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HW6 Q3
In [1]: import numpy as np
        import pickle
        import matplotlib.pyplot as plt
        import copy
        from tabulate import tabulate
        import pandas as pd
In [2]: with open('alice spelling.pkl','rb') as f:
           u = pickle. Unpickler(f)
            u.encoding = 'latin1'
            data = u.load()
        #Take a look at how the data looks, and let's make some helper functions.
        # data = pickle.load(open('alice spelling.pkl','rb'))
        vocab = np.unique(data['corpus'])
        V = len(vocab)
        ## CORRECT VS INCORRECT CORPUS
        ##For now, we will hold onto both the correct and incorrect corpuses. Later, you will only process the
         incorrect corpus, and the correct corpus is only used as a reference to check for recovery accuracy.
        def recovery rate(new corpus, correct corpus):
            wrong = 0
            for k in range(len(new_corpus)):
                if new corpus[k] != correct corpus[k]:
                    wrong += 1
            return 1.- wrong/(len(new corpus)+0.)
        print('current recovery rate', recovery rate(data['corpus'], data['corrupted corpus'] ))
        ## Probability of a word mispelling
        ## We will use the following function to predict whether a misspelled word was actually another word.
        # To avoid numerical issues, we make sure that the probablity is always something nonzero.
        def prob correct(word1,word2):
            SMALLNUM = 0.000001
            if len(word1) != len(word2): return SMALLNUM
            num wrong = np.sum(np.array([word1[k] == word2[k] for k in range(len(word1))]))
            return np.maximum(num_wrong / (len(word1)),SMALLNUM)
        # print('prob not misspelling alice vs alace', prob correct('alice', 'alice'))
        print('prob not misspelling alice vs alace', prob correct('alice', 'alace'))
        print('prob not misspelling alice vs earth', prob correct('alice', 'earth'))
        print('prob not misspelling machinelearning vs machinedreaming', prob correct('machinelearning', 'machin
        edreaming'))
        print('prob not misspelling machinelearning vs artificalintell', prob correct('machinelearning','artifi
        calintell'))
        ##HASHING
        #all of our objects should be vectors of length V or matrices which are V x V.
        #the kth word in the vocab list is represented by the kth element of the vector, and the relationship b
        etween the i,jth words is represented in the i,jth element in the matrix.
        # to easily go between the word indices and words themselves, we need to make a hash table.
        vocab hash = {}
        for k in range(len(vocab)):
            vocab hash[vocab[k]] = k
        \#now, to access the \$k\$th word, we do vocab[k]. To access the index of a word, we call vocab hash[word]
        d].
        current recovery rate 0.7716434266712013
        prob not misspelling alice vs alace 0.8
        prob not misspelling alice vs earth 1e-06
        prob not misspelling machinelearning vs artificalintell 1e-06
In [3]: ## FILL ME IN ##
        #WORD FREQUENCY
        #create an array of length V where V[k] returns the normalized frequency of word k in the entire data c
        # Do so by filling in this function.
        def get word prob(corpus):
            wordList,countArray = np.unique(corpus, return counts=True)
            totalWords = sum(countArray)
            word prob = np.zeros(len(wordList))
            for i in range(len(wordList)):
                word prob[i] = countArray[i] / totalWords
            return word prob
        word_prob = get_word_prob(data['corpus'])
        #report the answer of the following:
        print('prob. of "alice"', word prob[vocab hash['alice']])
        print('prob. of "queen"', word prob[vocab hash['queen']])
        print('prob. of "chapter"', word prob[vocab hash['chapter']])
        def getPrevWordAndCurrentWordDict():
            prevWordAndCurrentWordDict = {}
            prevWord = data['corpus'][0]
            for i in range(1,len(data['corpus'])):
                word = data['corpus'][i]
                if prevWord not in prevWordAndCurrentWordDict:
                    prevWordAndCurrentWordDict[prevWord] = {}
                    prevWordAndCurrentWordDict[prevWord][word] = 1
                elif word not in prevWordAndCurrentWordDict[prevWord]:
                    prevWordAndCurrentWordDict[prevWord] [word] = 1
                elif word in prevWordAndCurrentWordDict[prevWord]:
                    prevWordAndCurrentWordDict[prevWord] [word] += 1
                else:
                    print("Shouldn't happen")
                prevWord = word
