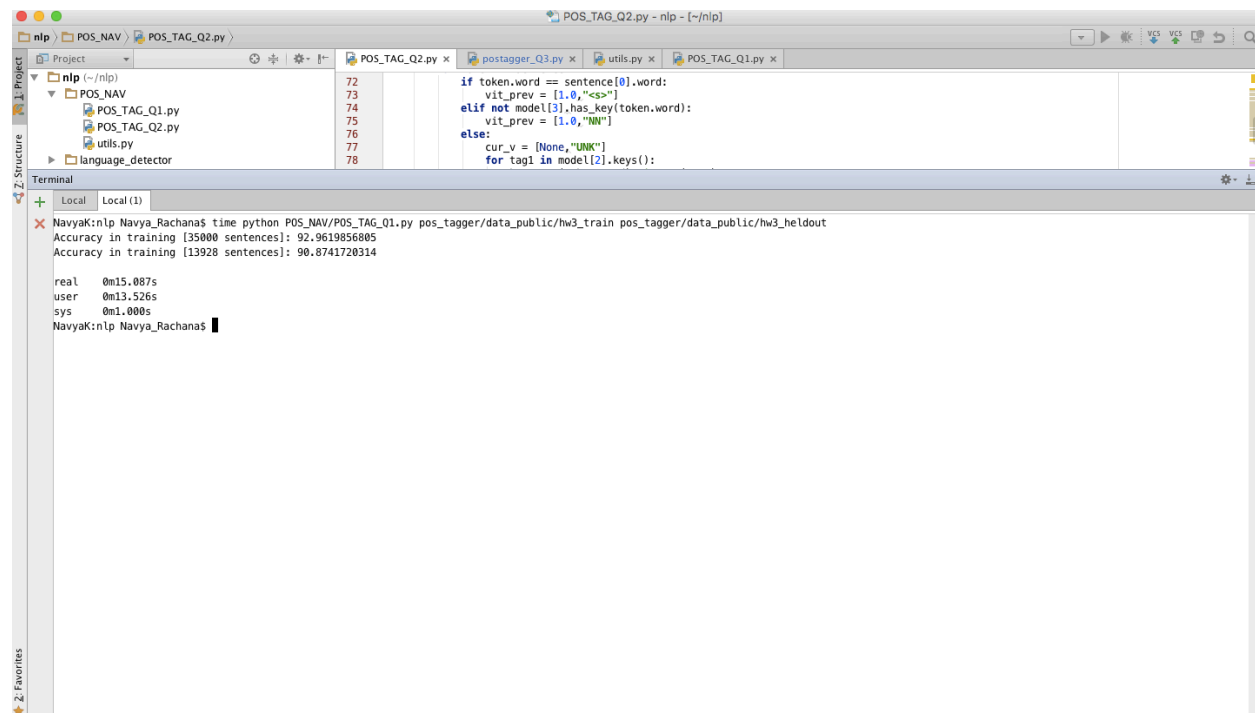


Question 2: Most Likely tag for POSTAGGER:

implementation:

- Initially the program counts all the tags, then the program finds the most likely tag for each word in the training set
- It then assigns the most likely tag to all the words in the testing set and “NN” for unseen words and then calculates the accuracy

Results: (LOWERCASING)



```
POS_TAG_Q2.py - nlp - [~/nlp]
Project
  nlp (~/nlp)
    POS_NAV
      POS_TAG_Q1.py
      POS_TAG_Q2.py
      utils.py
    language_detector
  POS_TAG_Q2.py x
  postagger_Q3.py x
  utils.py x
  POS_TAG_Q1.py x
Terminal
  Local Local (1)
  X NavyaK:nlp Navya_Rachana$ time python POS_NAV/POS_TAG_Q1.py pos_tagger/data_public/hw3_train pos_tagger/data_public/hw3_heldout
  Accuracy in training [35000 sentences]: 92.9619856805
  Accuracy in training [13928 sentences]: 90.8741720314
  real    0m15.087s
  user    0m13.526s
  sys     0m1.000s
  NavyaK:nlp Navya_Rachana$
```

