Fresh start

November 26, 2018

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In [1]: # these are all of our imports
        import re
        import string
        from nltk.corpus import stopwords
        from nltk.stem import WordNetLemmatizer as wnl
        from nltk.metrics.distance import edit_distance as lev
In [2]: # contractions can be miss categorized as incorrect when you strip and filter the data
        #fixes are uncontracted contractions
        #stop_en and stop_germ are the stopwords
contractions = ["´","'","'","it's", "he's","she's","that's", "what's", "there's", "aren't",\
                     "[newline]", "'m", "'ve", "n't", "'ll", "'re", "won't", "'d", "geht's", "gibt's", "'s", ' xd']
        fixes = ["'","'","'","it is", "he is","she is","that is", "what is", "there is","are not",\
" ", " am", " have", " not", " will", " are", "will not", " would", "geht es", "gibt es", "", ' ']
        stop_en = set(stopwords.words('english'))
        stop_germ = set(stopwords.words('german'))
In [3]: #opening our german and english dictionairies
        with open('german.dic', 'r', encoding='latin-1') as f:
            for row in f:
                 if len(row) >1:
                     de_dic[row.strip().lower()] = 1
        en_dic = \{\}
        with open('english.dic', 'r',) as f:
            for row in f:
                 en_dic[row.strip().lower()] = 1
In [4]: #words that are not in the english dictionary but should be- if we don't do this then these words will be erroneosly
        #labeled as misspells
        en_dic['anymore'] = 1
        en_dic['adhd'] = 1
        en_dic['asshole'] = 1
        en_dic['fucking'] = 1
        en_dic['porn'] = 1
        en_dic['fuck'] = 1
        en_dic['proud'] = 1
        en_dic['others'] = 1
        en_dic['mom'] = 1
        en_dic['ptsd'] = 1
        en_dic['europe']= 1
        en_dic['tumour'] = 1
        en_dic['tumours'] = 1
        en_dic['stats'] = 1
        en_dic['favourite'] = 1
        en_dic['boyfriend'] = 1
        en_dic['fortnite'] = 1 #qame name
        en_dic['bts'] = 1 # Korean pop group
        #we considered also doing this for acronyms because they are not accidental misspells they are just not in our dictionary
        #but utimately for the sake of the assignment and the dictionary given to us we decided not to input them into our dictionary
        \# en_dic['lol'] = 1 \#laugh out loud
        # en_dic['omg'] = 1 #oh my god
        \# en_dic['af'] = 1 \#as fuck
        # en_dic['tbh'] = 1 #to be honest
        # en_dic['bc'] = 1 #because
        # en_dic['idk'] = 1 #i don't know
        # en_dic['rn'] = 1 #right now
        # en_dic['ppl'] = 1 #people
        # en_dic['lmao'] = 1 #laugh my ass off
        \# en_dic['wtf'] = 1 \#what the fuck
        # en_dic['btw'] = 1 #by the way
        # en_dic['pls'] = 1 #please
        \# en_dic['thx'] = 1 \#thanks
        #en_dic['aml'] = 1 #anti-money laundering
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In [5]: # Terms not in German Dictionary that should be, we are all native English speakers
                #so there could be more than just this
                de_dic['sowas'] = 1
In [6]: # to open and read our tweets to begin the preprocessing step
                raw_text = open('tweets.csv').read()
                tab\_seperated = [item\_split('\t') for item in raw\_text.split('\n') if len(item\_split('\t')) >= 4]
In [7]: #Non-English or German characters, used to filter out foreign tweets
               non\_ende\_char = \left[ \text{"$a'', "$a'', "$a'', "$a'', "$e'', "$e'', "$e'', "$i'', "$i'', "$i'', "$o'', "$o'', "$o'', "$o'', "$o'', "$u'', 
In [8]: #Building our tweet filter
               filtered = []
               for i in tab_seperated:
                       for char in non_ende_char:
                               if char in i[4]:
                                      filtered.append(i[1])
In [9]: #we use regex to remove any websites, numbers, tagged twitter users, hashtags,
                #and basically anything that is not in the English and German alphabets
               data = \{\}
                for i in range(len(tab_seperated)):
                       #extracting the filtered tweets and IDs
                       if tab_seperated[i][1] not in filtered:
                               data[tab_seperated[i][1]] = tab_seperated[i][4]
                #Preprocessing and breaking apart common contractions
                data_index = data.copy()
                for tweet in data_index.keys():
                       data_index[tweet] = data_index[tweet].lower()
                       for i in range(len(contractions)):
                               if contractions[i] in data_index[tweet]:
                                       data_index[tweet] = data_index[tweet].replace(contractions[i], fixes[i])
                      data_index[tweet] = data_index[tweet].replace(contractions].

