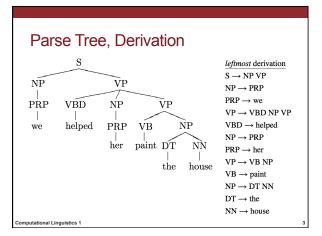
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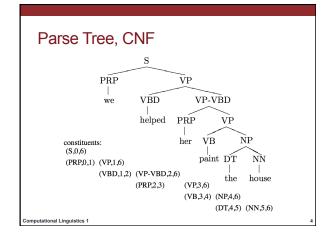
NUNERSITL OR BARRYLAND Kristy Hollingshead Seitz Institute for Advanced Computer Studies University of Maryland Lecture 17: 1 November 2011

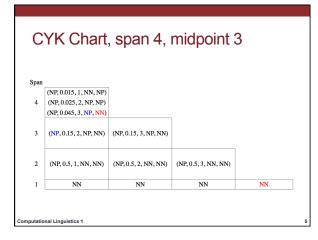


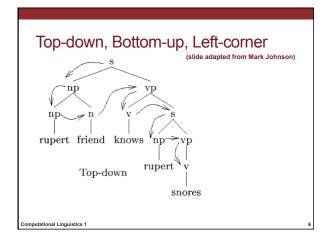
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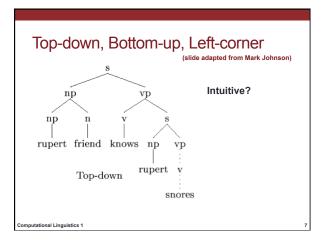
Context-sensitive grammar formalisms?

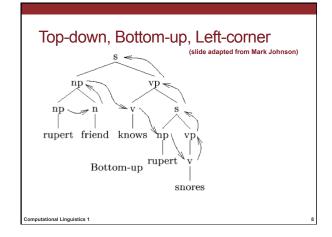


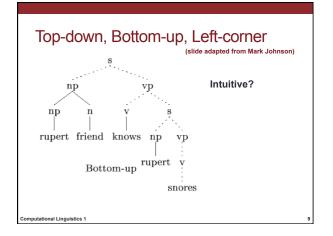


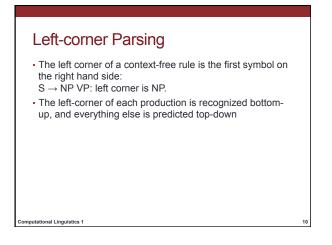


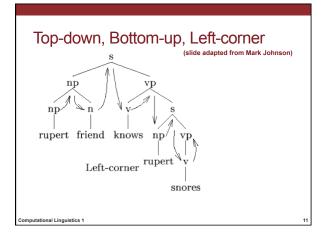


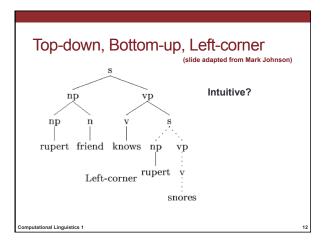


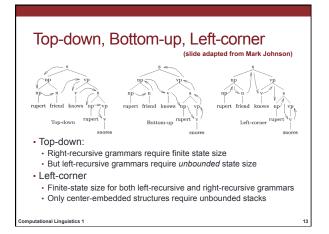






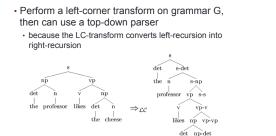




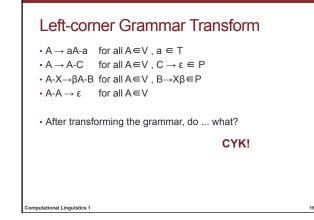


Top down,	Bottom-up,	Let	t-corr	iei
	grammars require finit e grammars require <i>u</i>			ze
 Left-corner 				
 Finite-state size 	for both left-recursive	and rig	ht-recursiv	/e gram
 Only center-em 	bedded structures req	uire unb	ounded et	ka alka
		ao ao	ounded 5	lacks
	es human behavior!		ounded 5	lacks
			ace requi	
which emulate	es human behavior!			
which emulate From	es human behavior!	Sp	ace requi	red
which emulate From	es human behavior! Strategy	Sp <i>Left</i>	ace requi <i>Center</i>	red <i>Right</i>
which emulate From	Strategy Top-down	Sp Left O(n)	ace requi Center O(n)	red <i>Right</i> <i>O</i> (1)

Building a Left-corner Parser?



the n np-np



Agenda

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- HW4, due Thursday
- Parsing algorithms
- Left-corner grammar transform
- Earley parsing
- Context-sensitive grammar formalisms

CKY: Analysis

- Since it's bottom up, CKY populates the table with a lot of "phantom constituents"
 - · Spans that are constituents, but cannot really occur in the context in which they are suggested
- Conversion of grammar to CNF adds additional nonterminal nodes
- · Leads to weak equivalence wrt original grammar
- Additional terminal nodes not (linguistically) meaningful: but can be cleaned up with post processing
- Is there a parsing algorithm for arbitrary CFGs that combines dynamic programming and top-down control?

