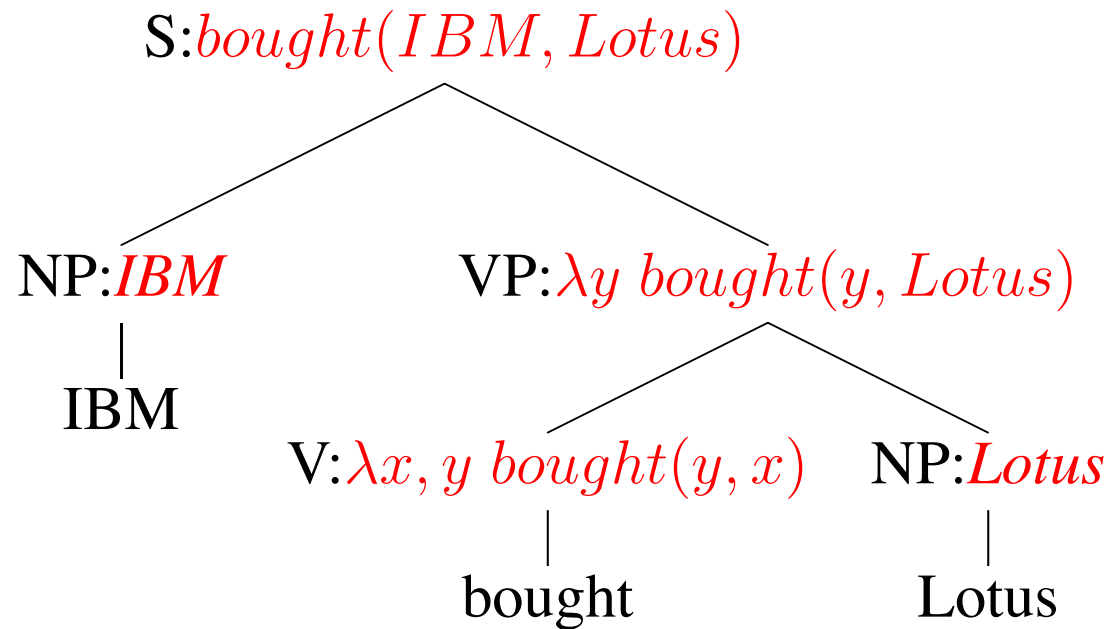
English: IBM bought Lotus

Japanese: *IBM Lotus bought*

English: Sources said that IBM bought Lotus yesterday

Japanese: *Sources yesterday IBM Lotus bought that said*

Syntax and Compositional Semantics



- Each syntactic non-terminal now has an associated **semantic expression**
- (We'll see more of this later in the course)

Context-Free Grammars

[[Hopcroft and Ullman 1979](#)]

A context free grammar $G = (N, \Sigma, R, S)$ where:

- N is a set of non-terminal symbols
- Σ is a set of terminal symbols
- R is a set of rules of the form $X \rightarrow Y_1 Y_2 \dots Y_n$
for $n \geq 0$, $X \in N$, $Y_i \in (N \cup \Sigma)$
- $S \in N$ is a distinguished start symbol

A Context-Free Grammar for English

$N = \{S, NP, VP, PP, DT, Vi, Vt, NN, IN\}$

$S = S$

$\Sigma = \{\text{sleeps, saw, man, woman, telescope, the, with, in}\}$

$R =$

| | | | |
|----|---------------|----|----|
| S | \Rightarrow | NP | VP |
| VP | \Rightarrow | Vi | |
| VP | \Rightarrow | Vt | NP |
| VP | \Rightarrow | VP | PP |
| NP | \Rightarrow | DT | NN |
| NP | \Rightarrow | NP | PP |
| PP | \Rightarrow | IN | NP |

| | | |
|----|---------------|-----------|
| Vi | \Rightarrow | sleeps |
| Vt | \Rightarrow | saw |
| NN | \Rightarrow | man |
| NN | \Rightarrow | woman |
| NN | \Rightarrow | telescope |
| DT | \Rightarrow | the |
| IN | \Rightarrow | with |
| IN | \Rightarrow | in |

Note: S=sentence, VP=verb phrase, NP=noun phrase, PP=prepositional phrase, DT=determiner, Vi=intransitive verb, Vt=transitive verb, NN=noun, IN=preposition

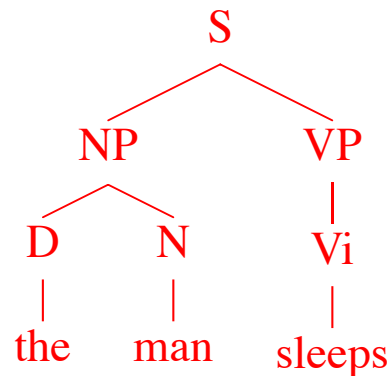
Left-Most Derivations

A left-most derivation is a sequence of strings $s_1 \dots s_n$, where

- $s_1 = S$, the start symbol
- $s_n \in \Sigma^*$, i.e. s_n is made up of terminal symbols only
- Each s_i for $i = 2 \dots n$ is derived from s_{i-1} by picking the left-most non-terminal X in s_{i-1} and replacing it by some β where $X \rightarrow \beta$ is a rule in R

For example: [S], [NP VP], [D N VP], [the N VP], [the man VP], [the man Vi], [the man sleeps]

Representation of a derivation as a tree:



