## ▼ Library used in this program

```
import nltk
nltk.download('brown')
nltk.download('universal_tagset')
import numpy as np
import pandas as pd
from nltk.corpus import brown
from tqdm import tqdm
from sklearn.model_selection import train_test_split, KFold
import seaborn as sns
import matplotlib.pyplot as plt

[nltk_data] Downloading package brown to /root/nltk_data...
[nltk_data] Unzipping corpora/brown.zip.
[nltk_data] Downloading package universal_tagset to /root/nltk_data...
[nltk_data] Unzipping taggers/universal_tagset.zip.
```

#### ▼ DataSet

```
# splitting the corpus into two for now
train corpus, test corpus = train test split(brown.tagged sents(tagset='universal'
class storage:
  #creating dataset for storing the computed data.
  def init (self):
    self.id = 0
    self.index2tag = dict()
    self.value2id = dict()
    self.values = set()
  def get length():
    return self.id
  def insert(self, value):
    self.index2tag[self.id] = value
    self.value2id[value] = self.id
    self.values.add(value)
    self.id += 1
  def retrive(self, key, method ='id'):
    if method == 'id':
      return self.index2tag[key]
    elif key in self.values:
      return self.value2id[key]
    else:
      return None
# val variable for hadling some cases
alnha - 0 000001
```

### → HMM Model

```
def get word(train corpus):
  words = storage()
  for sent in tqdm(train corpus):
    for word, tag in sent:
      if words.retrive(word.lower(), 'temp') == None:
        words.insert(word.lower())
  return words
def get tag():
  tag list = set([tag for words,tag in brown.tagged words(tagset='universal')])
  tags = storage()
  for tag in tqdm(tag list):
    if tags.retrive(tag, 'temp') == None:
      tags.insert(tag)
  return tags
words = get word(train corpus)
tags = get_tag()
                  45872/45872 [00:00<00:00, 50519.96it/s]
    100%|
    100%|
                    | 12/12 [00:00<00:00, 15787.84it/s]
def get hmm matrix(train corpus, words, tags, alpha):
  transmission matrix = np.zeros([tags.id,tags.id])
  emission matrix = np.zeros([tags.id,words.id])
  tags prob = np.zeros([tags.id])
  for sent in tqdm(train_corpus):
    for index in range(len(sent)):
      word = sent[index][0]
      tag = sent[index][1]
      word index = words.retrive(word.lower(), 'temp')
      tag index = tags.retrive(tag, 'temp')
      tags prob[tag index] +=1
      emission matrix[tag index,word index] +=1
      if index != len(sent) - 1:
        next_tag = tags.retrive(sent[index + 1][1], 'temp')
        transmission matrix[tag index,next tag] +=1
  transmission_matrix = np.divide((transmission_matrix+ alpha), (np.reshape(tags_p)
  emission matrix = np.divide((emission matrix+alpha), (np.reshape(tags prob,[-1,1
  tags_prob = np.divide(tags_prob, np.sum(tags_prob))
```

```
transmission_matrix[transmission_matrix == 0] = alpha
emission_matrix[emission_matrix == 0] =alpha
tags_prob[tags_prob == 0] = alpha
return transmission_matrix,emission_matrix,tags_prob
```

transmission\_matrix,emission\_matrix,tags\_prob = get\_hmm\_matrix(train\_corpus,words,

```
100% | 45872/45872 [00:03<00:00, 13794.08it/s]
```

