

# Structural Ambiguity

How different parse trees may be produced from the same sentence or phrase

- ▶ Attachment ambiguity: where a constituent may attach to the rest of the tree
  - ▶ I saw a man with a telescope
- ▶ Coordination ambiguity: How to group the arguments of a conjunction
  - ▶ Spicy rice and apples
- ▶ *Disambiguation* relies on applying additional knowledge
  - ▶ Of language, e.g., what verbs and nouns or prepositions go together
  - ▶ Of the real world
  - ▶ Of the context, such as prior sentences or conversations

# Jurafsky's Miniature Grammar, $\mathcal{L}_1$

## Omitting the lexicon

S  $\rightarrow$  NP VP

S  $\rightarrow$  Auxiliary-Verb NP VP

S  $\rightarrow$  VP

NP  $\rightarrow$  Pronoun

NP  $\rightarrow$  Proper-Noun

NP  $\rightarrow$  Determiner Nominal

Nominal  $\rightarrow$  Noun

Nominal  $\rightarrow$  Nominal Noun

Nominal  $\rightarrow$  Nominal PP

VP  $\rightarrow$  Verb

VP  $\rightarrow$  Verb NP

VP  $\rightarrow$  Verb NP PP

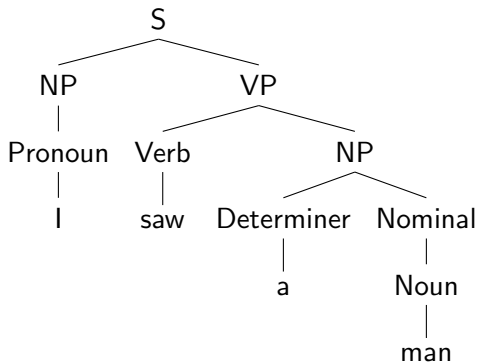
VP  $\rightarrow$  Verb PP

VP  $\rightarrow$  VP PP

PP  $\rightarrow$  Preposition NP

# Attachment Ambiguity: Setting the Stage

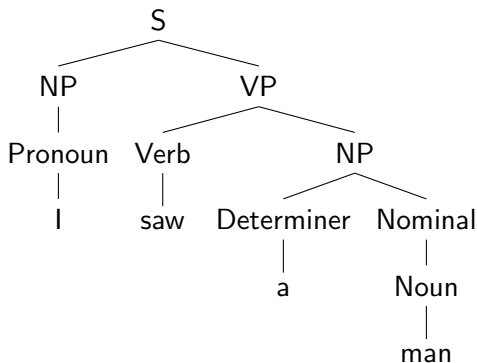
