#### University of Oslo Department of Informatics



INF4820: Algorithms for Artificial Intelligence and Natural Language Processing

Chart Parsing

Stephan Oepen & Erik Velldal

Language Technology Group (LTG)

October 28, 2015

#### Overview



#### Last Time

- ► Mid-Way Evaluation
- ► Forward Algorithm
- Quiz & Bonus Points
- Syntactic Structure

#### Today

- Context-Free Grammar
- Treebanks
- Probabilistic CFGs
- Syntactic Parsing
  - ▶ Naïve: Recursive-Descent
  - Dynamic Programming: CKY

## Recall: Question (2): Language Modelling



Group members at the Language Technology Group supervise a variety of topics for MSc projects in natural language processing.

Many candidate projects are available on-line.

Please make contact with us.

(2) What is the probability of the bi-gram language technology when ignoring case and punctuation, and using Laplace smoothing?

## Recall: Interpreting the Questions?



- ? *technology* following right after *language*  $\rightarrow P(B|A)$
- ? *language technology* occurring somewhere  $\rightarrow P(A, B)$
- ? *language* and *technology* occuring somewhere → P((A,B))

#### **Recall: Joint and Conditional Probabilities**

$$P(A,B) = P(A) \times P(B|A)$$

$$A \equiv \{w_{i-1} = language\} \ B \equiv \{w_i = technology\}$$

#### Alternatively: A Complex Event

$$A \equiv \{w_{i-1} = language \land w_i = technology\}$$

### **Recall: Syntactic Structures**



#### Constituency

- Words tends to lump together into groups that behave like single units: we call them *constituents*.
- ► *Constituency tests* give evidence for constituent structure:
  - ▶ interchangeable in similar syntactic environments.
  - can be co-ordinated
  - can be moved within a sentence as a unit
- (1) Kim read [a very interesting book about grammar]<sub>NP</sub>. Kim read [it]<sub>NP</sub>.
- (2) Kim [read a book] $_{VP}$ , [gave it to Sandy] $_{VP}$ , and [left] $_{VP}$ .
- (3) [Interesting books about grammar] I like.

### Recall: Grammar Aids Understanding



#### Formal grammars describe a language, giving us a way to:

► j	judge o	r predict well-	formedness
-----	---------	-----------------	------------

Kim was happy because \_\_\_\_\_ passed the exam.

Kim was happy because \_\_\_\_\_ final grade was an A.

make explicit structural ambiguities

Have her report on my desk by Friday! I like to eat sushi with { chopsticks | tuna }.

derive abstract representations of meaning

Kim gave Sandy a book.

Kim gave a book to Sandy.

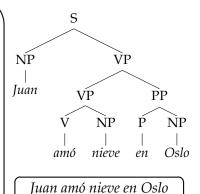
Sandy was given a book by Kim.

### A Grossly Simplified Example



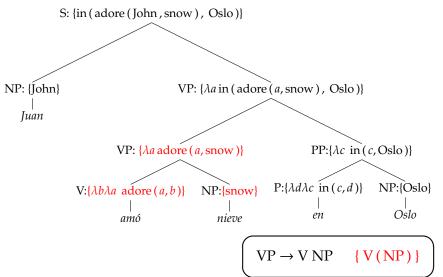
#### The Grammar of Spanish

```
S \rightarrow NP VP
                                   {VP(NP)}
VP \rightarrow V NP
                                     \{V(NP)\}
VP \rightarrow VP PP
                                   \{PP(VP)\}
PP \rightarrow P NP
                                     \{P(NP)\}
NP \rightarrow "nieve"
                                        { snow }
NP \rightarrow "Juan"
                                         { John }
NP \rightarrow "Oslo"
                                         {Oslo}
V \rightarrow "amó" { \lambda b \lambda a \text{ adore}(a, b) }
                           \{\lambda d\lambda c \text{ in } (c,d)\}
```



## Meaning Composition (Still Very Simplified)





### **Another Interpretation**



```
S: {adore (John, in (snow, Oslo)}
NP: {John} VP: {\lambda a adore (a, in (snow, Oslo)}
    Iuan
           V:\{\lambda b\lambda a \text{ adore } (a,b)\}\ NP:\{\text{in } (\text{snow, Oslo})\}\ 
                       amó
                                   NP:{snow}
                                                          PP:\{\lambda c \text{ in } (c, Oslo)\}
                                       nieve
                                                 P:\{\lambda d\lambda c \text{ in } (c,d)\}
                                                                          NP:{Oslo}
                                                                                Oslo
                                                           en
                                                    NP \rightarrow NP PP \{PP(NP)\}
```

#### **Context Free Grammars (CFGs)**



- ▶ Formal system for modeling constituent structure.
- Defined in terms of a lexicon and a set of rules
- Formal models of 'language' in a broad sense
  - natural languages, programming languages, communication protocols, . . .
- Can be expressed in the 'meta-syntax' of the Backus-Naur Form (BNF) formalism.
  - When looking up concepts and syntax in the Common Lisp HyperSpec, you have been reading (extended) BNF.
- ▶ Powerful enough to express sophisticated relations among words, yet in a computationally tractable way.

