

Speech and Language Processing

SLP Chapter 13 (continued)

Parsing

Today

- Parsing with CFGs
 - CKY
 - Earley
- Nitin: CCG

Dynamic Programming

- DP search methods fill tables with partial results and thereby
 - Avoid doing avoidable repeated work
 - Solve exponential problems in polynomial time (well, no not really)
 - Efficiently store ambiguous structures with shared sub-parts.
- We'll cover two approaches that roughly correspond to top-down and bottom-up approaches.
 - CKY
 - Earley

Conversion to CNF

- Procedure:

- Copy all conforming rules unchanged
- Convert terminals w/in rules to dummy non-terminals
- Convert unit productions (this is a flattening operation!)
- Binarize rules

- Example:

- $S \rightarrow VP$
- $VP \rightarrow \text{Verb}_{\text{trans}} NP$ // Copy this over (it already conforms)
- $VP \rightarrow V NP PP$
- $\text{INF-VP} \rightarrow \text{to VP}$

- Rule 4 becomes 4' and 4'':

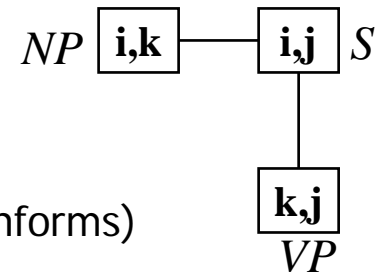
- $\text{INF-VP} \rightarrow \text{TO VP}$ // Convert terminals to dummy non-terminals
- $\text{TO} \rightarrow \text{to}$

- Rule 1 becomes 1' and 1'':

- $S \rightarrow \text{Verb}_{\text{trans}} NP$ // Convert unit productions
- $S \rightarrow V NP PP$

- Rules 3 and 1'' become:

- $S \rightarrow V \text{Double-Comp}$ // Binarize
- $VP \rightarrow V \text{Double-Comp}$
- $\text{Double-Comp} \rightarrow NP PP$



CYK Algorithm

- Classic example of dynamic programming paradigm
- Fill in upper-triangular matrix (chart) in bottom-up fashion using CNF grammar, where $\text{chart}(i,j) = \{A \mid A \rightarrow^* w_{i+1} \dots w_j\}$
- Note that cells include **sets** of non-terminals.

		TO:					
		1	2	3	4	5	6
FROM:	0	0-1	0-2	0-3	0-4	0-5	0-6
	1		1-2	1-3	1-4	1-5	1-6
	2			2-3	2-4	2-5	2-6
	3				3-4	3-5	3-6
	4					4-5	4-6
	5						5-6

CKY

- Build a table so that an A spanning from i to j in the input is placed in cell $[i, j]$ in the table.
- So a non-terminal spanning an entire string will sit in cell $[0, n]$
 - Hopefully an S
- If we build the table bottom-up, we'll know that the parts of the A must go from i to k and from k to j , for some k .

CKY

- Meaning that for a rule like $A \rightarrow B C$ we should look for a B in $[i,k]$ and a C in $[k,j]$.
- In other words, if we think there might be an A spanning i,j in the input... AND
 $A \rightarrow B C$ is a rule in the grammar THEN
- There must be a B in $[i,k]$ and a C in $[k,j]$ for some $i < k < j$

CKY

- So to fill the table loop over the cell $[i,j]$ values in some systematic way
 - What constraint should we put on that systematic search?
- For each cell, loop over the appropriate k values to search for things to add.

CKY Algorithm

```
function CKY-PARSE(words, grammar) returns table  
  
for  $j \leftarrow$  from 1 to LENGTH(words) do  
   $table[j - 1, j] \leftarrow \{A \mid A \rightarrow words[j] \in grammar\}$   
  for  $i \leftarrow$  from  $j - 2$  downto 0 do  
    for  $k \leftarrow i + 1$  to  $j - 1$  do  
       $table[i, j] \leftarrow table[i, j] \cup$   
         $\{A \mid A \rightarrow BC \in grammar,$   
           $B \in table[i, k],$   
           $C \in table[k, j]\}$ 
```