I saw a man



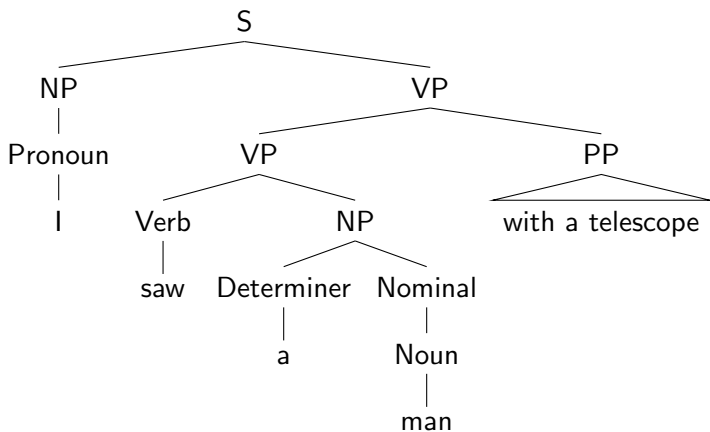
# Attachment Ambiguity: Example

I saw a man with a telescope

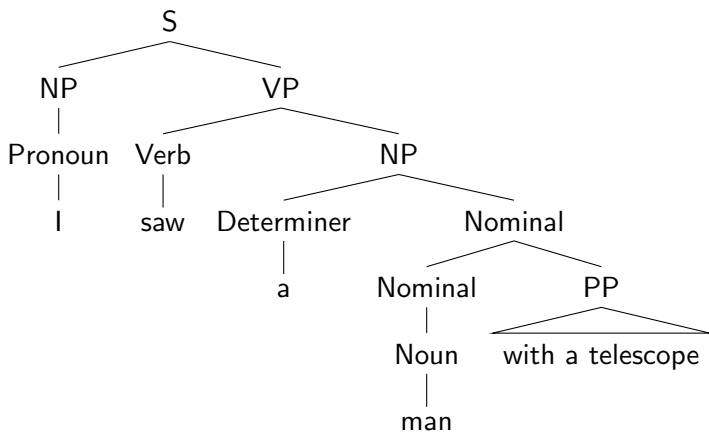
Modify the following tree for the above sentence



## Attachment Ambiguity: 1

I saw a man with a telescope

## Attachment Ambiguity: 2

I saw a man with a telescope

# Simple Coordination Productions

Add these to the earlier grammar

NP  $\rightarrow$  NP Conjunction NP

Nominal  $\rightarrow$  Nominal Conjunction Nominal

VP  $\rightarrow$  VP Conjunction VP

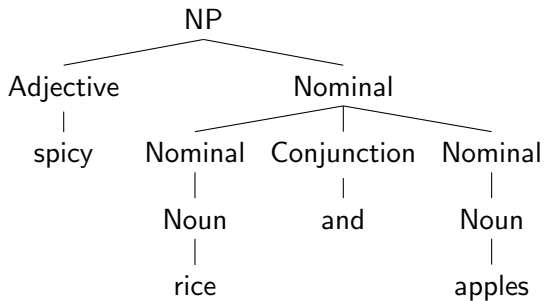
PP  $\rightarrow$  PP Conjunction PP

Also, for adjectives include

NP  $\rightarrow$  Adjective Nominal

# Coordination Ambiguity: 1

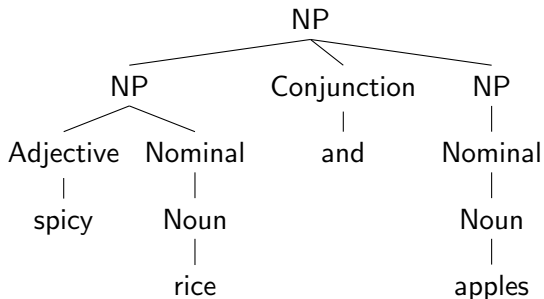
Spicy rice and apples





# Coordination Ambiguity: 2

Spicy rice and apples



# Sentences in Practice

A. A. Milne, Winnie the Pooh

## Eeyore's take on writing

“This writing business. Pencils and what-not. Over-rated, if you ask me. Silly stuff. Nothing in it.”

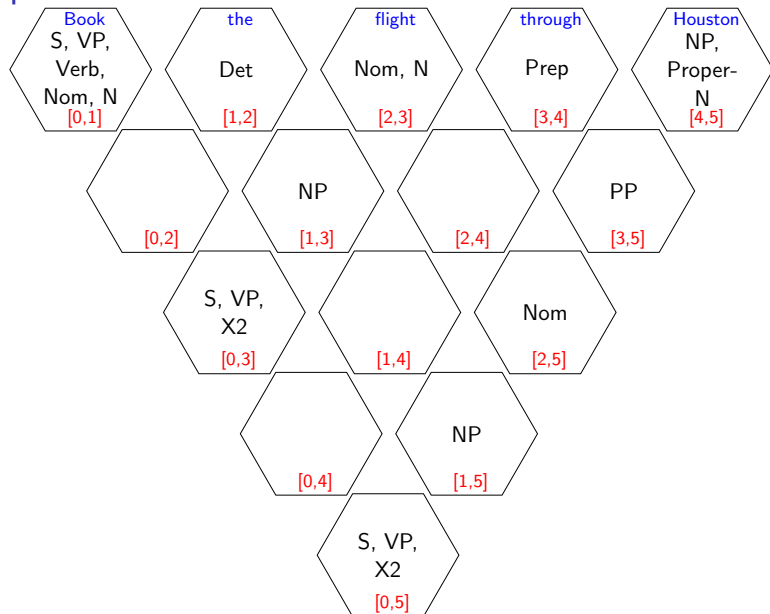
- ▶ Five sentences
- ▶ Do you identify verbs in them?
- ▶ What grammar would generate these sentences?

