EDAN₂₀

Language Technology http://cs.lth.se/edan20/

Chapter 12: Constituent Parsing

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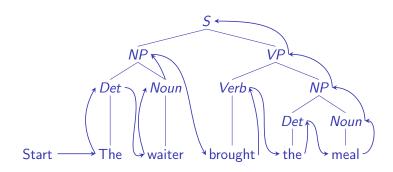


Parsing

Possible parsing strategies are top-down or bottom-up
Prolog uses a top-down exploration and backtracks in case of error
Ambiguity can produce two or more possible parse trees
It is necessary to use probabilistic or symbolic techniques to rank parse trees



Bottom-up Parsing





Shift and Reduce

The shift and reduce algorithm implements bottom-up parsing.

Two input arguments: the list of words to parse and the parsing goal.

The algorithm gradually reduces words, parts of speech, and phrases until it reaches the parsing goal.

The algorithm consists of a loop of two steps:

- Shift a word from the phrase or sentence to parse onto a stack;
- Apply a sequence of grammar rules to **reduce** elements of the stack until there is no more word in the list and the stack is reduced to the parsing goal.



DCG Rules

Nonterminal symbols

```
s --> np, vp.
np --> det, noun.
np --> np, pp.
vp --> verb, np.
vp --> verb, np, pp.
pp --> prep, np.
```

Terminal symbols

```
det --> [the].
det --> [a].
noun --> [waiter].
noun --> [meal].
noun --> [table].
noun --> [day].
verb --> [brought].
prep --> [to].
```

Shift and Reduce in Action

lt.	Stack	S/R	Word list
0			[the, waiter, brought, the, meal]
1	[the]	Shift	[waiter, brought, the, meal]
2	[det]	Reduce	[waiter, brought, the, meal]
3	[det, waiter]	Shift	[brought, the, meal]
4	[det, noun]	Reduce	[brought, the, meal]
5	[np]	Reduce	[brought, the, meal]
6	[np, brought]	Shift	[the, meal]
7	[np, v]	Reduce	[the, meal]
8	[np, v, the]	Shift	[meal]
9	[np, v, det]	Reduce	[meal]
10	[np, v, det, meal]	Shift	
11	[np, v, det, n]	Reduce	
12	[np, v, np]	Reduce	
13	[np, vp]	Reduce	
14	[s]	Reduce	

Backtracking May be Inefficient

Example:

The meal of the day

```
np --> npx. npx --> det, noun.
np --> npx, pp.
pp --> prep, np.
```



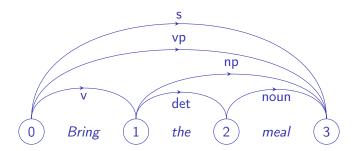
The Structure of a Chart

A chart is a data structure that avoids backtracking It uses classical grammar rules It is a graph (DAG) where nodes are intervals between words

- \bigcirc 0 Bring \bigcirc 1 the \bigcirc 2 meal \bigcirc 3
- \bigcirc The \bigcirc meal \bigcirc of \bigcirc the \bigcirc day \bigcirc 5



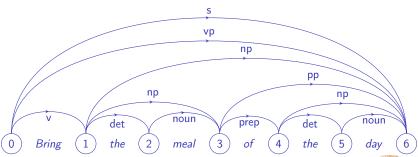
Parsing with a Chart





Charts Contain Alternative Parses

We can view rules vp --> v, np and vp --> v, np, pp in the chart





The Active Chart

The active chart stores constituents being parsed and marks the rules accordingly.

The rule:

```
np --> det noun •
```

is a completely parsed noun phrase: a determiner and a noun.

The arc is said to be inactive

The rules below are said to be active:

```
np --> det • noun A determiner has been found np --> • det noun We are seeking a noun phrase
```



The Earley Algorithm

```
Complexity of 0(N^3)
```

Three modules: the predictor, the scanner, and the completer.

They use phrase-structure rules as:

```
start --> • np
np --> det, noun.
np --> det, adj, noun.
np --> np, pp.
pp --> prep, np.
```



meal

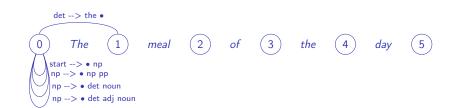
The Predictor

o The 1

start --> • np
np --> • np pp
np --> • det noun
np --> • det adj noun

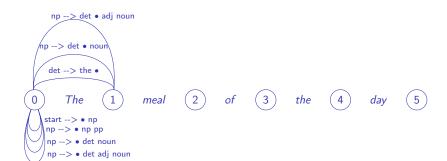
day

The Scanner



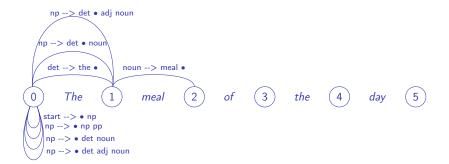


The Completer



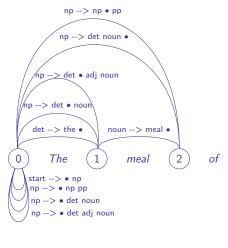


The Next Steps (I)





The Next Steps (II)



) the (4)



day (



The Prolog Database

Module	New chart entries
	Position 0
start	arc(start, ['.', np], 0, 0)
predictor	arc(np, [., d, n], 0, 0), arc(np, [., d, a, n], 0,
	0), arc(np, [., np, pp], 0, 0)
	Position 1
scanner	arc(d, [the, .], 0, 1)
completer	arc(np, [d, ., a, n], 0, 1), arc(np, [d, ., n], 0, 1)
predictor	
	Position 2
scanner	arc(n, [meal, .], 1, 2)
completer	arc(np, [d, n, .], 0, 2)
completer	arc(np, [np, ., pp], 0, 2), arc(start, [np,0, 2)
predictor	arc(pp, [., prep, np], 2, 2)

