

# Speech and Language Processing

## Chapter 12 Constituency Parsing

### Today

- Parsing with CFGs
  - Bottom-up, top-down
  - Ambiguity
  - CKY parsing
  - (Earley)
  - Shallow

## Parsing

- Parsing with CFGs refers to the task of assigning proper trees to input strings
- Proper here means a tree that covers **all and only the elements of the input** and **has an S at the top**
- It doesn't actually mean that the system can select the correct tree from among all the possible trees

9/25/2019

Speech and Language Processing - Jurafsky and Martin

3

## Parsing

- As with everything of interest, parsing involves a **search** which involves the making of choices
- We'll start with some basic (meaning bad) methods before moving on to the one that you need to know

9/25/2019

Speech and Language Processing - Jurafsky and Martin

4

## For Now

- **Assume...**
  - You have all the words already in some buffer
  - The input isn't POS tagged
  - We won't worry about morphological analysis
  - All the words are known
- These are all problematic in various ways, and would have to be addressed in real applications.

9/25/2019

Speech and Language Processing - Jurafsky and Martin

5

## Top-Down Search

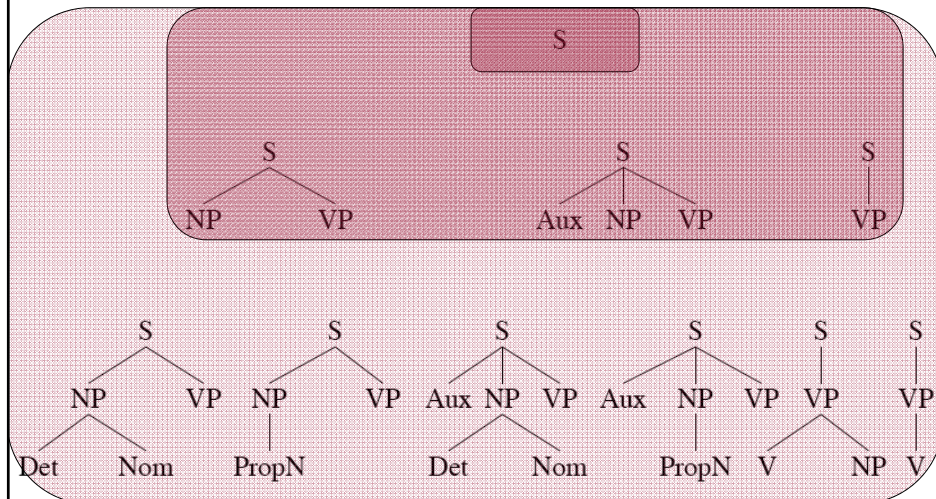
- Since we're trying to find trees rooted with an  $S$  (Sentences), why not start with the rules that give us an  $S$ .
- Then we can work our way down from there to the words.

9/25/2019

Speech and Language Processing - Jurafsky and Martin

6

## Top Down Space



9/25/2019

Speech and Language Processing - Jurafsky and Martin

7

## Bottom-Up Parsing

- Of course, we also want trees that cover the input words. So we might also start with trees that link up with the words in the right way.
- Then work your way up from there to larger and larger trees.