# Parsing with a Context-Free Grammar

## Cocke-Kasami-Younger (CKY) algorithm

- ▶ Apply dynamic programming
  - ▶ Build up solutions incrementally
  - ▶ Reusing them in larger solutions
- ▶ Convert to Chomsky Normal Form
- ▶ Each constituent is based on
  - ▶ A single terminal
  - ▶ Two nonterminals (constituents)
- ▶ Compute and store all possible constituents for each cell in a matrix
  - ▶ Allow duplicates to accommodate ambiguity
  - ▶ Store provenance of each value
- ▶ When we arrive at a cell the cells it relies upon are already computed
- ▶ The nonterminal in the final cell represents the constituent for the entire input (if any)
- ▶ Reconstruct parse tree from the provenance

# Example of a CKY Parse



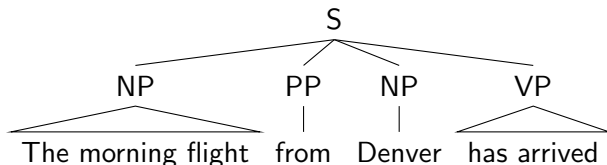
# Improving CKY for Practical Use

- ▶ Generalize to arbitrary grammars (not just Chomsky Normal Form)
  - ▶ Ensures parses produced reflect grammarians' intuitions
- ▶ In statistical parsing, accommodate probabilities to
  - ▶ Select likelier parses
  - ▶ Avoid exponentially many parses

## Partial or Shallow Parsing

Applicable when we don't need a complete parse to produce a valuable product

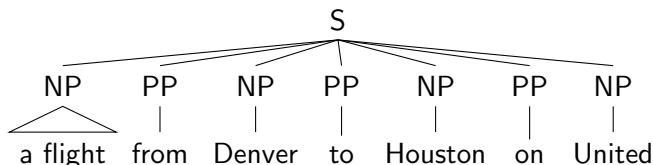
- ▶ Produce flat trees
  - ▶ Avoid decisions about nesting and ambiguity that a full parser must contend with
- ▶ *Chunking*: Identify constituents for nonoverlapping segments
- ▶ Exclude hierarchical structure (i.e., slightly above POS tagging)
  - ▶ [Pro I] [V saw] [NP a man] [PP with a telescope]



# Identifying Base Phrases

Alternative to chunking

- ▶ A base phrase (some variation in definitions)
  - ▶ Doesn't (recursively) contain constituents of the same type
  - ▶ Includes the headword and any prehead modifiers (or any post-head material)
  - ▶ Excludes post-head modifiers (to avoid attachment ambiguity)
    - ▶ Can be difficult to use as a result since boundaries are less clear
    - ▶ Can yield outcomes where an NP or PP may contain nothing other than its head



# Machine Learning for Chunking

An application of sequence learning

- ▶ Introduce  $2n + 1$  tags (given  $n$  chunk types)
  - ▶  $B_k$ : Beginning of chunk type  $k$
  - ▶  $I_k$ : Inside of chunk type  $k$
  - ▶  $O$ : Outside of all chunk types
  - ▶ No need for end of a chunk since the beginning of the next (or end of sentence) indicates its end
- ▶ Example of IOB chunking
 

