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Reg #: _____

Section: _____

National University of Computer and Emerging Sciences, Lahore Campus



Course:	Natural Language Processing	Course Code:	CS 535
Program:	BS(Computer Science)	Semester:	Spring 2018
Duration:	60 Minutes	Total Marks:	24
Paper Date:	12-April-18	Weight	13%
Section:	ALL	Page(s):	5
Exam:	Midterm 2 Solution		

Q1) A sentence can easily have more than one parse tree that is consistent with a given CFG. How do PCFGs and non-probability-based CFGs differ in terms of handling parsing ambiguity? [2 Marks]

Answer:

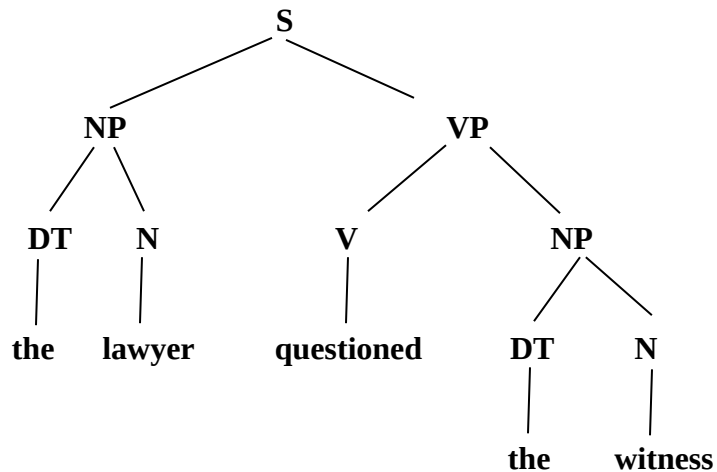
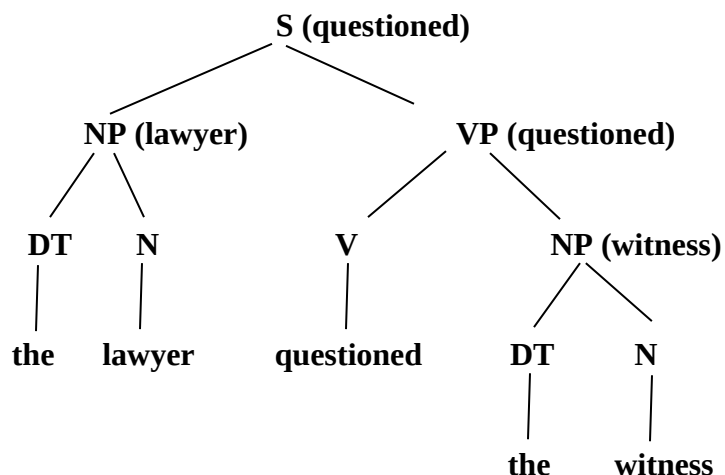
PCFGs define probability of each rule which can be used to find probability of parse tree. Tree with maximum probability is selected.

Q2) Which of the following is a false statement about PCFGs: [2 Marks]

- a) The rules impose independence assumptions that effect poor modeling of structural dependency across the tree.
- b) The rules do not model syntactic facts about particular words, which causes a variety of problems.
- c) The joint probability of a sentence, S, and a parses of it, T, is the same as the probability of the parse, T.

Answer: They are true

Q3) a) Which of the following trees is a lexicalized tree? [1 Mark]

Tree 1**Tree 2****Solution: Tree2**

b) For the trees above, when you count and estimate the probability for rules, which tree is most likely to encounter sparseness problem? [2 Marks]

Tree2

c) How can you alleviate sparseness problem encountered in estimating probability for parse trees? [2 Marks]

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smoothing

Q4) Consider a context-free grammar with the following rules (assume that S is the start symbol):

