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Parsing schemata (1)
How characterize a single parsing step?
During parsing, the parser produces trees (parse trees, partial results) and tries to combine them to new trees, until some tree rooted by the goal category (e.g. S) comes out.
- We can characterize parse trees
- We can characterize how new parse trees can be deduced from existing ones
- We can fix a goal: We want to deduce a tree with root S that spans the entire input sentence.

Parsing Schemata (2)
We characterize a parse tree rooted by some nonterminal X by the terminals X spans.

Example:
```
S
  |   |
  NP VP
  |   |
  NP VP
  |   |
  V Det N
  |   |
  I saw a man
```

We write parse trees/partial parse results in the form of items: \([X, i, j]\), meaning that X derives the terminals between i and j.
Here: \([S, 0, 4]\)

Parsing Schemata (3)
Sometimes, a category and its yield have been predicted but not yet recognized. To mark this, we can use dotted items or items with dotted productions:
1. •S signifies that S has been predicted;
2. S• signifies that S has been recognized;
3. \(A \rightarrow A_1 \ldots A_i \cdot A_{i+1} \ldots A_n\) signifies that the rhs of the production \(A \rightarrow A_1 \ldots A_n\) has been recognized up to \(A_i\) while the part from \(A_{i+1}\) to \(A_n\) has been predicted.

Parsing Schemata (4)
Parsing Schemata understand parsing as a deductive process. Deduction of new items from existing ones can be described using inference rules.

General form:
\[
\text{antecedent} \quad \text{consequent} \quad \text{side conditions}
\]

antecedent, consequent: lists of items
Application: if antecedent can be deduced and side condition holds, then the consequent can be deduced as well.
A parsing schema consists of
- deduction rules;
- an axiom (or axioms), can be written as a deduction rule with empty antecedent;
- and a goal item.

The parsing algorithm succeeds if, for a given input, it is possible to deduce the goal item.

Example: Unger (1)
Assume CFG without $\epsilon$-productions and without loops $A \Rightarrow A$.

```latex
\begin{align*}
\text{function unger}(w, X): & \\
\text{out} & := \text{false}; \\
\text{if } w = X, \text{ then out} & := \text{true} \\
\text{else for all } X \Rightarrow X_1 \ldots X_k: & \\
\text{for all } x_1, \ldots, x_k \in T^+ \text{ with } w = x_1 \ldots x_k: & \text{ Predict} \\
\text{if } \bigwedge_{i=1}^k \text{unger}(x_i, X_i): & \\
\text{then out} & := \text{true}; \\
\text{return out}
\end{align*}
```

Initial call: unger($w, S$)

Example: Unger (2)
An item needs to characterize
- a nonterminal category or a terminal symbol;
- its yield in the input string;
- whether the item is predicted or recognized.

Item form:
$[\bullet X, i, j]$ or $[X \bullet, i, j]$ with $X \in N \cup T, i, j \in \mathbb{N}, i \leq j$.

Example: Unger (3)
We start with the prediction that $S$ yields the whole input (question $S \Rightarrow w$):

Axiom:
$[\bullet S, 0, n]$, $|w| = n$

Whenever we encounter a terminal that matches the input, we can turn the predict item into a recognize item:

Scan:
$[\bullet a, i, i+1]$, $w_{i+1} = a$
Example: Unger (4)
Whenever we have predicted an A ∈ N and there is an A-production, we can predict this while partitioning the input according to its rhs (here we include the check for terminals):
Predict:
\[
\begin{align*}
\{A, i_0, i_k\} & \quad A \rightarrow A_1 \ldots A_k \in P; i_j < i_{j+1} \\
\{A_1, i_0, i_1\}, \ldots, \{A_k, i_{k-1}, i_k\} & \quad \text{if } A_j \in T \text{ then } i_j = i_{j-1} + 1 \\
& \quad \text{and } w_j = A_j
\end{align*}
\]

Once all predictions for the rhs are true (turned into recognized items), we can turn also the A-item into a recognize item:
Complete:
\[
\begin{align*}
\{X, i_0, i_k\} & \quad \text{A} \rightarrow \text{A}_1 \ldots \text{A}_k \in P
\end{align*}
\]

Example: Unger (5)
The goal is to find an S that spans the whole input:
Goal item \([S, 0, n]\) where \(n = |w|\).