DERIVATION

S

RULES USED

DERIVATION

S

NP VP

RULES USED

$S \rightarrow NP VP$

DERIVATION

S

NP VP

DT N VP

RULES USED

$S \rightarrow NP VP$

$NP \rightarrow DT N$

DERIVATION

S

NP VP

DT N VP

the N VP

RULES USED

$S \rightarrow NP VP$

$NP \rightarrow DT N$

$DT \rightarrow \text{the}$

DERIVATION

S

NP VP

DT N VP

the N VP

the dog VP

RULES USED

$S \rightarrow NP VP$

$NP \rightarrow DT N$

$DT \rightarrow \text{the}$

$N \rightarrow \text{dog}$

DERIVATION

S

NP VP

DT N VP

the N VP

the dog VP

the dog VB

RULES USED

$S \rightarrow NP VP$

$NP \rightarrow DT N$

$DT \rightarrow \text{the}$

$N \rightarrow \text{dog}$

$VP \rightarrow VB$

DERIVATION

S

NP VP

DT N VP

the N VP

the dog VP

the dog VB

the dog laughs

RULES USED

$S \rightarrow NP VP$

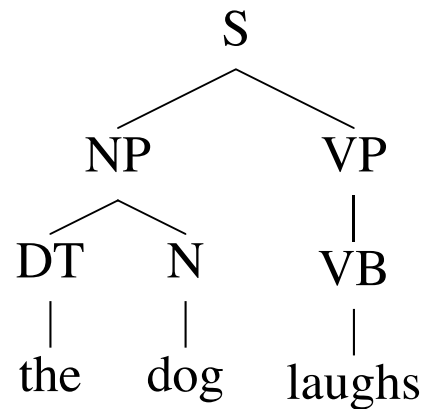
$NP \rightarrow DT N$

$DT \rightarrow \text{the}$

$N \rightarrow \text{dog}$

$VP \rightarrow VB$

$VB \rightarrow \text{laughs}$



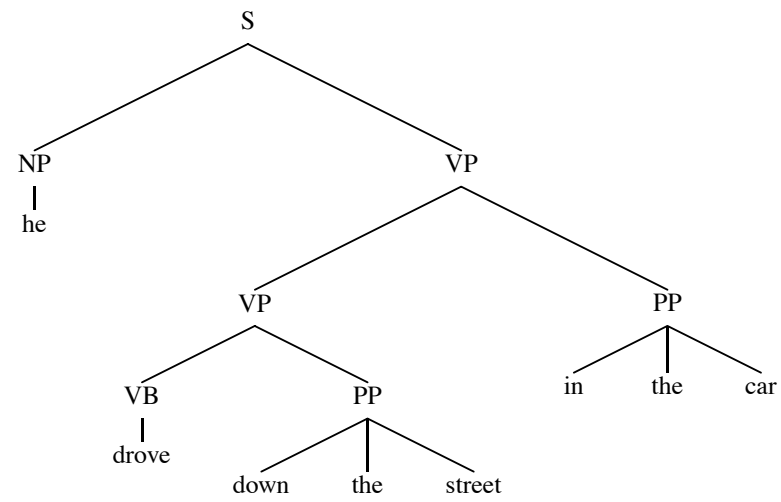
Properties of CFGs

- A CFG defines a set of possible derivations
- A string $s \in \Sigma^*$ is in the *language* defined by the CFG if there is at least one derivation which yields s
- Each string in the language generated by the CFG may have more than one derivation (“ambiguity”)