Double-click (or enter) to edit

```
# Viterbi algorithm
def get pos(sent list, transmission matrix, emission matrix, tags prob, words, tags
  if len(sent list) == 0:
    return []
  seq score matrix = np.zeros([tags.id, len(sent list)])
  back pointer = np.zeros([tags.id, len(sent list)])
  # First step in viterbi Intialization
  word id = words.retrive(sent list[0].lower(), 'temp')
  for i in range(tags.id):
    if word id == None:
      seq score matrix[i,0] = tags prob[i] * alpha
    else:
      seq_score_matrix[i,0] = tags_prob[i] * emission_matrix[i,word_id]
    back_pointer[i,0] = 0
  # Second step is Iteration
  for p in range(len(sent_list)):
    if p! = 0:
      for i in range(tags.id):
        word_id = words.retrive(sent_list[p].lower(), 'temp')
        transmission vector = np.multiply(seq score matrix[:, p-1], transmission matrix[:, p-1], transmission
        tag max arg = np.argmax(transmission vector)
        # print(len(transmission_vector
        back_pointer[i,p] = tag_max_arg
        # print(back_pointer)
        if word_id == None:
          seq_score_matrix[i,p] = transmission_vector[tag_max_arg] * alpha
        else:
          seq_score_matrix[i,p] = transmission_vector[tag_max_arg] * emission_matrix
  # Third Step is Sequence Identification
  tag_index = np.zeros([len(sent_list)])
  tag index[-1] = np.argmax(seg score matrix[:, len(sent list)-1])
  # print(back pointer)
  # print(tag index)
```

```
# print(tag index)
  return [tags.retrive(index, 'id') for index in tag_index]
import re
sent = 'the man saw bank near river bank'
sent = re.findall( r'\w+|[^\s\w]+', sent)
print(get_pos(sent, transmission_matrix,emission_matrix,tags_prob, words, tags))
    ['DET', 'NOUN', 'VERB', 'NOUN', 'ADP', 'NOUN', 'NOUN']
def prediction(test corpus, transmission matrix, emission matrix, tags prob, words
  confusion matrix = np.zeros([tags.id, tags.id], dtype=np.int32)
  for test sent in test corpus:
   test tag = [item[1] for item in test sent]
    test token = [item[0] for item in test sent]
   predicted_tag = get_pos(test_token,transmission_matrix, emission_matrix, tags_|
    for (predicted, test) in zip(predicted_tag, test_tag):
      confusion matrix[tags.retrive(predicted, 'temp'), tags.retrive(test, 'temp')] .
  return confusion matrix
```

for i in reversed(range(len(sent list)-1)):

tag\_index[i] = back\_pointer[int(tag\_index[i+1]), int(i+1)]

### Cross Validation

```
def cross_validation():
    tags = get_tag()
    confusion_matrix = np.zeros([tags.id,tags.id], dtype=np.int32)
    dataset = np.array(brown.tagged_sents(tagset='universal'))
    kfold = KFold(n_splits=5,shuffle=True)
    kfold.get_n_splits(dataset)

for train, test in kfold.split(dataset):
    train_courpus = dataset[train]
    test_courpus = dataset[test]
    print("Train Data Size: ",len(train))
    print("Test Data Size: ",len(test))
    words = get_word(train_corpus)
```

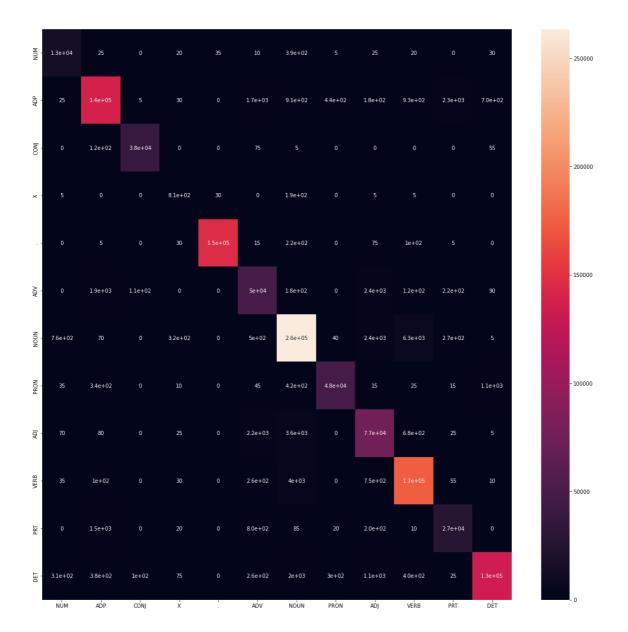
```
transmission_matrix,emission_matrix,tags_prob = get_hmm_matrix(train_corpus,wo
   confusion matrix += prediction(test corpus, transmission matrix, emission matrix
  return confusion matrix
confusion matrix = cross validation()
                  | 12/12 [00:00<00:00, 84307.62it/s]
    /usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:4: VisibleDeprec
      after removing the cwd from sys.path.
    Train Data Size:
                      45872
    Test Data Size:
                      11468
                     45872/45872 [00:00<00:00, 52212.35it/s]
    100%|
    100%|
                     45872/45872 [00:03<00:00, 13956.03it/s]
    Train Data Size:
                      45872
    Test Data Size:
                      11468
                     45872/45872 [00:00<00:00, 52418.30it/s]
    100%|
    100%
                     45872/45872 [00:03<00:00, 14100.38it/s]
    Train Data Size:
                      45872
    Test Data Size:
                      11468
    100%|
                     45872/45872 [00:00<00:00, 52221.21it/s]
                     45872/45872 [00:03<00:00, 13901.27it/s]
    100%|
    Train Data Size:
                      45872
    Test Data Size:
                      11468
                     45872/45872 [00:00<00:00, 51917.43it/s]
    100%|
                     45872/45872 [00:03<00:00, 13971.56it/s]
    100%
    Train Data Size:
                       45872
    Test Data Size:
                      11468
                     45872/45872 [00:00<00:00, 52393.04it/s]
    100%
    100%|
                      45872/45872 [00:03<00:00, 13939.98it/s]
```

