

## Probability estimation used in Tree Tagger: Suffix Lexicon

- ◆ Probability of unknown words are estimated according to its suffix,  $p(t | w)$

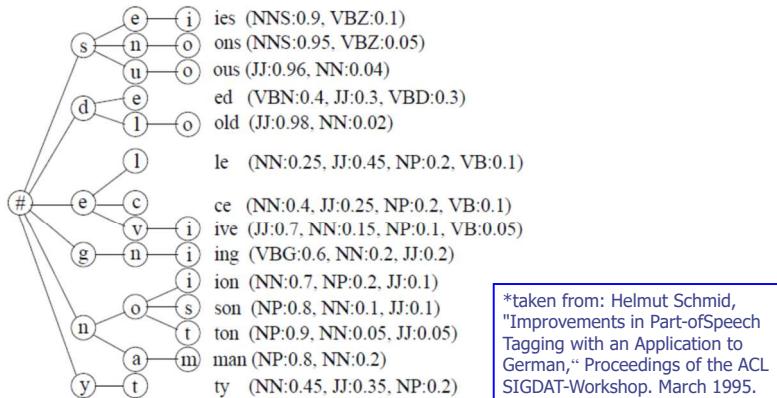


Figure 2: A sample suffix tree of maximal length 3.

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## Contents of this lecture

- ◆ Morphological Analysis
  - Tokenization
  - Part-of-speech tagging
- ◆ Syntactic Parsing Algorithms
  - Phrase structure parsing
  - Word dependency parsing

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## Estimation of the probabilities 確率値の推定法

- ◆ Unsupervised learning (no tagged data)
  - Baum-Welch algorithm (Hidden Markov Model): 隠れマルコフモデル
- ◆ Supervised learning (with tagged data)
  - Maximum likelihood(最尤推定)
  - Decision Trees(決定木)
  - Maximum Entropy model(最大エントロピー法)
  - Conditional Random Fields(CRF: 条件付き確率場)
  - Recurrent Neural Networks(再帰型NN) + CRF

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## Formal Grammars(形式文法)

- ◆  $G=(N, T, P, S)$ 
  - $N$ : a finite set of nonterminal symbols (phrases)
  - $T$ : a finite set of terminal symbols (words)
  - $P$ : a finite set of grammar rules
  - $S$ : start symbol ( $S \in N$ )
- ◆ Grammar rule:  $\alpha \rightarrow \beta$  (文法規則の形)
  - Context free grammar(文脈自由文法):  
 $A \rightarrow \beta$  ( $\beta \in (N \cup T)^*$ )
  - Regular grammar(正規文法):  
 $A \rightarrow B$  or  $A \rightarrow aB$   
( $A, B \in N$ ,  $a \in T$ )

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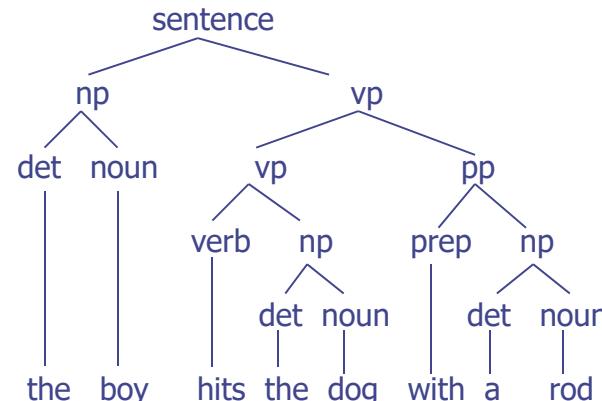
## A sample context-free grammar

- (a) sentence → np, vp
- (b) np → det, noun
- (c) vp → verb, np
- (d) vp → vp, pp
- (e) pp → prep, np
- (f) det → [the] | [a]
- (g) noun → [boy] | [dog] | [rod]
- (h) verb → [hits]
- (i) prep → [with]

Note:  
• [the] means “the” is  
a terminal symbol  
• “|” means “or”

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## A Phrase Structure Tree 句構造木



