# CMPT-825 Natural Language Processing

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# Why are parsing algorithms important?

- ► A linguistic theory is implemented in a formal system to generate the set of grammatical strings and rule out ungrammatical strings.
- ▶ Such a formal system has computational properties.
- ▶ One such property is a simple decision problem: given a string, can it be generated by the formal system (recognition).
- ▶ If it is generated, what were the steps taken to recognize the string (parsing).

# Why are parsing algorithms important?

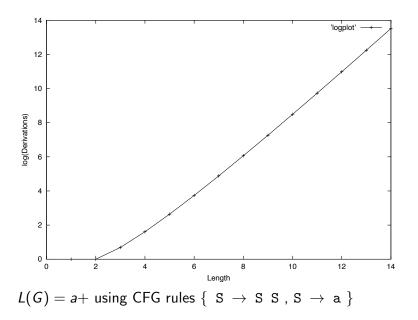
- ▶ Consider the recognition problem: find algorithms for this problem for a particular formal system.
- ▶ The algorithm must be decidable.
- ▶ Preferably the algorithm should be polynomial: enables computational implementations of linguistic theories.
- ▶ Elegant, polynomial-time algorithms exist for formalisms like CFG

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## Number of derivations

CFG rul	es $\{$ S $ ightarrow$ S S , S	$\rightarrow$ a $\}$
n:a <sup>n</sup>	number of parses	
1	1	
2	1	
3	2	
4	5	
5	14	
6	42	
7	132	
8	429	
9	1430	
10	4862	
11	16796	

# Number of derivations grows exponentially



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# Syntactic Ambiguity: (Church and Patil 1982)

- Algebraic character of parse derivations
- ► Power Series for grammar for coordination type of grammars (more general than PPs):

```
N \rightarrow natural | language | processing | course N \rightarrow N N
```

- ▶ We write an equation for algebraic expansion starting from N
- ➤ The equation represents generation of each string in the language as the terms, and the number of different ways of generating the string as the coefficients:

# **CFG** Ambiguity

- ► Coefficients in previous equation equal the number of parses for each string derived from *E*
- ► These ambiguity coefficients are Catalan numbers:

$$Cat(n) = \frac{1}{n+1} \left( \begin{array}{c} 2n \\ n \end{array} \right)$$

 $\qquad \qquad \bullet \quad \begin{pmatrix} a \\ b \end{pmatrix} \text{ is the binomial coefficient}$ 

$$\left(\begin{array}{c} a \\ b \end{array}\right) = \frac{a!}{\left(b!(a-b)!\right)}$$

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#### Catalan numbers

- ▶ Why Catalan numbers? Cat(n) is the number of ways to parenthesize an expression of length *n* with two conditions:
  - 1. there must be equal numbers of open and close parens
  - 2. they must be properly nested so that an open precedes a close

#### Catalan numbers

For an expression of with n ways to form constituents there are a total of 2n choose n parenthesis pairs, e.g. for n = 2,

$$\begin{pmatrix} 4 \\ 2 \end{pmatrix} = 6:$$
 a(bc), a)bc(, )a(bc, (ab)c, )ab(c, ab)c(

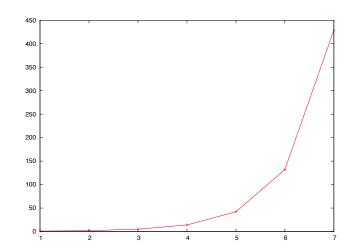
- ▶ But for each valid parenthesis pair, additional n pairs are created that have the right parenthesis to the left of its matching left parenthesis, from e.g. above: a)bc(, )a(bc, )ab(c, ab)c(
- ▶ So we divide 2n choose n by n + 1:

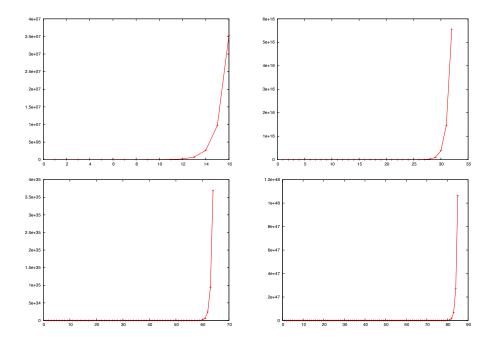
$$Cat(n) = \frac{\binom{2n}{n}}{n+1}$$

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#### Catalan numbers

n	catalan(n)
1	1
2	2
3	5
4	14
5	42
6	132
7	429
8	1430
9	4862
10	16796





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# Syntactic Ambiguity

► Cat(n) also provides exactly the number of parses for the sentence: John saw the man on the hill with the telescope (generated by the grammar given below, a different grammar will have different number of parses)

number of parse trees = Cat(2 + 1) = 5. With 8 PPs: Cat(9) = 4862 parse trees

## Syntactic Ambiguity

- For grammar on previous page, number of parse trees = Cat(2 + 1) = 5.
- ▶ Why Cat(2+1)?
  - ▶ For 2 PPs, there are 4 things involved: VP, NP, PP-1, PP-2
  - We want the items over which the grammar imposes all possible parentheses
  - ► The grammar is structured in such a way that each combination with a VP or an NP reduces the set of items over which we obtain all possible parentheses to 3
  - ► This can be viewed schematically as VP \* NP \* PP-1 \* PP-2
    - 1. (VP (NP (PP-1 PP-2)))
    - 2. (VP ((NP PP-1) PP-2))
    - 3. ((VP NP) (PP-1 PP-2))
    - 4. ((VP (NP PP-1)) PP-2)
    - 5. (((VP NP) PP-1) PP-2)
  - ▶ Note that combining PP-1 and PP-2 is valid because PP-1 has an NP inside it.

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# Syntactic Ambiguity

▶ Other sub-grammars are simpler. For chains of adjectives: cross-eyed pot-bellied ugly hairy professor We can write the following grammar, and compute the power series:

$$ADJP 
ightarrow adj \ ADJP \mid \epsilon$$
 
$$ADJP = 1 + adj + adj^2 + adj^3 + \dots$$

# Syntactic Ambiguity

▶ Now consider power series of combinations of sub-grammars:

```
S = NP \cdot VP ( The number of products over sales ... ) ( is near the number of sales ... )
```

► Both the NP subgrammar and the VP subgrammar power series have Catalan coefficients

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# Syntactic Ambiguity

▶ The power series for the S  $\rightarrow$  NP VP grammar is the multiplication:

$$(N \sum_{i} Cat_{i} (P N)^{i}) \cdot (is \sum_{i} Cat_{j} (P N)^{j})$$

▶ In a parser for this grammar, this leads to a cross-product:

$$L \times R = \{ (I, r) | I \in L \& r \in R \}$$

# Syntactic Ambiguity

► A simple change:

```
Is ( The number of products over sales ... ) ( near the number of sales ... )  = \text{Is } N \sum_{i} Cat_{i} (PN)^{i} \cdot (\sum_{j} Cat_{j} (PN)^{j}) 
 = \text{Is } N \sum_{i} \sum_{j} Cat_{i} Cat_{j} (PN)^{i+j} 
 = \text{Is } N \sum_{i+j} Cat_{i+j+1} (PN)^{i+j}
```

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## Dealing with Ambiguity

- ► A CFG for natural language can end up providing exponentially many analyses, approx n!, for an input sentence of length n
- Much worse than the worst case in the part of speech tagging case, which was  $n^m$  for m distinct part of speech tags
- ▶ If we actually have to process all the analyses, then our parser might as well be exponential
- ► Typically, we can directly use the compact description (in the case of CKY, the chart or 2D array, also called a *forest*)