Example: Unger (6)
Sample CFG: \(S \rightarrow aSb \mid ab\)
Input word \(w = aaabbb\). Deduced items:
\[
\begin{align*}
\{S, 0, n\} & \quad \text{axiom} \\
\{a, 0, 1\}, \{S, 1, 5\}, \{b, 5, 6\} & \quad \text{predict} \\
\{a, 0, 1\}, \{a, 1, 2\}, \{S, 2, 4\}, \{b, 4, 5\}, \{b, 5, 6\} & \quad \text{scan, predict, scan} \\
\{a, 1, 2\}, \{a, 2, 3\}, \{b, 3, 4\}, \{b, 4, 5\} & \quad \text{scan, predict, scan} \\
\{a, 2, 3\}, \{b, 3, 4\} & \quad \text{scan, scan} \\
\{S, 2, 4\} & \quad \text{complete} \\
\{S, 1, 5\} & \quad \text{complete} \\
\{S, 0, 6\} & \quad \text{complete}
\end{align*}
\]

Example: Unger (7)
Soundness and completeness of Ungers’s algorithm:
Assume that we don’t have the check on the terminals. Then for all \(X \in N \cup T, i, j \in [0..n]\) with \(i < j\):

- \(\{X, i, j\}\) if \(S \Rightarrow_{\alpha} \alpha \beta\) for some \(\alpha, \beta \in (N \cup T)^*\) such that \(|\alpha| \leq i, |\beta| \leq n - j\);
- \([X, i, j]\) if \(S \Rightarrow_{\alpha} \alpha \beta \Rightarrow_{\alpha w_{i+1} \ldots w_j \beta}\) for some \(\alpha, \beta \in (N \cup T)^*\) such that \(|\alpha| \leq i, |\beta| \leq n - j\);

This can be shown by induction on the parsing schema:

1. Show that claim holds for axiom.
2. Show for every deduction rule that, if the claim holds for the antecedent, then it also holds for the consequent.
Example: Top-Down (1)
Assume CFG without $\epsilon$-productions and without loops $A \Rightarrow A$.

function top-down($w, \alpha$):
    out := false;
    if $w = \alpha = \epsilon$, then out := true;
    else if $w = aw'$ and $\alpha = a\alpha'$,
        then out := top-down($w', \alpha'$)
    else if $\alpha = X\alpha'$ with $X \in N$,
        then for all $X \rightarrow X_1\ldots X_k$:
            if top-down($w, X_1\ldots X_k\alpha'$) then out := true;
    return out

Example: Top-Down (2)
The items must encode
• the remaining input or, alternatively, the position up to which
  the input has been parsed, and
• the remaining sentential form
⇒ item form $[\alpha, i]$ with $\alpha \in (N \cup T)^*, 0 \leq i \leq n$

Example: Top-Down (3)
Whenever we have the next input terminal as topmost symbol of the stack (left element of $\alpha$), we can scan it:

Scan: $[\alpha, i] \quad w_{i+1} = a$

Whenever the topmost stack symbol is $A$ and there is an $A$-production $A \rightarrow \gamma$, we can predict this (here with check on length of sentential form):

Predict: $[A\alpha, i] \quad A \rightarrow \gamma \in P, |\gamma\alpha| \leq n - i$

Example: Top-Down (4)
Axiom is the whole input $w$ as remaining input (i.e., only the part up to position 0 has been parsed) and a stack containing $S$:

Axiom: $[S, 0]$

The goal item is an empty stack with the input up to position $n$ already parsed:

Goal: $[\epsilon, n]$ with $|w| = n$
Example: Top-Down (5)

CFG $S \rightarrow aSb \mid ab$, input $w = aaabbb$:

Deduced items (only successful ones are listed):

- $[S, 0]$ axiom
- $[aSb, 0]$ predict
- $[Sb, 1]$ scan
- $[aSbb, 1]$ predict
- $[Sbb, 2]$ scan
- $[abbb, 2]$ predict
- $[bbb, 3], [bb, 4], [b, 5], [\epsilon, 6]$ scan

Implementation issues (1)

When dealing with natural languages, we are in general faced with highly ambiguous grammars.

1. On the one hand, strings can have more than one analysis. Consequently, we need to find some way to branch and pursue all of them.
2. On the other hand, different analyses can have common sub-analyses for certain substrings. In order to avoid computing these sub-analyses several times, we need to find some way to reuse (partial) parse trees that we have already found.

⇒ we have to store intermediate parsing results, make sure we pursue all of them and retrieve them if needed in order to reuse them in a different context.

Example: Top-Down (6)

How about soundness and completeness?

There is a direct correspondence between leftmost derivations of a $w$ and parses in a top down parser:

- Soundness:
  If $[\alpha, i]$, then $S \Rightarrow w_1 \ldots w_i \alpha$.

- Completeness:
  If $S \Rightarrow w_1 \ldots w_i \gamma$ is a leftmost derivation where $\gamma \in (N \cup T)^*$ such that $\gamma = A\gamma'$, $A \in N$ or $\gamma = \epsilon$, then $[\gamma, i]$.

Implementation issues (2)

Computation sharing (tabulation) is particularly easy when using parsing schemata:

- During parsing, we deduce new trees from already existing trees (partial results), present as items.
- The same item can be used in different deductions (and has to be calculated only once).