CKY Parsing

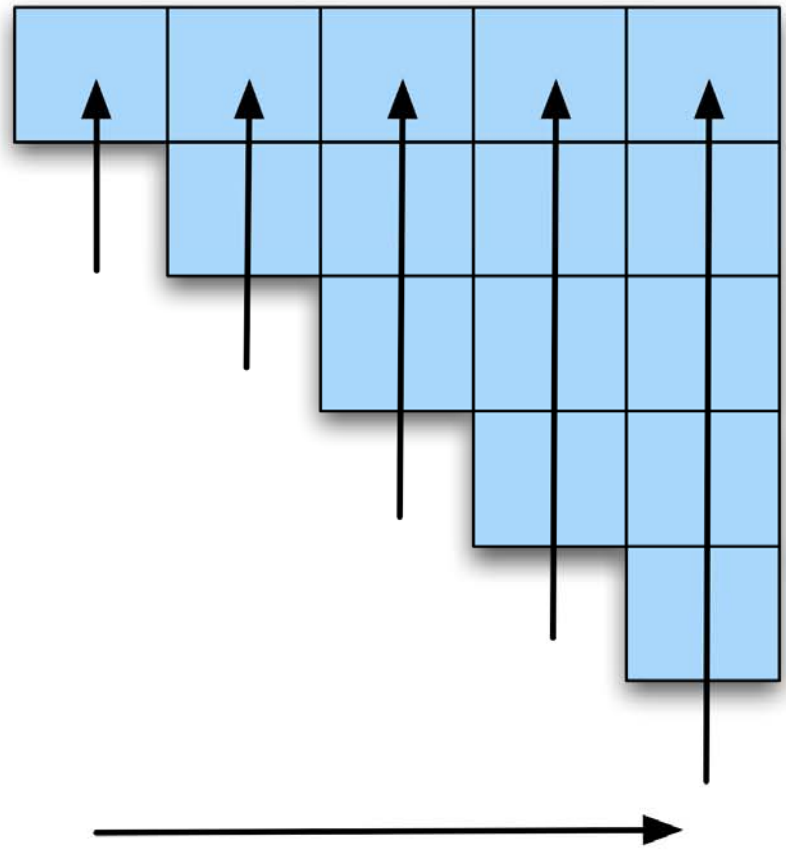
- Is that really a parser?

Note

- We arranged the loops to fill the table a column at a time, from left to right, bottom to top.
 - This assures us that whenever we're filling a cell, the parts needed to fill it are already in the table (to the left and below)
 - It's somewhat natural in that it processes the input a left to right a word at a time
 - Known as online

Example

<i>Book</i>	<i>the</i>	<i>flight</i>	<i>through</i>	<i>Houston</i>
S, VP, Verb Nominal, Noun [0,1]		S,VP,X2 [0,3]		S,VP,X2 [0,5]
	Det [1,2]	NP [1,3]		NP [1,5]
		Nominal, Noun [2,3]		Nominal [2,5]
			Prep [3,4]	PP [3,5]
				NP, Proper- Noun [4,5]



Example

<i>Book</i>	<i>the</i>	<i>flight</i>	<i>through</i>	<i>Houston</i>
S, VP, Verb, Nominal, Noun [0,1]	[0,2]	S,VP,X2 [0,3]	[0,4]	[0,5]
	Det [1,2]	NP [1,3]	[1,4]	[1,5]
		Nominal, Noun [2,3]	[2,4]	Nominal [2,5]
			Prep [3,4]	[3,5]
				NP, Proper- Noun [4,5]

Filling column 5

Example

<i>Book</i>	<i>the</i>	<i>flight</i>	<i>through</i>	<i>Houston</i>
S, VP, Verb, Nominal, Noun [0,1]	[0,2]	S,VP,X2 [0,3]	[0,4]	[0,5]
	Det [1,2]	NP [1,3]	[1,4]	NP [1,5]
		Nominal, Noun [2,3]	[2,4]	[2,5]
			Prep ← [3,4]	PP [3,5]
				NP, Proper- Noun [4,5]