9/25/2019

Speech and Language Processing - Jurafsky and Martin

8

## Bottom-Up Search

Book that flight

9/25/2019

Speech and Language Processing - Jurafsky and Martin

9

## Bottom-Up Search

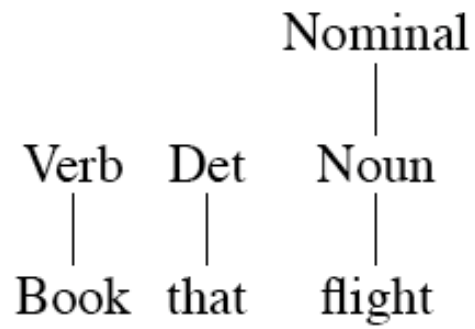
Verb	Det	Noun
Book	that	flight

9/25/2019

Speech and Language Processing - Jurafsky and Martin

10

# Bottom-Up Search

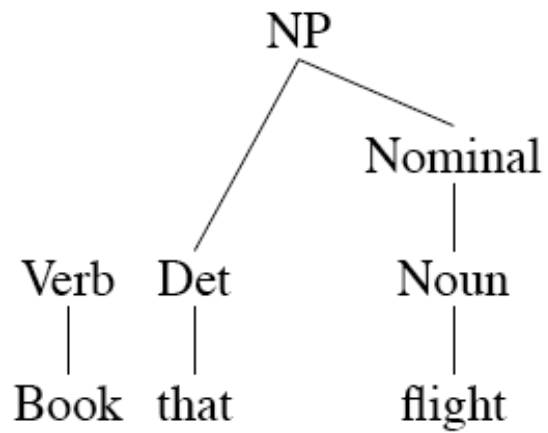


9/25/2019

Speech and Language Processing - Jurafsky and Martin

11

# Bottom-Up Search

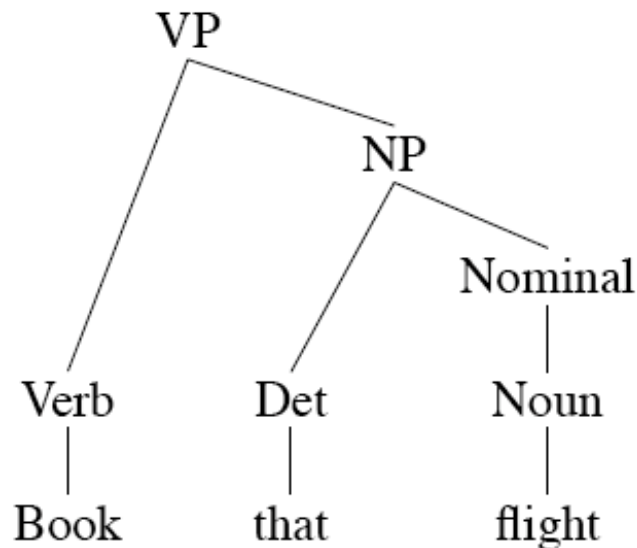


9/25/2019

Speech and Language Processing - Jurafsky and Martin

12

## Bottom-Up Search



9/25/2019

Speech and Language Processing - Jurafsky and Martin

13

## Top-Down and Bottom-Up

- **Top-down**
  - Only searches for trees that can be answers (i.e. S's)
  - But also suggests trees that are not consistent with any of the words
- **Bottom-up**
  - Only forms trees consistent with the words
  - But suggests trees that make no sense globally

9/25/2019

Speech and Language Processing - Jurafsky and Martin

14

## Control

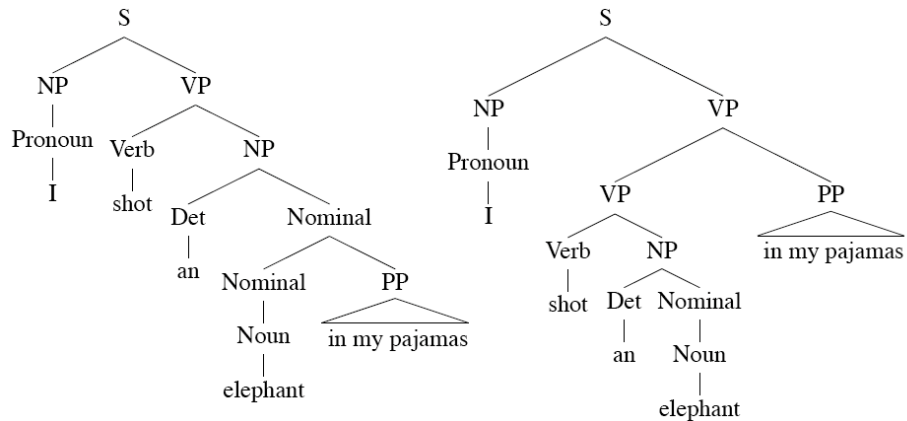
- Of course, in both cases we left out how to keep track of the search space and how to make choices
  - Which node to try to expand next
  - Which grammar rule to use to expand a node
- One approach is called backtracking.
  - Make a choice, if it works out then fine
  - If not then back up and make a different choice

## Problems

- Even with the best filtering, backtracking methods are doomed because of two inter-related problems
  - Ambiguity
  - Shared subproblems



# Ambiguity



9/25/2019

Speech and Language Processing - Jurafsky and Martin

17

## Example types of ambiguity

- POS
- Attachment
  - PP
  - Coordination (*old dogs and cats*)

9/25/2019

Speech and Language Processing - Jurafsky and Martin

18

## Shared Sub-Problems

- No matter what kind of search (top-down or bottom-up or mixed) that we choose.
  - We don't want to redo work we've already done.
  - Unfortunately, naïve backtracking will lead to duplicated work.

