

CS114 Lecture 10

Parsing and Partial Parsing

Thanks for Jurafsky & Martin & Prof. Pustejovksy for slides

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
- Partial parsing returns a set of constituents or small trees
 - Does not require a single S at the top

3/3/14 Speech and Language

Top-Down and Bottom-Up

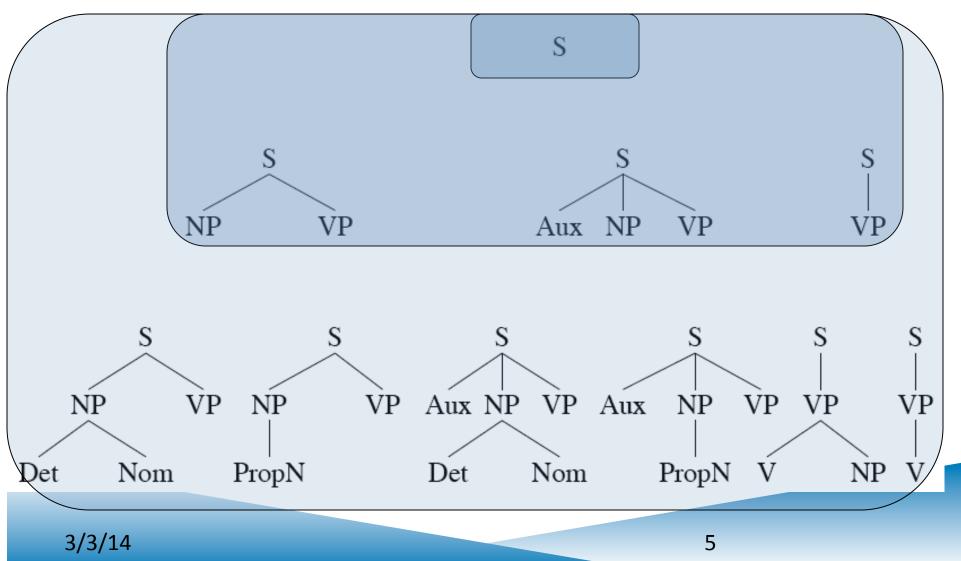
- Top-down
 - Only finds trees that form sentences
 - Finds trees that don't match the words
- Bottom-up
 - Only finds trees consistent with the words
 - Builds subtrees that can't combine to make sentences

Top-Down Search

- Sentences: trees rooted with S
- Start with S, expand nodes working downwards
- Try to reach the right set of words

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Top Down Space



Speech and Language
Processing - Jurafsky and

Bottom-Up Parsing

- Input must be matched exactly very hard
- Start by combining words into small trees
- Work your way up, combining small trees into larger trees
- Try to get a single tree rooted at S

Bottom-Up Search

Grammar:

S ♦ NP VP

S♦ Aux NP VP

S ♦ VP

NP ♦ **Det Nom**

Nom \Diamond Noun

Nom \Diamond Noun Nom

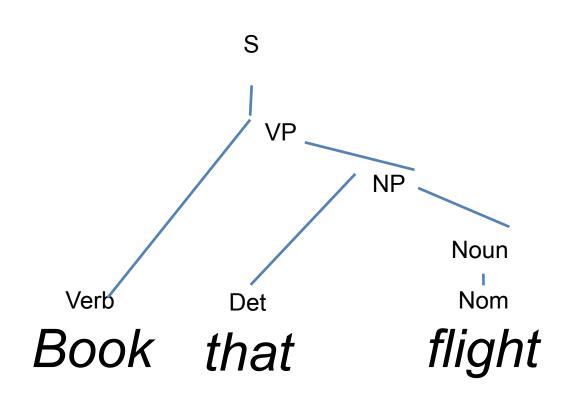
Nom \Diamond Nom PP

NP ♦ **Proper-Noun**

VP ♦ **Verb**

VP ♦ **Verb NP**

PP ♦ Prep NP



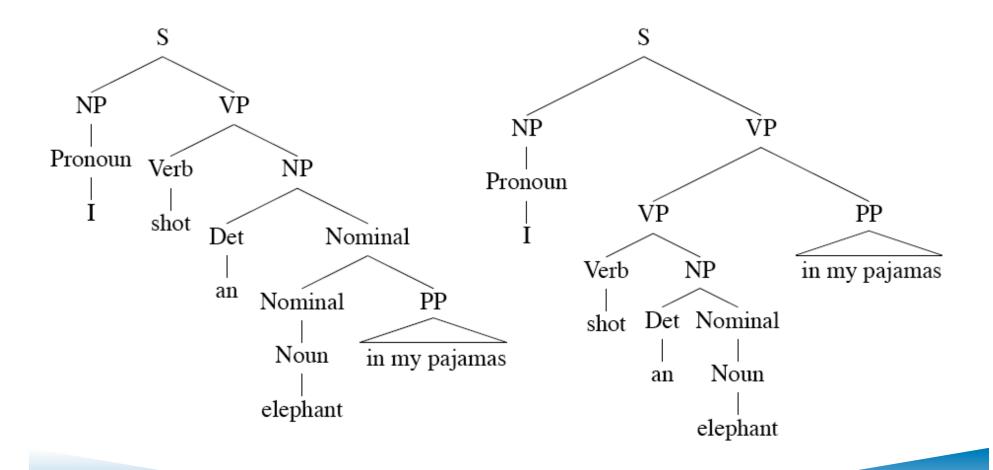
Control

- How to explore the search space?
 - Which node to try to expand next
 - Which grammar rule to use to expand a node
 - Wrong choice leads to unsolvable problem
- 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

- Backtracking is not good enough
 - Ambiguity
 - Shared subproblems

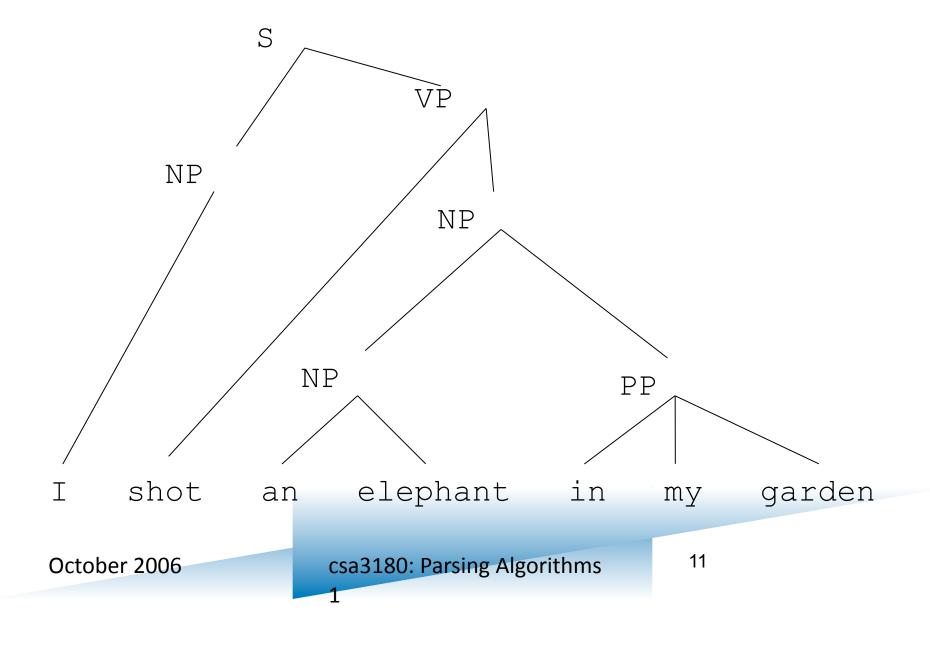
Ambiguity



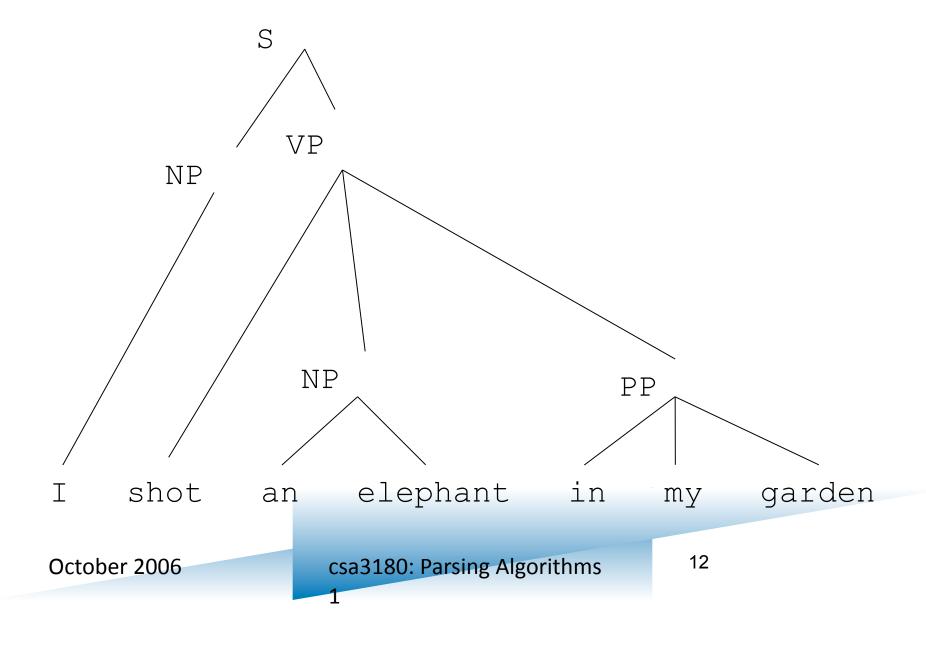
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The elephant is in the garden



I fired from the garden



Shared Sub-Problems

- No matter what kind of search (top-down or bottom-up or mixed) that we choose.
 - Redoing work is wasted effort
 - Naïve backtracking means duplicated work.
 - Dynamic Programming...

