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Fall 2017

### Homework 3

Handed Out: 10/27/2017 Due: 10pm, 11/17/2017

Please submit an archive of your solution (including code) on Compass by 10:00pm on the due date. Please document your code where necessary.

# Getting started

All files that are necessary to do the assignment are contained in a tarball which you can get from:

```
https://courses.engr.illinois.edu/cs447/HW/cs447_hw3.tar.gz
```

You need to unpack this tarball (tar -zxvf cs447\_hw3.tar.gz) to get a directory that contains the code and data you will need for this homework.

### **NLTK**

For this assignment, you are expected to use python 3.5.2 and NLTK 3.2.2, this configuration is available on the EWS machines and you are expected to test your codes on EWS machines before making a submission. The provided scripts work only with NLTK 3.2.2 (installed independently or as a part of Canopy). If you do not want to use Canopy and already have a more recent version of NLTK installed<sup>1</sup>, you can install pip first , and then use the following commands:

```
sudo pip uninstall nltk
sudo pip install -Iv nltk==3.2.2
```

This will tell pip to uninstall your current version of NLTK and then download and install NLTK 3.2.2. Please use the instructions in the following section in order to setup Canopy on the EWS machines (if you haven't already).

## Setting up NLTK on Canopy and using it on python 3.5.2

You will need to log in on EWS machines or ssh to (your-net-id)@remlnx.ews.illinois.edu via X servers, for example a complete command could be ssh ss46@remlnx.ews.illinois.edu.

After connecting, you need to first type module load canopy/py35-2.1.3

In the next step, type **canopy\_cli**. It will start setting up the environment and copying relevant libraries to your home directory. This process will take around 20 mins. The default canopy environment is python 2.7. Therefore after this process, you will need to switch the default environment into python 3.5.

In the pop-up configuration window, take the following steps: Edit - Preferences - Python - Python Interpreter used by Canopy - Uncheck "Use Default Canopy Environment (User)" box - Create New Environment-name (name it something convenient like nlp) - Click select EDM Bundle File - Select canopy-2.1.3.3542-python35-full-rh6-64.bundle. This process will switch the default python to 3.5.2 and start copying relevant libraries. This will take around 40 mins.

Now you have the canopy version of python 3.5.2 located in

/home/(yourNetId)/.local/share/canopy/edm/envs/(name of your environment)/bin/python You can now access the canopy version of python 3.5.2 by typing

/home/(yourNetId)/.local/share/canopy/edm/envs/(name of your environment)/bin/python For example, I can run python 3.5.2 with NLTK using

/home/ss46/.local/share/canopy/edm/envs/nlp/bin/python

<sup>&</sup>lt;sup>1</sup>NLTK 3.2.5, for example, will not successfully execute our scripts

# Part 1: Writing a Context-Free Grammar (4 points)

### 1.1 Goal

Your first task is to write a context-free grammar that can parse a set of short sentences. You will be using parsing code from the NLTK for this portion of the assignment, and your grammar does not need to be in Chomsky Normal Form.

#### 1.2 Data

The file sentences.txt contains an evaluation set of 25 short sentences.

## 1.3 Provided code

We have provided a script (hw3\_nltkcfg.py) that reads your grammar from mygrammar.cfg and tries to parse the evaluation set. The script expects the sentences.txt file to be in the same directory, and will write its output to a new file, hw3\_cfg\_out.txt. Call the script from the command line with:

```
python hw3_nltkcfg.py
```

For each sentence, the script will provide a brief message in the console (whether the parse was a SUCCESS or a FAILURE), along with the number of successful parse trees. More importantly, each successful parse tree will be written to the output file. When evaluating your grammar, you could use these parse trees to find ways to improve your grammar.

# 1.4 What you need to implement

You don't need to write (or submit) any actual Python code for this part of the assignment. Instead, your task is to define the rules for your CFG in mygrammar.cfg. This file already contains the lexical rules required for these sentences (i.e., words are mapped to their POS tags), along with a start symbol for the grammar that generates the nonterminal symbol S.

## Please do not change or delete the existing rules.

Instead, you will need to add rules such as:

```
S -> NP VP
NP -> DT N
N -> NN | NNS | ...
etc.
```

You may define your own inventory of nonterminals, but please do not change the existing rules. Your grade will depend in part on your grammar's ability to successfully parse the evaluation sentences; however, we care more about your ability to define and explain a linguistically meaningful grammar (see below).

### Implementation hints

While it's possible to define a very simple grammar that parses every sentence, such as:

```
S -> LEX | LEX S
LEX -> NN | NNS | COMMA | VB | ...
```

that's not the point of this part of the assignment, and such a solution would receive zero credit. Similarly, a set of rules like:

```
S -> NN CC NN VBD FS
S -> PRP VBP NNS CC NNS FS
etc.
```

where each (extremely flat) rule generates the POS tag sequence for a particular sentence in the evaluation data is also not allowed. Instead, your grammar should use a set of nonterminals similar to those of the Penn Treebank parse trees you've seen in class. You can find a description of the POS tags used in this assignment at https://www.ling.upenn.edu/courses/Fall\_2003/ling001/penn\_treebank\_pos.html.

### 1.5 What to submit

Along with your completed grammar file (mygrammar.cfg), you should submit your output file (hw3\_cfg\_out.txt). Additionally, you must answer the following discussion questions on Compass when you submit your assignment. These questions can be found by navigating to  $Course\ Content\ \to Homeworks\ \to Homework\ 3$   $\to Homework\ 3\ CFG\ Discussion\ Questions$ 

- 1. Describe how your grammar analyzes noun phrases. (0.5pts)
- 2. Describe how your grammar analyzes verb phrases. (0.5pts)
- 3. Which sentence has the most parses (please include the number of parses)? What about the sentences (besides their length) contributes to the large number of parses? (1.0pts)

# 1.6 What you will be graded on

You will be graded on your grammar's ability to successfully parse the evaluation sentences (1.0 pts) and on the quality<sup>2</sup> of your grammar (1.0 pts). Your answers to the discussion questions on Compass provide the rest of the grade (2.0 pts).

