Name: Parthvi Shah NetID: pss434

```
In [ ]: import pandas as pd
   import numpy as np
   import os
   import matplotlib.pyplot as plt
```

Question 1:

```
In []: train_data = pd.read_csv('sample_data/spam_train (1).txt', header= None)
    test_data = pd.read_csv('sample_data/spam_test.txt', header= None)

In []: trainn = train_data.iloc[:,0]
    label = trainn.str.split(' ').str[0]
    email = trainn.str.split(' ').str[1:]

In [33]: train_size = int(len(train_data)*0.8)
    X_train = email[:train_size]
    X_val = email[:train_size]
    Y_train = label[:train_size]
    Y_val = label[:train_size:]
    X_train.shape

Out[33]: (4000,)

In []: test = test_data.iloc[:,0]
    label_test = test.str.split(' ').str[0]
    email_test = test.str.split(' ').str[1:]
```

If we had not created Validation Set, Our model could risk overfitting. We need to hypertune the parameters, in this case, the number of iterations. After finding the best model and itertions, we will fit our model to the test.

Question 2:

```
In [ ]: def preprocess(data):
          vocab = {}
          data = data.to_dict()
          for key,values in data.items():
              temp = set()
              for v in values:
                  temp.add(v)
              for t in temp:
                  if t in vocab.keys():
                      vocab[t] += 1
                  else:
                      vocab[t] = 1
              temp.clear()
          for key,value in list(vocab.items()):
            if (vocab[key] < 30):
              vocab.pop(key)
          return vocab
```

```
In [ ]: vocab_train = preprocess(X_train)
```

```
In []: def featurevectorl(data, vocab):
    vocablist = list(vocab)
    sorted_items = sorted(vocablist)
    data = data.to_dict()
    results = []
    for item in data.values():
        featurevector = [0 for i in range(0, len(vocablist))]
        for i in item:
        if i in vocablist:
            one_hot_index = sorted_items.index(i)
            featurevector[one_hot_index] = 1
        results.append(featurevector)
        return np.asarray(results)
```

```
In [ ]: fvarr_train = featurevector1(X_train, vocab_train)
```

Question 3:

```
In [ ]: def perceptron_train(data, fvarr, vocab):
          data = data.astype(int)
          data = list(data.replace([0,1],[-1,1]))
          vocablist = list(vocab)
          w = np.zeros(len(fvarr[0]))
          iteration = 0
          mistake = 0
          All True = False
          while not All True:
            iteration += 1
            k = 0
            for i,x in enumerate(fvarr):
              if (int(np.dot(fvarr[i], w)*float(data[i]))) <= 0:</pre>
                k += 1
                w += np.dot(data[i], fvarr[i])
                mistake += 1
            if k==1:
              All True = True
            else:
              All True = False
          return w, mistake, iteration
```

```
In []: def perceptron_test(w, vector, label, n):
    label = label.astype(int)
    label = list(label.replace([0,1],[-1,1]))
    errors = []
    error = 0
    for i,x in enumerate(vector):
        if (int(np.dot(vector[i], w)*float(label[i]))) < 0:
            error += 1
        fraction = error/len(vector)</pre>
```

Question 4:

```
In [ ]: w,k,iteration = perceptron_train(Y_train, fvarr_train, vocab_train)
    fraction,error = perceptron_test(w, fvarr_train, Y_train, iteration)

In [48]: print('The number of mistakes in training set:',k,'& the number of iterations to converge:'
    ,iteration)

The number of mistakes in training set: 448 & the number of iterations to converge: 11
```

```
In [49]: print('Fraction of error with training set', fraction)
```

```
In [ ]: | vocab_val = preprocess(X_val)
           fvarr val = featurevector1(X val, vocab train)
  In [ ]: fraction,errors = perceptron_test(w, fvarr_val, Y_val, iteration)
 In [52]: print('Fraction of error with validation set', fraction)
           Fraction of error with validation set 0.013
Question 5:
  In [ ]: vtk = list(sorted(vocab train.keys()))
           wl = list(w)
           s = dict(zip(vtk, wl))
  In [ ]: sorted_x = sorted(s.items(), key=lambda kv: kv[1], reverse=False)
  In [ ]: mostnegative = []
           mostpositive = []
           for i in range(1,16):
             mostnegative.append(sorted_x[i])
             mostpositive.append(sorted x[-i])
 In [56]: mostnegative , mostpositive
 Out[56]: ([('reserv', -15.0),
             ('prefer', -14.0),
             ('copyright', -13.0),
             ('i', -12.0),
             ('still', -12.0),
             ('technolog', -12.0),
             ('but', -11.0),
             ('comput', -11.0),
             ('recipi', -11.0),
             ('someth', -11.0),
             ('which', -11.0),
('coupl', -10.0),
             ('date', -10.0),
             ('url', -10.0),
            ('execut', -9.0)],
[('sight', 22.0),
('click', 18.0),
('these', 16.0),
             ('remov', 16.0),
             ('market', 16.0),
             ('our', 15.0),
             ('deathtospamdeathtospam', 14.0),
             ('most', 13.0),
             ('yourself', 12.0),
             ('present', 12.0),
             ('parti', 12.0),
             ('ever', 12.0),
('pleas', 11.0),
             ('guarante', 11.0),
             ('check', 11.0)])
```