Related notion: chart parsing. The chart is the structure that contains all intermediate results computed so far.
Implementation issues (3)

Chart parsing:
We have two structures,
• the chart \(C\)
• and an agenda \(A\).
Both are initialized as empty.

• We start by computing all items that are axioms, i.e., that can
  be obtained by applying rules with empty antecedents.
• Starting from these items, we extend the set \(C\) as far as
  possible by subsequent applications of the deduction rules.
• The agenda contains items that are waiting to be used in
  further deduction rules. It avoids multiple applications of the
  same instance of a deduction rule.

\(C = A = \emptyset\)
for all items \(I\) resulting form a rule application with empty
antecedent set:
  add \(I\) to \(C\) and to \(A\)
repeat until \(A = \emptyset\):
  remove an item \(I\) from \(A\)
  for all items \(I'\) resulting from a rule application
    with antecedents \(I\) and items from \(C\):
      if \(I' \notin C\)
        then add \(I'\) to \(C\) and to \(A\)
if there is a goal item in \(C\)
  then output true
else output false

Implementation issues (4)

Example: Unger Parsing: The chart is a \((n+1) \times (n+1)\) table
where \(n\) is the length of the input.
• Whenever an item \([X, i, j]\) is predicted, we enter it into the
  chart.
• Whenever an item \([X, i, j]\) is completed, we replace the
  predicted item by the completed one.

A chart is sometimes depicted as a directed graph where the
positions between terminals are the nodes and for every completed
\([\alpha, i, j]\) there is a directed edge from \(i\) to \(j\) with edge label \(\alpha\).

Implementation issues (5)

Relation to hypergraphs:
• The deduction rules describe dependencies between items.
• Every application of a deductions rules can be seen as an
  hyperedge (i.e., an edge from a set of items to a set of items,
  namely from the set of antecedents to the set of consequents.
• The set of items together with these hyperedges forms a
  hypergraph.

Control structures: any topological order of the nodes of this
hypergraph is a possible order of the computations during parsing.

Topological order = for every hyperedge from \{\(x_1, \ldots, x_k\}\} to
\{\(y_1, \ldots, y_m\)\}, the items \(x_1, \ldots, x_k\) must be visited (generated) before
visiting \(y_1, \ldots, y_m\).
Complexity (1)

For a given grammar $G$ and an input sentence $w \in T^*$, we call the recognition problem the task to decide whether $w \in L(G)$ or not.

- **Fixed recognition problem**: Assume a given grammar $G$ (fixed). Then decide for a given input word $w$ if $w \in L(G)$. In this case, the complexity of the problem is given only with respect to the size of the input sentence $w$, i.e., the size of the grammar is taken to be a constant. This is also sometimes called the word recognition problem.

- **Universal recognition problem**: Decide for an input grammar $G$ and an input word $w$ if $w \in L(G)$. In this case, we have to investigate the complexity of the problem in the size of the input sentence $w$ and the grammar $G$.

Complexity (2)

In real natural language applications, we often deal with very large grammars: Grammars extracted from treebanks for instance can easily have much more than 10,000 productions.

- The average sentence length in natural languages is somewhere between 20 and 30.
- Therefore, for natural language processing, the complexity of the universal recognition problem is an important factor.

Complexity (3)

We distinguish between the time and the space complexity.

We distinguish the following different complexity classes:

- **P (PTIME)**: problems that can be solved deterministically in an amount of time that is polynomial in the size of the input. I.e., there are constants $c$ and $a$ such that the problem can be solved in an amount of time $\leq cn^a$ where $n$ the size of the input.
  Notation: $O(n^a)$.

- **NP**: problems whose positive solutions can be verified in polynomial time given the right information, or equivalently, whose solutions can be non-deterministically found in polynomial time.

Complexity (4)

- **NP-complete**: the hardest problems in NP. A problem is NP-complete if any problem in NP can be transformed into it in polynomial time.

The question whether the two classes P and NP are equal or not is an open question. Most people think however that NP is larger.
The specification of parsing algorithms via deduction rules facilitates the computation of the complexity of an algorithm.

In order to determine the time complexity, we have to calculate the maximal number of (different) rule applications that is possible.

This depends on the most complex deduction rule in our parsing schema.

Example: Unger. Most complex rule: complete

\[
\text{\begin{align*}
\text{A, } i_0, i_k \rightarrow [A_1, i_0, i_1, \ldots, A_k, i_{k-1}, i_k] \to A \to A_1 \ldots A_k \in P
\end{align*}}
\]

Complexity \(O(n^{k+1})\) where \(k\) the maximal length of a righthand side in the grammar.

Conclusion

Parsing Schemata

• characterize partial parsing results via items;
• characterize parsing as a deductive process;
• allow to separate the proper algorithm from data structures and control structures;
• facilitate the proof of soundness and completeness of an algorithm;
• facilitate comparisons between different algorithms;
• make the complexity of an algorithm more visible;
• facilitate tabulation and computation sharing.

References