# Part 2: Parsing with Probabilistic Context-Free Grammars (6 points)

## 2.1 Goal

Your second task is to implement the CKY algorithm for parsing using a probabilistic context-free grammar (PCFG). Given a PCFG, the algorithm uses dynamic programming to return the most-likely parse for an input sentence.

### 2.2 Data

We have given you a text file (toygrammar.pcfg) containing a toy grammar that generates sentences similar to the running examples seen in class:

```
the woman eats the sushi with the tuna the woman eats some sushi with the chopsticks the woman eats some sushi with a man etc.
```

Each binary rule in the grammar is stored on a separate line, in the following format:

```
prob P -> LC RC
```

where prob is the rule's probability, P is the left-hand side of the rule (a parent nonterminal), and LC and RC are the left and right children, respectively.

For unary rules, we only have a single child C and the line has format:

```
prob P -> C
```

We provide code for reading this file format to produce the equivalent PCFG object in Python.

<sup>&</sup>lt;sup>2</sup>See the **Implementation hints** in section 1.4

### 2.3 Provided code

We provide the module hw3\_pcfg.py, which contains several classes that may be useful for chart parsing. You should look over the source code yourself, but a brief summary of these classes includes:

Rule: a grammatical Rule has a probability and a parent category, and is extended by UnaryRule and BinaryRule.

UnaryRule: a UnaryRule is a Rule with only a single child (word or nonterminal).

BinaryRule: a BinaryRule is a Rule with two children.

Item: an Item is a chart entry that stores a label and a Viterbi probability. It is extended by LeafItem and InternalItem.

LeafItem: a LeafItem is a chart entry that corresponds to a leaf in the parse tree. For a leaf, the label is a word and the Viterbi probability is 1.0.

InternalItem: an InternalItem is a chart entry that corresponds to a nonterminal with a particular span in the parse tree. Its label is a nonterminal symbol, and its Viterbi probability is the probability of the best subtree rooted by that symbol (for this span). Its number of parses counts the number of possible trees rooted at the label for this span. Additionally, an InternalItem maintains a tuple of pointers to its children (if the children were generated by a BinaryRule) or single child (if generated by a UnaryRule). This tuple is a backpointer for the Viterbi parse.

PCFG: a PCFG maintains a collection of Rules; for your CKY algorithm, the rules are sorted by their right-hand-side in the ckyRules dictionary.

## 2.4 What you need to implement

Your task is to finish implementing the CKY method in PCFG:

CKY(self, sentence): given sentence (a list of word strings), return the root of the Viterbi parse tree (i.e. an InternalItem with label TOP<sup>3</sup> whose probability is the Viterbi probability). By recursing on the children of this item, we should be able to get the complete Viterbi tree. If no such tree exists, return None (a parse failure).

In order to finish the CKY method implementation, you must also modify the constructor for InternalItem to correctly set the number of parses. We also provide the stubs for the Chart and Cell classes to help guide you in your implementation of the CKY chart.

### Implementation hints

In Python, to obtain a tuple for a singleton value v, use (v,) instead of (v). You will need this for the InternalItem's child pointer when applying unary rules.

While our grammar is small, we still want you to implement your CKY algorithm using log probabilities (using math.log(prob) with the default base). It is good practice for implementing an actual NLP system.

You will probably want to define submethods for your CKY algorithm, as well as data structures to represent the parse chart or individual cells (each covering a particular span). You are welcome to add any functionality you need, as long as your CKY method returns an InternalItem object as we have defined it.

When filling the cells of your chart, you should add all possible items that can generate the span using a binary rule before adding items that can generate the span using a unary rule. Keep in mind that you only check for unary rules once per cell; our grammar does not allow chains of unary rules within a cell.

 $<sup>^3</sup>$ The start symbol for the grammar

#### 2.4.1 Test script

We have provided a hardcoded test script to check your implementation. After you have implemented your CKY algorithm in hw3\_pcfg.py, run

```
python hw3_pcfg_test.py
```

to evaluate your parser.

### 2.5 What to submit

The only file you need to submit for this part is your completed hw3\_pcfg.py program. We will evaluate your code using a test harness similar to the one provided (all files will be in the same format and we will initialize your PCFG in the same way, but we may use a different grammar and test sentences).

# 2.6 What you will be graded on

The grade for your implementation of the CKY algorithm will be based on returning the correct Viterbi trees (2.0 pts) and probabilities (2.0 pts) for our test harness, as well as the correct number of parse trees for each sentence (2.0 pts).

# Submission guidelines

You should submit your solution as a compressed tarball on Compass; to do this, save your files in a directory called abc123\_hw3 (where abc123 is your NetID) and create the archive from its parent directory (tar -czvf abc123\_hw3.tar.gz abc123\_hw3). Please include the following files:

- 1. mygrammar.cfg: your final CFG for parsing the sentences from Part 1
- 2. hw3\_cfg\_out.txt: the list of trees produced by your grammar on the evaluation sentences in Part 1 (the output of hw3\_nltkcfg.py)
- 3. hw3\_pcfg.py: your completed Python module for parsing using CKY and PCFGs in Part 2

Additionally, you must answer the following discussion questions on Compass when you submit your assignment. These questions can be found by navigating to Course Content  $\rightarrow$  Homework 3 CFG Discussion Questions