## Review

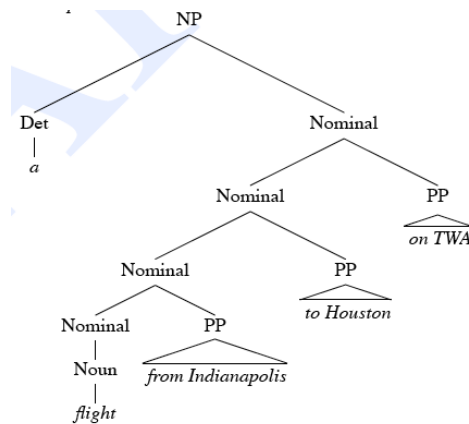
- Formal Grammars
  - CFG – what, why, why not?
  - Dependency
  - Treebanks
- Parsing with CFGs
  - Bottom-up, top-down
  - Ambiguity

## *“The old dog the footsteps of the young.”*

S → NP VP	VP → V
S → Aux NP VP	VP → V PP
S → VP	PP → Prep NP
NP → Det Nom	N → old   dog   footsteps   young
NP → PropN	V → dog   eat   sleep   bark   meow
Nom → Adj N	Aux → does   can
Nom → N	Prep → from   to   on   of
Nom → N Nom	PropN → Fido   Felix
Nom → Nom PP	Det → that   this   a   the
VP → V NP	Adj → old   happy   young

## Shared Sub-Problems

- Consider
  - A flight from Indianapolis to Houston on TWA



## Shared Sub-Problems

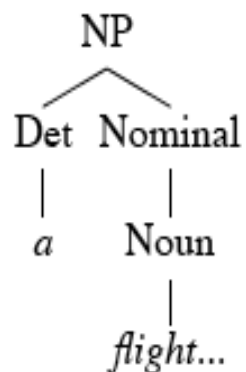
- Assume a top-down parse making choices among the various Nominal rules.
- In particular, between these two
  - Nominal -> Noun
  - Nominal -> Nominal PP
- Statically choosing the rules in this order leads to the following bad results...

9/25/2019

Speech and Language Processing - Jurafsky and Martin

23

## Shared Sub-Problems

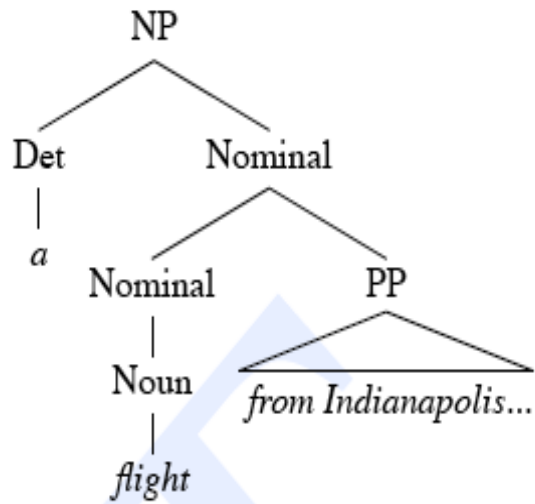


9/25/2019

Speech and Language Processing - Jurafsky and Martin

24

# Shared Sub-Problems

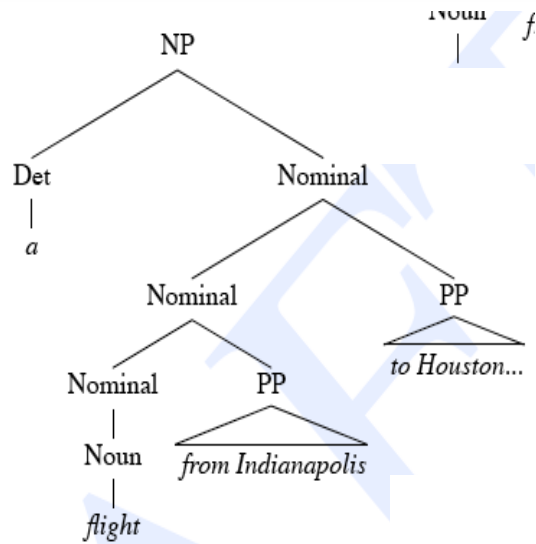


9/25/2019

Speech and Language Processing - Jurafsky and Martin

25

# Shared Sub-Problems

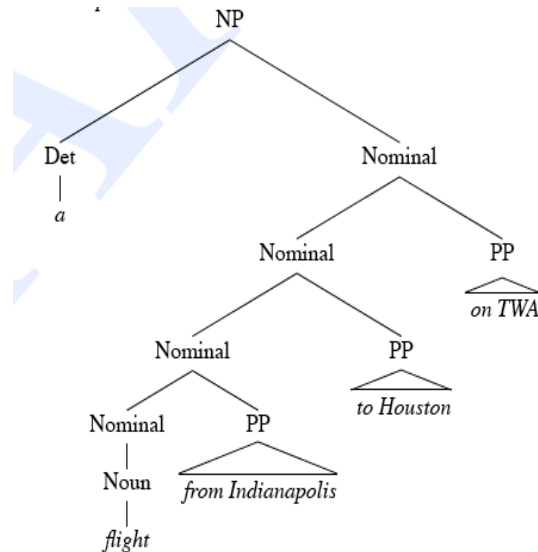


9/25/2019

Speech and Language Processing - Jurafsky and Martin

26

## Shared Sub-Problems



9/25/2019

Speech and Language Processing - Jurafsky and Martin

27

## Dynamic Programming

- DP search methods fill tables with partial results and thereby
  - Avoid doing avoidable repeated work
  - Solve exponential problems in polynomial time
  - Efficiently store ambiguous structures with shared sub-parts.
- Two approaches roughly correspond to bottom-up and top-down approaches.
  - CKY
  - Earley