data_index[tweet] = re.sub('https?[^\s]+', ' ', data_index[tweet])

data_index[tweet] = re.sub('[0+][^\s]+', ' ', data_index[tweet])

data_index[tweet] = re.sub(r'\w+\.[^\s]+', ' ', data_index[tweet])

data_index[tweet] = re.sub(r'\w+\.[^\s]+', ' ', data_index[tweet])

data_index[tweet] = re.sub(r'[^a-zäöüßs]', ' ', data_index[tweet])
                       data_index[tweet] = re.sub(r'[^\w\s]', ' ', data_index[tweet])
                #Here we are building our dict of terms and frequenceis, as well as a dict with sorted docIDs
                terms = {}
                term index = {}
                for num,tweet in data_index.items():
                       for word in tweet.split():
                               if word in term_index:
                                       term_index[word].append(num)
                               elif word not in term_index:
                                      term_index[word] = [num]
                               if word in terms:
                                       terms[word] += 1
                               else:
                                       terms[word] = 1
                for key in term_index.keys():
                        term_index[key] = sorted(term_index[key])
In [10]: #this function determines if a word is German or English
                 #Given on number of tweets within the postings list based on stop words and foreign characters
                 def is_language(term):
                                 de_char = ['ä','ö','ü','ß']
                                 de score = 0
                                 en_score = 0
                                 for post in term_index[term]:
                                         for i in data_index[post].strip().split():
                                                 for char in de_char:
                                                         if char in i:
                                                                de_score +=1
                                                 if i in stop_germ:
                                                         de_score += 1
                                                 if i in stop_en:
                                                       en_score += 1
                                 if de_score>en_score:
                                        return 'german'
                                 elif de_score<en_score:
                                        return 'english'
                                         return None
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In [11]: #This function is used to find which words are the most frequent in there respective languages
         def top_freq(dict):
                 freq = []
                 for term, val in dict.items():
                     freq.append((val, term))
                 freq = sorted(freq)[::-1]
                 freq_de = []
                 freq_en = []
                 for i,j in freq:
                     if is_language(j) == 'german':
                         freq_de.append(j)
                     else:
                         freq_en.append(j)
                 return freq_de, freq_en
In [12]: #Generated the english and german terms, sorted by descending frequency
         top_de, top_en = top_freq(terms)
In [13]: #This function is used to find out if the tweet is English or German
         #Here we use a refined search to better identify if a tweet is english or Germn
         #given stop words and our most frequent words occuring in English or German
         #If it fails to identify, it will return none
         def language(post):
                 de_char = ['ä','ö','ü','ß']
                 de_score = 0
                 en score = 0
                 for i in data_index[post].split():
                     for char in de_char:
                             if char in i:
                                 de_score +=1
                     if i in stop_germ or i in top_de[:200]:
                         de_score += 1
                     if i in stop_en or i in top_en[:200]:
                        en_score += 1
                 if de_score>en_score:
                     return 'german'
                 elif en_score>de_score:
                     return 'english'
                     return None
In [14]: #This function finds misspelled words
         def get_misspells():
                 #Here we generate a list of terms and list to partition them in
                 words =sorted([key for key in terms.keys()])
                 de = []
                 en = []
                 correct_en = []
                 correct_de = []
                 for word in words:
                    g_count = 0
                     e_count = 0
                     e_posts = []
                     g_posts = []
                     {\it \#here we sort correct words separate correct\_addicitonal\ terms}
                     if word in en_dic:
                         correct_en.append(word)