Accuracy in training [35000 sentences]: 92.9619856805

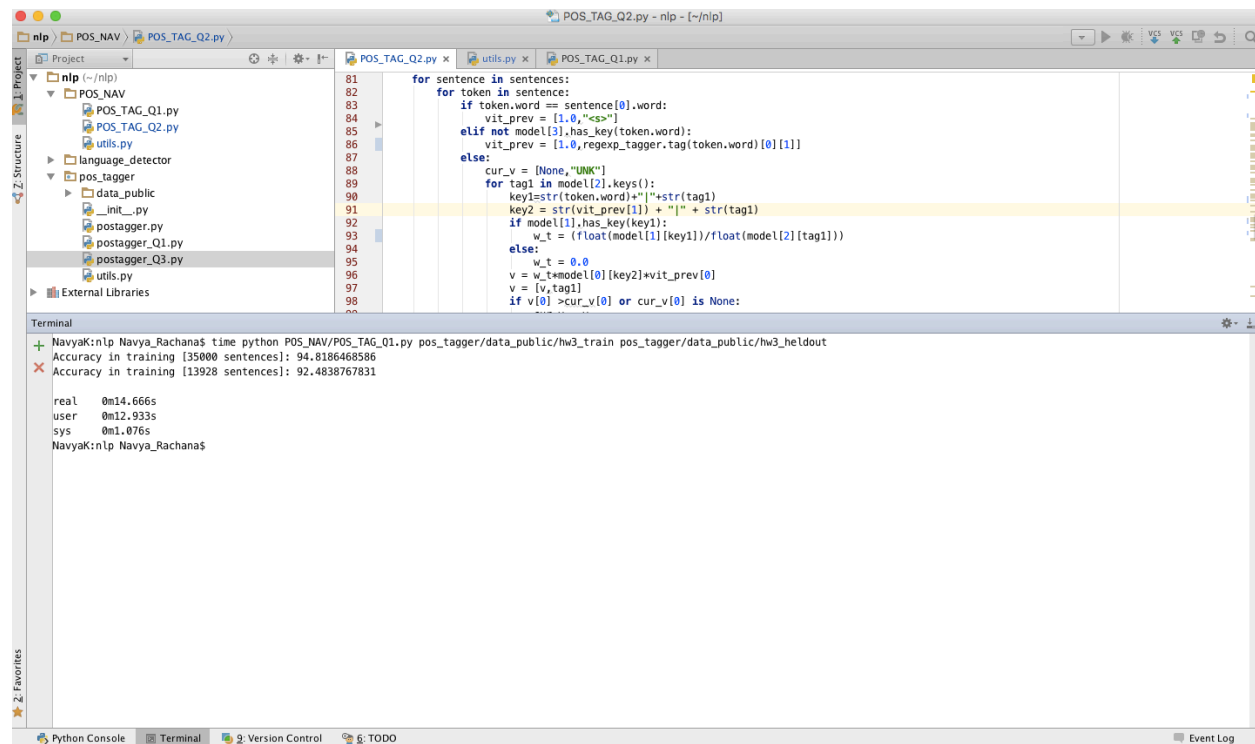
Accuracy in training [13928 sentences]: 90.8741720314

real 0m15.087s

user 0m13.526s

sys 0m1.000s

Results: (NO LOWERCASING)



```
POS_TAG_Q2.py - nlp - [~/nlp]

Project
  nlp (~/nlp)
    POS_NAV
      POS_TAG_Q1.py
      POS_TAG_Q2.py
      utils.py
    language_detector
    pos_tagger
      data_public
        __init__.py
        postagger.py
        postagger_Q1.py
        postagger_Q3.py
      utils.py
    External Libraries

Terminal
+ NavyaK:nlp Navya_Rachana$ time python POS_NAV/POS_TAG_Q1.py pos_tagger/data_public/hw3_train pos_tagger/data_public/hw3_heldout
Accuracy in training [35000 sentences]: 94.8186468586
Accuracy in training [13928 sentences]: 92.4838767831

real    0m14.666s
user    0m12.933s
sys      0m1.076s
NavyaK:nlp Navya_Rachana$
```

Accuracy in training [35000 sentences]: 94.8186468586

Accuracy in training [13928 sentences]: 92.4838767831

real 0m14.666s

user 0m12.933s

sys 0m1.076s

Question 3: HMM POS Tagging

Implementation:

Creating Model:

- The function initially calculates all the counts of words, tags, word|tag and tag|tag
- It then calculates the probabilities of word|tag and tag|tag
- The model then returns the probabilities and the tags count probabilities

Predicting test results:

- The Predict model calculates the Viterbi matrix in two ways: If the word is the start of the sentence it would calculate the from the Viterbi (<s>|<s>) which is taken as 1
- If the word is not the first word of the sentence it then calculates the Viterbi of the previous word|tags and finally get the result