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## Parsing Algorithms 統語解析アルゴリズム

### Context free grammar parsing

Concurrent

- CKY
- Chart Parsing
- Bottom-up
- Top-down
- = Earley parsing

Backtracking

- Top-down
- Bottom-up
- (Shift-reduce)
- Left-corner

Generalized LR parsing  
(Tomita algorithm)

### Dependency parsing

Deterministic      Concurrent

- Transition-based
- (Shift-reduce)
- CKY
- (Eisner algorithm)

### Global optimization

MST (Maximum Spanning Tree algorithm)

ILP (Integer Linear Programming)

## Concurrent Parsing Algorithm

### ◆ CKY(Cocke-Kasami-Younger) algorithm

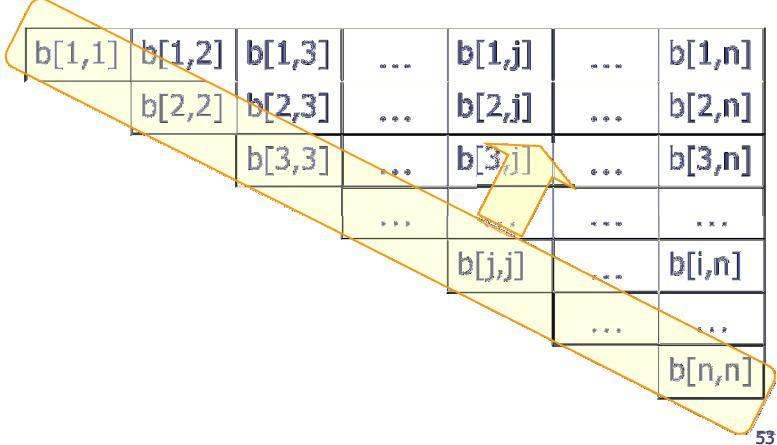
- Assumes grammar rules to be in Chomsky Normal Form (right-hand side of rules have no more than two children)
  - ◆ Grammar rules:  $A \rightarrow B, C$  or  $A \rightarrow [a]$
- CKY algorithm is an application of bottom-up dynamic programming to grammar rules
  - ◆ Box  $b[i,j]$  ( $1 \leq i \leq j \leq n$ ) is defined as the set of phrases that span between  $i$ -th and  $j$ -th words of the input sentence (length  $n$ )
  - ◆ Construction of  $b[i,j]$  by bottom-up DP method:  
 $b[i,j] = \{A \mid \text{rule } A \rightarrow B, C \text{ exists, and}$   
 $B \in b[i,k] \wedge C \in b[k+1,j] \text{ } (i \leq k \leq j-1)\}$

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## CKY Table

- Table construction goes toward right-upwards



## CKY Table

	b[i,i]	b[i,i+1]	...	b[i,j-1]	b[i,j]	
		...		b[i+1,j]		b[i+1,j]
			...			b[i+1,j]
					b[j,j]	

$b[i,j] = \{A \mid \text{rule } A \rightarrow B,C \text{ exists,}$   
 and  
 $B \in b[i,k] \wedge C \in b[k+1,j]$   
 $(i \leq k \leq j-1) \}$

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## CKY Table

	b[i,i]	b[i,i+1]	...	b[i,j-1]	b[i,j]	
		...		b[i+1,j]		b[i+1,j]
			...	b[i+2,j]		b[i+2,j]
				b[j,j]		

$$b[i,i] \cup b[i+1,j] \Rightarrow b[i,j]$$

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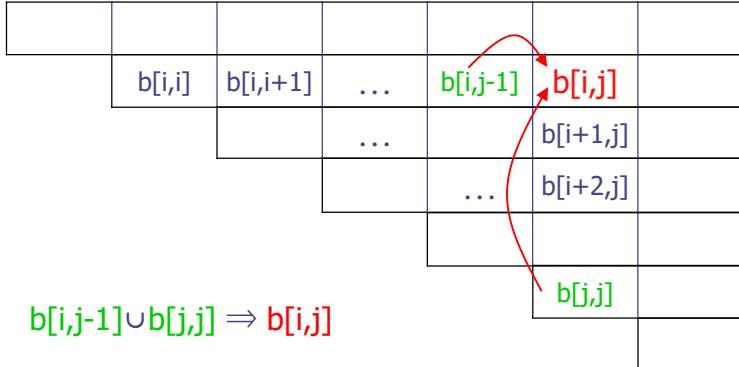
## CKY Table

