Recognition: Given a (context-free) grammar G and a string of words w, determine whether $w \in L(G)$. Parsing: If $w \in L(G)$, produce the (tree) structure that is assigned by G to w. General requirements for a parsing algorithm:

- Generality: the algorithm must be applicable to any grammar
- Completeness: the algorithm must produce *all* the results in case of ambiguity
- Efficiency
- Flexibility: a good algorithm can be easily modified

Parameters that define different parsing algorithms:

Orientation: Top-down vs. bottom-up vs. mixed

- Direction: Left-to-right vs. right-to-left vs. mixed (e.g., island-driven)
- Handling multiple choice: Dynamic programming vs. parallel processing vs. backtracking

Search: Breadth-first or Depth-first

Example: An example grammar

 $D \rightarrow the$ $N \rightarrow cat$ $N \rightarrow hat$ $P \rightarrow in$ $NP \rightarrow D N$ $PP \rightarrow P NP$ $NP \rightarrow NP PP$

Example sentences:

the cat in the hat the cat in the hat in the hat Assumptions:

- The grammar is given in Chomsky Normal Form: each rule is either of the form A → B C (where A, B, C are non-terminals) or of the form A → a (where a is a terminal).
- The string to recognize is $w = w_1 \cdots w_n$.
- A set of indices {0, 1, ..., n} is defined to point to positions between the input string's words:

0 the 1 cat 2 in 3 the 4 hat 5

The CYK algorithm

- Bottom-up, chart-based recognition algorithm for grammars in CNF
- To recognize a string of length *n*, uses a *chart*: a bi-dimensional matrix of size $n \times (n + 1)$
- Invariant: a non-terminal A is stored in the [i, j] entry of the chart iff A ⇒ w_{i+1} · · · w_j
- Consequently, the chart is triangular. A word *w* is recognized iff the start symbol *S* is in the [0, *n*] entry of the chart
- The idea: build all constituents up to the *i*-th position before constructing the *i* + 1 position; build smaller constituents before constructing larger ones.

```
for j := 1 to n do
for all rules A \rightarrow w_j do
chart[j-1,j] := chart[j-1,j] \cup \{A\}
for i := j-2 downto 0 do
for k := i+1 to j-1 do
for all B \in chart[i,k] do
for all C \in chart[k,j] do
for all rules A \rightarrow B C do
chart[i,j] := chart[i,j] \cup \{A\}
if S \in chart[0,n] then accept else reject
```

Example: The CYK algorithm

0 the 1 cat 2 in 3 the 4 hat 5

1 2 3 4 5 0 1 2 3 4 Extensions:

- Parsing in addition to recognition
- Support for ϵ -rules
- General context-free grammars (not just CNF)

- To provide a unified framework for discussing various parsing algorithms we use *parsing schemata*, which are generalized schemes for describing the principles behind specific parsing algorithms.
- This is a generalization of the *parsing as deduction* paradigm.
- A parsing schema consists of four components:
 - a set of items
 - a set of axioms
 - a set of deduction rules
 - a set of goal items

Given a grammar $G = \langle \Sigma, V, S, P \rangle$ and a string $w = w_1 \cdots w_n$: Items: [i, A, j] for $A \in V$ and $0 \le i, j \le n$ (state that $A \stackrel{*}{\Rightarrow} w_{i+1} \cdots w_j$) Axioms: [i, A, i + 1] when $A \rightarrow w_{i+1} \in P$ Goals: [0, S, n]Inference rules: [i, B, i] [i, C, k]

$$\frac{[i, B, j] \quad [j, C, \kappa]}{[i, A, \kappa]} \quad A \to B C$$

Example: Deduction example

$D \rightarrow the$	$NP \rightarrow D N$
N ightarrow cat	$PP \rightarrow P NP$
N ightarrow hat	$NP \rightarrow NP PP$
$P \rightarrow in$	

0 the 1 cat 2 in 3 the 4 hat 5

Example: Deduction example [3, D, 4] [4, N, 5][0, D, 1] [1, N, 2] [2, P, 3][0, NP, 2] [3, NP, 5] [2, PP, 5][0, NP, 5]

Items: $[\alpha \bullet, j]$ (state that $\alpha w_{j+1} \cdots w_n \stackrel{*}{\Rightarrow} w_1 \cdots w_n$) Axioms: $[\bullet, 0]$ Goals: $[S \bullet, n]$ Inference rules:

Shift $\frac{[\alpha \bullet, j]}{[\alpha w_{j+1} \bullet, j+1]}$ Reduce $\frac{[\alpha \gamma \bullet, j]}{[\alpha B \bullet, j]} \quad B \to \gamma$

Bottom-up deduction: example

Item form: $[\bullet\beta, j]$ (state that $S \stackrel{*}{\Rightarrow} w_1 \cdots w_j\beta$) Axioms: $[\bullet S, 0]$ Goals: $[\bullet, n]$ Inference rules:

Scan
$$\frac{[\bullet w_{j+1}\beta, J]}{[\bullet\beta, j+1]}$$
Predict
$$\frac{[\bullet\beta\beta, j]}{[\bullet\gamma\beta, j]} \quad B \to \gamma$$