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Earley Parsing Algorithm

- One advantage of top-down over bottom-up is that one never builds constituents that cannot be rooted
- · Earley parsing motivation
- · Only want to build categories that can be rooted
- · Use a top-down filter
- Use a chart parsing approach
- Dynamic programming algorithm (surprise)
- Allows arbitrary CFGs
- · Fills a chart in a single sweep over the input

Earley Parsing: Chart, States

- Chart is an array of length N + 1, where N = number of words
- · Chart entries represent states:
- · Completed constituents and their locations
- · In-progress constituents
- · Predicted constituents
- · Each state contains three items of information:
- A grammar rule
- Information about progress made in completing the sub-tree represented by the rule
- Span of the sub-tree

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Chart Entries: State Examples

• S \rightarrow • VP [0,0]

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- A VP is predicted at the start of the sentence
- NP → Det Nominal [1,2]
- An NP is in progress; the Det goes from 1 to 2
- VP \rightarrow V NP [0,3]

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A VP has been found starting at 0 and ending at 3

Earley in a nutshell

- · Start by predicting S
- Step through chart:
- New predicted states are created from current states
 New incomplete states are created by advancing existing states as
- new constituents are discovered
- States are completed when rules are satisfied
- Termination: look for S $\rightarrow \alpha \bullet [0, N]$

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Earley Algorithm function EARLEY-PARSE(words, grammar) returns chart ENQUEUE((γ → • S, [0,0]), chart[0]) for i − 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) else COMPLETER(state) else COMPLETER(state) end

COMPLE end end return(chart) Computational Linguistics 1

Earley Algorithm

```
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
```

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Earley Example

Input: Book that flight

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• Desired end state: $S \rightarrow \alpha \cdot [0,3]$

 Meaning: S spanning from 0 to 3, completed rule 	
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Ea	arley: 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 gramm same for all inputs; they		

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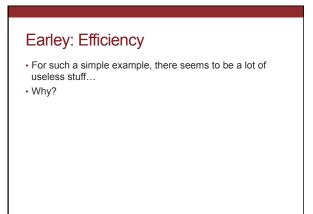
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S12	$Verb \rightarrow book \bullet$	[0,1]	Scanner
S13	$VP \rightarrow Verb \bullet$	[0,1]	Completer
S14	$VP \rightarrow Verb \bullet NP$	[0,1]	Completer
S15	$VP \rightarrow Verb \bullet NP PP$	[0,1]	Completer
S16	$VP \rightarrow Verb \bullet PP$	[0,1]	Completer
S17	$S \rightarrow VP \bullet$	[0,1]	Completer
S18	$VP \rightarrow VP \bullet PP$	[0,1]	Completer
S19	$NP \rightarrow \bullet Pronoun$	[1,1]	Predictor
S20	$NP \rightarrow \bullet Proper-Noun$	[1,1]	Predictor
S21	$NP \rightarrow \bullet Det Nominal$	[1,1]	Predictor
S22	$PP \rightarrow \bullet Prep NP$	[1,1]	Predictor

Ear	ley: Chart[2] and	Chart[3]	
S23	$Det \rightarrow that \bullet$	[1,2]	Scanner
S24	$NP \rightarrow Det \bullet Nominal$	[1,2]	Completer
S25	Nominal $\rightarrow \bullet$ Noun	[2,2]	Predictor
S26	$Nominal \rightarrow \bullet Nominal Noun$	[2,2]	Predictor
S27	Nominal $\rightarrow \bullet$ Nominal PP	[2,2]	Predictor
S28	Noun \rightarrow flight •	[2,3]	Scanner
S29	Nominal \rightarrow Noun •	[2,3]	Completer
S30	$NP \rightarrow Det Nominal \bullet$	[1,3]	Completer
S31	$Nominal \rightarrow Nominal \bullet Noun$	[2,3]	Completer
S32	Nominal \rightarrow Nominal \bullet PP	[2,3]	Completer
S33	$VP \rightarrow Verb NP \bullet$	[0,3]	Completer
S34	$VP \rightarrow Verb NP \bullet PP$	[0,3]	Completer
S35	$PP \rightarrow \bullet Prep NP$	[3,3]	Predictor
S36	$S \rightarrow VP \bullet$	[0,3]	Completer
S37	$VP \rightarrow VP \bullet PP$	[0,3]	Completer
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	,	backpointers		
Chart[1]	S12	$Verb \rightarrow book \bullet$	[0,1]	Scanner
Chart[2]	S23	Det ightarrow that ullet	[1,2]	Scanner
Chart[3]	S28	$Noun \rightarrow flight \bullet$	[2,3]	Scanner
	S29	$Nominal \rightarrow Noun \bullet$	[2,3]	(S28)
	S30	$NP \rightarrow Det Nominal \bullet$	[1,3]	(S23, S29)
	\$33	$VP \rightarrow Verb NP \bullet$	[0,3]	(\$12, \$30)
	S36	$S \rightarrow VP \bullet$	[0,3]	(\$33)



Back to Ambiguity

- Did we solve it?
- No: both CKY and Earley return multiple parse trees...
 Plus: compact encoding with shared sub-trees
- Plus: work deriving shared sub-trees is reused
- Minus: neither algorithm tells us which parse is correct

Ambiguity

- Why don't humans usually encounter ambiguity?
- · How can we improve our models?

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Agenda: Summary

- HW4, due Thursday
- Parsing algorithms
 - Earley parsing
 - Left-corner grammar transform
- Next time: context-sensitive grammar formalisms

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