	b[i,i]	b[i,i+1]	...	b[i,j-1]	b[i,j]	
		...		b[i+1,j]		b[i+1,j]
			...		b[i+2,j]	b[i+2,j]
				b[j,j]		

$$b[i,i+1] \cup b[i+2,j] \Rightarrow b[i,j]$$

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## CKY Table



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## Sample Process of CKY parsing algorithm

input sentence: The boy hits the dog with a rod.

Grammar rules:

sentence  $\rightarrow$  np, vp  
np  $\rightarrow$  det, noun  
vp  $\rightarrow$  verb, np  
vp  $\rightarrow$  vp, pp  
pp  $\rightarrow$  prep, np  
det  $\rightarrow$  [the] | [a]  
noun  $\rightarrow$  [boy] | [dog] | [rod] | [hits]  
verb  $\rightarrow$  [hits]  
prep  $\rightarrow$  [with]

[→ CKY.ppt](#)

CKY parsing algorithm requires  $O(n^3)$  processing time, while the backtracking algorithms needs  $O(c^n)$  time in worst case.  
Time complexity of Chart parsing is  $O(n^3)$

## Chart Parsing

- ◆ An extension of CKY parsing to a general context-free grammars
  - CKY parsing handles grammar rules of Chomsky normal form (right-hand side of grammar rule has at most 2 non-terminal symbols)
  - General rules can have 3 or more non-terminal symbols on the right-hand side  
⇒ introduce special notation indicating which part has been analyzed

Example: For a rule like  $vp \rightarrow \text{verb } np \text{ pp}$   
the following notations are used ("·" shows until which part has been analyzed):

$vp \rightarrow \text{verb} \cdot np \text{ pp}, \quad vp \rightarrow \text{verb } np \cdot pp$

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## Bottom-up Chart Parsing

- ◆ Mainly consisting of two operations:
  - Invocation (of a new rule): If B is in  $b[i,j]$ , and there are rules like  $A \rightarrow B C \dots D$  having B as the first symbol on the right-hand, then for each rule add a dotted rule like  $A \rightarrow B \cdot C \dots D$ , in  $b[i,j]$
  - Update (of a dotted rule): If there is a dotted rule,  $A \rightarrow B \dots \cdot C D \dots E$ , in  $b[i,j]$ , and there is a completed symbol C in  $b[j+1,k]$ , then add  $A \rightarrow B \dots C \cdot D \dots E$  in  $b[i,k]$ .  
If the dotted rule included in  $b[i,j]$  is  $A \rightarrow B \dots \cdot C$  (having C as the final symbol), add a completed symbol A in  $b[i,k]$ .

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## Bottom-up Chart Parsing for “the boy hits the dog with a rod” (Initial state)

det							
the	n						
boy	verb						
hits	det						
	the	n					
	dog	prep					
	with	det					
	a	n					
(1) sentence → np vp							
(2) np → det n							
(3) vp → verb np							
(4) vp → verb np pp							
(5) pp → prep, np							

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## Steps of Bottom-up Chart Parsing

det	np
np → det · n	
the	n

boy

(2) np → det n

## Bottom-up Chart Parsing for “the boy hits the dog with a rod”

det	np			s			s
np → det · n	s → np · vp						
the	n						
boy	v			vp			
	vp → v · np			vp → v np · pp			
	vp → v · np pp						
hits	det	np					vp
	np → det · n	s → np · vp					
the	n						
dog	prep						pp
	pp → prep · np						
with	det	np					
	np → det · n	s → np · vp					
a	n						
rod							
(1) s → np vp							
(2) np → det n							
(3) vp → v np							
(4) vp → v np pp							
(5) pp → prep, np							

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