            return prevWordAndCurrentWordDict
        ## FILL ME IN ##
        # Pr(word | prev word)
        # Using the uncorrupted corpus, accumulate the conditional transition probabilities. Do so via this for
        # pr(word | prev) = max(# times 'prev' preceded 'word' , 1) / # times prev appears
        # where again, we ensure that this number is never 0 with some small smoothing.
        def get_transition_matrix(corpus):
            transition_matrix = np.zeros((len(vocab),len(vocab)))
            wordList,countArray = np.unique(corpus, return_counts=True)
            prevWordAndCurrentWordDict = getPrevWordAndCurrentWordDict()
            for word in range(len(vocab)):
                wordString = wordList[word]
                for prevWord in range(len(vocab)):
                    prevWordString = wordList[prevWord]
                    prevWordPreceded = 0
                    if wordString in prevWordAndCurrentWordDict[prevWordString]:
                        prevWordPreceded = prevWordAndCurrentWordDict[prevWordString] [wordString]
                    occurences = max(prevWordPreceded, 1)
                    transition matrix[word] [prevWord] = occurences / countArray[prevWord]
            return transition matrix
        transition_matrix = get_transition_matrix(data['corpus'])
        print('prob. of "the alice"', transition matrix[vocab hash['alice'], vocab hash['the']])
        print('prob. of "the queen"', transition_matrix[vocab_hash['queen'], vocab_hash['the']])
        print('prob. of "the chapter"', transition_matrix[vocab_hash['hatter'], vocab_hash['the']])
        prob. of "alice" 0.014548615047424706
        prob. of "queen" 0.002569625514869818
        prob. of "chapter" 0.0009069266523069947
        prob. of "the alice" 0.0006105006105006105
        prob. of "the queen" 0.03968253968253968
        prob. of "the chapter" 0.031135531135531136
In [4]: | #The prior probabilities are just the word frequencies
        prior = word prob
        #write a function that returns the emission probability of a potentially misspelled word, by comparing
         its probabilities against every word in the correct vocabulary
        def get emission(mword):
            emission_prob = np.zeros(len(vocab))
            for index, word in enumerate(vocab):
                emission_prob[index] = prob_correct(mword, word)
            return emission prob
        #find the 10 closest words to 'abice' and report them
        idx = np.argsort(get emission('abice'))[::-1]
        print([vocab[j] for j in idx[:10]])
        ['abide', 'alice', 'above', 'voice', 'alive', 'twice', 'thick', 'dance', 'stick', 'prize']
In [5]: #now we reduce our attention to a small segment of the corrupted corpus
        corrupt_corpus = data['corrupted_corpus'][:1000]
        correct corpus = data['corpus'][:1000]
In [6]: def normalize(vector):
            return vector/np.sum(vector)
In [7]: | # encode the HMM spelling corrector.
        # To debug, you can see the first hundred words of both the corrupted and proposed corpus,
        # to see if spelling words got corrupted.
        # report the recovery rate of the proposed (corrected) corpus.
        totalStates = len(transition_matrix)
        node values fwd = np.zeros((len(corrupt corpus), totalStates))
        for i, sequence val in enumerate(correct corpus):
            if (i == 0):
                word prob = get word prob(data['corpus'])
                start probs = word prob[vocab hash[sequence val]]
                emission = get emission(sequence val)
                firstStateBeforeNormalisation = start probs * emission
                node_values_fwd[i, :] = normalize(firstStateBeforeNormalisation)
                emission = get emission(sequence val)
                nextStateBeforeNormalization = np.multiply(emission,np.dot(transition matrix ,node values fwd[i
        -1, :]))
                node values fwd[i, :] = normalize(nextStateBeforeNormalization)
        totalStates = len(transition matrix)
        node_values_bwd = np.zeros((len(corrupt_corpus), totalStates))
        for i, e in reversed(list(enumerate(corrupt corpus))):
            if (i == len(corrupt corpus)-1):
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nextStateBeforeNormalization = np.multiply(emission,np.dot(transition matrix ,node values bwd[i

word prob = get word prob(data['corpus'])

emission = get emission(sequence val)

emission = get emission(sequence val)

+---+

+1, :]))

start_probs = word_prob[vocab_hash[sequence_val]]

firstStateBeforeNormalisation = start probs * emission

node_values_bwd[i, :] = normalize(firstStateBeforeNormalisation)

node_values_bwd[i, :] = normalize(nextStateBeforeNormalization)