The Prolog Database

```
Position 3
            arc(prep, [of, .], 2, 3)
scanner
            arc(pp, [prep, ., np], 2, 3)
completer
            arc(np, [., d, n], 3, 3), arc(np, [., d, a, n], 3,
predictor
            3), arc(np, [., np, pp], 3, 3)
            Position 4
            arc(d, [the, .], 3, 4)
scanner
            arc(np, [d, ., a, n], 3, 4), arc(np, [d, ., n], 3, 4)
completer
predictor
            Г٦
            Position 5
            arc(n, [day, .], 4, 5)
scanner
            arc(np, [d, n, .], 3, 5)
completer
completer
            arc(np, [np, ., pp], 3, 5), arc(pp, [prep, np, .], 2,
            5)
            arc(np, [np, pp, .], 0, 5)
completer
            arc(np, [np, ., pp], 0, 5), arc(start, [np, ..., 0]
completer
```

Probabilistic Context-Free Grammars

$$P(T,S) = \prod_{rule(i) \text{producing } T} P(rule(i)).$$

where

$$P(lhs \rightarrow rhs_i | lhs) = \frac{Count(lhs \rightarrow rhs_i)}{\sum\limits_{i} Count(lhs \rightarrow rhs_j)}.$$



An Example of PCFG

Rul	es		Р	Rules			Р
S	>	np vp	0.8	det	>	the	1.0
S	>	vp	0.2	noun	>	waiter	0.4
np	>	det noun	0.3	noun	>	meal	0.3
np	>	det adj noun	0.2	noun	>	day	0.3
np	>	pronoun	0.3	verb	>	bring	0.4
np	>	np pp	0.2	verb	>	slept	0.2
vp	>	v np	0.6	verb	>	brought	0.4
vp	>	v np pp	0.1	pronoun	>	he	1.0
vp	>	v pp	0.2	prep	>	of	0.6
vp	>	V	0.1	prep	>	to	0.4
pp	>	prep np	1.0	adj	>	big	410

Parse Trees of Bring the meal of the day

Parse trees

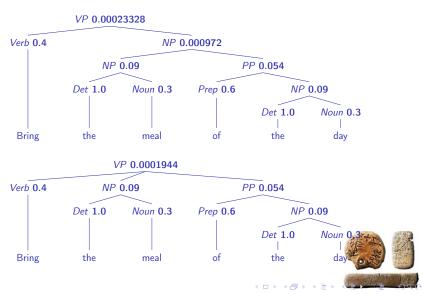


Computing the Probabilities

```
P(T_1, \text{Bring the meal of the day}) = P(vp \rightarrow v, np) \times P(v \rightarrow Bring) \times P(np \rightarrow np, pp) \times P(np \rightarrow det, noun) \times P(det \rightarrow the) \times P(noun \rightarrow meal) \times P(pp \rightarrow prep, np) \times P(prep \rightarrow of) \times P(np \rightarrow det, noun) \times P(det \rightarrow the) \times P(noun \rightarrow day) = 0.6 \times 0.4 \times 0.2 \times 0.3 \times 1.0 \times 0.3 \times 1.0 \times 0.6 \times 0.3 \times 1.0 \times 0.3 = 0.00023328,
```

$$P(T_2, \text{Bring the meal of the day}) = \\ P(\textit{vp} \rightarrow \textit{v}, \textit{np}, \textit{pp}) \times P(\textit{v} \rightarrow \textit{Bring}) \times P(\textit{np} \rightarrow \textit{det}, \textit{noun}) \times \\ P(\textit{det} \rightarrow \textit{the}) \times P(\textit{noun} \rightarrow \textit{meal}) \times P(\textit{pp} \rightarrow \textit{prep}, \textit{np}) \times P(\textit{prep} \rightarrow \textit{of}) \times \\ P(\textit{np} \rightarrow \textit{det}, \textit{noun}) \times P(\textit{det} \rightarrow \textit{the}) \times P(\textit{noun} \rightarrow \textit{day}) = \\ 0.1 \times 0.4 \times 0.3 \times 1.0 \times 0.3 \times 1.0 \times 0.6 \times 0.3 \times 1.0 \times 0.3 = 0.000$$

Computing the Probabilities



PCF Grammars Ignore Lexical Preferences

$$\frac{P(T1|\text{Bring the meal of the day})}{P(T2|\text{Bring the meal of the day})} = \frac{P(T1|\text{Bring the meal to the table})}{P(T2|\text{Bring the meal to the table})},$$

$$= \frac{P(Vp \to V, np) \times P(Np \to Np, pp)}{P(Vp \to V, Np, pp)}$$

PCF grammars do not take into account the lexicon and the attachment preferences of of and to.



Parser Evaluation

Constituent parsing

```
\mathsf{Recall} = \frac{\mathsf{Number} \ \mathsf{of} \ \mathsf{correct} \ \mathsf{constituents} \ \mathsf{generated} \ \mathsf{by} \ \mathsf{the} \ \mathsf{parser}}{\mathsf{Number} \ \mathsf{of} \ \mathsf{constituents} \ \mathsf{in} \ \mathsf{the} \ \mathsf{manually} \ \mathsf{bracketed} \ \mathsf{corpus}}
```

 $Precision = \frac{Number\ of\ correct\ constituents\ generated\ by\ the\ parser}{Total\ number\ of\ constituents\ generated\ by\ the\ parser}$

Bracketing	Crossing brackets				
(((bring) (the meal))	(of the day))	()()
((bring) ((the meal) (of the day)))	()()