Example

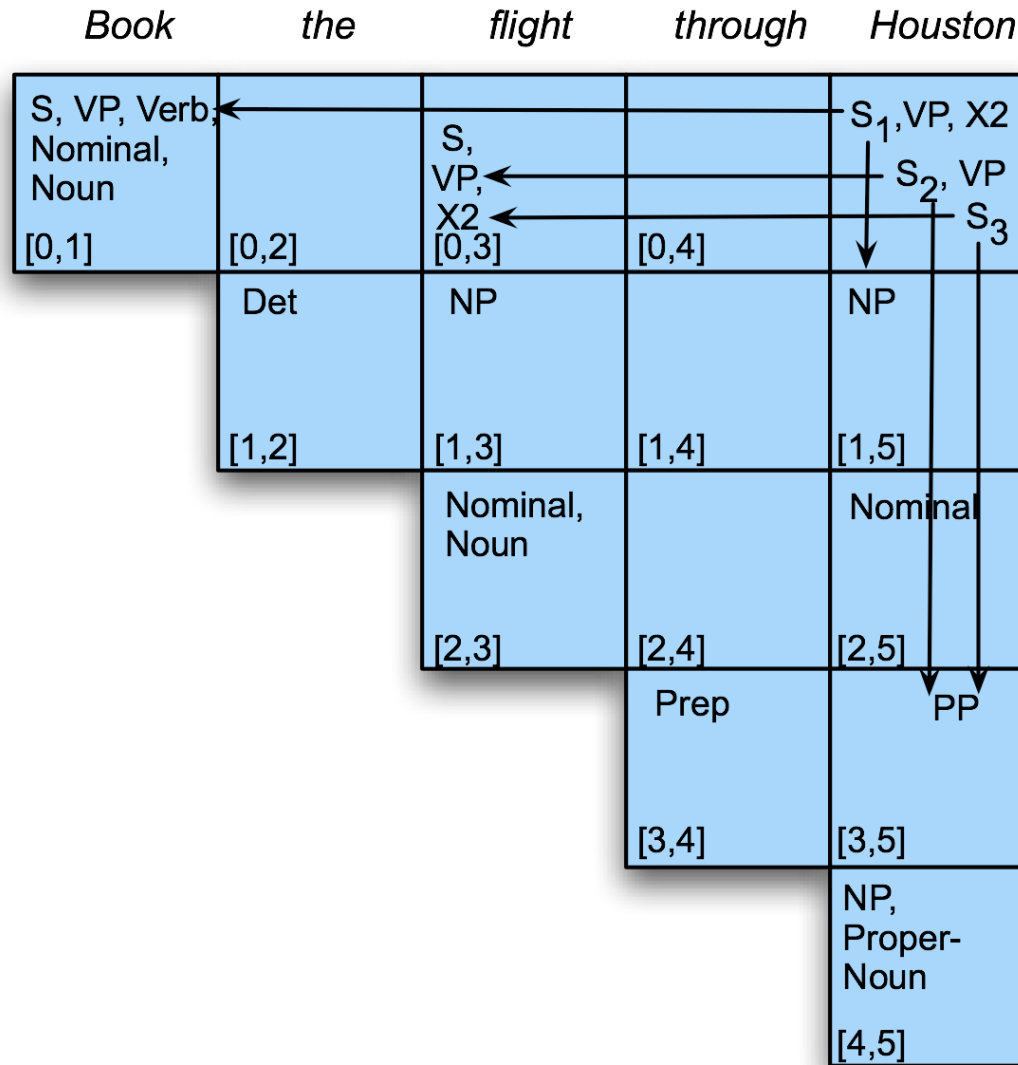
<i>Book</i>	<i>the</i>	<i>flight</i>	<i>through</i>	<i>Houston</i>
S, VP, Verb, Nominal, Noun [0,1]		S,VP,X2 [0,3]		
	Det [1,2]	NP [1,3]		NP [1,5]
		Nominal, Noun [2,3]		Nominal [2,5]
			Prep [3,4]	PP [3,5]
				NP, Proper- Noun [4,5]

Diagram illustrating the parse tree structure for the sentence "Book the flight through Houston". The table shows the words and their corresponding syntactic categories and spans. Arrows indicate dependencies: a horizontal arrow from the "Nominal" in the "flight" cell to the "Nominal" in the "Houston" cell, and a vertical arrow from the "Nominal" in the "Houston" cell to the "PP" in the "through" cell.