9/25/2019

Speech and Language Processing - Jurafsky and Martin

28

## CKY Parsing

- First we'll limit our grammar to epsilon-free, binary rules (more later)
- Consider the rule  $A \rightarrow BC$ 
  - If there is an  $A$  somewhere in the input then there must be a  $B$  followed by a  $C$  in the input.
  - If the  $A$  spans from  $i$  to  $j$  in the input then there must be some  $k$  st.  $i < k < j$ 
    - Ie. The  $B$  splits from the  $C$  someplace.

9/25/2019

Speech and Language Processing - Jurafsky and Martin

29

## Problem

- What if your grammar isn't binary?
  - As in the case of the TreeBank grammar?
- Convert it to binary... any arbitrary CFG can be rewritten into Chomsky-Normal Form automatically.
- What does this mean?
  - The resulting grammar accepts (and rejects) the same set of strings as the original grammar.
  - **But** the resulting derivations (trees) are different.

9/25/2019

Speech and Language Processing - Jurafsky and Martin

30

## Problem

- More specifically, we want our rules to be of the form

$A \rightarrow B C$

Or

$A \rightarrow w$

*That is, rules can expand to either 2 non-terminals or to a single terminal.*

## Binarization Intuition

- Eliminate chains of unit productions.
- Introduce new intermediate non-terminals into the grammar that distribute rules with length  $> 2$  over several rules.

So...  $S \rightarrow A B C$  turns into

$S \rightarrow X C$  and

$X \rightarrow A B$

Where  $X$  is a symbol that doesn't occur anywhere else in the the grammar.



# Sample L1 Grammar

Grammar	Lexicon
$S \rightarrow NP VP$	$Det \rightarrow that \mid this \mid a$
$S \rightarrow Aux NP VP$	$Noun \rightarrow book \mid flight \mid meal \mid money$
$S \rightarrow VP$	$Verb \rightarrow book \mid include \mid prefer$
$NP \rightarrow Pronoun$	$Pronoun \rightarrow I \mid she \mid me$
$NP \rightarrow Proper-Noun$	$Proper-Noun \rightarrow Houston \mid NWA$
$NP \rightarrow Det Nominal$	$Aux \rightarrow does$
$Nominal \rightarrow Noun$	$Preposition \rightarrow from \mid to \mid on \mid near \mid through$
$Nominal \rightarrow Nominal Noun$	
$Nominal \rightarrow Nominal PP$	
$VP \rightarrow Verb$	
$VP \rightarrow Verb NP$	
$VP \rightarrow Verb NP PP$	
$VP \rightarrow Verb PP$	
$VP \rightarrow VP PP$	
$PP \rightarrow Preposition NP$	

9/25/2019

Speech and Language Processing - Jurafsky and Martin

33

# CNF Conversion

$\mathcal{L}_1$ Grammar	$\mathcal{L}_1$ in CNF
$S \rightarrow NP VP$	$S \rightarrow NP VP$
$S \rightarrow Aux NP VP$	$S \rightarrow XI VP$
	$XI \rightarrow Aux NP$
$S \rightarrow VP$	$S \rightarrow book \mid include \mid prefer$
	$S \rightarrow Verb NP$
	$S \rightarrow X2 PP$
	$S \rightarrow Verb PP$
	$S \rightarrow VP PP$
$NP \rightarrow Pronoun$	$NP \rightarrow I \mid she \mid me$
$NP \rightarrow Proper-Noun$	$NP \rightarrow TWA \mid Houston$
$NP \rightarrow Det Nominal$	$NP \rightarrow Det Nominal$
$Nominal \rightarrow Noun$	$Nominal \rightarrow book \mid flight \mid meal \mid money$
$Nominal \rightarrow Nominal Noun$	$Nominal \rightarrow Nominal Noun$
$Nominal \rightarrow Nominal PP$	$Nominal \rightarrow Nominal PP$
$VP \rightarrow Verb$	$VP \rightarrow book \mid include \mid prefer$
$VP \rightarrow Verb NP$	$VP \rightarrow Verb NP$
$VP \rightarrow Verb NP PP$	$VP \rightarrow X2 PP$
	$X2 \rightarrow Verb NP$
$VP \rightarrow Verb PP$	$VP \rightarrow Verb PP$
$VP \rightarrow VP PP$	$VP \rightarrow VP PP$
$PP \rightarrow Preposition NP$	$PP \rightarrow Preposition NP$

9/25/2019

Speech and Language Processing - Jurafsky and Martin

34

## CKY

- So let's build a table so that an  $A$  spanning from  $i$  to  $j$  in the input is placed in cell  $[i,j]$  in the table.
- So a non-terminal spanning an entire string will sit in cell  $[0, n]$ 
  - Hopefully an  $S$
- If we build the table bottom-up, we'll know that the parts of the  $A$  must go from  $i$  to  $k$  and from  $k$  to  $j$ , for some  $k$ .