Input: 0 the 1 cat 2 in 3 the 4 hat 5

Example: Top-down deduction	n
[• <i>NP</i> ,0] [• <i>NP PP</i> ,0] [• <i>D N PP</i> ,0] [• <i>the N PP</i> ,0] [• <i>N PP</i> ,1] [• <i>cat PP</i> ,1] [• <i>PP</i> ,2] [• <i>P NP</i> ,2] [• <i>in NP</i> ,2] [• <i>in NP</i> ,2] [• <i>in NP</i> ,2] [• <i>hNP</i> ,3] [• <i>D N</i> ,3] [• <i>the N</i> ,3] [• <i>hx</i> ,4] [• <i>hat</i> ,4]	axiom predict $NP \rightarrow NP PP$ predict $NP \rightarrow D N$ predict $D \rightarrow$ the scan predict $N \rightarrow$ cat scan predict $PP \rightarrow P NP$ predict $P \rightarrow$ in scan predict $NP \rightarrow D N$ predict $D \rightarrow$ the scan predict $D \rightarrow$ the scan predict $N \rightarrow$ hat
0,5	scan

if Parse(S,0) then accept else reject

Two inherent constraints:

- The root of the tree is S
- The yield of the tree is the input word

Example:

 $S \rightarrow NP VP$ $S \rightarrow Aux NP VP$ $S \rightarrow VP$ $VP \rightarrow Verb$ $NP \rightarrow Det Nominal Aux \rightarrow does$ $NP \rightarrow Proper-Noun$ Nominal \rightarrow Noun Nominal \rightarrow Nominal PP $PP \rightarrow Prep NP$

 $Det \rightarrow that | this | a$ Noun \rightarrow book |flight | meal *Verb* \rightarrow *book* | *include* | *includes* $Prep \rightarrow from \mid to \mid on$ $VP \rightarrow Verb NP$ Proper-Noun \rightarrow Houston | TWA

Example: Derivation tree S VPNP Nominal Verb Det Noun book that flight

An example derivation tree



An example derivation tree



When expanding the top-down search space, which local trees are created?

To reduce "blind" search, add bottom-up filtering. Observation: when trying to Parse(β , j), where $\beta = B\gamma$, the parser succeeds only if $B \stackrel{*}{\Rightarrow} w_{j+1}\beta$. Definition: A word w is a **left-corner** of a non-terminal B iff $B \stackrel{*}{\Rightarrow} w\beta$ for some β .

if Parse(S,0) then accept else reject

Even with bottom-up filtering, top-down parsing suffers from the following problems:

- Left recursive rules can cause non-termination: $NP \rightarrow NP PP$.
- Even when parsing terminates, it might take exponentially many steps.
- Constituents are computed over and over again









Reduplication:

Constituent	#
a flight	4
from Chicago	3
to Houston	2
on TWA	1
a flight from Chicago	3
a flight from Chicago to Houston	2
a flight from Chicago to Houston on TWA	1

When expanding the bottom-up search space, which local trees are created?

Bottom-up parsing suffers from the following problems:

- All possible analyses of every substring are generated, even when they can never lead to an *S*, or can never combine with their neighbors
- ϵ -rules can cause performance degradation
- Reduplication of effort

- Dynamic programming: partial results are stored in a chart
- Combines top-down predictions with bottom-up scanning
- No reduplication of computation
- Left-recursion is correctly handled
- ϵ -rules are handled correctly
- Worst-case complexity: $O(n^3)$

Basic concepts:

Dotted rules: if $A \to \alpha \beta$ is a grammar rule then $A \to \alpha \bullet \beta$ is a dotted rule.

- Edges: if $A \to \alpha \bullet \beta$ is a dotted rule and i, j are indices into the input string then $[i, A \to \alpha \bullet \beta, j]$ is an edge. An edge is **passive** (or **complete**) if $\beta = \epsilon$, **active** otherwise.
- Actions: The algorithm performs three operations: *scan*, *predict* and *complete*.

- scan: read an input word and add a corresponding complete edge to the chart.
- predict: when an active edge is added to the chart, predict all possible edges that can follow it
- complete: when a complete edge is added to the chart, combine it with appropriate active edges

rightsisters: given an active edge $A \to \alpha \bullet B\beta$, return all dotted rules $B \to \bullet \gamma$ leftsisters: given a complete edge $B \to \gamma \bullet$, return all dotted edges $A \to \alpha \bullet B\beta$

combination:

$$[i, A \to \alpha \bullet B\beta, k] * [k, B \to \gamma \bullet, j] = [i, A \to \alpha B \bullet \beta, j]$$

Item form:
$$[i, A \to \alpha \bullet \beta, j]$$
 (state that $S \stackrel{*}{\Rightarrow} w_1 \cdots w_i A \gamma$, and
also that $\alpha \stackrel{*}{\Rightarrow} w_{i+1} \cdots w_j$)
Axioms: $[0, S' \to \bullet S, 0]$
Goals: $[0, S' \to S \bullet, n]$

Inference rules:

Scan
$$[i, A \to \alpha \bullet w_{j+1}\beta, j]$$
 $[i, A \to \alpha w_{j+1} \bullet \beta, j+1]$ Predict $[i, A \to \alpha \bullet B\beta, j]$
 $[j, B \to \bullet \gamma, j]$ $B \to \gamma$ Complete $[i, A \to \alpha \bullet B\beta, k]$
 $[k, B \to \gamma \bullet, j]$

$$[i, A \to \alpha B \bullet \beta, j]$$

Parse ::
enteredge(
$$[0, S' \rightarrow \bullet S, 0]$$
)
for j := 1 to n do
for every rule $A \rightarrow w_j$ do
enteredge($[j-1, A \rightarrow w_j \bullet, j]$)

if $S' \to S \bullet \in C[0,n]$ then accept else reject

```
enteredge(i,edge,j) ::
    if edge ∉ C[i,j] then /* occurs check */
    C[i,j] := C[i,j] ∪ {edge}
    if edge is active then /* predict */
      for edge' ∈ rightsisters(edge) do
        enteredge([j,edge',j])
    if edge is passive then /* complete */
      for edge' ∈ leftsisters(edge) do
        for k such that edge' ∈ C[k,i] do
        enteredge([k,edge'*edge,j])
```