Example

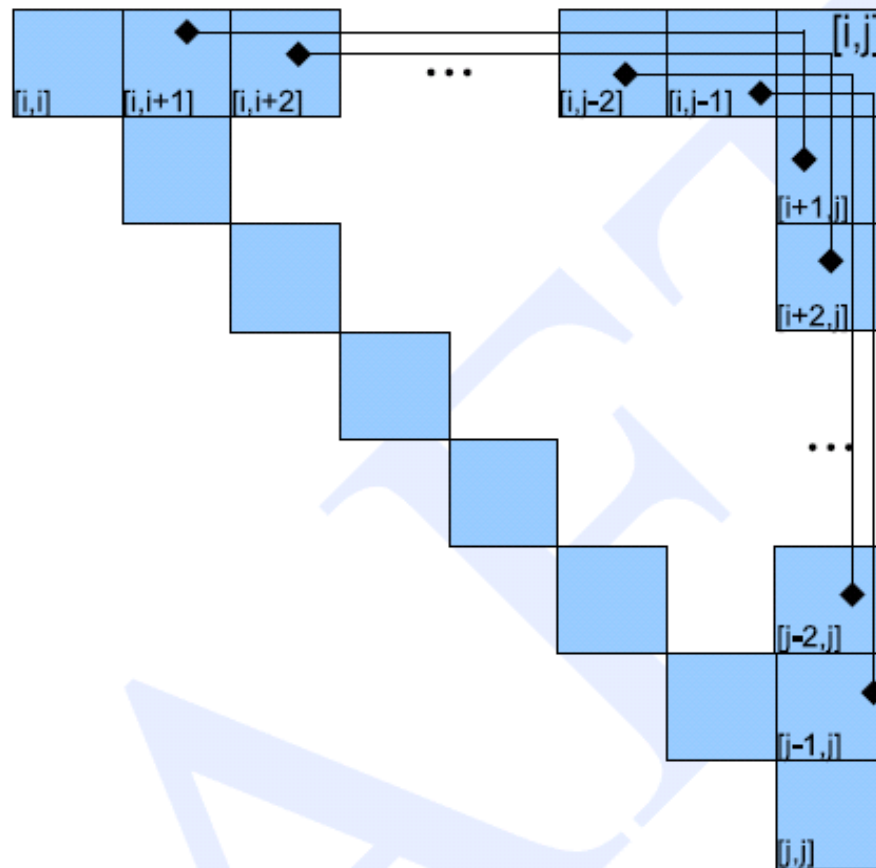
<i>Book</i>	<i>the</i>	<i>flight</i>	<i>through</i>	<i>Houston</i>
S, VP, Verb, Nominal, Noun [0,1]	[0,2]	S,VP,X2 [0,3]	[0,4]	[0,5]
	Det ←	NP ←		NP
	[1,2]	[1,3]	[1,4]	[1,5]
		Nominal, Noun		Nominal
		[2,3]	[2,4]	[2,5]
			Prep	PP
			[3,4]	[3,5]
				NP, Proper- Noun
				[4,5]

Example



How does Dynamic Programming Provide Savings?

- Multiple ways to fill cells with the same non-terminal—but only **one** copy is included in the cell.



This shows all the ways to fill the $[i, j]$ cell

Disadvantages of CYK

- Primary disadvantage: Conversion to CNF.
 - Leads to “weak equivalence” w.r.t. original grammar
 - Introduces non-linguistically relevant categories e.g., X₂
 - Complicates semantic analysis
 - Adds a level of “hidden” complexity. Parse time is Gn^3 (where G is size of grammar) but conversion to CNF throws in another factor of G : G^2n^3 — same as Earley!
- Since it's bottom up, CKY populates the table with a lot of phantom constituents.
 - Segments that by themselves are constituents but cannot really occur in the context in which they are being suggested.
 - To avoid this we can switch to a top-down control strategy
 - Or we can add some kind of filtering that blocks constituents where they can not happen in a final analysis.
- Is there a parsing approach that adopts dynamic programming and accepts arbitrary CFGs?