### CFGs (Formally, this Time)



Formally, a CFG is a quadruple:  $G = \langle C, \Sigma, P, S \rangle$ 

- ► *C* is the set of categories (aka *non-terminals*),
  - ▶ {S, NP, VP, V}
- $\triangleright$   $\Sigma$  is the vocabulary (aka *terminals*),
  - ► {Kim, snow, adores, in}
- ▶ *P* is a set of category rewrite rules (aka *productions*)

$$S \rightarrow NP \ VP$$
  $NP \rightarrow Kim$   $VP \rightarrow V \ NP$   $NP \rightarrow snow$   $V \rightarrow adores$ 

- ▶  $S \in C$  is the *start symbol*, a filter on complete results;
- ▶ for each rule  $\alpha \to \beta_1, \beta_2, ..., \beta_n \in P$ :  $\alpha \in C$  and  $\beta_i \in C \cup \Sigma$

#### **Generative Grammar**



#### Top-down view of generative grammars:

- ▶ For a grammar G, the language  $\mathcal{L}_G$  is defined as the set of strings that can be derived from S.
- ► To derive  $w_1^n$  from S, we use the rules in P to recursively rewrite S into the sequence  $w_1^n$  where each  $w_i \in \Sigma$
- ► The grammar is seen as generating strings.
- Grammatical strings are defined as strings that can be generated by the grammar.
- ► The 'context-freeness' of CFGs refers to the fact that we rewrite non-terminals without regard to the overall context in which they occur.

#### **Treebanks**



#### Generally

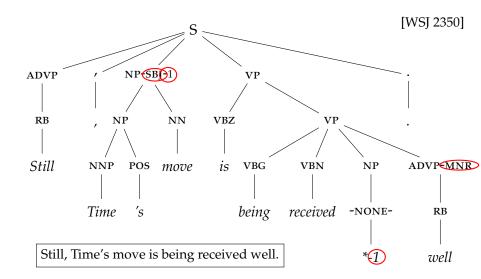
- A treebank is a corpus paired with 'gold-standard' (syntactic) analyses
- Can be created by manual annotation or selection among outputs from automated processing (plus correction).

#### Penn Treebank (Marcus et al., 1993)

- About one million tokens of Wall Street Journal text
- Hand-corrected PoS annotation using 45 word classes
- Manual annotation with (somewhat) coarse constituent structure

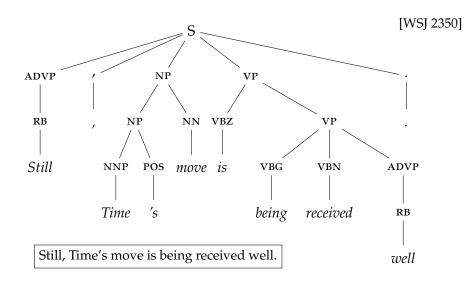
## One Example from the Penn Treebank





#### **Elimination of Traces and Functions**





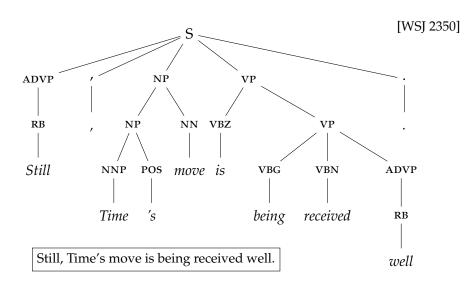
#### Probabilitic Context-Free Grammars



- We are interested, not just in which trees apply to a sentence, but also to which tree is most likely.
- Probabilistic context-free grammars (PCFGs) augment CFGs by adding probabilities to each production, e.g.
  - $S \rightarrow NP VP \qquad 0.6$  $S \rightarrow NP VP PP \qquad 0.4$
- ► These are conditional probabilities the probability of the right hand side (RHS) given the left hand side (LHS)
  - ▶  $P(S \rightarrow NP VP) = P(NP VP|S)$
- We can learn these probabilities from a treebank, again using Maximum Likelihood Estimation.