Question 6:

```
In [ ]: def average_perceptron(data, fvarr, vocab):
          data = data.astype(int)
          data = list(data.replace([0,1],[-1,1]))
          vocablist = list(vocab)
          w = np.zeros(len(fvarr[0]))
          #Number of mistakes
          k = 0
          mistake = 0
          #Number of passes through the data
          iteration = 0
          All True = False
          weightstotal = 0
          while not All_True:
            iteration += 1
            k = 0
            for i,x in enumerate(fvarr):
              if (int(np.dot(fvarr[i], w)*float(data[i]))) <= 0:</pre>
               k += 1
                w += np.dot(data[i], fvarr[i])
                weightstotal+= w
                mistake += 1
              else:
                w=w
                weightstotal += w
            if k==1:
              All_True = True
            else:
             All_True = False
          w = weightstotal/(iteration*len(fvarr))
          return w,mistake,iteration
```

```
In [ ]: w,k,iteration = average_perceptron(Y_train, fvarr_train, vocab_train)
```

Question 7:

```
In [ ]: N = [100, 200, 400, 800, 2000, 4000]
```

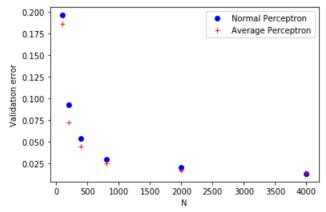
```
In [64]: for i in N:
    vocab_trainnl = preprocess(X_train[:i])
    fvarr_trainnl = featurevectorl(X_train[:i], vocab_trainnl)
    w,k,iteration = perceptron_train(Y_train[:i], fvarr_trainnl[:i], vocab_trainnl)
    wa,ka,iterationa = average_perceptron(Y_train[:i], fvarr_trainnl[:i], vocab_trainnl)
    fvarr_val = featurevectorl(X_val,vocab_trainnl)

    fractionnormal,errorsnormal = perceptron_test(w, fvarr_val, Y_val, iteration)
    fractionaverage,errorsaveraged = perceptron_test(wa, fvarr_val, Y_val, iterationa)

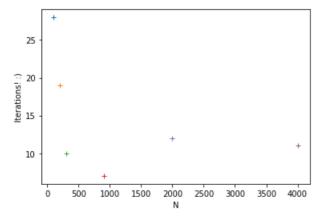
    plt.plot(i,fractionnormal, 'bo', color='blue', label="Normal Perceptron" if i == 100 else
    '')

    plt.plot(i, fractionaverage, 'r+', color='red', label = 'Average Perceptron' if i == 100
    else '')

    plt.xlabel('N')
    plt.ylabel('Validation error')
    plt.legend()
    plt.show()
```



Question 8:



```
In [ ]: def tryperceptron train(data, fvarr, vocab, iteration):
          data = data.astype(int)
          data = list(data.replace([0,1],[-1,1]))
          vocablist = list(vocab)
          w = np.zeros(len(fvarr[0]))
          itera = 0
          mistake = 0
          All_True = False
          while not All_True and itera < iteration :</pre>
            itera += 1
            k = 0
            for i,x in enumerate(fvarr):
              if (int(np.dot(fvarr[i], w)*float(data[i]))) <= 0:</pre>
                 k += 1
                 w += np.dot(data[i], fvarr[i])
                mistake += 1
             if k==0:
              All_True = True
             else:
              All True = False
          return w, mistake, iteration
```