Problems during Implementation

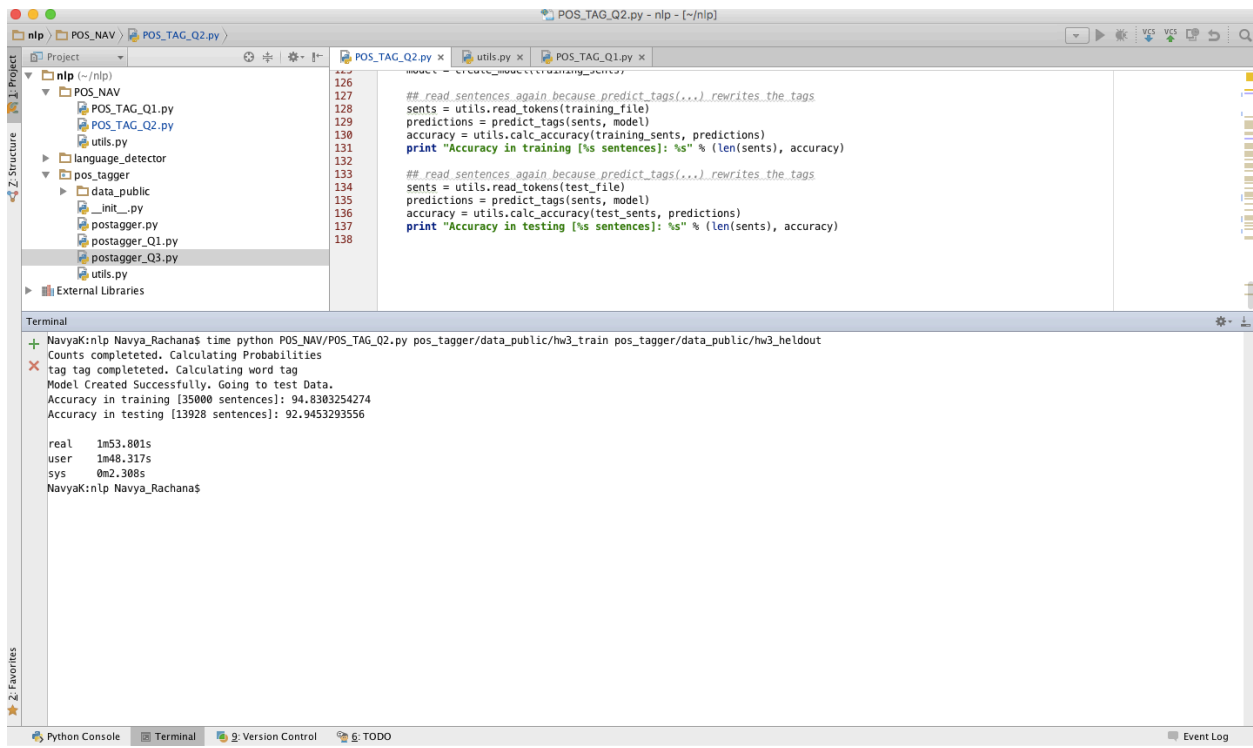
- Calculating the dictionaries for all the counts and probabilities
- Smoothing when probabilities aren't found
- Storing a dictionary of dictionaries in model
- Accessing the dictionaries passed in model from creating model function
- Logic for the Viterbi Calculation

Extra credit Unseen words:

For unseen words Regular Expression POS tagger from nltk is implemented with the following patterns.

```
patterns = [  
    (r'.*ing$', 'VBG'),  
    (r'.*ed$', 'VBD'),  
    (r'.*es$', 'VBZ'),  
    (r'.*ould$', 'MD'),  
    (r'.*\'s$', 'NN$'),  
    (r'.*s$', 'NNS'),  
    (r'^-?[0-9]+(.[0-9]+)?$', 'CD'), # cardinal num.  
    (r'.*', 'NN') # nouns (default)  
]
```

Results: (LOWERCASING)



```
126
127
128 ## read sentences again because predict_tags(...) rewrites the tags
129 sents = utils.read_tokens(training_file)
130 predictions = predict_tags(sents, model)
131 accuracy = utils.calc_accuracy(training_sents, predictions)
132 print "Accuracy in training [%s sentences]: %s" % (len(sents), accuracy)
133
134 ## read sentences again because predict_tags(...) rewrites the tags
135 sents = utils.read_tokens(test_file)
136 predictions = predict_tags(sents, model)
137 accuracy = utils.calc_accuracy(test_sents, predictions)
138 print "Accuracy in testing [%s sentences]: %s" % (len(sents), accuracy)
```

```
NavyaK:nlp Navya_Rachana$ time python POS_NAV/POS_TAG_Q2.py pos_tagger/data_public/hw3_train pos_tagger/data_public/hw3_heldout
Counts completed. Calculating Probabilities
tag tag completed. Calculating word tag
Model Created Successfully. Going to test Data.
Accuracy in training [35000 sentences]: 94.8303254274
Accuracy in testing [13928 sentences]: 92.9453293556

real    1m53.801s
user    1m48.317s
sys      0m2.308s
NavyaK:nlp Navya_Rachana$
```