Parsing so far

- Goals
- Define a language
- A grammar defines a language L over token set (terminals)
- A parser determines whether a particular string of tokens is in the language
- Determine the constituent structure of a string of tokens (assuming it's in the language)
- Determine the meaning (we're not there yet)

Parsing so far

A Information

- Categories (nonterminals) of the tokens
- Which adjacent nonterminals form tokens (rules)
- What additional conditions must be met in order for a constituent to be formed (e.g. features, words)

Parsing Types

- Next:
 - Chunking
 - Partial parsing
- · We'll come back to Parsing as Search
 - CFGS
 - Top down, bottom up
 - Earley's algorithm
 - Probabalistic CFGs

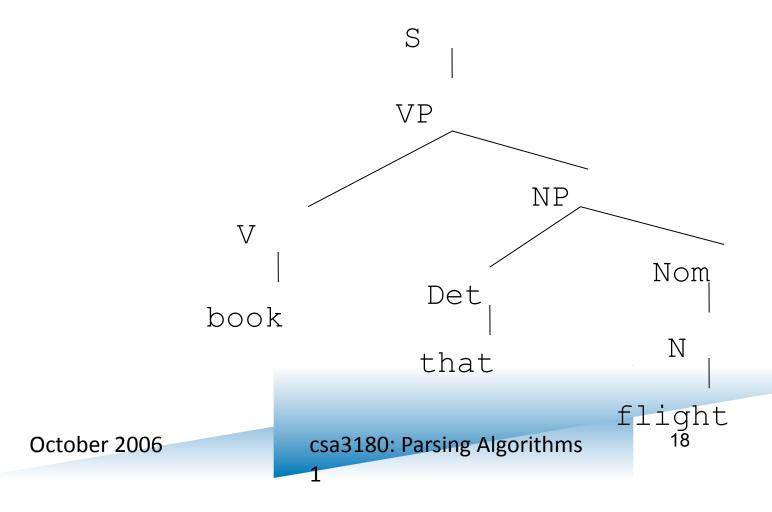
Unification Grammars

Motivation: Parsing is hard

- Example (wsj_0001):
 - Pierre Vinken, 61 years old, will join the board as a nonexecutive director Nov. 29.
 - Mr. Vinken is chairman of Elsevier N.V., the Dutch publishing group.
 - Pierre/NNP Vinken/NNP ,/, 61/CD years/NNS old/JJ ,/, will/MD join/VB the/DT board/NN as/IN a/DT nonexecutive/JJ director/NN Nov./NNP 29/CD ./.
 - Mr./NNP Vinken/NNP is/VBZ chairman/NN of/IN Elsevier/NNP N.V./NNP ,/, the/DT Dutch/NNP publishing/ VBG group/NN ./.

Parsing Problem

 Given grammar G and sentence A discover all valid parse trees for G that exactly cover A



Slide modified from Steven Bird's Shallow (Chunk) Parsing

Goal: divide a sentence into a sequence of chunks.

- Chunks are non-overlapping regions of a text [I] saw [a tall man] in [the park].
- Chunks are non-recursive
 - -Cannot contain other chunks
- Chunks are non-exhaustive
 - Not all words are included in chunks

Slide modified from Steven Bird's Chunk Parsing Examples

Noun-phrase chunking:

[I] saw [a tall man] in [the park].

Verb-phrase chunking:

The man who [was in the park] [saw me].

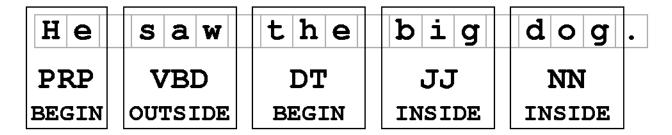
- Question answefing:
 - What [Spanish explorer] discovered [the Mississippi River]?