DERIVATION

S

RULES USED



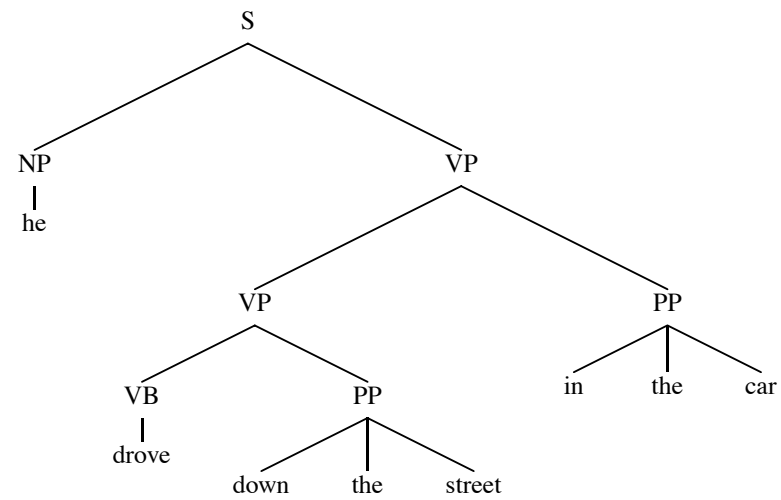
DERIVATION

S

NP VP

RULES USED

$S \rightarrow NP VP$



DERIVATION

S

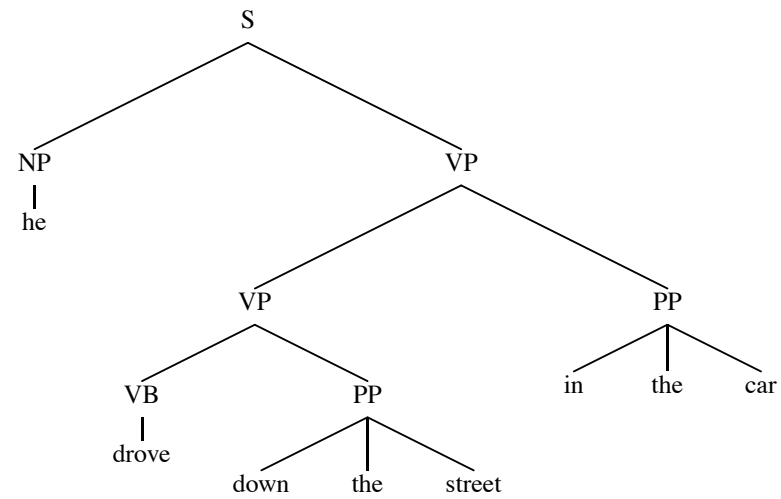
NP VP

he VP

RULES USED

$S \rightarrow NP VP$

$NP \rightarrow he$



DERIVATION

S

NP VP

he VP

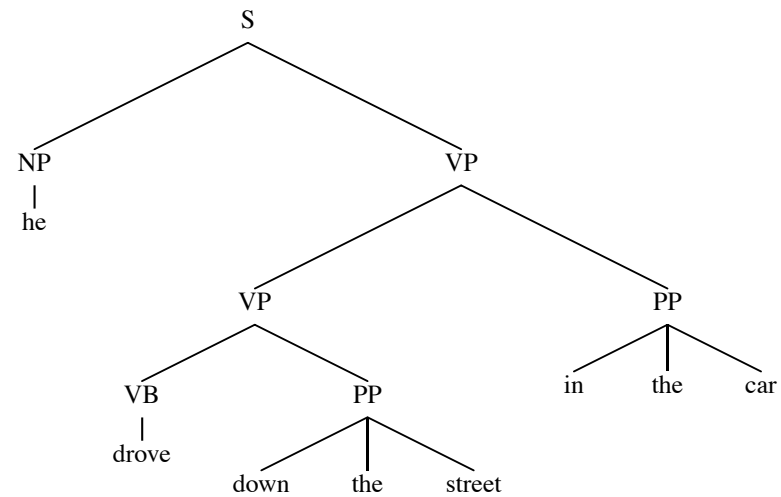
he VP PP

RULES USED

$S \rightarrow NP VP$

$NP \rightarrow he$

$VP \rightarrow VP PP$



DERIVATION

S

NP VP

he VP

he VP PP

he VB PP PP

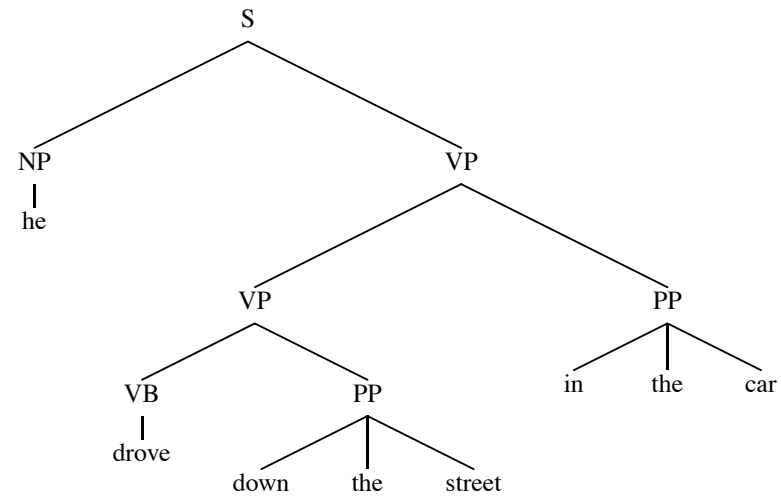
RULES USED

$S \rightarrow NP VP$

$NP \rightarrow he$

$VP \rightarrow VP PP$

$VP \rightarrow VB PP$



DERIVATION

S

NP VP

he VP

he VP PP

he VB PP PP

he drove PP PP

RULES USED

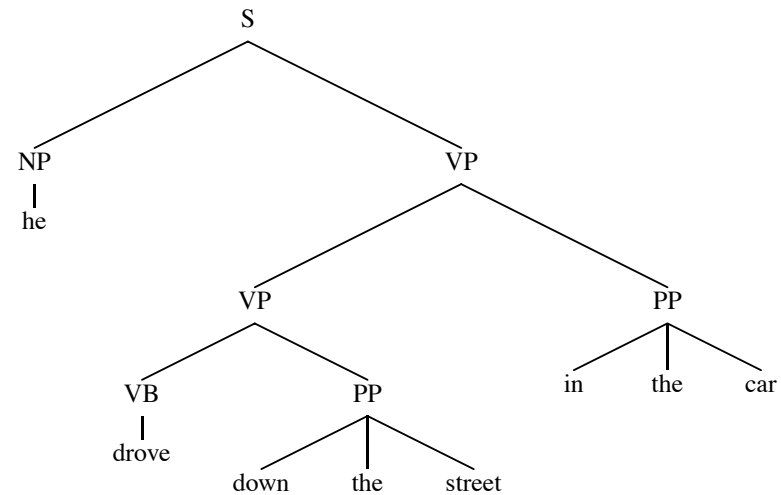
$S \rightarrow NP VP$

$NP \rightarrow he$

$VP \rightarrow VP PP$

$VP \rightarrow VB PP$

$VB \rightarrow drove$



DERIVATION

S

NP VP

he VP

he VP PP

he VB PP PP

he drove PP PP

he drove down the street PP

RULES USED

$S \rightarrow NP VP$

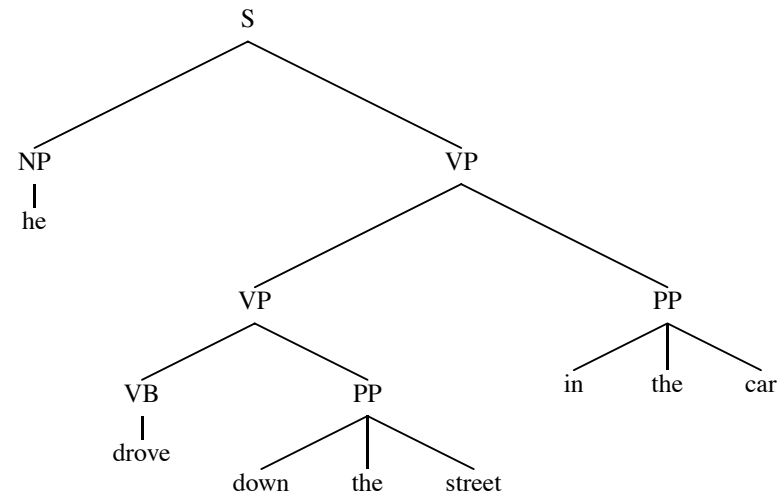
$NP \rightarrow he$

$VP \rightarrow VP PP$

$VP \rightarrow VB PP$

$VB \rightarrow drove$

$PP \rightarrow down\ the\ street$



DERIVATION

S

NP VP

he VP

he VP PP

he VB PP PP

he drove PP PP

he drove down the street PP

he drove down the street in the car

RULES USED

$S \rightarrow NP VP$

$NP \rightarrow he$

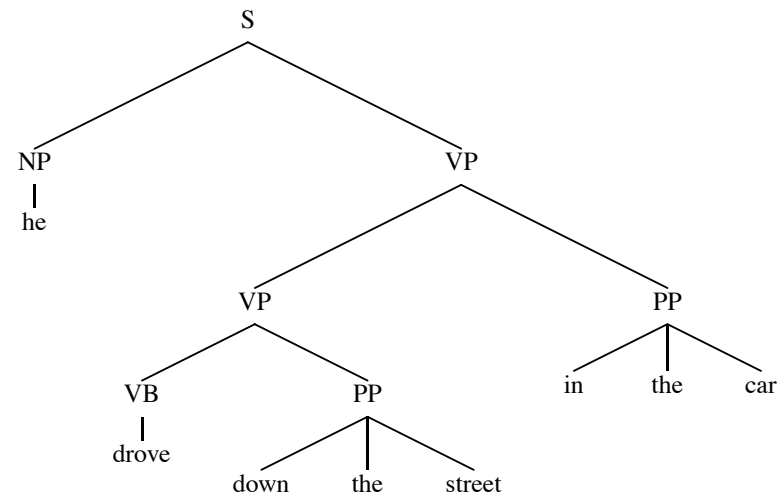
$VP \rightarrow VP PP$

$VP \rightarrow VB PP$

$VB \rightarrow drove$

$PP \rightarrow down\ the\ street$

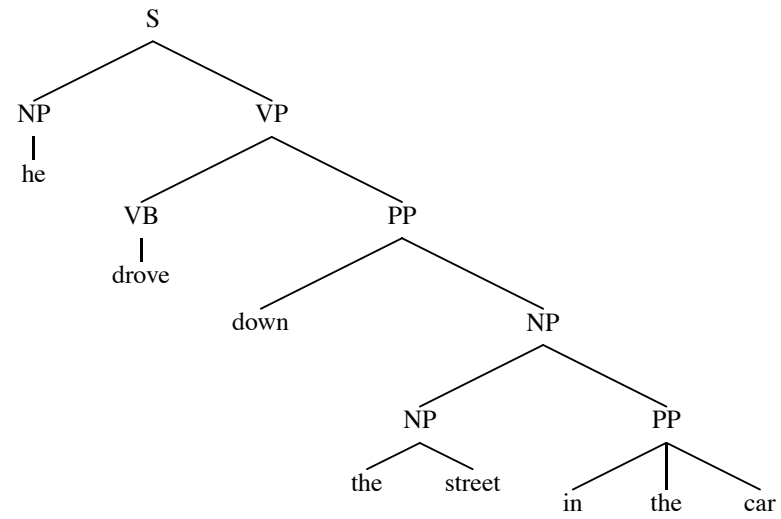
$PP \rightarrow in\ the\ car$



DERIVATION

S

RULES USED



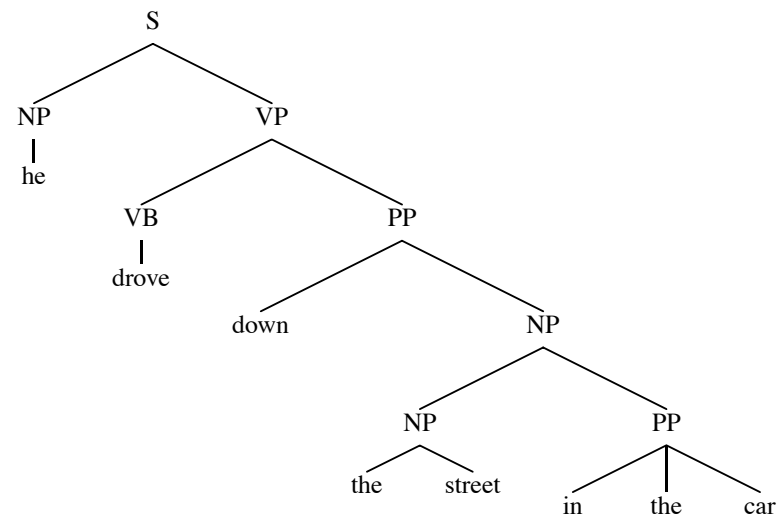
DERIVATION

S

NP VP

RULES USED

$S \rightarrow NP VP$



DERIVATION

S

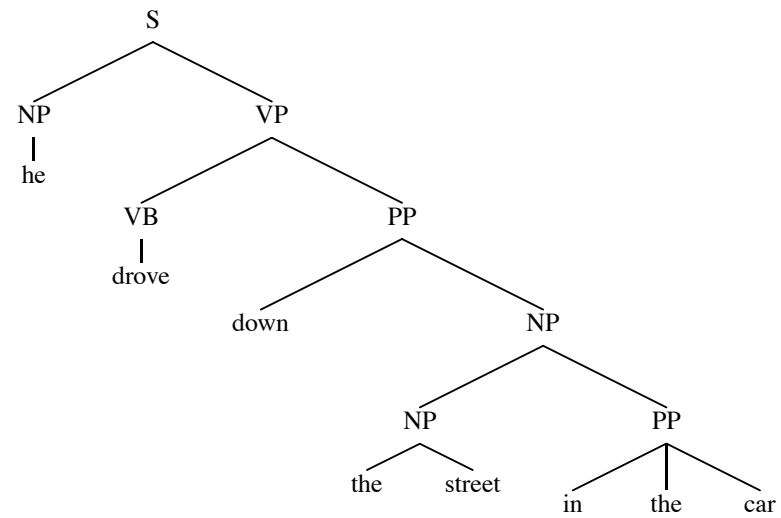
NP VP

he VP

RULES USED

$S \rightarrow NP VP$

$NP \rightarrow he$



DERIVATION

S

NP VP

he VP

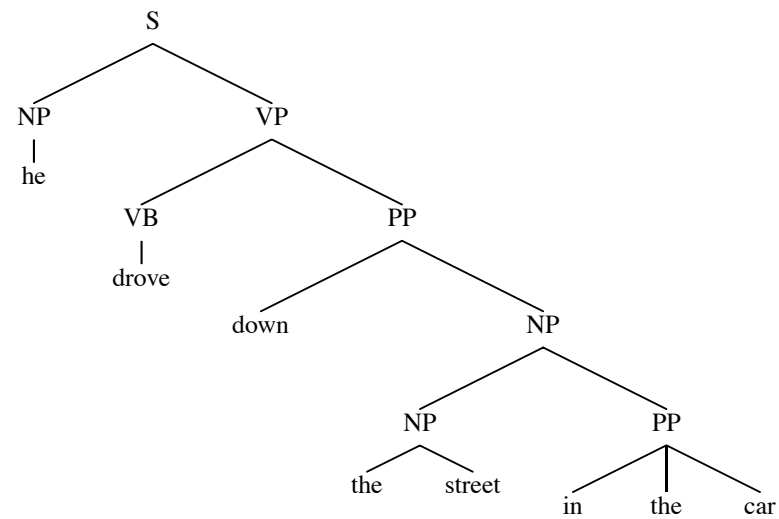
he VB PP

RULES USED

$S \rightarrow NP VP$

$NP \rightarrow he$

$VP \rightarrow VB PP$



DERIVATION

S

NP VP

he VP

he VB PP

he drove PP