9/25/2019

Speech and Language Processing - Jurafsky and Martin

35

## CKY

- Meaning that for a rule like  $A \rightarrow B C$  we should look for a  $B$  in  $[i,k]$  and a  $C$  in  $[k,j]$ .
- In other words, if we think there might be an  $A$  spanning  $i,j$  in the input... AND  $A \rightarrow B C$  is a rule in the grammar THEN
- There must be a  $B$  in  $[i,k]$  and a  $C$  in  $[k,j]$  for some  $i < k < j$

9/25/2019

Speech and Language Processing - Jurafsky and Martin

36

## CKY

- So to fill the table loop over the cell  $[i,j]$  values in some systematic way
  - What constraint should we put on that systematic search?
  - For each cell, loop over the appropriate  $k$  values to search for things to add.

9/25/2019

Speech and Language Processing - Jurafsky and Martin

37

## Note

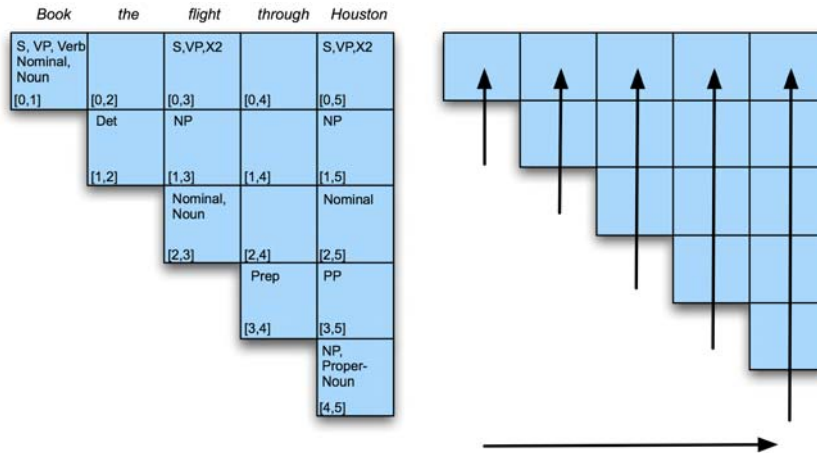
- We arranged the loops to fill the table a column at a time, from left to right, bottom to top.
  - This assures us that whenever we're filling a cell, the parts needed to fill it are already in the table (to the left and below)
  - It's somewhat natural in that it processes the input a left to right a word at a time
    - Known as online