Shallow Parsing: Motivation

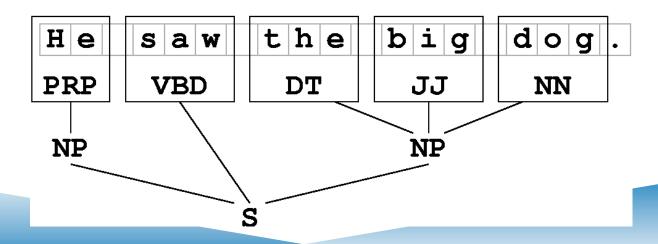
- Locating information
 - e.g. index a document collection on its noun phrases
- Ignoring information
 - Generalize in order to study higher-level patterns
 - e.g. phrases involving "gave" in Penn treebank:
 - gave NP; gave up NP in NP; gave NP up; gave NP help; gave
 NP to NP
 - Sometimes a full parse has too much structure

Slide modified from Steven Bird's Representation

BIO (or IOB)

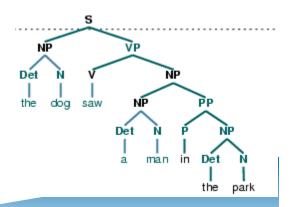


Trees



Comparison with Full Syntactic

- Parsing often an intermediate stage
 - later stages draw on the parse for their own purposes
- Full parsing sufficient, often not necessary
 - Often more information than we need
- . Shallow parsing is an easier problem
- Less structure, no recursion



Slide modified from Steven Bird's Chunks and Constituency

Constituents: [[a tall man] [in [the park]]].

Chunks: [a tall man] in [the park].

- A constituent is part of some higher unit in the hierarchical syntactic parse
- Chunks are not constituents
 - Constituents are recursive
 - Chunks do not cross major constituent boundaries (why?)

Slide modified from Steven Bird's Chunk Parsing in NLTK

- Chunk parsers usually ignore lexical content
 - Only need to look at part-of-speech tags
- Possible steps in chunk parsing
 - Chunking, unchunking
 - Chinking
 - Merging, splitting

Chunking

 Define a regular expression that matches the sequences of tags in a chunk

A simple noun phrase chunk regexp:

```
(Note that <NN.*> matches any tag starting with NN) <DT>? <II>* <NN.*>
```

Chunk all matching subsequences:

```
the/DT little/JJ cat/NN sat/VBD on/IN the/DT mat/NN
```

[the/DT little/JJ cat/NN] sat/VBD on/IN [the/DT mat/NN]

If matching subsequences overlap, first 1 gets priority

Unchunking

- Remove any chunk with a given pattern
 - e.g., unChunkRule('<NN|DT>+', 'Unchunk NNDT')
 - Combine with Chunk Rule < NN | DT | JJ>+
- · Chunk all matching subsequences:
 - Input:

the/DT little/JJ cat/NN sat/VBD on/IN the/DT mat/NN

Apply chunk rule

[the/DT little/JJ cat/NN] sat/VBD on/IN [the/DT mat/NN]

Apply unchunk rule

[the/DT little/JJ cat/NN] sat/VBD on/IN the/DT mat/NN

Chinking

- . A *chink* is a subsequence of the text that is not a chunk.
- Define a regular expression that matches the sequences of tags in a chink

A simple chink regexp for finding NP chunks: (<VB.?>|<IN>)+

- First apply chunk rule to chunk everything
 - Input: the/DT little/JJ cat/NN sat/VBD on/IN the/DT mat/NN
 - ChunkRule('<.*>+', 'Chunk everything')

[the/DT little/JJ cat/NN sat/VBD on/IN the/DT mat/NN]

Apply Chink rule above:

[the/DT little/JJ cat/NN] sat/VBD on/IN [the/DT mat/NN]

Chunk Chunk

Merging

- Combine adjacent chunks into a single chunk
 - Define a regular expression that matches the sequences of tags on both sides of the point to be merged
- Example:
 - Merge a chunk ending in JJ with a chunk starting with NN MergeRule('<JJ>', '<NN>', 'Merge adjs and nouns')

[the/DT little/JJ] [cat/NN] sat/VBD on/IN the/DT mat/NN [the/DT little/JJ cat/NN] sat/VBD on/IN the/DT mat/NN

Splitting is the opposite of merging

Cascaded Chunking

- Goal: create chunks that include other chunks
- Examples:
 - PP consists of preposition + NP
 - VP consists of verb followed by PPs or NPs

Slide modified from Steven Bird's Cascaded Chunking

Next Assignment

- Use the nltk chunker and write chunking rules and apply it to the Treebank data.
- Evaluate your performance on the "tagged" treebank data (which is already chunked).
- There are 200 sentences. Develop the grammar using the first 150 sentences and test on the last 50 sentences.

Example

- Write rules using POS to produce the chunks
- Section 7.3, NLTK book
- Evaluate performance
- Definitely > 70% fmeasure, shoot for >90%

```
[ A/DT form/NN ]

of/IN

[ asbestos/NN ]

once/RB used/VBN to/TO make/VB

[ Kent/NNP ]

[ cigarette/NN filters/NNS ]

has/VBZ caused/VBN

[ a/DT high/JJ percentage/NN ]

of/IN
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[cancer/NN deaths/NNS]

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