#### Dealing with Ambiguity

- Solutions to this problem:
  - ▶ CKY algorithm: computes all parses in  $\mathcal{O}(n^3)$  time. Problem is that worst-case and average-case time is the same.
  - ▶ Earley algorithm: computes all parses in  $\mathcal{O}(n^3)$  time for arbitrary CFGs,  $\mathcal{O}(n^2)$  for unambiguous CFGs, and  $\mathcal{O}(n)$  for so-called bounded-state CFGs (e.g.  $S \to aSa \mid bSb \mid aa \mid bb$  which generates palindromes over the alphabet a, b). Also, average case performance of Earley is better than CKY.
  - ▶ Deterministic parsing: only report one parse. Two options: top-down (LL parsing) or bottom-up (LR or shift-reduce) parsing

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#### Shift-Reduce Parsing

- ► Every CFG has an equivalent pushdown automata: a finite state machine which has additional memory in the form of a stack
- ▶ Consider the grammar:  $NP \rightarrow Det\ N,\ Det \rightarrow the,\ N \rightarrow dogs$
- Consider the input: the dogs
- ▶ shift the first word *the* into the stack, check if the top *n* symbols in the stack matches the right hand side of a rule in which case you can **reduce** that rule, or optionally you can shift another word into the stack

#### Shift-Reduce Parsing

- ightharpoonup reduce using the rule Det o the, and push Det onto the stack
- ▶ shift dogs, and then reduce using  $N \rightarrow dogs$  and push N onto the stack
- ▶ the stack now contains Det, N which matches the rhs of the rule  $NP \rightarrow Det\ N$  which means we can reduce using this rule, pushing NP onto the stack
- ▶ If *NP* is the start symbol and since there is no more input left to shift, we can accept the string
- ► Can this grammar get stuck (that is, there is no shift or reduce possible at some stage while parsing) on a valid string?
- ▶ What happens if we add the rule  $NP \rightarrow dogs$  to the grammar?

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#### Shift-Reduce Parsing

- ► Sometimes humans can be "led down the garden-path" when processing a sentence (from left to right)
- Such garden-path sentences lead to a situation where one is forced to backtrack because of a commitment to only one out of many possible derivations
- ► Consider the sentence:

  The emergency crews hate most is domestic violence.
- ► Consider the sentence:

  The horse raced past the barn fell

#### Shift-Reduce Parsing

- ▶ Once you process the word *fell* you are forced to reanalyze the previous word *raced* as being a verb inside a *relative clause*: raced past the barn, meaning the horse that was raced past the barn
- ▶ Notice however that other examples with the same structure but different words do not behave the same way.
- ► For example: the flowers delivered to the patient arrived

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- ► Earley Parsing is a more advanced form of CKY parsing with two novel ideas:
  - ► A *dotted rule* as a way to get around the explicit conversion of a CFG to Chomsky Normal Form
  - ► Do not explore every single element in the CKY parse chart. Instead use goal-directed search
- ► Since natural language grammars are quite large, and are often modified to be able to parse more data, avoiding the explicit conversion to CNF is an advantage
- ► A dotted rule denotes that the right hand side of a CF rule has been partially recognized/parsed
- ▶ By avoiding the explicit  $n^3$  loop of CKY, we can parse some grammars more efficiently, in time  $n^2$  or n.
- ► Goal-directed search can be done in any order including left to right (more psychologically plausible)

- ▶  $S \rightarrow \bullet NP \ VP$  indicates that once we find an NP and a VP we have recognized an S
- $S \rightarrow NP$  VP indicates that we've recognized an NP and we need a VP
- $S \rightarrow NP \ VP$  indicates that we have a complete S
- Consider the dotted rule S → •NP VP and assume our CFG contains a rule NP → John
  Because we have such an NP rule we can predict a new dotted rule NP → John