# overall accuracy

```
#overall accuracy
total_examples = np.sum(confusion_matrix)
correct_predictions = np.trace(confusion_matrix)
print('The overall accuracy of the hmm model is:', correct_predictions * 100 / total
```

The overall accuracy of the hmm model is: 96.0176570693066

```
# plotting the heat map
plt.figure(figsize = (20, 20))
tag_list = [tags.retrive(i, 'id') for i in range(tags.id)]
confusion_figure = sns.heatmap(confusion_matrix, annot=True, xticklabels=tag_list,
```



```
per_pos_dict = {'tag': [], 'precision': [], 'recall': [], 'f1-score': []}
for tag_id in range(tags.id):
    per_pos_dict['precision'].append(confusion_matrix[tag_id, tag_id] / np.sum(confusion_per_pos_dict['recall'].append(confusion_matrix[tag_id, tag_id] / np.sum(confusion_per_pos_dict['tag'].append(tags.retrive(tag_id, 'id'))
    per_pos_dict['f1-score'].append(2 * per_pos_dict['precision'][tag_id] * per_pos_der_pos_df = pd.DataFrame(per_pos_dict)
    per_pos_df.to_csv('hmm_per_pos_accuracy.csv')
```

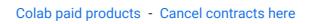
```
print(per_pos_df)
                           recall f1-score
         tag
              precision
    0
         MUM
               0.959971 0.915474 0.937195
    1
         ADP
               0.950139 0.968298 0.959132
    2
                         0.994389 0.993935
        CONJ
               0.993482
    3
           Χ
               0.775120 0.589091 0.669421
    4
               0.996934
                         0.999556
                                   0.998243
    5
         \mathsf{ADV}
               0.908718 0.893963 0.901280
    6
        NOUN
               0.960876 0.956444
                                   0.958655
    7
        PRON
               0.959453
                         0.983518
                                   0.971336
    8
         ADJ
               0.920172 0.914870 0.917513
    9
        VERB
               0.970720 0.952881
                                   0.961718
    10
         PRT
               0.909153 0.900604
                                   0.904858
    11
         DET
               0.964596 0.985229 0.974803
Overall precision = np.sum(per pos df['precision'])/12
Overall recall = np.sum(per pos df['recall'])/12
f1_score = 2 * Overall_precision * Overall_recall / ( Overall_precision + Overall_
```

```
f_half_score = 1.25 * Overall_precision * Overall_recall / ( (0.25*Overall_precision))
f2_score = 5 * Overall_precision * Overall_recall / ( (4*Overall_precision) + Overall_precision) + Overall_recall / ( (4*Overall_precision) + Overall_precision) + Overall_precision * Overall_
```

- ▼ Fbeta = ((1 + beta^2) \* Precision \* Recall) / (beta^2 \* Precision + Recall)
  - F0.5-Measure (beta=0.5): More weight on precision, less weight on recall.
  - F2-Measure (beta=2.0): Less weight on precision, more weight on recall
  - F1-Measure (beta=1.0): Balance the weight on precision and recall.

```
print("Precision: ",Overall_precision)
print("Recall: ",Overall_recall)
print("F1_score: ",f1_score)
print("F0.5_score: ",f_half_score)
print("F2 score: ",f2 score)
```

Precision: 0.9391110574859298 Recall: 0.9211929771128807 F1 score: 0.9300657255970051 F0.5 score: 0.935471893707086 F2\_score: 0.9247216838236312



×