

**Home work 3 (NLP)**  
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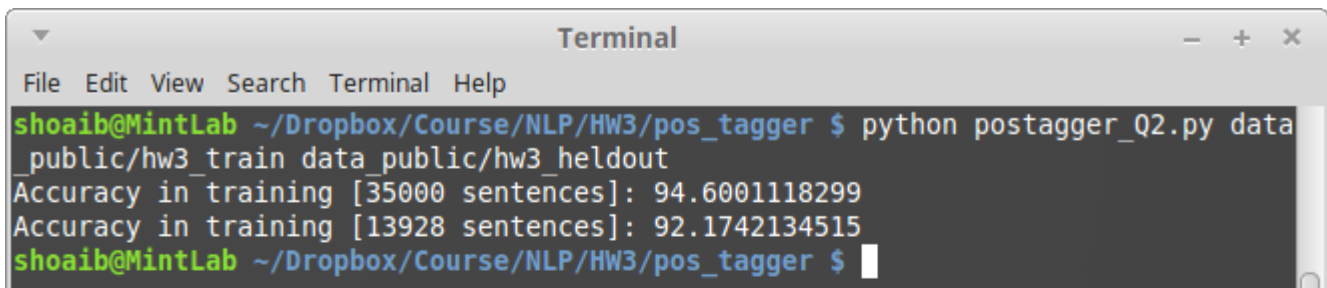
Q1:

- |               |               |   |
|---------------|---------------|---|
| 1. Atlanta/NN | Should be NNP | (Proper Noun)                                       |
| 2. Dinner/NNS | Should be NN  | (Noun, Singular)                                    |
| 3. Have/VB    | Should be VBP | (Verb, non 3 <sup>rd</sup> person singular present) |
| 4. Can/VBP    | Should be MD  | (Modal)   |

Q2 and Q3:

**Findings and results:**

For Q2, the accuracy is 94.60% and 92.17% on training data and test data respectively.  
It takes around 0.35 seconds to create the model for this question on my machine.

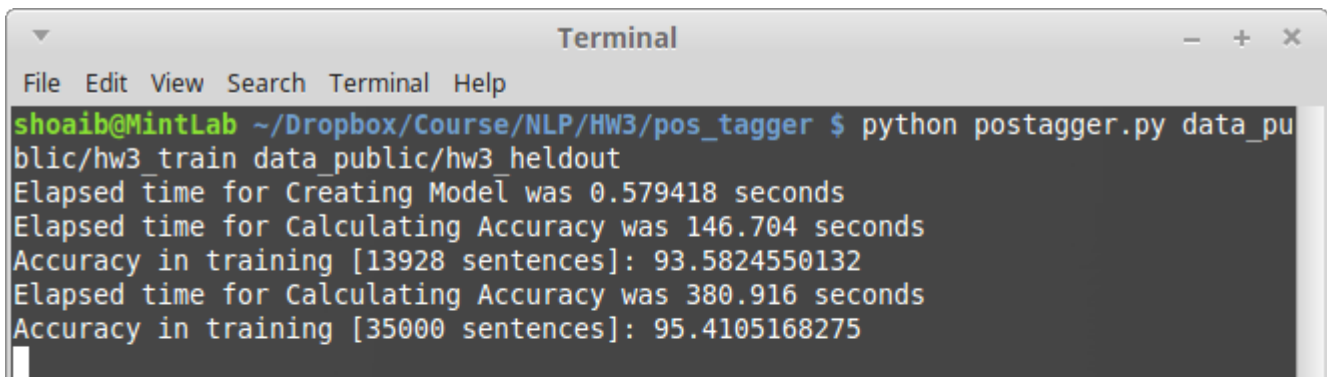


```
Terminal
File Edit View Search Terminal Help
shoaib@MintLab ~/Dropbox/Course/NLP/HW3/pos_tagger $ python postagger_Q2.py data_public/hw3_train data_public/hw3_heldout
Accuracy in training [35000 sentences]: 94.6001118299
Accuracy in training [13928 sentences]: 92.1742134515
shoaib@MintLab ~/Dropbox/Course/NLP/HW3/pos_tagger $
```

For Q3, the accuracy is 95.41% and 93.20% on training data and test data respectively, with lowering the cases and smoothing only.

I have been able to increase the accuracy to 93.58% for the testing dataset so far, using some hard coded morphology for the unknown word.

It takes around 0.50-0.60 seconds to create the model for this question on my machine, and 8/9 minutes to run the whole prediction process. I have modified the code to show the time.



```
Terminal
File Edit View Search Terminal Help
shoaib@MintLab ~/Dropbox/Course/NLP/HW3/pos_tagger $ python postagger.py data_public/hw3_train data_public/hw3_heldout
Elapsed time for Creating Model was 0.579418 seconds
Elapsed time for Calculating Accuracy was 146.704 seconds
Accuracy in training [13928 sentences]: 93.5824550132
Elapsed time for Calculating Accuracy was 380.916 seconds
Accuracy in training [35000 sentences]: 95.4105168275

```

**How you deal with unknown words:**

I have made a list of the unknown words. I have fixed some rules for CD, JJ, WDT and a few other tags. For example:

- If the word is a number I have set the tag to CD.
- If the word is a combination of numbers and alphabet, I have set the tag to JJ

**Problems faced during the implementation:**

The multiplication process is a bit slower, so I used log and addition, which I believe makes the program a bit faster.

For prediction using the Viterbi matrix, the runtime is slow. I was using two max() functions, and it was taking around 13/14 minutes for the whole process. I modified the code to remove one max() function, now the code runs in 8/9 minutes.

**A small error analysis (what errors does your tagger make? is it hard to distinguish VBN vs. VBD?)**

I have checked just more than 11,000 errors. Please note that, in my following calculations, the counts contain a certain part of speech wrongly tagged as another one, or another part of speech wrongly tagged as the certain one.

In that the predictor was unable to predict between **VBN** and **VBD** 682 times. So This doesn't look that significant. However, for all the verbs:

It was unable to predict between **any verb** and **any noun** 1402 times.

It was unable to predict between **any verb** and **adjective** 665 times.

It was unable to predict between **any verb** and **other verbs, or any other parts of speech** 3193 times.

This count includes the aforementioned two counts.

So I assume, verbs are confused with another verb or another part of speech somewhat evenly.

Whereas, for nouns:

It was unable to predict between **any verb** and **any noun** 1402 times.

It was unable to predict between **any noun** and **adjective** 1937 times.

It was unable to predict between **any noun** and **other nouns, or any other parts of speech** 7151 times.

This count includes the aforementioned two counts.

So, a noun is more likely to be confused as other nouns, as well as other parts of speech.