                         continue
                     elif word in de_dic:
                         correct_de.append(word)
                         continue
                     *pos tags for lemmatization generation
                     tags = ['n','v','a','s','r']
                     en_lemma = \{\}
                     for tag in tags:
                         lemma = wnl.lemmatize(wnl,word=word, pos=tag)
                         en_lemma[lemma] = 1
                       #German lemmatiziaton/stemming
                     de lemma = word
                 #here we search post by post to determin if term in index
                 # is German or English, to see if in the context it is a misspell
                     for post in term_index[word]:
                         if language(post) == 'german':
                             ## get german misspelling count
                             if de_lemma not in de_dic:
                                 if len([i for i in en_lemma.keys() if i in en_dic]) == 0:
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g_count += 1
                                     g_posts.append(post)
                             else:
                                 correct_de.append(word)
                         elif language(post) == 'english':
                             ## get german misspelling count
                             if len([i for i in en_lemma.keys() if i in en_dic]) == 0:
                                 e_count += 1
                                 e_posts.append(post)
                             else:
                                 correct_en.append(word)
                     #Here if the refined methon is not able to solve,
                     #It falls back to the likelihood of based on occurance in German or English Tweets
                         else:
                             lang = is_language(word)
                             if lang == 'german':
                                  ## get german misspelling count
                                 if de_lemma not in de_dic:
                                     if len([i for i in en_lemma.keys() if i in en_dic]) == 0:
                                         g_count += 1
                                         g_posts.append(post)
                                 else:
                                     correct_de.append(word)
                             if lang == 'english':
                                  ## get english misspelling count
                                 if len([i for i in en_lemma.keys() if i in en_dic]) == 0:
                                     e_count += 1
                                     e_posts.append(post)
                                 else:
                                     correct_en.append(word)
                     if g_count > 0:
                         de.append((g_count, word, g_posts))
                     if e_count > 0:
                         en.append((e_count, word, e_posts))
                 #retruning list of misspells and correct lists
                 return sorted(de)[::-1], sorted(en)[::-1], correct_de, correct_en
In [15]: #Here we generate our lists of misspells
         de_mis, en_mis, terms_de, terms_en = get_misspells()
In [16]: #This function calculates the Damerau distance of the misspelled words that were found in the function above
         def en_damerau(word):
             #here we reduce the workload by reducing same characters
             #that occur more than twice in a row to just two characters to increase accuracy
             term = ''
             for i in range(len(word)):
                 try:
                     if word[i] != word[i+2]:
                         term += word[i]
                 except:
                     term += word[i]
             alphabet = "abcdefghijklmnopqrstuvwxyz"
             possible = {}
             #we are genearting a dictionary of terms that are 1 damerau distance from the term
             chunks = [(term[:i], term[i:])for i in range(len(term) + 1)]
             for chunk1, chunk2 in chunks:
                 if chunk2:
                     #subtraction
                     possible[chunk1+chunk2[1:]] = 1
                     for char in alphabet:
                         #substitution
                         possible[chunk1+char+chunk2[1:]] = 1
                 if len(chunk2) > 1:
                     #transposition
                     possible[chunk1+chunk2[1]+chunk2[0]+chunk2[2:]] = 1
                 for char in alphabet:
                     #addition
                     possible[chunk1+char+chunk2] = 1
             return possible
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In [17]: #This function finds the suggested terms based on the damerau distance for English words
         def en_suggested(term):
             """Some terms such as the following can be force edited
                They are slang to represent expressions"""
               if term == 'kinda':
         #