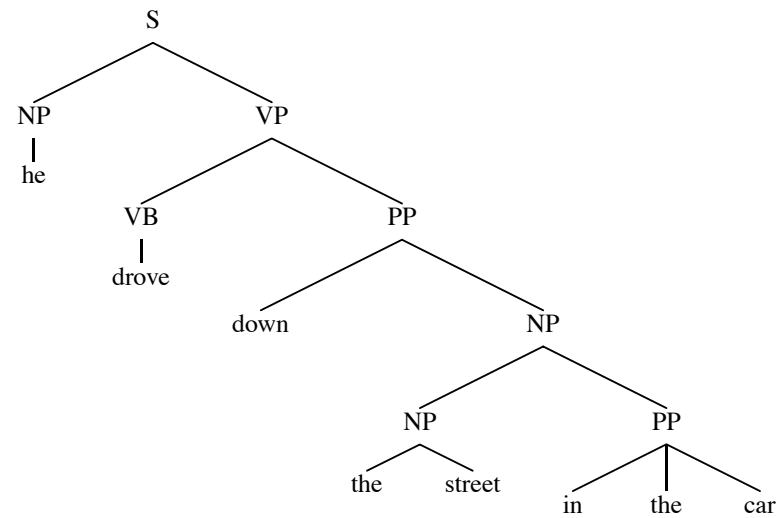
RULES USED

$S \rightarrow NP VP$

$NP \rightarrow he$

$VP \rightarrow VB PP$

$VB \rightarrow drove$



DERIVATION

S

NP VP

he VP

he VB PP

he drove PP

he drove down NP

RULES USED

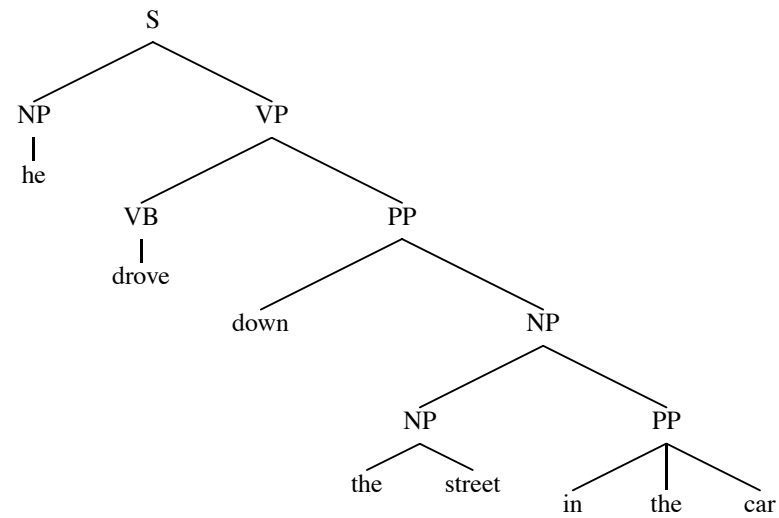
$S \rightarrow NP VP$

$NP \rightarrow he$

$VP \rightarrow VB PP$

$VB \rightarrow drove$

$PP \rightarrow down NP$



DERIVATION

S

NP VP

he VP

he VB PP

he drove PP

he drove down NP

he drove down NP PP

RULES USED

$S \rightarrow NP VP$

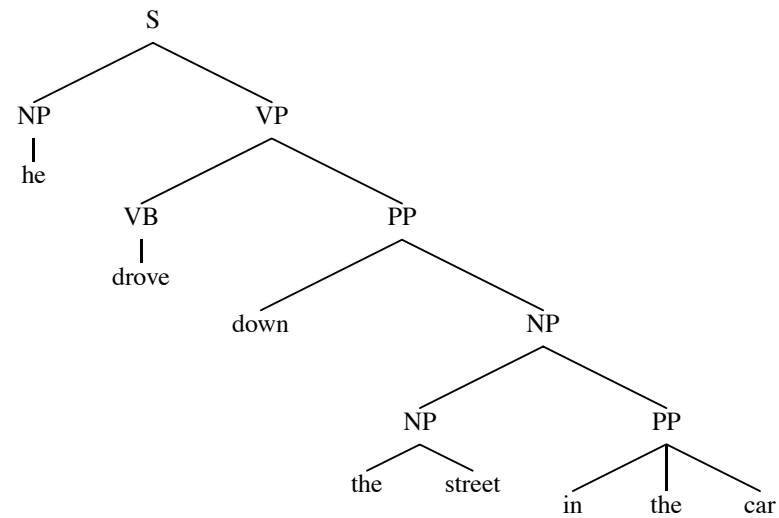
$NP \rightarrow he$

$VP \rightarrow VB PP$

$VB \rightarrow drove$

$PP \rightarrow down NP$

$NP \rightarrow NP PP$



DERIVATION

S

NP VP

he VP

he VB PP

he drove PP

he drove down NP

he drove down NP PP

he drove down the street PP

RULES USED

$S \rightarrow NP VP$

$NP \rightarrow he$

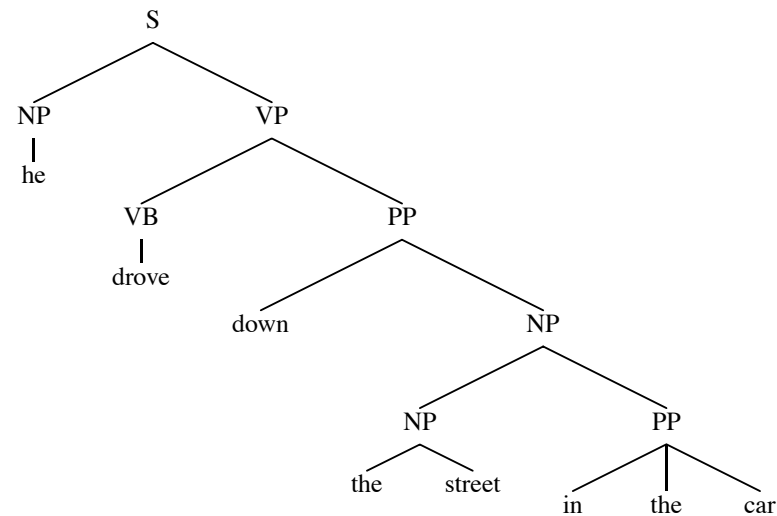
$VP \rightarrow VB PP$

$VB \rightarrow drove$

$PP \rightarrow down NP$

$NP \rightarrow NP PP$

$NP \rightarrow the street$



DERIVATION

S

NP VP

he VP

he VB PP

he drove PP

he drove down NP

he drove down NP PP

he drove down the street PP

he drove down the street in the car

RULES USED

$S \rightarrow NP VP$

$NP \rightarrow he$

$VP \rightarrow VB PP$

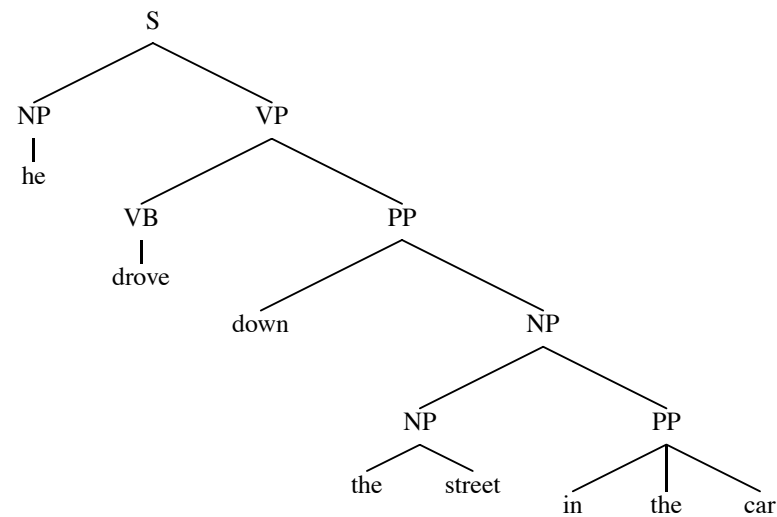
$VB \rightarrow drove$

$PP \rightarrow down NP$

$NP \rightarrow NP PP$

$NP \rightarrow the\ street$

$PP \rightarrow in\ the\ car$



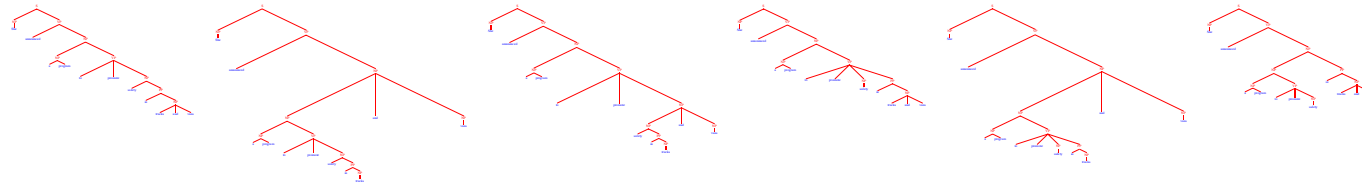
The Problem with Parsing: Ambiguity

INPUT:

She announced a program to promote safety in trucks and vans



POSSIBLE OUTPUTS:



And there are more...

A Brief Overview of English Syntax

Parts of Speech:

- Nouns
(Tags from the *Brown corpus*)
 - NN = singular noun e.g., man, dog, park
 - NNS = plural noun e.g., telescopes, houses, buildings
 - NNP = proper noun e.g., Smith, Gates, IBM
- Determiners
 - DT = determiner e.g., the, a, some, every
- Adjectives
 - JJ = adjective e.g., red, green, large, idealistic

A Fragment of a Noun Phrase Grammar

$\bar{N} \Rightarrow NN$