## **Estimating PCFGs (1/3)**





### Estimating PCFGs (2/3)



```
RB \rightarrow Still
                                                                 AVP \rightarrow RB
(S
                                                                 | \cdot | \rightarrow 
    (ADVP (RB "Still"))
                                                                 NNP \rightarrow Time
    (|.| ".")
                                                                 POS \rightarrow 's
    (NP
                                                                 NP \rightarrow NNP POS
         (NP (NNP "Time") (POS "'s"))
                                                                 NN \rightarrow move
         (NN "move"))
                                                                 NP \rightarrow NP NN
      (VP
                                                                 VBZ \rightarrow is
         (VBZ "is")
                                                                 VBG \rightarrow being
         (VP
                                                                 VBN \rightarrow received
              (VBG "being")
                                                                 RB \rightarrow well
              (VP
                                                                 VP \rightarrow VBN ADVP
                 (VBN "received")
                                                                 VP \rightarrow VBG VP
                 (ADVP (RB "well")))))
                                                                 \backslash . \rightarrow .
    (\. "."))
                                                                 S \rightarrow ADVP \mid NP VP \setminus
                                                                 START \rightarrow S
```

## **Estimating PCFGs (3/3)**



Once we have counts of all the rules, we turn them into probabilities.

$$P(S \to ADVP \mid, \mid NP \mid VP \mid) \approx \frac{C(S \to ADVP \mid, \mid NP \mid VP \mid)}{C(S)}$$

$$= \frac{50}{1150}$$

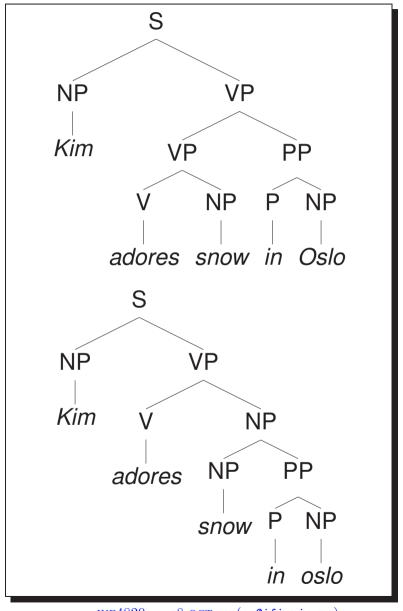
$$= 0.0435$$

# Parsing with CFGs: Moving to a Procedural View

 $S \rightarrow NP \ VP$   $VP \rightarrow V \mid V \ NP \mid VP \ PP$   $NP \rightarrow NP \ PP$   $PP \rightarrow P \ NP$   $NP \rightarrow Kim \mid snow \mid Oslo$   $V \rightarrow adores$   $P \rightarrow in$ 

# **All Complete Derivations**

- are rooted in the start symbol S;
- label internal nodes with categories  $\in C$ , leafs with words  $\in \Sigma$ ;
- instantiate a grammar rule  $\in P$  at each local subtree of depth one.





# Recursive Descend: A Naïve Parsing Algorithm

## **Control Structure**

- top-down: given a parsing goal  $\alpha$ , use all grammar rules that rewrite  $\alpha$ ;
- successively instantiate (extend) the right-hand sides of each rule;
- for each  $\beta_i$  in the RHS of each rule, recursively attempt to parse  $\beta_i$ ;
- $\bullet$  termination: when  $\alpha$  is a prefix of the input string, parsing succeeds.

# (Intermediate) Results

- Each result records a (partial) tree and remaining input to be parsed;
- $\bullet$  complete results consume the full input string and are rooted in S;
- whenever a RHS is fully instantiated, a new tree is built and returned;
- all results at each level are combined and successively accumulated.

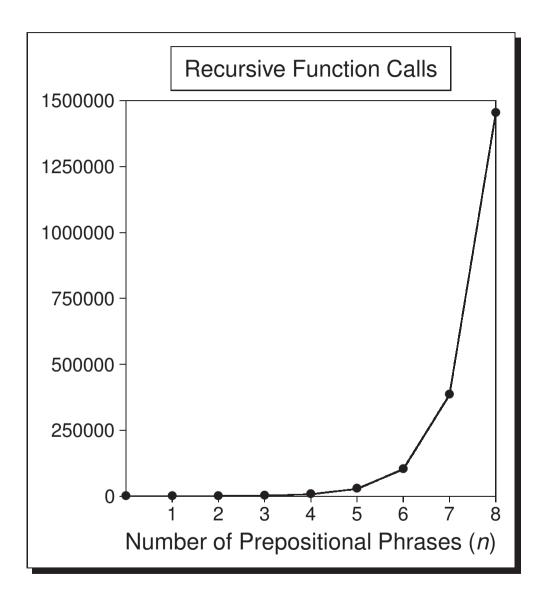


# The Recursive Descent Parser

```
(defun parse (input goal)
  (if (equal (first input) goal)
      (let ((edge (make-edge :category (first input))))
        (list (make-parse :edge edge :input (rest input))))
      (loop
            for rule in (rules-deriving goal)
            append (extend-parse (rule-lhs rule) nil (rule-rhs rule) input))))
```



