

CS 378 Lecture 12

Constituency Parsing

Announcements

- Midterm: Weds - Sat of next week
- A3 due Tues
- A2 back soon

Recap

Viterbi: dynamic programming alg to
compute $\max_{\bar{y}} P(\bar{y}, \bar{x})$ for an HMM
(similar alg today for parsing: CKY)

Beam search: approximates Viterbi by
only preserving K states at a timestep

Today Constituency intro
Probabilistic context-free grammars
CKY algorithm

Constituency see slides

Context-free grammars

{ N	T	S	R }
nonterminals	terminals	↓	↓
S, NP, VP, PP	~ words	start	rules
DT, NN, VPD, }	the	symbol	
IN, NNS }	children	S	
preterminals ←	ate		
	cake		
	with		
	spoon		

binary

$S \rightarrow NP VP \quad 1.0$

$VP \rightarrow VBD \ NP \quad 1.0$

$NP \rightarrow DT \ NN \quad 0.5$

$NP \rightarrow DT \ NNS \quad 0.5$

unary

$DT \rightarrow the \quad 1.0$

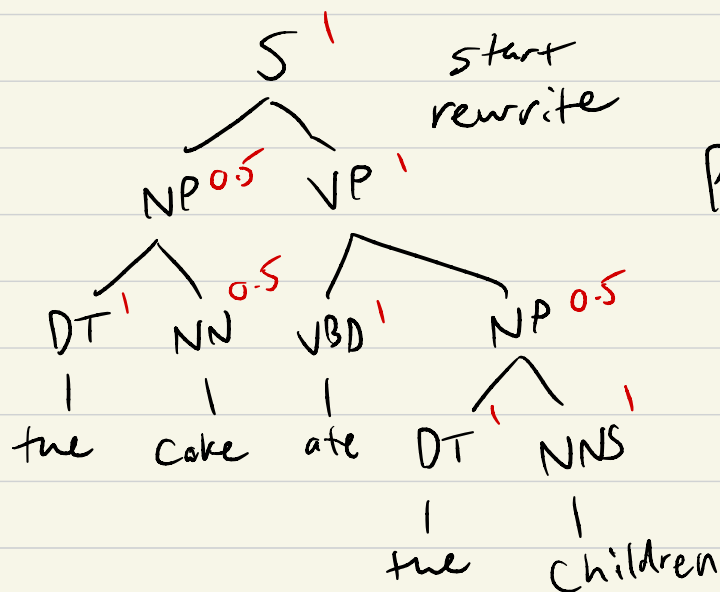
$NNS \rightarrow children \quad 1.0$

$NN \rightarrow cake \quad 0.5$

$NN \rightarrow spoon \quad 0.5$

$VBD \rightarrow ate \quad 1.0$

CFG defines a set of trees



$$P(\text{tree}) = \frac{1}{8}$$

Probabilistic CFGs

Each rule has a prob.

Probs normalize per parent

$$P(\text{rule} | \text{NP}) = \begin{cases} \text{NP} \rightarrow \text{DT NN} & 0.5 \\ \text{NP} \rightarrow \text{DT NNS} & 0.5 \end{cases}$$

($\sim$ HMM transitions)

$$P(T) = \prod_{\text{rules}} P(\text{rule} | \text{parent}(\text{rule}))$$

Building a parser

Input: treebank (sentences w/ labeled trees)

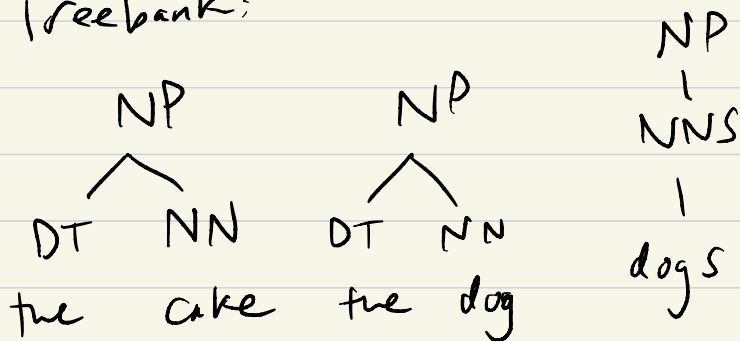
Output: model + way to compute $\arg\max_T P(T|\bar{x})$

- ① Grammar preprocessing (binarization) next time
- ② Computing grammar + getting probs
- ③ Parsing

