CSE250 HW4 Q3

October 30, 2018

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In [18]: import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
         from numpy.linalg import *
         import math
In [114]: unigram = pd.read_csv('hw4_unigram.txt', sep=" ", header=None) #df for unigram
          unigram.columns = ['count(w)']
          vocab = np.loadtxt(fname = "hw4_vocab.txt", dtype='str') #array of vocabs
         bigram = pd.read_csv('hw4_bigram.txt', sep="\t", header=None)
                                                                          #df for bigram
          bigram.columns = ['index(w1)', 'index(w2)', 'count(w1,w2)']
0.1 (a)
In [115]: # compute unigram probabilities
          unigram.loc[:, 'pu(w)'] = unigram['count(w)'].apply(lambda x: x/sum(unigram['count(w
          unigram.head()
Out[115]:
            count(w)
                         pu(w)
          0 25223698 0.308490
          1 3021866 0.036958
          2 3021866 0.036958
             3855375 0.047152
             3667333 0.044852
In [86]: letters = ['A']
        output_names = [name for name in vocab if (name[0] in letters)]
In [87]: def FindWordIndex(vocab, words):
             indexes = np.zeros((len(words),1))
             for i in range(len(words)):
                 indexes[i] = np.argwhere(vocab == words[i] )
             return indexes
         indexes = FindWordIndex(vocab, output_names)
         def GetPu(indexes,unigram):
             Pu = np.zeros((indexes.shape[0],1))
```

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for i in range(indexes.shape[0]):
                Pu[i] = unigram.values[int(indexes[i,0]),1]
            return Pu
        Pu = GetPu(indexes,unigram)
In [88]: # tokens start with A and their unigram probabilities
        TableA = pd.DataFrame(data = output_names)
        TableA.columns = ['word start with A']
        TableA.loc[:,'Pu(w)'] = Pu
        TableA
Out[88]:
           word start with A
                                 Pu(w)
                            A 0.018407
        0
         1
                          AND 0.017863
         2
                           AT 0.004313
         3
                          AS 0.003992
         4
                          AN 0.002999
        5
                          ARE 0.002990
        6
                       ABOUT 0.001926
        7
                       AFTER 0.001347
        8
                         ALSO 0.001310
        9
                          ALL 0.001182
        10
                          A. 0.001026
        11
                          ANY 0.000632
        12
                     AMERICAN 0.000612
                      AGAINST 0.000596
        13
        14
                     ANOTHER 0.000428
        15
                        AMONG 0.000374
         16
                          AGO 0.000357
         17
                    ACCORDING 0.000348
         18
                          AIR 0.000311
         19
               ADMINISTRATION 0.000292
        20
                       AGENCY 0.000280
        21
                       AROUND 0.000277
        22
                    AGREEMENT 0.000263
        23
                      AVERAGE 0.000259
        24
                        ASKED 0.000258
        25
                      ALREADY 0.000249
        26
                         AREA 0.000231
        27
                     ANALYSTS 0.000226
        28
                    ANNOUNCED 0.000227
        29
                       ADDED 0.000221
        30
                    ALTHOUGH 0.000214
        31
                       AGREED 0.000212
        32
                       APRIL 0.000207
         33
                         AWAY 0.000202
```