9/25/2019

Speech and Language Processing - Jurafsky and Martin

38

# Example

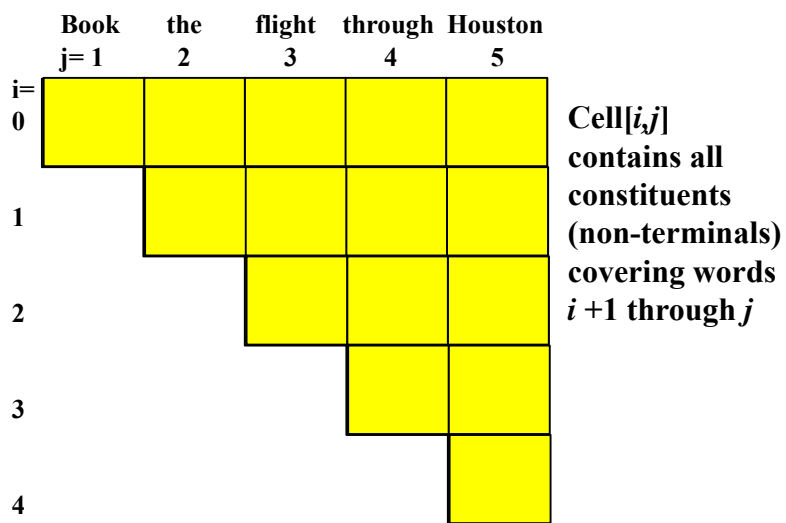


9/25/2019

Speech and Language Processing - Jurafsky and Martin

39

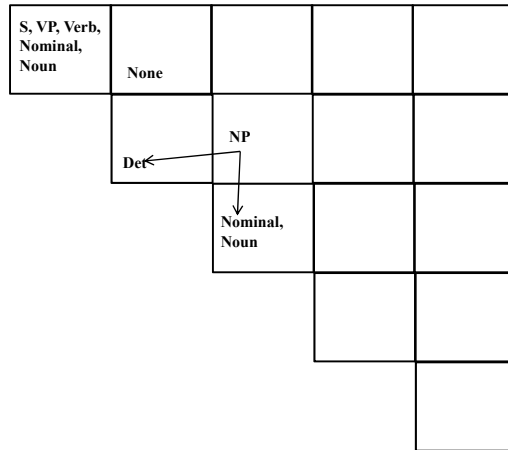
# CKY Parser



40

# CKY Parser

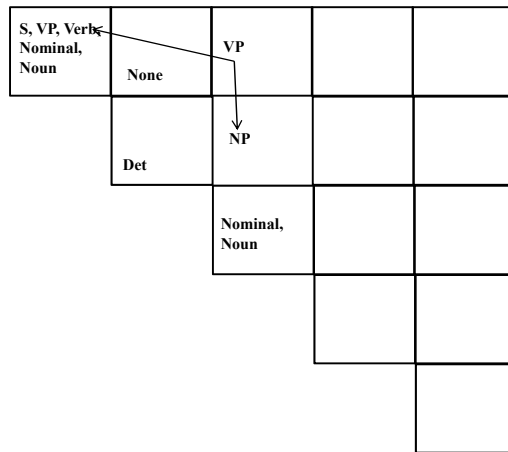
Book the flight through Houston



41

# CKY Parser

Book the flight through Houston



42

# CKY Parser

Book the flight through Houston

S, VP, Verb, Nominal, Noun	None	S VP		
	Det	NP		
		Nominal, Noun		

43

# CKY Parser

Book the flight through Houston

S, VP, Verb, Nominal, Noun	None	S VP, X2		
	Det	NP		
		Nominal, Noun		

44

# CKY Parser

Book the flight through Houston

S, VP, Verb, Nominal, Noun	None	S VP	None	
	Det	NP	None	
		Nominal, Noun	None	
			Prep	

45

# CKY Parser

Book the flight through Houston

S, VP, Verb, Nominal, Noun	None	S VP	None	
	Det	NP	None	
		Nominal, Noun	None	
			Prep ← pp	
				NP ProperNoun

46

# CKY Parser

Book the flight through Houston

S, VP, Verb, Nominal, Noun	None	S VP	None	
	Det	NP	None	
		Nominal Noun	None	Nominal
			Prep	PP
				NP ProperNoun

47

# CKY Parser

Book the flight through Houston

S, VP, Verb, Nominal, Noun	None	S VP	None	
	Det	NP	None	NP
		Nominal, Noun	None	Nominal
			Prep	PP
				NP ProperNoun

48



# CKY Parser

Book the flight through Houston

S, VP, Verb, Nominal, Noun	None	S VP	None	VP
		NP	None	NP
	Det	Nominal, Noun	None	Nominal
			Prep	PP
				NP ProperNoun

49

# CKY Parser

Book the flight through Houston

S, VP, Verb, Nominal, Noun	None	S VP	None	S VP
		NP	None	NP
	Det	Nominal, Noun	None	Nominal
			Prep	PP
				NP ProperNoun

50

# CKY Parser

Book the flight through Houston

S, VP, Verb, Nominal, Noun	None	S VP←	None	YP \$ VP
	Det	NP	None	NP
		Nominal, Noun	None	Nominal
			Prep	↓ PP
				NP ProperNoun

51

# CKY Parser

Book the flight through Houston

S, VP, Verb, Nominal, Noun	None	S VP←	None	\$ VP S VP
	Det	NP	None	NP
		Nominal, Noun	None	Nominal
			Prep	↓ PP
				NP ProperNoun

52

# CKY Parser

Book the flight through Houston

S, VP, Verb, Nominal, Noun	None	S VP ←	None	S VP S X2 VP S
	Det	NP	None	NP
		Nominal, Noun	None	Nominal
			Prep	PP
				NP ProperNoun