I	saw	a	man	with	a	telescope
$B_{NP}$	$B_{VP}$	$B_{NP}$	$I_{NP}$	$B_{PP}$	$I_{PP}$	$I_{PP}$
$[NP]I$	$[VP]saw$	$[NP]a$	$man$	$[PP]with$	$a$	$telescope$
- ▶ Training data: from existing treebanks
  - ▶ Identify head words of a constituent
  - ▶ Include head and prehead words within the constituent
  - ▶ Exclude post-head words



## Evaluation Metrics for Chunking

- ▶ Correct chunk: whose tag (label) and segment are correct
- ▶ Metrics adopted from information retrieval

$$\text{Precision, } P = \frac{\text{Number of correct chunks identified}}{\text{Number of chunks identified}}$$

$$\text{Recall, } R = \frac{\text{Number of correct chunks identified}}{\text{Number of (correct) chunks existing}}$$

$$\text{F-measure, } F_{\beta} = \frac{(\beta^2 + 1)PR}{\beta^2 P + R}$$

$$F_1, F_1 = \frac{2PR}{P + R}$$

- ▶ F-measure trades off precision and recall
  - ▶  $F_1$  gives equal importance to precision and recall

# Labeled Recall and Precision to Evaluate Parsers

- ▶ Like recall and precision but
  - ▶ Based on counting correct constituents identified
  - ▶ Correctness with respect to a ground truth *reference* parse tree
- ▶ Recall
  - ▶ How many of the correct constituents are discovered
- ▶ Precision
  - ▶ How many of the constituents discovered are correct

# Cross Brackets

A metric specific to comparing parse trees

- ▶ A measure of error
- ▶ The number of constituents for which
  - ▶ The reference parse has a bracketing ((A B) C)
  - ▶ The hypothesis parse has a bracketing (A (B C))
- ▶ On the Wall Street Journal treebank, modern parsers yield
  - ▶ Recall 90%
  - ▶ Precision 90%
  - ▶ Cross-bracketing 1%
- ▶ Extended metrics for comparing parsers using different grammars

# Human Parsing

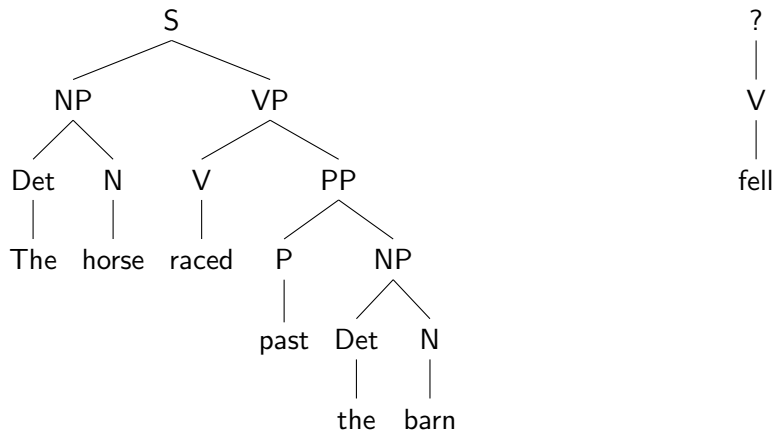
## Psycholinguistics

- ▶ Studies of human processing ease
  - ▶ Delay in reading
  - ▶ Eye gaze fixation (dwell) time
- ▶ Garden-path sentences
  - ▶ Prefix (initial portion) is ambiguous
  - ▶ That is, temporarily ambiguous while reading
  - ▶ A higher preferred parse of the prefix doesn't lead to a parse of the entire sentence

# The Horse Raced Past the Barn Fell: Problematic Sentence

A complete sentence followed by an extra verb

The first part gets a likely parse that offers no clear attachment for the final verb



# The Horse Raced Past the Barn Fell: Correct

Raced is part of a reduced relative clause modifying “The horse”