# **Quantifying the Complexity of the Parsing Task**



Kim adores snow (in Oslo)<sup>n</sup>

n	trees	calls
0	1	46
1	2	170
2	5	593
3	14	2,093
4	42	7,539
5	132	27,627
6	429	102,570
7	1430	384,566
8	4862	1,452,776
i	i	i



# **Top-Down vs. Bottom-Up Parsing**

# **Top-Down (Goal-Oriented)**

- Left recursion (e.g. a rule like 'VP → VP PP') causes infinite recursion;
- search is uninformed by the (observable) input: can hypothesize many unmotivated sub-trees, assuming terminals (words) that are not present;
- $\rightarrow$  assume bottom-up as basic search strategy for remainder of the course.

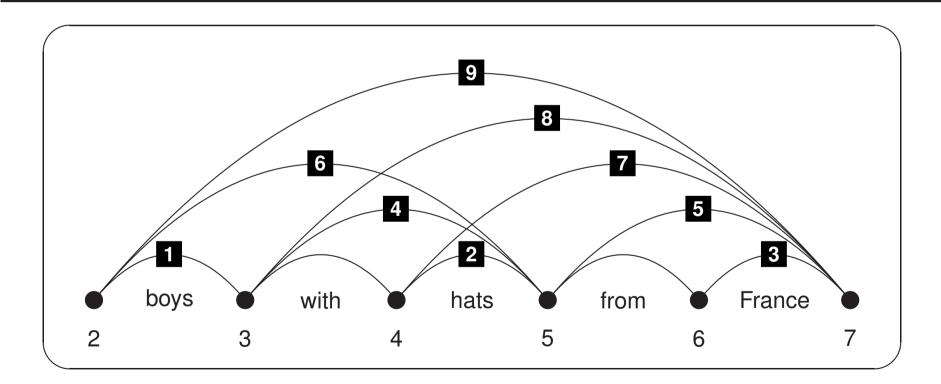
## **Bottom-Up (Data-Oriented)**

- unary (left-recursive) rules (e.g. 'NP → NP') would still be problematic;
- lack of parsing goal: compute all possible derivations for, say, the input adores snow; however, it is ultimately rejected since it is not sentential;
- availability of partial analyses desirable for, at least, some applications.



# A Key Insight: Local Ambiguity

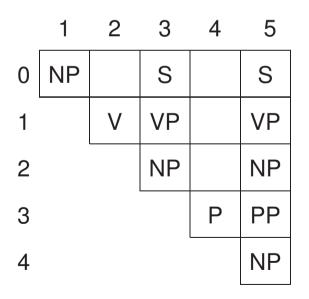
- For many substrings, more than one way of deriving the same category;
- NPs: 1 | 2 | 3 | 6 | 7 | 9; PPs: 4 | 5 | 8; 9 = 1 + 8 | 6 + 5;
- parse forest a single item represents multiple trees [Billot & Lang, 89].





# The CKY (Cocke, Kasami, & Younger) Algorithm

$$\begin{aligned} &\text{for } (0 \leq i < |\textit{input}|) \text{ do} \\ &\textit{chart}_{[i,i+1]} \leftarrow \{\alpha \,|\, \alpha \rightarrow \textit{input}_i \in P\}; \\ &\text{for } (1 \leq l < |\textit{input}|) \text{ do} \\ &\text{for } (0 \leq i < |\textit{input}| - l) \text{ do} \\ &\text{for } (1 \leq j \leq l) \text{ do} \\ &\text{if } (\alpha \rightarrow \beta_1 \,\beta_2 \in P \land \beta_1 \in \textit{chart}_{[i,i+j]} \land \beta_2 \in \textit{chart}_{[i+j,i+l+1]}) \text{ then} \\ &\textit{chart}_{[i,i+l+1]} \leftarrow \textit{chart}_{[i,i+l+1]} \cup \{\alpha\}; \end{aligned}$$





# **Limitations of the CKY Algorithm**

# **Built-In Assumptions**

- Chomsky Normal Form grammars:  $\alpha \to \beta_1\beta_2$  or  $\alpha \to \gamma$  ( $\beta_i \in C$ ,  $\gamma \in \Sigma$ );
- breadth-first (aka exhaustive): always compute all values for each cell;
- rigid control structure: bottom-up, left-to-right (one diagonal at a time).

# **Generalized Chart Parsing**

- Liberate order of computation: no assumptions about earlier results;
- active edges encode partial rule instantiations, 'waiting' for additional (adjacent and passive) constituents to complete: [1, 2, VP → V • NP];
- parser can fill in chart cells in any order and guarantee completeness.