Earley Parsing

- Allows arbitrary CFGs
- Top-down control
- Fills a table in a single sweep over the input
 - Table is length $N+1$; N is number of words
 - Table entries represent
 - Completed constituents and their locations
 - In-progress constituents
 - Predicted constituents

States

- The table-entries are called states and are represented with **dotted-rules**.

$S \rightarrow \cdot VP$

A VP is predicted

$NP \rightarrow Det \cdot Nominal$

An NP is in progress

$VP \rightarrow V NP \cdot$

A VP has been found

States/Locations

- $S \rightarrow \bullet VP$ [0,0]
 - A VP is predicted at the start of the sentence
- $NP \rightarrow Det \bullet Nominal$ [1,2]
 - An NP is in progress; the Det goes from 1 to 2
- $VP \rightarrow V NP \bullet$ [0,3]
 - A VP has been found starting at 0 and ending at 3

Earley

- As with most dynamic programming approaches, the answer is found by looking in the table in the right place.
- In this case, there should be an S state in the final column that spans from 0 to N and is complete. That is,
 - $S \rightarrow \alpha \bullet [0, N]$
- If that's the case you're done.

Earley

- So sweep through the table from 0 to N...
 - New predicted states are created by starting top-down from S
 - New incomplete states are created by advancing existing states as new constituents are discovered
 - New complete states are created in the same way.

Earley

- More specifically...
 1. *Predict* all the states you can upfront
 2. Read a word
 1. Extend states based on matches
 2. Generate new predictions
 3. Go to step 2
 3. When you're out of words, look at the chart to see if you have a winner

Core Earley Code

function EARLEY-PARSE(*words*, *grammar*) **returns** *chart*

ENQUEUE($(\gamma \rightarrow \bullet S, [0, 0])$, *chart*[0])

for $i \leftarrow$ **from** 0 **to** LENGTH(*words*) **do**

for each *state* **in** *chart*[*i*] **do**

if INCOMPLETE?(*state*) **and**

 NEXT-CAT(*state*) is not a part of speech **then**

 PREDICTOR(*state*)

elseif INCOMPLETE?(*state*) **and**

 NEXT-CAT(*state*) is a part of speech **then**

 SCANNER(*state*)

else

 COMPLETER(*state*)

end

end

return(*chart*)

Earley Code

```
procedure PREDICTOR( $(A \rightarrow \alpha \bullet B \beta, [i, j])$ )  
  for each  $(B \rightarrow \gamma)$  in GRAMMAR-RULES-FOR( $B, grammar$ ) do  
    ENQUEUE( $(B \rightarrow \bullet \gamma, [j, j])$ ,  $chart[j]$ )  
end
```

```
procedure SCANNER( $(A \rightarrow \alpha \bullet B \beta, [i, j])$ )  
  if  $B \subset PARTS-OF-SPEECH(word[j])$  then  
    ENQUEUE( $(B \rightarrow word[j], [j, j+1])$ ,  $chart[j+1]$ )
```

```
procedure COMPLETER( $(B \rightarrow \gamma \bullet, [j, k])$ )  
  for each  $(A \rightarrow \alpha \bullet B \beta, [i, j])$  in  $chart[j]$  do  
    ENQUEUE( $(A \rightarrow \alpha B \bullet \beta, [i, k])$ ,  $chart[k]$ )  
end
```

Example

- Book that flight
- We should find... an S from 0 to 3 that is a completed state...