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0.2 (b)
In [157]: #
          bigram['sum count(w1,w2)'] = bigram['count(w1,w2)'].groupby(bigram['index(w1)']).tra
          bigram['pb(w2|w1)'] = (bigram['count(w1,w2)']/bigram['sum count(w1,w2)'])
          bigram.head()
Out [157]:
             index(w1)
                        index(w2)
                                   count(w1,w2)
                                                  sum count(w1,w2)
                                                                     pb(w2|w1)
                                 1
                                         7355976
                                                           25223698
                                                                      0.291630
          1
                     1
                                 3
                                            5645
                                                           25223698
                                                                      0.000224
          2
                                 4
                     1
                                          647219
                                                           25223698
                                                                      0.025659
          3
                     1
                                 5
                                         2373160
                                                           25223698
                                                                      0.094085
          4
                     1
                                         1801245
                                                           25223698
                                                                      0.071411
In [153]: #find index of THE
          index_the = np.argwhere(vocab == "THE" )
          bigram_the = bigram[(index_the[0,0]-1)*496+1:(index_the[0,0])*499-33]
          bigram_the_5 = bigram_the.sort_values('pb(w2|w1)', ascending=False).head(5)
In [155]: word1 = vocab[0]
          word2 = vocab[69]
          word3 = vocab[78]
          word4 = vocab[72]
          word5 = vocab[60]
          likely_words = [word1,word2,word3,word4,word5]
          bigram_the_5['words to follow THE'] = likely_words
In [156]: # 5 most likely words to follow THE and their conditional probabilities
          bigram_the_5
Out[156]:
                index(w1)
                           index(w2)
                                       count(w1,w2)
                                                     sum count(w1,w2)
                                                                        pb(w2|w1) \
                                            2371132
                                                               3855375
                                                                         0.615020
          993
                        4
                                    1
                        4
                                              51556
          1058
                                   70
                                                               3855375
                                                                         0.013372
                        4
          1064
                                   79
                                              45186
                                                               3855375
                                                                         0.011720
          1060
                        4
                                   73
                                              44949
                                                               3855375
                                                                         0.011659
          1050
                        4
                                   61
                                              36439
                                                               3855375
                                                                         0.009451
               words to follow THE
          993
                              <UNK>
          1058
                                 U.
          1064
                             FIRST
          1060
                           COMPANY
          1050
                                NF.W
0.3 (c)
In [122]: # calculate Lu
          words = ['LAST','WEEK','THE','STOCK','MARKET','FELL','BY','ONE','HUNDRED','POINTS']
          index_c = FindWordIndex(vocab, words)
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Pu_c = GetPu(index_c,unigram)
                         1_u = np.log(np.prod(Pu_c))
                         print("Lu is: ", l_u)
                      -64.50944034364878
Lu is:
In [212]: # calculate Lb
                         def Pb(vocab,first,second,bigram):
                                    first_idx = np.argwhere(vocab == first)[0,0]
                                    second_idx = np.argwhere(vocab == second )[0,0]
                                    i1 = bigram.index[bigram['index(w1)'] == first_idx+1]
                                    i2 = bigram.index[bigram['index(w2)'] == second_idx+1]
                                    if bool(set(i1).intersection(i2)):
                                              i = list(set(i1).intersection(i2))[0]
                                             pb = bigram.values[i,4]
                                             print("Not found in corpus", first, second)
                                   return pb
                         1_b_c = np.log(Pb(vocab,'<s>','LAST',bigram)*Pb(vocab,'LAST','WEEK',bigram)* Pb(vocab)
                                                      *Pb(vocab, 'THE', 'STOCK', bigram) *Pb(vocab, 'STOCK', 'MARKET', bigram) *Pb(vocab
                                                      *Pb(vocab, 'FELL', 'BY', bigram) *Pb(vocab, 'BY', 'ONE', bigram) *Pb(vocab, 'ONE',
                                                      *Pb(vocab, 'HUNDRED', 'POINTS', bigram))
                         print("Lb is: ", l_b_c)
Lb is:
                      -44.740469213403735
       Lb > Lu, thus the bigram model yields higher log likelihood.
0.4 (d)
In [213]: # calculate Lu
                         words_d = ['THE','NINETEEN','OFFICIALS','SOLD','FIRE','INSURANCE']
                         index_d = FindWordIndex(vocab, words_d)
                         Pu_d = GetPu(index_d,unigram)
                         l_u_d = np.log(np.prod(Pu_d))
                         print("Lu is: ", l_u_d)
Lu is:
                      -41.64345971649364
In [214]: # calculate Lb
                         1_b_d = np.log(Pb(vocab,'<s>','THE',bigram)*Pb(vocab,'THE','NINETEEN',bigram)* Pb(vocab,'THE','NINETEEN',bigram)*
                                                      *Pb(vocab, 'OFFICIALS', 'SOLD', bigram) *Pb(vocab, 'SOLD', 'FIRE', bigram) *Pb(vocab, 'SOLD', 'FIRE', bigram) *Pb(vocab, 'SOLD', bigram) *Pb(vocab, 'SOLD', bigram) *Pb(vocab, bigram) *
```

Not found in corpus NINETEEN OFFICIALS

UnboundLocalError: local variable 'pb' referenced before assignment

For the sentence in part (d), NINETEEN OFFICIALS is not observed in the training corpus, thus the Pb of this pair is zero. This will cause the log likelihood of bigram model to be minus infinity.