 $\bar{N} \Rightarrow NN \bar{N}$
 $\bar{N} \Rightarrow JJ \bar{N}$
 $\bar{N} \Rightarrow \bar{N} \bar{N}$
 $NP \Rightarrow DT \bar{N}$

$NN \Rightarrow \text{box}$
 $NN \Rightarrow \text{car}$
 $NN \Rightarrow \text{mechanic}$
 $NN \Rightarrow \text{pigeon}$

$DT \Rightarrow \text{the}$
 $DT \Rightarrow \text{a}$

$JJ \Rightarrow \text{fast}$
 $JJ \Rightarrow \text{metal}$
 $JJ \Rightarrow \text{idealistic}$
 $JJ \Rightarrow \text{clay}$

Generates:

a box, the box, the metal box, the fast car mechanic, . . .

Prepositions, and Prepositional Phrases

- Prepositions

IN = preposition e.g., of, in, out, beside, as

An Extended Grammar

| | | | | | | | | | |
|-----------|---------------|-----------|-----------|----|---------------|----------|----|---------------|------------|
| \bar{N} | \Rightarrow | NN | | NN | \Rightarrow | box | JJ | \Rightarrow | fast |
| \bar{N} | \Rightarrow | NN | \bar{N} | NN | \Rightarrow | car | JJ | \Rightarrow | metal |
| \bar{N} | \Rightarrow | JJ | \bar{N} | NN | \Rightarrow | mechanic | JJ | \Rightarrow | idealistic |
| \bar{N} | \Rightarrow | \bar{N} | \bar{N} | NN | \Rightarrow | pigeon | | | |
| NP | \Rightarrow | DT | \bar{N} | | | | IN | \Rightarrow | in |
| PP | \Rightarrow | IN | NP | DT | \Rightarrow | the | IN | \Rightarrow | under |
| \bar{N} | \Rightarrow | \bar{N} | PP | DT | \Rightarrow | a | IN | \Rightarrow | of |
| | | | | | | | IN | \Rightarrow | on |
| | | | | | | | IN | \Rightarrow | with |
| | | | | | | | IN | \Rightarrow | as |

Generates:

in a box, under the box, the fast car mechanic under the pigeon in the box, . . .

Verbs, Verb Phrases, and Sentences

- Basic Verb Types

Vi = Intransitive verb e.g., sleeps, walks, laughs

Vt = Transitive verb e.g., sees, saw, likes

Vd = Ditransitive verb e.g., gave

- Basic VP Rules

VP \rightarrow Vi

VP \rightarrow Vt NP

VP \rightarrow Vd NP NP

- Basic S Rule

S \rightarrow NP VP

Examples of VP:

sleeps, walks, likes the mechanic, gave the mechanic the fast car,
gave the fast car mechanic the pigeon in the box, . . .