$S \rightarrow NP VP$

$NP \rightarrow DT NN$

$NP \rightarrow NP PP$

$PP \rightarrow IN NP$

$VP \rightarrow VB NP$

$DT \rightarrow \text{the}$

$NN \rightarrow \text{man}$

$NN \rightarrow \text{dog}$

$NN \rightarrow \text{cat}$

$NN \rightarrow \text{park}$

$VB \rightarrow \text{saw}$

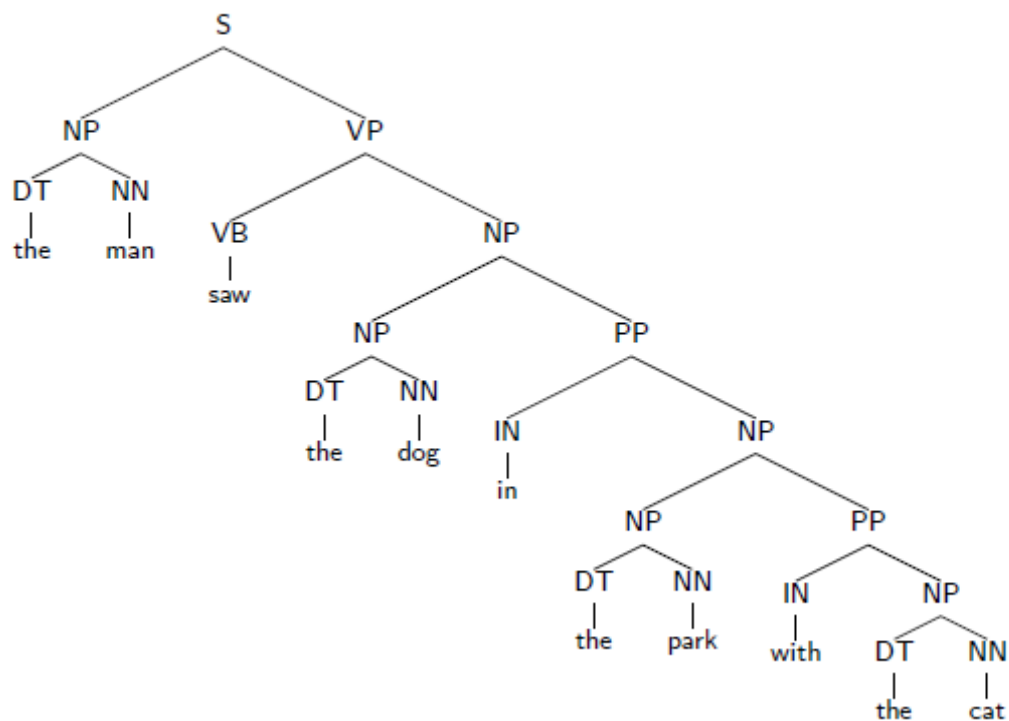
$IN \rightarrow \text{in}$

$IN \rightarrow \text{with}$

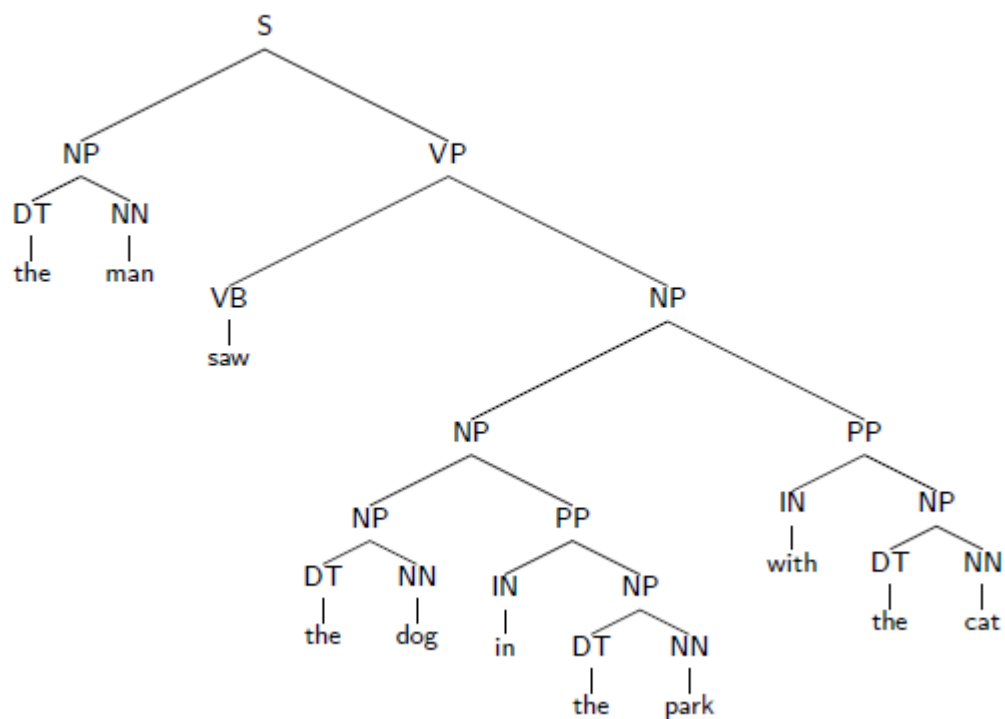
$IN \rightarrow \text{under}$

How many parse trees for “the man saw the dog in the park with the cat”? Draw all the parse trees for this sentence. [5 Marks]

Two parse trees, parse tree 1:



Two parse trees, parse tree 2:



Q3) Consider a trigram HMM, as introduced in class. We saw that the Viterbi algorithm could be used to find $\max_{y_1, \dots, y_{n+1}} p(x_1, \dots, x_n; y_1, \dots, y_{n+1})$ where the max is taken over all sequences y_1, \dots, y_{n+1} such that $y_i \in K$ for $i = 1 \dots n$, and $y_{n+1} = \text{STOP}$. (Recall that K is the set of possible tags in the HMM.) In a trigram tagger we assume that p takes the form

$$p(x_1, \dots, x_n; y_1, \dots, y_{n+1}) = \prod_{i=1}^{n+1} q(y_i | y_{i-1}, y_{i-2})$$

Recall that we have assumed in this definition that $y_0 = y_{-1} = y_{-2} = *$, and $y_{n+1} = \text{STOP}$. The Viterbi algorithm is shown in figure below.

Input: A sentence x_1, \dots, x_n , parameters $q(s|u, v)$ and $e(x|s)$