Accuracy in training [35000 sentences]: 94.8303254274

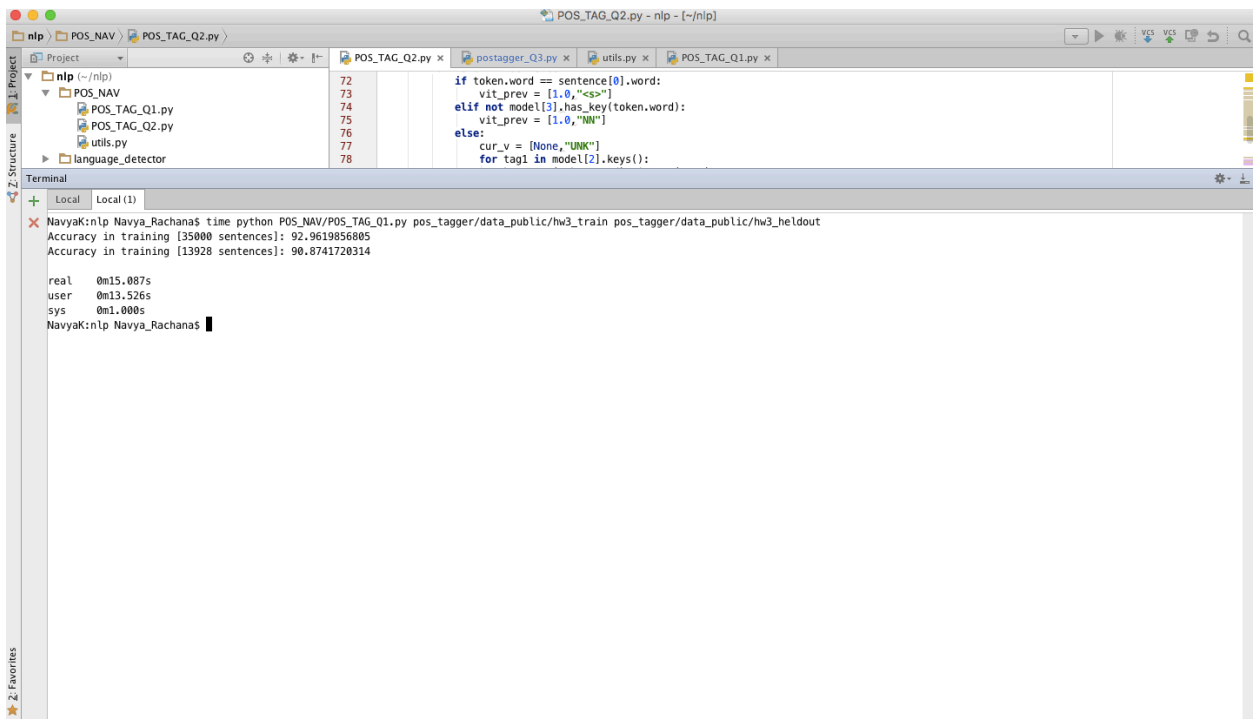
Accuracy in testing [13928 sentences]: 92.9453293556

real 1m53.801s

user 1m48.317s

sys 0m2.308s

Results: (NO LOWERCASING)



The screenshot shows a Jupyter Notebook environment with a project named 'nlp'. The file explorer on the left shows a directory structure with files: POS_TAG_Q1.py, POS_TAG_Q2.py, utils.py, and language_detector. The main editor displays a Python script for POS tagging. The terminal output shows the execution of the script, which reports the following results:

```
Accuracy in training [35000 sentences]: 92.9619856805
Accuracy in training [13928 sentences]: 90.8741720314

real    0m15.087s
user    0m13.526s
sys     0m1.000s
NavyaK:nlp Navya_Rachana$
```

Accuracy in training [35000 sentences]: 96.3441454525

Accuracy in testing [13928 sentences]: 94.2741499328

real 1m33.712s

user 1m32.218s

sys 0m0.990s

Observation:

1. Lowercasing the words did not help in improving the performance in fact the performance decreased in both cases.