Examples of S:

the man sleeps, the dog walks, the dog likes the mechanic, the dog
in the box gave the mechanic the fast car,. . .

PPs Modifying Verb Phrases

A new rule:

VP \rightarrow VP PP

New examples of VP:

sleeps in the car, walks like the mechanic, gave the mechanic the fast car on Tuesday, . . .

Complementizers, and SBARs

- Complementizers

COMP = complementizer e.g., that

- SBAR

SBAR → COMP S

Examples:

that the man sleeps, that the mechanic saw the dog . . .

More Verbs

- New Verb Types

V[5] e.g., said, reported

V[6] e.g., told, informed

V[7] e.g., bet

- New VP Rules

VP → V[5] SBAR

VP → V[6] NP SBAR

VP → V[7] NP NP SBAR

Examples of New VPs:

said that the man sleeps

told the dog that the mechanic likes the pigeon

bet the pigeon \$50 that the mechanic owns a fast car

Coordination

- A New Part-of-Speech:

CC = Coordinator e.g., and, or, but

- New Rules

NP → NP CC NP

\bar{N} → \bar{N} CC \bar{N}

VP → VP CC VP

S → S CC S

SBAR → SBAR CC SBAR

Sources of Ambiguity

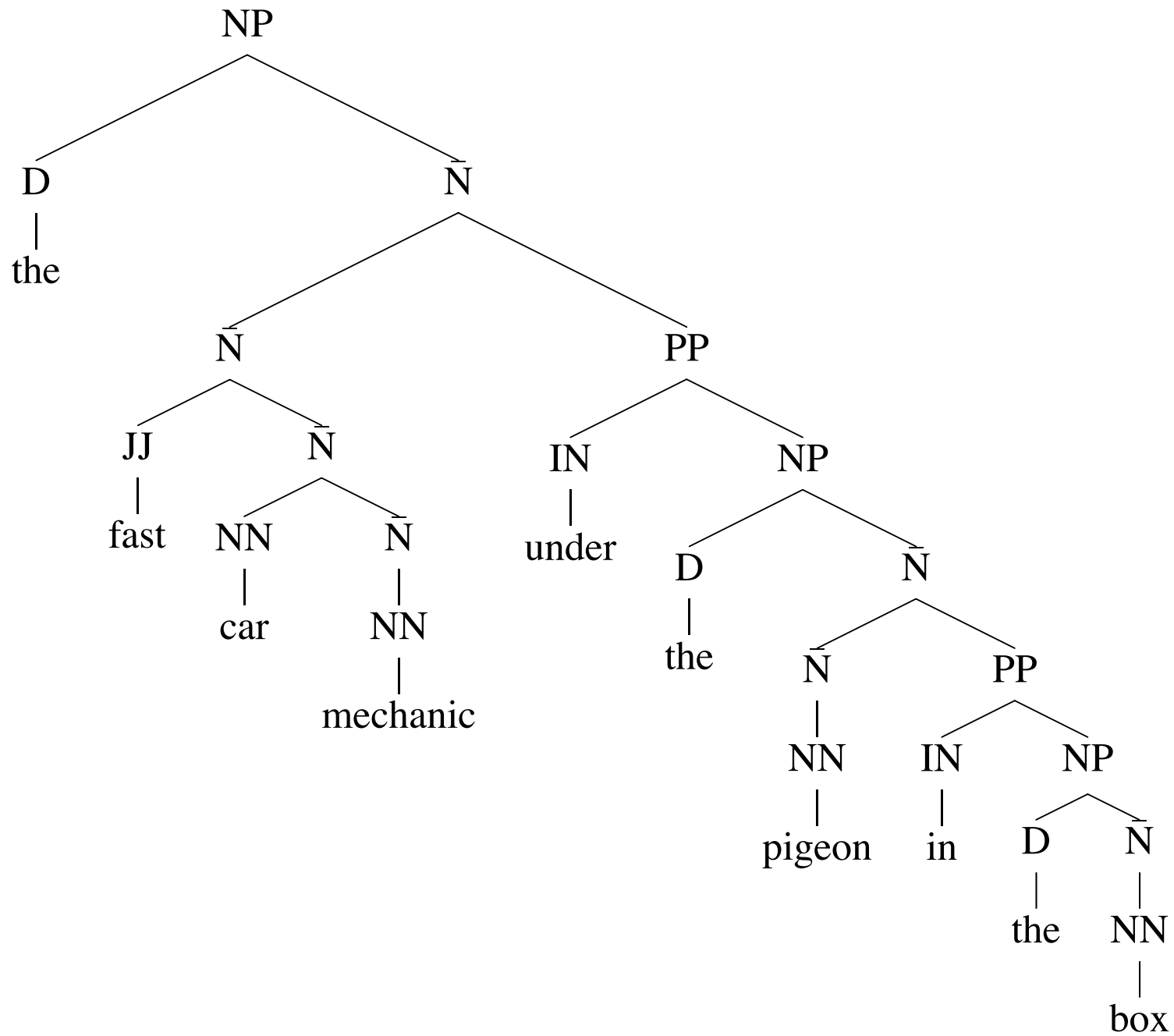
- Part-of-Speech ambiguity

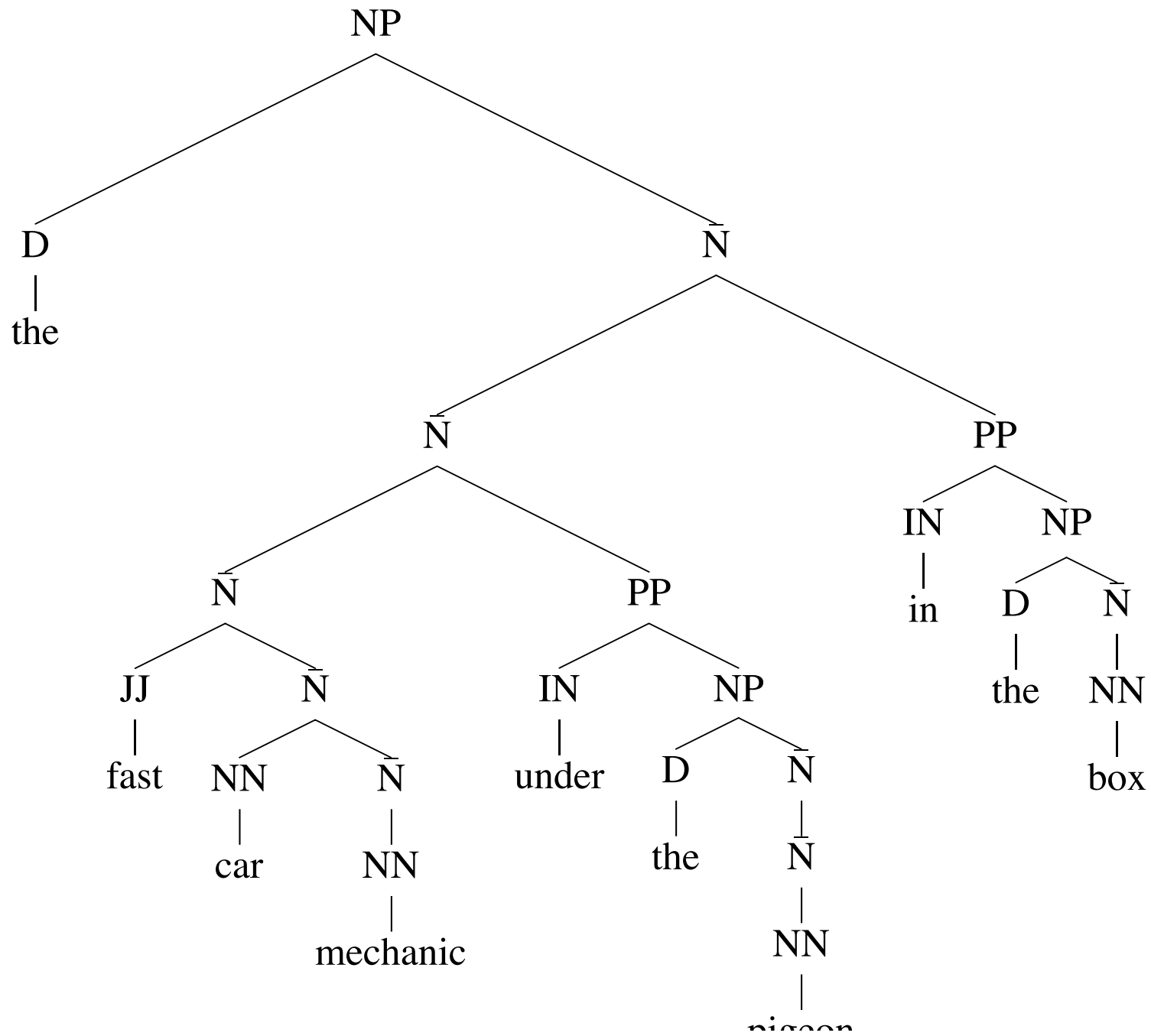
NNS → walks

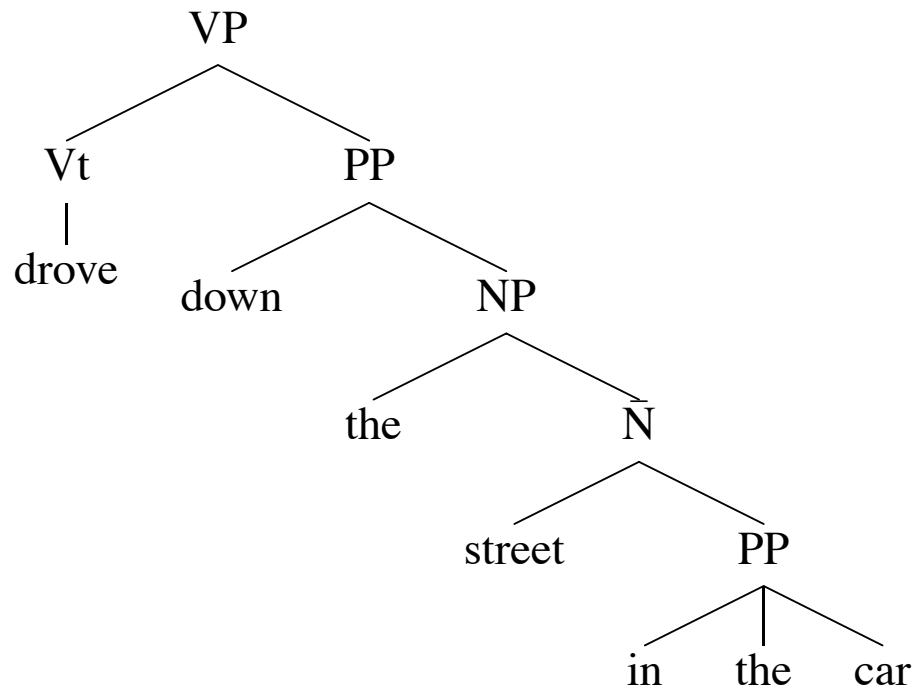
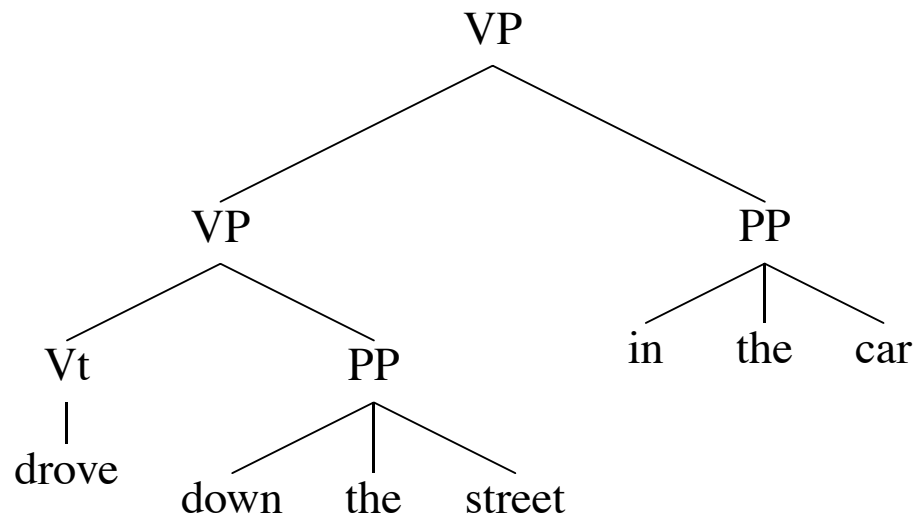
Vi → walks

- Prepositional Phrase Attachment

the fast car mechanic under the pigeon in the box



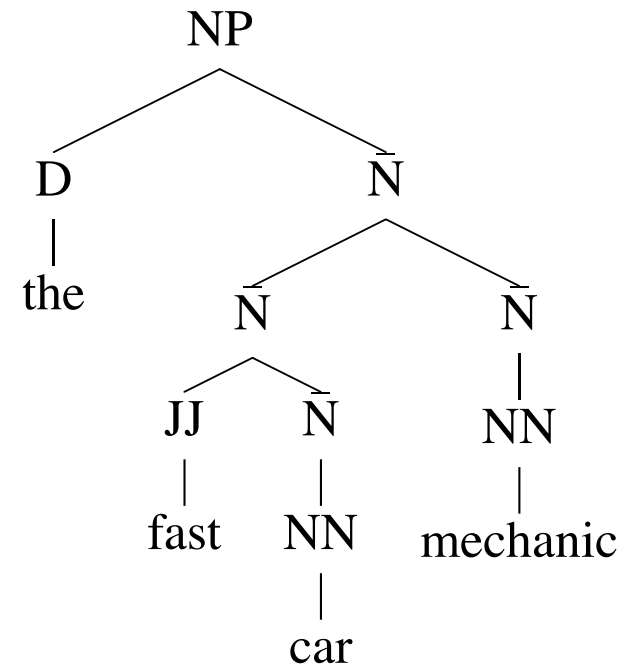
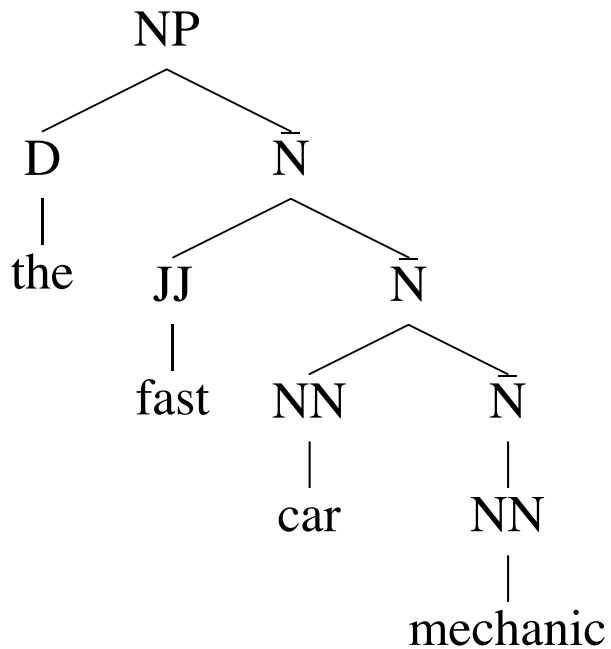