Question 9:

```
In [ ]: def tryaverage perceptron(data, fvarr, vocab, iteration):
          data = data.astype(int)
          data = list(data.replace([0,1],[-1,1]))
          vocablist = list(vocab)
          w = np.zeros(len(fvarr[0]))
          #Number of mistakes
          k = 0
          itera = 0
          #Number of passes through the data
          All True = False
          weightstotal = 0
          while not All True and itera < iteration:</pre>
              itera += 1
              for i,x in enumerate(fvarr):
                if (np.dot(fvarr[i] ,w)*data[i]) > 0:
                  weightstotal += w
                else:
                  k += 1
                  w += np.dot(data[i], fvarr[i])
                  weightstotal+= w
              if k==0:
                All_True = True
              else:
                All_True = False
          w = weightstotal/(itera*len(fvarr))
          return w,k,iteration
```

```
In [93]: N = [10,15,20,50,100]
for i in N:
    wa,ka,iterationa = tryaverage_perceptron(Y_train, fvarr_train, vocab_train, i)
    fractiontrya,errortrya = perceptron_test(wa, fvarr_val, Y_val, iteration)
    print('Validation error on Average peceptron with iteration',iterationa,'is:',fractiontry
    a)
```

```
Validation error on Average peceptron with iteration 10 is: 0.015 Validation error on Average peceptron with iteration 15 is: 0.011 Validation error on Average peceptron with iteration 20 is: 0.012 Validation error on Average peceptron with iteration 50 is: 0.012 Validation error on Average peceptron with iteration 100 is: 0.012
```

```
In [94]:

for i in N:

wa,ka,iterationa = tryperceptron_train(Y_train, fvarr_train, vocab_train, i)

fractiontrya,errortrya = perceptron_test(wa, fvarr_val, Y_val, iteration)

print('Validation error on Normal peceptron with iteration',iterationa,'is:',fractiontrya)

Validation error on Normal peceptron with iteration 10 is: 0.013

Validation error on Normal peceptron with iteration 15 is: 0.013

Validation error on Normal peceptron with iteration 20 is: 0.013

Validation error on Normal peceptron with iteration 50 is: 0.013

Validation error on Normal peceptron with iteration 100 is: 0.013
```

So According to me, With the right number of iterations, both the algorithm work similarly at least for this data set! But I believe normal perceptron works a little better since it uses lesser number of iterations to get such a less validation error. Average perceptron works better when the number of iterations is 15.

I would use average perceptron with 15 iterations as that gives the least validation error.

Question 11:

```
In [ ]: vocab_train = preprocess(email)
    fvarr_train = featurevector1(email, vocab_train)
    w,k,iteration = tryaverage_perceptron(label, fvarr_train, vocab_train, 15)
    fvarr_test = featurevector1(email_test, vocab_train)
    fraction,error = perceptron_test(w, fvarr_test, label_test, iteration)
In [96]: print('Test set error on Average peceptron with 15 iterations is ',fraction)
```

Test set error on Average peceptron with 15 iterations is 0.016

Question 10: Optional

```
In [ ]: def tryingxpreprocess(data):
          vocab = {}
          data = data.to_dict()
          for key,values in data.items():
               temp = set()
              for v in values:
                  temp.add(v)
              for t in temp:
                  if t in vocab.keys():
                      vocab[t] += 1
                  else:
                      vocab[t] = 1
              temp.clear()
          for key,value in list(vocab.items()):
            if (vocab[key] < 15):
              vocab.pop(key)
          return vocab
```

```
In [ ]: vocab_train = tryingxpreprocess(email)
    fvarr_train = featurevector1(email, vocab_train)
    w,k,iteration = tryaverage_perceptron(label, fvarr_train, vocab_train, 11)
    fvarr_test = featurevector1(email_test, vocab_train)
    fraction,error = perceptron_test(w, fvarr_test, label_test, iteration)
```

```
In [99]: print('Test set error on changong X and applying Average peceptron with 11 iterations is ',
    fraction)
```

Test set error on changong X and applying Average peceptron with 11 iterations is 0.018

If I change my X from 30 to 10, I see an increase in the error rate.