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- ▶ If we have the dotted rule:  $NP \rightarrow \bullet$  John and the next input symbol on our *input tape* is the word John we can **scan** the input and create a new dotted rule  $NP \rightarrow John \bullet$
- ▶ Consider the dotted rule  $S \to \bullet NP$  VP and  $NP \to John \bullet$  Since NP has been completely recognized we can **complete**  $S \to NP \bullet VP$
- These three steps: predictor, scanner and completer form the Earley parsing algorithm and can be used to parse using any CFG without conversion to CNF Note that we have not accounted for € in the scanner

- ▶ A state is a dotted rule plus a span over the input string, e.g.  $(S \rightarrow NP \bullet VP, [4, 8])$  implies that we have recognized an NP
- ▶ We store all the states in a *chart* in *chart[j]* we store all states of the form:  $(A \to \alpha \bullet \beta, [i, j])$ , where  $\alpha, \beta \in (N \cup T)^*$

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- Note that  $(S \to NP \bullet VP, [0,8])$  implies that in the chart there are two states  $(NP \to \alpha \bullet, [0,8])$  and  $(S \to \bullet NP VP, [0,0])$  this is the *completer* rule, the heart of the Earley parser
- ▶ Also if we have state  $(S \rightarrow \bullet NP \ VP, [0, 0])$  in the chart, then we always *predict* the state  $(NP \rightarrow \bullet \alpha, [0, 0])$  for all rules  $NP \rightarrow \alpha$  in the grammar

```
S \rightarrow NP VP
NP \rightarrow Det N \mid NP PP \mid John
Det \rightarrow the
N \rightarrow cookie \mid table
VP \rightarrow VP PP \mid V NP \mid V
V \rightarrow ate
PP \rightarrow P NP
P \rightarrow on
```

Consider the input: 0 John 1 ate 2 on 3 the 4 table 5 What can we predict from the state  $(S \rightarrow \bullet NP VP, [0, 0])$ ? What can we complete from the state  $(V \rightarrow ate \bullet, [1, 2])$ ?

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```
enqueue(state, j):
    input: state = (A → α • β, [i, j])
    input: j (insert state into chart[j])
    if state not in chart[j] then
        chart[j].add(state)
    end if
predictor(state):
    input: state = (A → B • C, [i, j])
    for all rules C → α in the grammar do
        newstate = (C → • α, [j, j])
        enqueue(newstate, j)
    end for
```

```
scanner(state, tokens):
    input: state = (A → B • a C, [i, j])
    input: tokens (list of input tokens to the parser)
    if tokens[j] == a then
        newstate = (A → B a • C, [i, j + 1])
        enqueue(newstate, j+1)
    end if
completer(state):
    input: state = (A → B C •, [j, k])
    for all rules X → Y • A Z, [i, j] in chart[j] do
        newstate = (X → Y A • Z, [i, k])
        enqueue(newstate, k)
    end for
```

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```
earley(tokens[0 . . . N], grammar):
      for each rule S \rightarrow \alpha where S is the start symbol do
        add (S \rightarrow \bullet \alpha, [0, 0]) to chart [0]
      end for
      for 0 \le j \le N+1 do
        for state in chart[j] that has not been marked do
           mark state
           if state = (A \rightarrow \alpha \bullet B \beta, [i,j]) then
              predictor(state)
           else if state = (A \rightarrow \alpha \bullet b \beta, [i,j]), j < N+1 then
              scanner(state, tokens)
              completer(state)
           end if
        end for
      end for
      return yes if chart [N+1] has a final state
```

```
    isIncomplete(state):
        if state is of type (A → α •, [i, j]) then
            return False
        end if
        return True
    nextCategory(state):
        if state == (A → B • ν C, [i, j]) then
            return ν (ν can be terminal or non-terminal)
        else
            raise error
        end if
```

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```
    isFinal(state):
        input: state = (A → α •, [i, j])
        cond1 = A is a start symbol
        cond2 = isIncomplete(state) is False
        cond3 = j is equal to length(tokens)
        if cond1 and cond2 and cond3 then
        return True
        end if
        return False
    isToken(category):
        if category is terminal symbol then
        return True
        end if
        return False
```