

CMPT-413: Computational Linguistics

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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

Top-down, depth-first, left to right parsing

```
S \rightarrow NP VP
NP \rightarrow Det N
NP \rightarrow Det N PP
VP \rightarrow V
VP \rightarrow V NP
VP \rightarrow V NP PP
PP \rightarrow P NP
NP \rightarrow I
Det \rightarrow a | the
  V \rightarrow saw
  N \rightarrow park \mid dog \mid man \mid telescope
  P \rightarrow in \mid with
```

Top-down, depth-first, left to right parsing

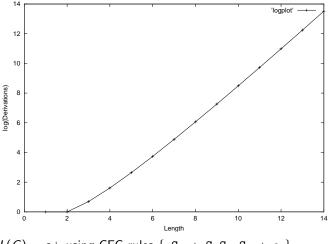
- ► Consider the input string: the dog saw a man in the park
- ► S ... (S (NP VP)) ... (S (NP Det N) VP) ... (S (NP (Det the) N) VP) ... (S (NP (Det the) (N dog)) VP) ...
- (S (NP (Det the) (N dog)) VP) ... (S (NP (Det the) (N dog)) (VP V NP PP)) ... (S (NP (Det the) (N dog)) (VP (V saw) NP PP)) ...
- ► (S (NP (Det the) (N dog)) (VP (V saw) (NP Det N) PP)) ...
- (S (NP (Det the) (N dog)) (VP (V saw) (NP (Det a) (N man)) (PP (P in) (NP (Det the) (N park)))))

Number of derivations

CFG rules $\{$ S \rightarrow S S , S \rightarrow a $\}$

n:a ⁿ	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



$$L(G) = a + \text{ using CFG rules } \{ S \rightarrow S S, S \rightarrow a \}$$

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:

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

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

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,

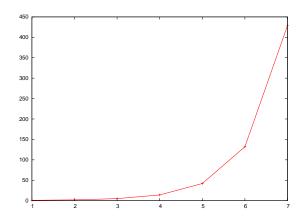
$$\begin{pmatrix} 4 \\ 2 \end{pmatrix} = 6:$$
a(bc), a)bc(,)a(bc, (ab)c,)ab(c, ab)c(

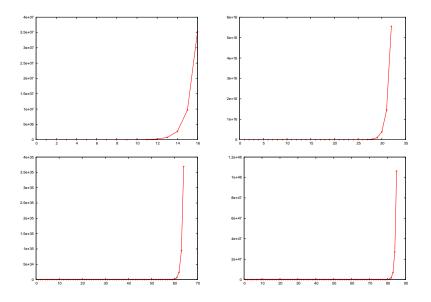
- ▶ But for each valid parenthesis pair, additional n pairs are created that have the right parenthesis to the left of its matching left parenthesis, from e.g. above: a)bc(,)a(bc,)ab(c, ab)c(
- ▶ So we divide 2n choose n by n + 1:

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

Catalan numbers

n	catalan(n)
1	1
2	2
3	5
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 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)

$$S o NP VP o VP o VP o VP PP \ NP o John \mid Det N NP o NP PP \ NP o NP o NP o NP PP \ NP o VP o VNP V o Saw \ P o on \mid with \ NP o NP$$

- ► 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.

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:

$$\begin{array}{lll} \textit{ADJP} & \rightarrow & \textit{adj ADJP} & \mid \; \epsilon \\ \\ \textit{ADJP} & = & 1 + \textit{adj} + \textit{adj}^2 + \textit{adj}^3 + \dots \end{array}$$

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

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

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

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

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

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_{i} Cat_{j} (PN)^{j})
     = \text{ Is } N \sum_{i} \sum_{i} \textit{Cat}_{i} \; \textit{Cat}_{j} (\; P \; N \;)^{i+j}
     = \quad \text{Is} \; \; \mathcal{N} \; \sum_{\scriptscriptstyle i \perp i} \mathsf{Cat}_{i+j+1} \; (\; P \; \mathcal{N} \;)^{i+j}
```

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

- 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

- 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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- 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

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- ► Consider the sentence:

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- ► 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

- 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.

- 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

- 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

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- ▶ 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 → •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

- If we have the dotted rule: NP → 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 → John •
- ▶ 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 ϵ in the scanner

- A state is a dotted rule plus a span over the input string, e.g.
 (S → NP 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 $\alpha, \beta \in (N \cup T)^*$

- 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 $(S \to \bullet NP \ VP, [0,0])$ in the chart, then we always *predict* the state $(NP \to \bullet \ \alpha, [0,0])$ for all rules $NP \to \alpha$ in the grammar

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

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 $(V \rightarrow ate \bullet, [1, 2])$?

enqueue(state, j):

```
input: state = (A \rightarrow \alpha \bullet \beta, [i,j])

input: j (insert state into chart[j])

if state not in chart[j] then

chart[j].add(state)

end if
```

enqueue(state, j): **input:** state = $(A \rightarrow \alpha \bullet \beta, [i, j])$ **input:** *i* (insert state into chart[*i*]) **if** state not in chart[j] **then** chart[i].add(state) end if predictor(state): **input:** state = $(A \rightarrow B \bullet C, [i, j])$ **for** all rules $C \rightarrow \alpha$ in the grammar **do** newstate = $(C \rightarrow \bullet \alpha, [i, j])$ enqueue(newstate, i) end for

scanner(state, tokens):

```
input: state = (A \rightarrow B \bullet a C, [i,j])

input: tokens (list of input tokens to the parser)

if tokens[j] == a then

newstate = (A \rightarrow B \ a \bullet C, [i,j+1])

enqueue(newstate, j+1)

end if
```

scanner(state, tokens):

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

```
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 < j < 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)
           else
              completer(state)
           end if
        end for
     end for
     return yes if \operatorname{chart}[N+1] has a final state
```

isIncomplete(state):

```
if state is of type (A \to \alpha \bullet, [i,j]) then return False end if return True
```

```
isIncomplete(state):
     if state is of type (A \rightarrow \alpha \bullet, [i, j]) then
        return False
     end if
     return True
nextCategory(state):
     if state == (A \rightarrow B \bullet \nu C, [i,j]) then
        return \nu (\nu can be terminal or non-terminal)
     else
        raise error
     end if
```

isFinal(state):

```
input: state = (A \rightarrow \alpha \bullet, [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
```

isFinal(state):

```
input: state = (A \rightarrow \alpha \bullet, [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