Two analyses for: John was believed to have been shot by Bill

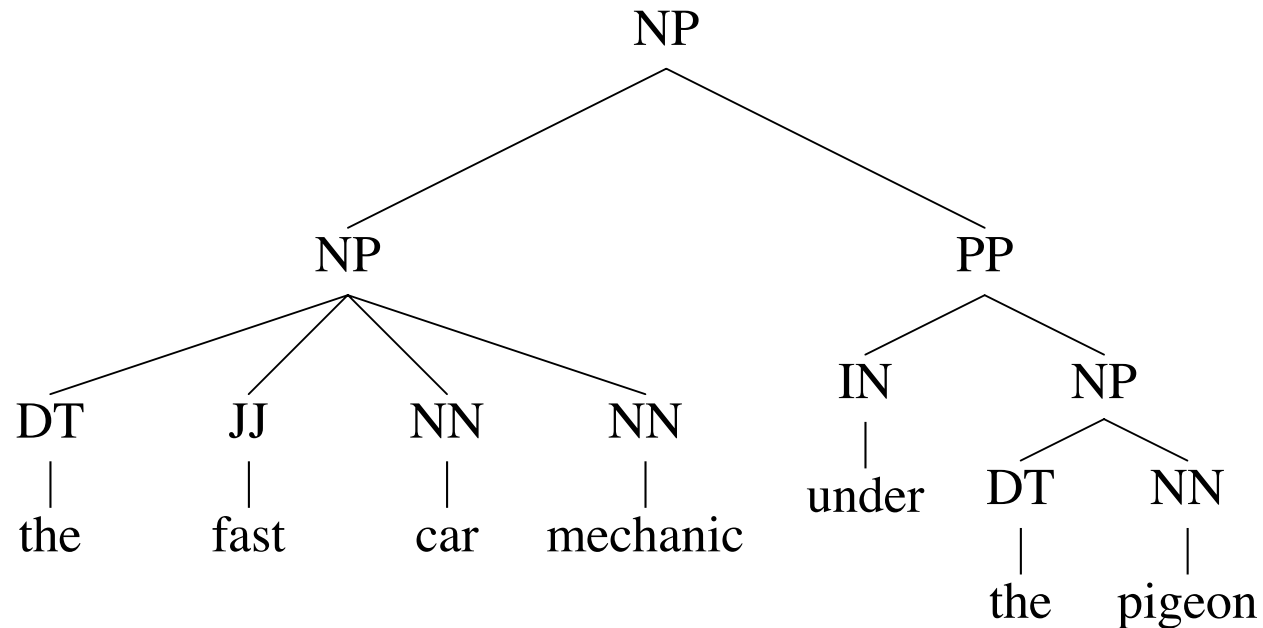
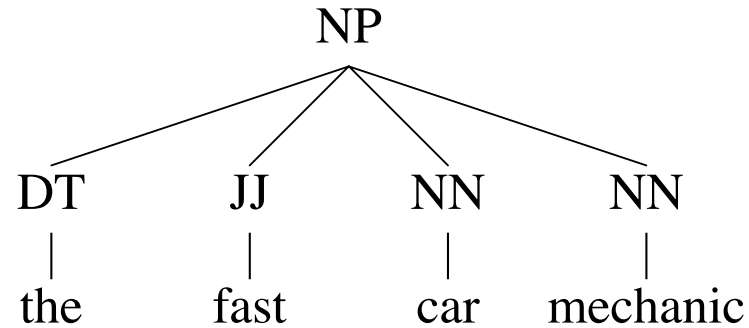
Sources of Ambiguity: Noun Premodifiers

- Noun premodifiers:



A Funny Thing about the Penn Treebank

Leaves NP premodifier structure flat, or underspecified:



A Probabilistic Context-Free Grammar

| | | | | |
|----|---------------|----|----|-----|
| S | \Rightarrow | NP | VP | 1.0 |
| VP | \Rightarrow | Vi | | 0.4 |
| VP | \Rightarrow | Vt | NP | 0.4 |
| VP | \Rightarrow | VP | PP | 0.2 |
| NP | \Rightarrow | DT | NN | 0.3 |
| NP | \Rightarrow | NP | PP | 0.7 |
| PP | \Rightarrow | P | NP | 1.0 |

| | | | |
|----|---------------|-----------|-----|
| Vi | \Rightarrow | sleeps | 1.0 |
| Vt | \Rightarrow | saw | 1.0 |
| NN | \Rightarrow | man | 0.7 |
| NN | \Rightarrow | woman | 0.2 |
| NN | \Rightarrow | telescope | 0.1 |
| DT | \Rightarrow | the | 1.0 |
| IN | \Rightarrow | with | 0.5 |
| IN | \Rightarrow | in | 0.5 |

- Probability of a tree with rules $\alpha_i \rightarrow \beta_i$ is $\prod_i P(\alpha_i \rightarrow \beta_i | \alpha_i)$

DERIVATION

RULES USED

PROBABILITY

S

DERIVATION

S

NP VP

RULES USED

$S \rightarrow NP VP$

PROBABILITY

1.0

DERIVATION

S

NP VP

DT N VP

RULES USED

$S \rightarrow NP VP$

$NP \rightarrow DT N$

PROBABILITY

1.0

0.3

DERIVATION

S

NP VP

DT N VP

the N VP

RULES USED

$S \rightarrow NP VP$

$NP \rightarrow DT N$

$DT \rightarrow \text{the}$

PROBABILITY

1.0

0.3

1.0

DERIVATION

S

NP VP

DT N VP

the N VP

the dog VP

RULES USED

$S \rightarrow NP VP$

$NP \rightarrow DT N$

$DT \rightarrow \text{the}$

$N \rightarrow \text{dog}$

PROBABILITY

1.0

0.3

1.0

0.1

DERIVATION

S

NP VP

DT N VP

the N VP

the dog VP

the dog VB

RULES USED

$S \rightarrow NP VP$

$NP \rightarrow DT N$

$DT \rightarrow \text{the}$

$N \rightarrow \text{dog}$

$VP \rightarrow VB$

PROBABILITY

1.0

0.3

1.0

0.1

0.4

| DERIVATION | RULES USED | PROBABILITY |
|----------------|-------------------------|-------------|
| S | $S \rightarrow NP VP$ | 1.0 |
| NP VP | $NP \rightarrow DT N$ | 0.3 |
| DT N VP | $DT \rightarrow the$ | 1.0 |
| the N VP | $N \rightarrow dog$ | 0.1 |
| the dog VP | $VP \rightarrow VB$ | 0.4 |
| the dog VB | $VB \rightarrow laughs$ | 0.5 |
| the dog laughs | | |

$$\text{TOTAL PROBABILITY} = 1.0 \times 0.3 \times 1.0 \times 0.1 \times 0.4 \times 0.5$$