53

# CKY Parser

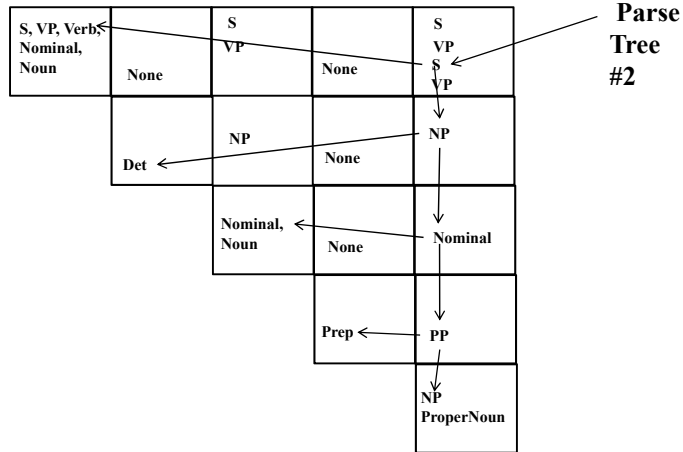
Book the flight through Houston

S, VP, Verb, Nominal, Noun	None	S VP ←	None	S ← VP S VP	Parse Tree #1
	Det	NP	None	NP	
		Nominal, Noun	None	Nominal	
			Prep ←	PP	
				NP ProperNoun	

