#### EDAN20

Language Technology

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

Chapter 12: Constituent Parsing

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September 19, 2016

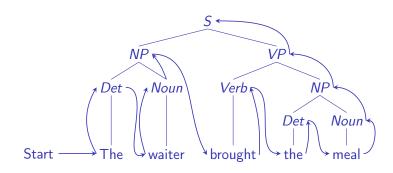


#### Parsing

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.



#### 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.
```



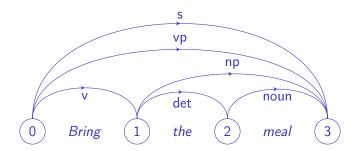
#### The Structure of a Chart

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

- $\bigcirc$  Bring  $\bigcirc$  the  $\bigcirc$  meal  $\bigcirc$  3
- $\bigcirc$  The  $\bigcirc$  meal  $\bigcirc$  of  $\bigcirc$  the  $\bigcirc$  day  $\bigcirc$  5



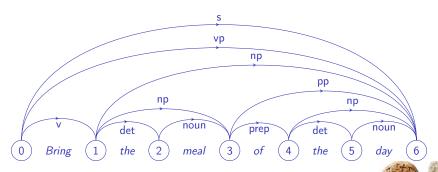
### Parsing with a Chart





#### 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 O(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

0 The 1

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

of (3) the

(4

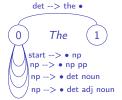
day





meal

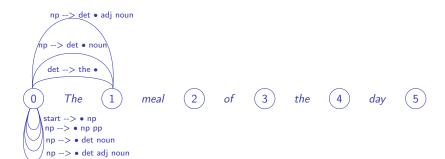
#### The Scanner



of 3 the 4 day 5

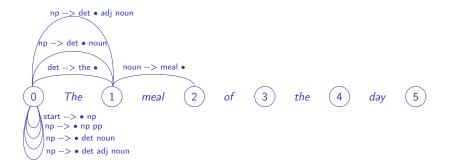


### The Completer



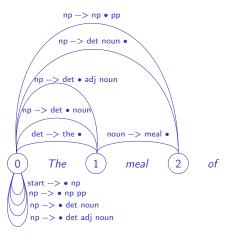


# The Next Steps (I)





# The Next Steps (II)



3 the

4

day

(5



## 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,)      |
| 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

| Rules |   | 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

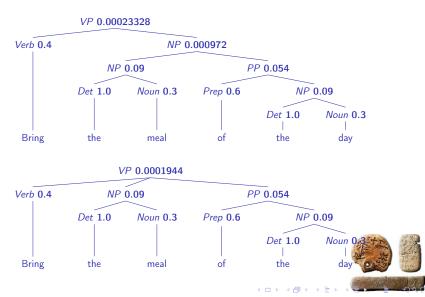


### Computing the Probabilities

```
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(vp \rightarrow v, np, pp) \times P(v \rightarrow Bring}) \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}) = 0.1 \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 = 0.0001
```

### 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\ of\ correct\ constituents\ generated\ by\ the\ parser}}{\mathsf{Number\ of\ constituents\ in\ the\ manually\ bracketed\ corpus}}.
```

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

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