Definitions: Define K to be the set of possible tags. Define $K_{-1} = K_0 = \{*\}$, and $K_k = K$ for $k = 1 \dots n$.

Initialization: Set $\pi(0, *, *) = 1$

Algorithm: For $k = 1 \dots n$,

For $u \in K_{k-1}, v \in K_k$,

$\pi(k, u, v) = \max_{w \in K_{k-2}} (\pi(k-1, w, u) \times q(v | w, u) \times e(x_k | v))$

Return $\max_{u \in K_{n-1}, v \in K_n} (\pi(n, u, v) \times q(\text{STOP} | u, v))$

Now consider a four-gram tagger, where p takes the form

$$p(x_1, \dots, x_n; y_1, \dots, y_{n+1}) = \prod_{i=1}^{n+1} q(y_i | y_{i-1}, y_{i-2}, y_{i-3})$$

We have assumed in this definition that $y_0 = y_{-1} = y_{-2} = *$, and $y_{n+1} = \text{STOP}$. In the box on next page, give a version of the Viterbi algorithm that takes as input a sentence x_1, \dots, x_n , and finds $\max_{y_1, \dots, y_{n+1}} p(x_1, \dots, x_n; y_1, \dots, y_{n+1})$ for a four-gram tagger, as defined above equation.

[10 Marks]

Input: A sentence x_1, \dots, x_n , parameters $q(w|t, u, v)$ and $e(x|s)$

Definitions: Define K to be the set of possible tags. Define $K_{-2} = K_{-1} = K_0 = \{*\}$, and $K_k = K$ for $k = 1 \dots n$.

Initialization: Set $\pi(0, *, *) = 1$

Algorithm: For $k = 1 \dots n$,

For $t \in K_{k-2}, u \in K_{k-1}, v \in K_k$,

$\pi(k, t, u, v) = \max_{w \in K_{k-3}} (\pi(k-2, w, t, u) \times q(v|w, t, u) \times e(x_k|v))$

Return $\max_{t \in K_{n-2}, u \in K_{n-1}, v \in K_n} (\pi(n, t, u, v) \times q(STOP|t, u, v))$