Chart [0]

S0	$\gamma \rightarrow \bullet S$	[0,0]	Dummy start state
S1	$S \rightarrow \bullet NP VP$	[0,0]	Predictor
S2	$S \rightarrow \bullet Aux NP VP$	[0,0]	Predictor
S3	$S \rightarrow \bullet VP$	[0,0]	Predictor
S4	$NP \rightarrow \bullet Pronoun$	[0,0]	Predictor
S5	$NP \rightarrow \bullet Proper-Noun$	[0,0]	Predictor
S6	$NP \rightarrow \bullet Det Nominal$	[0,0]	Predictor
S7	$VP \rightarrow \bullet Verb$	[0,0]	Predictor
S8	$VP \rightarrow \bullet Verb NP$	[0,0]	Predictor
S9	$VP \rightarrow \bullet Verb NP PP$	[0,0]	Predictor
S10	$VP \rightarrow \bullet Verb PP$	[0,0]	Predictor
S11	$VP \rightarrow \bullet VP PP$	[0,0]	Predictor

Note that given a grammar, these entries are the same for all inputs; they can be pre-loaded.

Chart [1]

S12	<i>Verb</i> → <i>book</i> •	[0,1]	Scanner
S13	<i>VP</i> → <i>Verb</i> •	[0,1]	Completer
S14	<i>VP</i> → <i>Verb</i> • <i>NP</i>	[0,1]	Completer
S15	<i>VP</i> → <i>Verb</i> • <i>NP PP</i>	[0,1]	Completer
S16	<i>VP</i> → <i>Verb</i> • <i>PP</i>	[0,1]	Completer
S17	<i>S</i> → <i>VP</i> •	[0,1]	Completer
S18	<i>VP</i> → <i>VP</i> • <i>PP</i>	[0,1]	Completer
S19	<i>NP</i> → • <i>Pronoun</i>	[1,1]	Predictor
S20	<i>NP</i> → • <i>Proper-Noun</i>	[1,1]	Predictor
S21	<i>NP</i> → • <i>Det Nominal</i>	[1,1]	Predictor
S22	<i>PP</i> → • <i>Prep NP</i>	[1,1]	Predictor

Charts[2] and [3]

S23	<i>Det</i> → <i>that</i> •	[1,2]	Scanner
S24	<i>NP</i> → <i>Det</i> • <i>Nominal</i>	[1,2]	Completer
S25	<i>Nominal</i> → • <i>Noun</i>	[2,2]	Predictor
S26	<i>Nominal</i> → • <i>Nominal Noun</i>	[2,2]	Predictor
S27	<i>Nominal</i> → • <i>Nominal PP</i>	[2,2]	Predictor
S28	<i>Noun</i> → <i>flight</i> •	[2,3]	Scanner
S29	<i>Nominal</i> → <i>Noun</i> •	[2,3]	Completer
S30	<i>NP</i> → <i>Det Nominal</i> •	[1,3]	Completer
S31	<i>Nominal</i> → <i>Nominal</i> • <i>Noun</i>	[2,3]	Completer
S32	<i>Nominal</i> → <i>Nominal</i> • <i>PP</i>	[2,3]	Completer
S33	<i>VP</i> → <i>Verb NP</i> •	[0,3]	Completer
S34	<i>VP</i> → <i>Verb NP</i> • <i>PP</i>	[0,3]	Completer
S35	<i>PP</i> → • <i>Prep NP</i>	[3,3]	Predictor
S36	<i>S</i> → <i>VP</i> •	[0,3]	Completer
S37	<i>VP</i> → <i>VP</i> • <i>PP</i>	[0,3]	Completer

Details

- As with CKY that isn't a parser until we add the backpointers so that each state knows where it came from.

Efficiency

- For such a simple example, there seems to be a lot of useless stuff in there.
- Why?
 - It's predicting things that aren't consistent with the input
 - That's the flipside to the CKY problem.

Back to Ambiguity

- Did we solve it?

Ambiguity

- No...
 - Both CKY and Earley will result in multiple **S** structures for the **[0,N]** table entry.
 - They both efficiently store the sub-parts that are shared between multiple parses.
 - And they obviously avoid re-deriving those sub-parts.
 - But neither can tell us which one is right.

Ambiguity

- In most cases, humans don't notice incidental ambiguity (lexical or syntactic). It is resolved on the fly and never noticed.
- Can be modeled with probabilities—something that would be covered in an advanced NLP course.