54

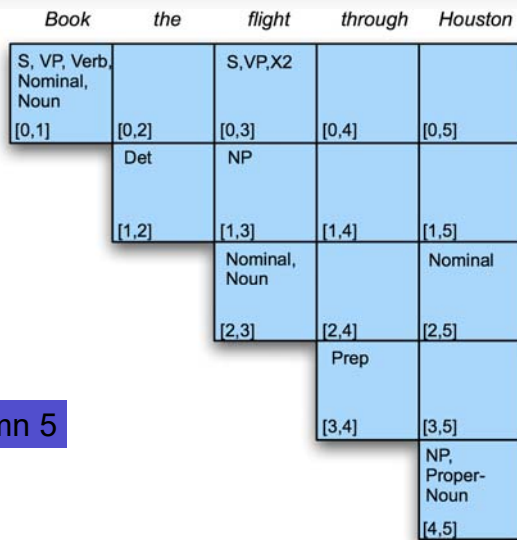
# CKY Parser

Book the flight through Houston



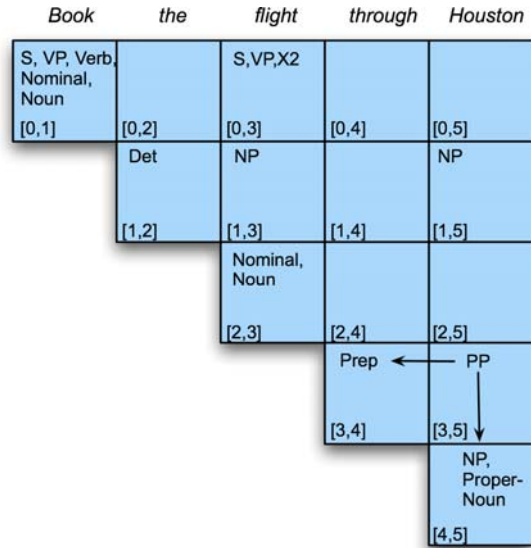
55

# Example



Filling column 5

# Example

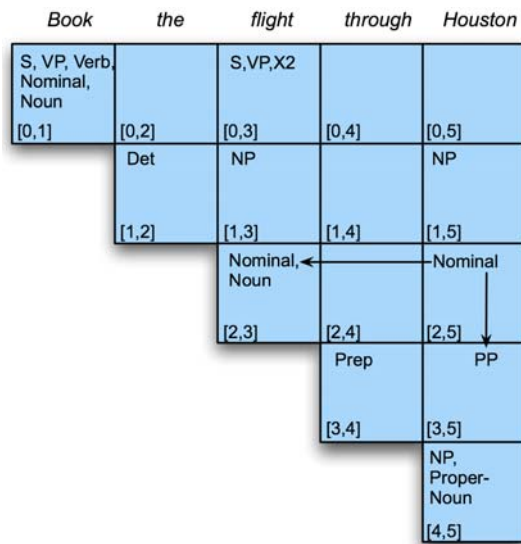


9/25/2019

Speech and Language Processing - Jurafsky and Martin

57

# Example

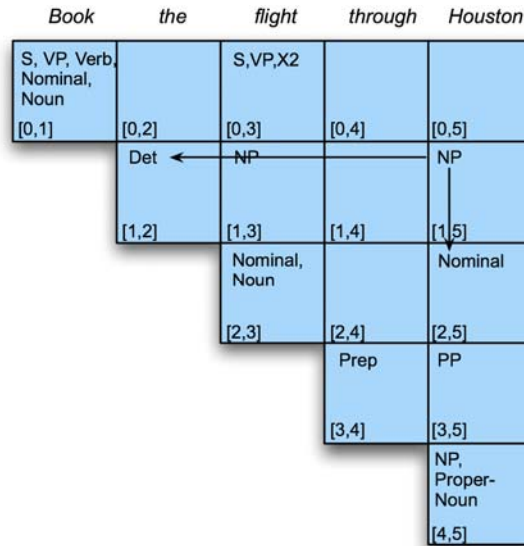


9/25/2019

Speech and Language Processing - Jurafsky and Martin

58

# Example

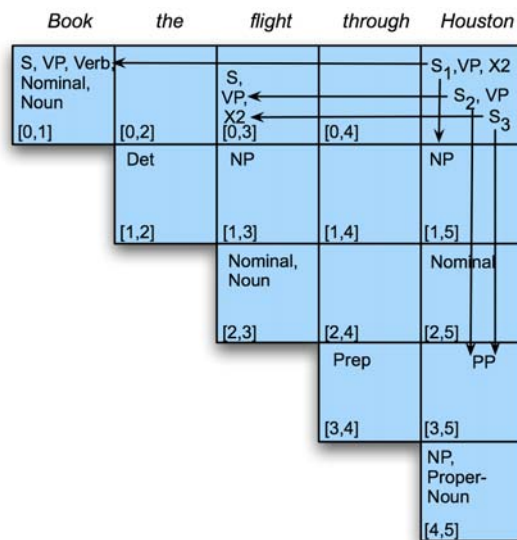


9/25/2019

Speech and Language Processing - Jurafsky and Martin

59

# Example



9/25/2019

Speech and Language Processing - Jurafsky and Martin

60

## CKY Notes

- Since it's bottom up, CKY populates the table with a lot of phantom constituents.
  - Segments that by themselves are constituents but cannot really occur in the context in which they are being suggested.
  - To avoid this we can switch to a top-down control strategy
  - Or we can add some kind of filtering that blocks constituents where they can not happen in a final analysis.

9/25/2019

Speech and Language Processing - Jurafsky and Martin

61

## Earley Parsing

- Allows arbitrary CFGs
- Top-down control
- Fills a table in a single sweep over the input
  - Table is length  $N+1$ ;  $N$  is number of words
  - Table entries represent
    - Completed constituents and their locations
    - In-progress constituents
    - Predicted constituents

9/25/2019

Speech and Language Processing - Jurafsky and Martin

62

## Back to Ambiguity

- Did we solve it?

## Ambiguity

- No...
  - Both CKY and Earley will result in multiple **S** structures for the **[0,N]** table entry.
  - They both efficiently store the sub-parts that are shared between multiple parses.
  - And they obviously avoid re-deriving those sub-parts.
  - But neither can tell us which one is right.



## Ambiguity

- In most cases, humans don't notice incidental ambiguity (lexical or syntactic). It is resolved on the fly and never noticed.
  - I ate the spaghetti with chopsticks
  - I ate the spaghetti with meatballs
- We'll try to model that with probabilities.

9/25/2019

Speech and Language Processing - Jurafsky and Martin

65

## Shallow or Partial Parsing

- Sometimes we don't need a complete parse tree
  - Information extraction
  - Question answering
- But we would like more than simple POS sequences

66

## Chunking

- Find major but unembedded constituents like NPs, VPs, AdjPs, PPs
  - Most common task: NP chunking of base NPs
  - [NP I] saw [NP the man] on [NP the hill] with [NP a telescope]
  - No attempt to identify full NPs – no recursion, no post-head words
  - No overlapping constituents
  - E.g., if we add PPs or VPs, they may consist only of their heads, e.g. [PP on]

## Approaches: RE Chunking

- Use regexps to identify constituents, e.g.
  - NP → (DT) NN\* NN
  - Find longest matching chunk
  - Hand-built rules
  - No recursion but can cascade to approximate true CF parser, aggregating larger and larger constituents

## Approaches: Tagging for Chunking

- Require annotated corpus
  - Train classifier to classify each element of input in sequence (e.g. IOB Tagging)
    - B (beginning of sequence)
    - I (internal to sequence)
    - O (outside of any sequence)
    - No end-of-chunk coding – it's implicit
    - Easier to detect the beginning than the end
- Book/B\_VP that/B\_NP flight/I\_NP quickly/O

## Summary and Limitations

- Sometimes shallow parsing is enough for task
- Performance quite accurate

## Distribution of Chunks in CONLL Shared Task

Label	Category	Proportion (%)	Example
<i>NP</i>	Noun Phrase	51	<i>The most frequently cancelled flight</i>
<i>VP</i>	Verb Phrase	20	<i>may not arrive</i>
<i>PP</i>	Prepositional Phrase	20	<i>to Houston</i>
<i>ADVP</i>	Adverbial Phrase	4	<i>earlier</i>
<i>SBAR</i>	Subordinate Clause	2	<i>that</i>
<i>ADJP</i>	Adjective Phrase	2	<i>late</i>

## Summing Up

- Parsing as search: what search strategies to use?
  - Top down
  - Bottom up
  - How to combine?
- How to parse as little as possible
  - Dynamic Programming
- Shallow Parsing