

# CS 378 Lecture 12: Constituency Parsing

## Announcements

- A3 due in one week
- Monday: talk by Preslav Nakov
- CRFs: see video on course website

Recap Viterbi: dynamic programming algo to compute  $\max_{\bar{y}} P(\bar{y}, \bar{x})$

(similar algo today: CKY)

Beam search: faster approximation to Viterbi by ruling out low-scoring things

Today Constituency intro  
Probabilistic context free  
grammars  
CKY algorithm (start)

Constituency see slides

Context-free grammars

$\{N$	$T$	$S$	$R$
nonterminals	terminals	start	rules
$S, NP, VP, PP...$	words	symbol	
tags (preterminals)		$S$	
$DT, NN, IN, NNS$			

# Rules

binary

probs

unary

$S \rightarrow NP VP \quad 1$

$DT \rightarrow the \quad 1$

$VP \rightarrow VBD NP \quad 1$

$NNS \rightarrow children \quad 1$

$NP \rightarrow DT NN \quad \frac{1}{2}$

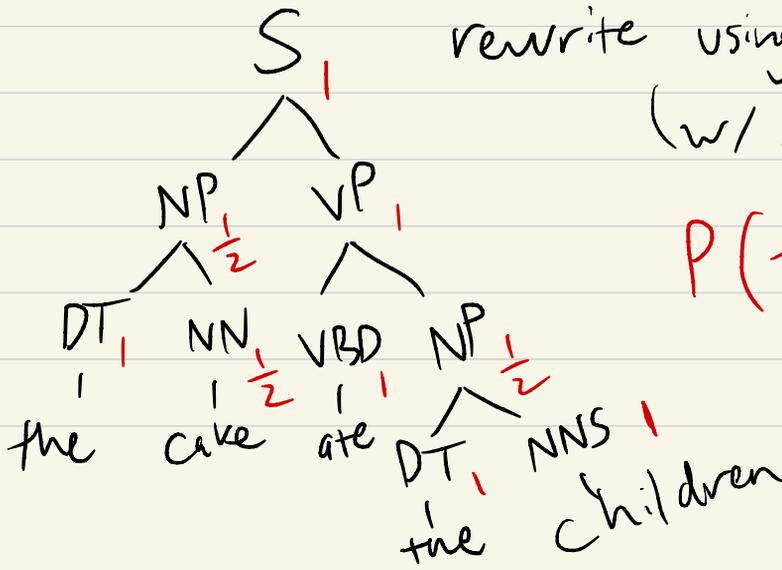
$NN \rightarrow cake \quad \frac{1}{2}$

$NP \rightarrow DT NNS \quad \frac{1}{2}$

$NN \rightarrow spoon \quad \frac{1}{2}$

$VBD \rightarrow ate \quad 1$

CFGs define a set of trees



rewrite using a rule  
(w/ S on LHS)

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

## Probabilistic CFG (PCFG)

Each rule has a prob.

Probs normalize per parent

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

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

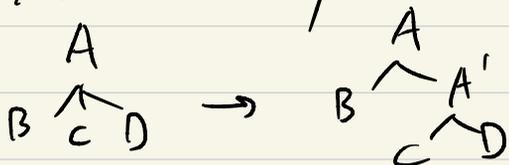
## Building a Parser

Input: treebank (sentences w/ labeled trees)

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

# ① Grammar preprocessing

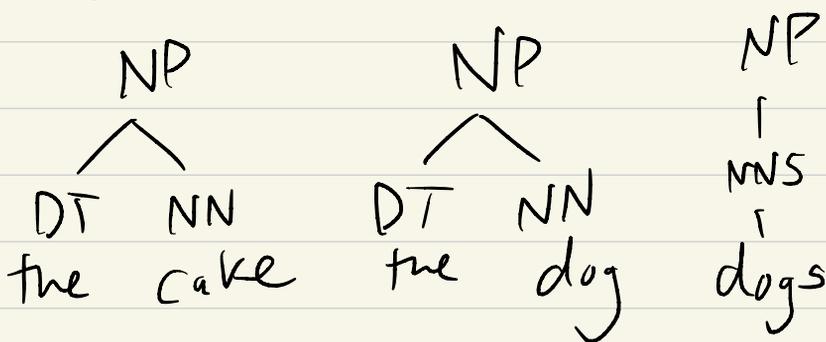
- Make every tree binary + unary



# ② Computing probs. for grammar

maximum likelihood estimate (MLE) probs

Treebank:



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

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

$$P(w | NNS) = 1.0 \text{ dogs}$$

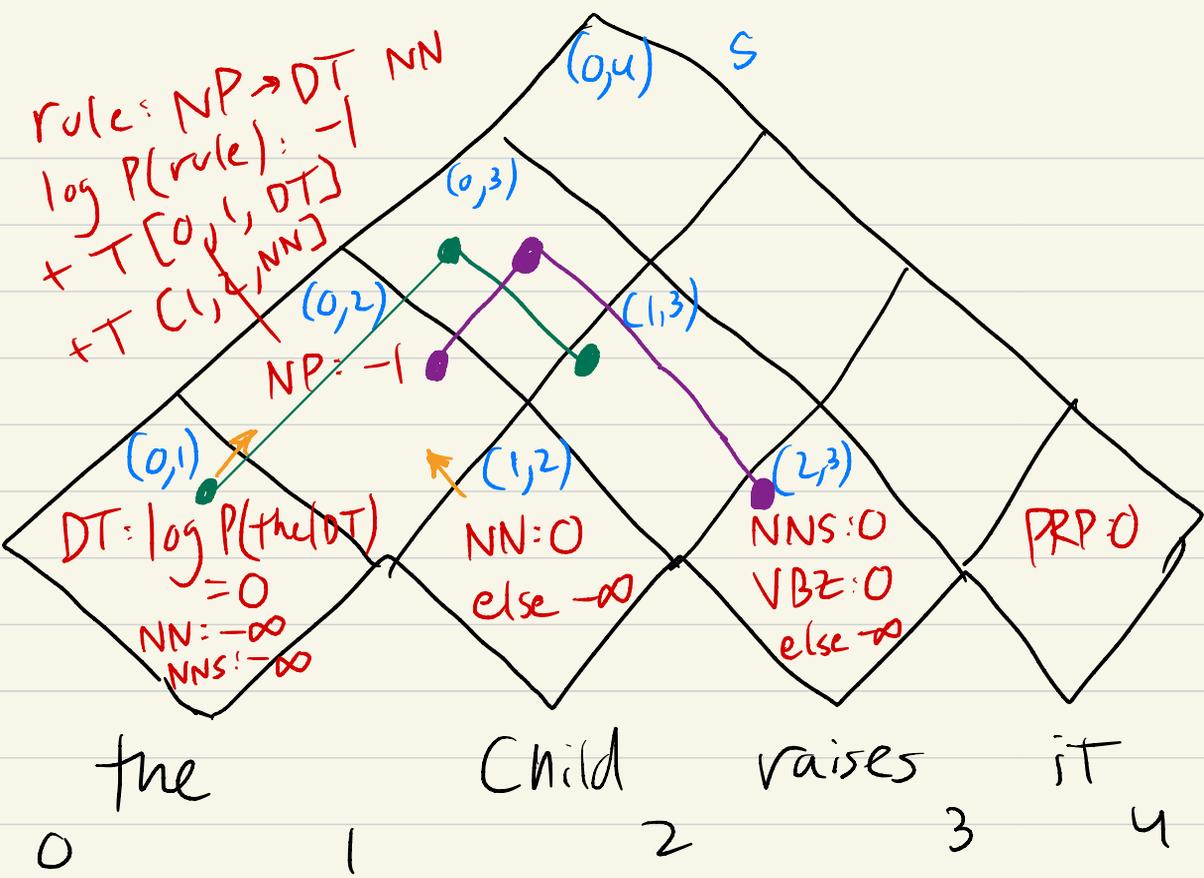
### ③ CKY algorithm

Inputs: PCFG  
sentence  $\bar{x}$

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

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

$T(i, j, X) = \text{score (log prob)} of the best way to build X over  $\bar{x}_i \dots \bar{x}_j$  (span  $(i, j)$ )$



Grammar:

- $S \rightarrow NP \quad VP \quad |$
  - $NP \rightarrow DT \quad NN \quad 1/2$
  - $NP \rightarrow NN \quad NNS \quad 1/2$
  - $VP \rightarrow VBZ \quad PRP \quad |$
  - $DT \rightarrow \text{the} \quad |$
  - $NN \rightarrow \text{child} \quad |$
  - $NNS \rightarrow \text{raises} \quad |$
  - $VBZ \rightarrow \text{raises} \quad |$
  - $PRP \rightarrow \text{it} \quad |$
- $\log(1/2) = -1$

CKY:

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

(loop over all  $X \in \text{nonterminals}$ )

Recursive case:

$T(i, j, X)$  loop over all  $i_s + j_s + X_s$   
loop over split points  $k$

$$= \max_{k=i < k < j} \max_{X \rightarrow X_1 X_2} \left[ \log P(X \rightarrow X_1 X_2) \right.$$

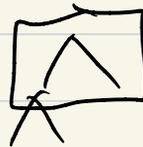
$$\left. + T[i, k, X_1] + T[k, j, X_2] \right]$$

(0, 3)

0   1   2   3  
  k   k



$$(0, 1) + (1, 3)$$



$$(0, 2) + (2, 3)$$