                  return "kind of"
              if term == 'gonna':
                  return 'going to
              if term == 'wanna':
                  return 'want to'
             suggestions = en_damerau(term)
             suggested = []
             #Lemmatizing the suggestions to get a more accurate reference
             for i in suggestions.keys():
                 tags = ['n','v','a','s','r']
                 lemmas = []
                 for tag in tags:
                    lemmas.append(wnl.lemmatize(wnl,word=i, pos=tag))
                 lemmas = set(lemmas)
                 for j in lemmas:
                     if j in en_dic:
                        suggested.append(i)
             #Refining the words that were suggested using edited distance again
             refined = [word for word in suggested if lev(term, word) == min(lev(term,word) for word in suggested)]
                 #further refining the search
                 best = sorted([(terms[word],word) for word in refined if word in terms_en])[::-1]
                 return [i for j,i in best][:3]
             except:
                 return refined
In [18]: \#This\ function\ finds\ the\ damerau\ distance\ for\ the\ misspelled\ German\ words
             #here we reduce the workload by reducing same characters
             #that occur more than twice in a row to just two characters to increase accuracy
             term = ''
             for i in range(len(word)):
                 try:
                    if word[i] != word[i+2]:
                         term += word[i]
                 except:
                     term += word[i]
             alphabet = "abcdefghijklmnopqrstuvwxyzäöüß"
             possible = {}
             #we are genearting a dictionary of terms that are 1 damerau distance from the term
             chunks = [(term[:i], term[i:])for i in range(len(term) + 1)]
             for chunk1, chunk2 in chunks:
                 if chunk2:
                     #subtraction
                     possible[chunk1+chunk2[1:]] = 1
                     for char in alphabet:
                         #substitution
                         possible[chunk1+char+chunk2[1:]] = 1
                 if len(chunk2) > 1:
                     #transposition
                     possible[chunk1+chunk2[1]+chunk2[0]+chunk2[2:]] = 1
                 for char in alphabet:
                     #addition
                     possible[chunk1+char+chunk2] = 1
             return possible
In [19]: ##This function finds the suggested terms based on the damerau distance for German words
         def de_suggested(term):
             suggestions = de_damerau(term)
             suggested = []
             for i in suggestions:
                if i in de_dic:
                     suggested.append(i)
             #Refining the words that were suggested using edited distance again
             refined = [word for word in suggested if lev(term, word) == min(lev(term,word) for word in suggested)]
             try:
                 #further refining the search
                 best = sorted([(terms[word],word) for word in refined if word in de_dic and word in terms_de])[::-1]
                 return [i for j,i in best][:3]
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except:
                    return refined
In [20]: #This gives us the top misspelled English words
           top_mis_en = []
          for count, word, posts in en_mis[:10]:
               top_mis_en.append((word, count, en_suggested(word)))
In [21]: #it should be noted that even though some of the outputted words were not intentional misspells they were not in our dictionary
          #so they were considered misspells we ultimately decided to leave them out in order to comply to the constraints
           #of the dictionarys that were given to us in the assignment
          top_mis_en
Out[21]: [('kinda', 2290, ['kind', 'linda']),
            ('bc', 712, ['be', 'by', 'b']),
            ('gonna', 492, ['donna', 'gonne', 'gona']), ('lol', 395, ['vol', 'lo', 'pol']),
            ('omg', 377, ['om', 'og']),
           ('wanna', 330, ['anna', 'canna', 'hanna']),
('rn', 297, ['in', 'an', 'on']),
('tbh', 284, ['th', 'tch']),
('idk', 261, ['id', 'ink', 'ilk']),
('ppl', 217, ['pol', 'pal'])]
In [22]: #This gives us the top misspelled German words
          top_mis_de = []
          for count, word, posts in de_mis[:10]:
               top_mis_de.append((word, count, de_suggested(word)))
In [23]: #Top 10 German misspells
          top_mis_de
Out[23]: [('nen', 424, ['ren', 'nn']),
           ('nochmal', 294, ['nochmals']),
            ('nem', 233, ['dem', 'neu', 'nm']),
            ('erstmal', 216, ['erstmals']),
           ('lol', 214, ['hol', 'mol', 'aol']),
('daß', 180, ['saß', 'maß', 'dax']),
           ('gibts', 164, ['gibt', 'gifts']),
('nich', 151, ['nicht', 'noch', 'sich']),
            ('zb', 147, ['zu', 'ob', 'tb']),
            ('vllt', 146, [])]
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