① Make every tree binary

② Read off grammar + count rules

Treebank:



$$P(w | NN) = \begin{cases} 1/2 & \text{cake} \\ 1/2 & \text{dog} \end{cases}$$

$$P(w | NNS) = \begin{cases} 1.0 & \text{dogs} \end{cases}$$

$$P(r | NP) = \begin{cases} 2/3 & \text{DT NN} \\ 1/3 & \text{NNS} \end{cases}$$

This is the maximum likelihood estimate of params for this data

③ CKY algorithm

Input: PCFG
sentence $\bar{x}$

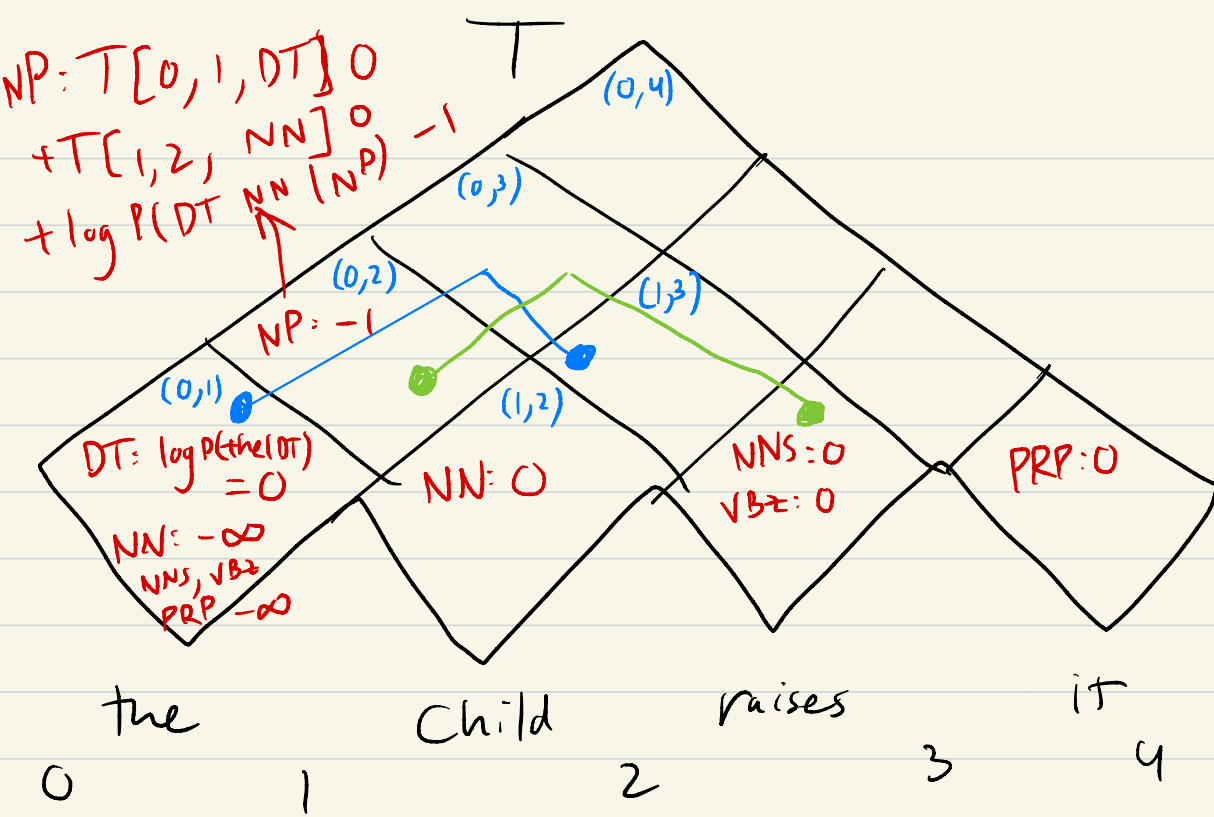
Output: $\underset{T}{\operatorname{argmax}} P(T|\bar{x})$
most likely tree for $\bar{x}$
 $= \underset{T}{\operatorname{argmax}} P(T, \bar{x})$

dynamic program: track the best score
for each nonterminal over each
span of the sentence

$T(i, j, X) = \text{score } (\sim \log \text{ prob}) \text{ of the}$
best way to build symbol
 X over the span (i, j)

sent len
+ sent len
+ number of
symbols

python list indices



Grammar:

DT $\rightarrow$ the
 NN $\rightarrow$ child
 NNS $\rightarrow$ raises
 VBZ $\rightarrow$ raises
 PRP $\rightarrow$ it

$S \rightarrow NP VP$
 $NP \rightarrow DT NN$
 $NP \rightarrow NN NNS$
 $VP \rightarrow VBZ PRP$

$$\log(1/2) = -1$$

CKY

$$\text{base: } T(i, i+1, X) = \log P(w_i | X)$$

recursive:

$$T(i, j, X) = \max_{\substack{k: i < k < j \\ \text{split point}}} \max_{\substack{X \rightarrow X_1 X_2 \\ \text{rule}}} \left[\log P(X \rightarrow X_1 X_2) + T[i, k, X_1] + T[k, j, X_2] \right]$$



$$\begin{aligned} & \text{cell } (0, 3) \\ & (0, 1) + (1, 3) \\ & (0, 2) + (2, 3) \end{aligned}$$