## lab3 TFIDF 2

## September 25, 2023

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[1]: from sklearn.feature_extraction.text import TfidfVectorizer
      from sklearn.metrics.pairwise import cosine_similarity
[94]: import pandas as pd
[31]: documentA='the man went out for a walk'
      documentB='the children sat around the fire'
[32]: #create bag of words
      bagofwordsA=documentA.split()
      bagofwordsB=documentB.split()
[33]: bagofwordsA
[33]: ['the', 'man', 'went', 'out', 'for', 'a', 'walk']
[34]: bagofwordsB
[34]: ['the', 'children', 'sat', 'around', 'the', 'fire']
[35]: #to find the unique values
      uniquewords=set(bagofwordsA).union(set(bagofwordsB))
      uniquewords
[35]: {'a',
       'around',
       'children',
       'fire',
       'for',
       'man',
       'out',
       'sat',
       'the',
       'walk',
       'went'}
[36]: | #Next,we'll create a dictionary of words and their occurence for each documnet
       ⇔in the corpus
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[37]: numofwordsA=dict.fromkeys(uniquewords,0)
      numofwordsA
[37]: {'fire': 0,
       'walk': 0,
       'for': 0,
       'man': 0,
       'went': 0,
       'children': 0,
       'a': 0,
       'sat': 0,
       'the': 0,
       'around': 0,
       'out': 0}
[38]: #TF of each term for document 1
[39]: for word in bagofwordsA:
          if word in numofwordsA:
              numofwordsA[word]+=1
      numofwordsA
[39]: {'fire': 0,
       'walk': 1,
       'for': 1,
       'man': 1,
       'went': 1,
       'children': 0,
       'a': 1,
       'sat': 0,
       'the': 1,
       'around': 0,
       'out': 1}
[40]: #TF of each terms for document 2
[41]: numofwordsB=dict.fromkeys(uniquewords,0)
      numofwordsB
[41]: {'fire': 0,
       'walk': 0,
       'for': 0,
       'man': 0,
       'went': 0,
       'children': 0,
       'a': 0,
       'sat': 0,
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'the': 0,
       'around': 0,
       'out': 0}
[42]: for word in bagofwordsB:
          if word in numofwordsB:
              numofwordsB[word]+=1
      numofwordsB
[42]: {'fire': 1,
       'walk': 0,
       'for': 0,
       'man': 0,
       'went': 0,
       'children': 1,
       'a': 0,
       'sat': 1,
       'the': 2,
       'around': 1,
       'out': 0}
[48]: #removing the stop words
      import nltk
      from nltk.corpus import stopwords
      nltk.download('stopwords')
     [nltk_data] Downloading package stopwords to
     [nltk_data]
                      C:\Users\ASUS\AppData\Roaming\nltk_data...
     [nltk_data]
                    Package stopwords is already up-to-date!
[48]: True
[50]: stop_words=stopwords.words('english')
[51]: stop_words
[51]: ['i',
       'me',
       'my',
       'myself',
       'we',
       'our',
       'ours',
       'ourselves',
       'you',
       "you're",
       "you've",
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"you'll",
"you'd",
'your',
'yours',
'yourself',
'yourselves',
'he',
'him',
'his',
'himself',
'she',
"she's",
'her',
'hers',
'herself',
'it',
"it's",
'its',
'itself',
'they',
'them',
'their',
'theirs',
'themselves',
'what',
'which',
'who',
'whom',
'this',
'that',
"that'll",
'these',
'those',
'am',
'is',
'are',
'was',
'were',
'be',
'been',
'being',
'have',
'has',
'had',
'having',
'do',
'does',
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'did',
'doing',
'a',
'an',
'the',
'and',
'but',
'if',
'or',
'because',
'as',
'until',
'while',
'of',
'at',
'by',
'for',
'with',
'about',
'against',
'between',
'into',
'through',
'during',
'before',
'after',
'above',
'below',
'to',
'from',
'up',
'down',
'in',
'out',
'on',
'off',
'over',
'under',
'again',
'further',
'then',
'once',
'here',
'there',
'when',
'where',
'why',
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'how',
'all',
'any',
'both',
'each',
'few',
'more',
'most',
'other',
'some',
'such',
'no',
'nor',
'not',
'only',
'own',
'same',
'so',
'than',
'too',
'very',
's',
't',
'can',
'will',
'just',
'don',
"don't",
'should',
"should've",
'now',
'd',
'11',
'm',
'0',
're',
've',
'y',
'ain',
'aren',
"aren't",
'couldn',
"couldn't",
'didn',
"didn't",
'doesn',
"doesn't",
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'hadn',
       "hadn't",
       'hasn',
       "hasn't",
       'haven',
       "haven't",
       'isn',
       "isn't",
       'ma',
       'mightn',
       "mightn't",
       'mustn',
       "mustn't",
       'needn',
       "needn't",
       'shan',
       "shan't",
       'shouldn',
       "shouldn't",
       'wasn',
       "wasn't",
       'weren',
       "weren't",
       'won',
       "won't",
       'wouldn',
       "wouldn't"]
[70]: \#Calculating\ the\ TF\ s\ (1+log(tf))
      #but we are not using that equation in here
      #we are using "tfDict[word]=count/float(bagofwordsCount)"
[65]: def computeTF(wordDict,bagofwords):
          import math
          tfDict={}
          bagofwordsCount=len(bagofwords)
          for word,count in wordDict.items():
              tfDict[word] = count
              tfDict[word] = count/float(bagofwordsCount)
          return tfDict
[66]: #the following lines compute the term frequency for each of our documents
      tfA=computeTF(numofwordsA,bagofwordsA)
      tfB=computeTF(numofwordsB,bagofwordsB)
```

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[67]: tfA
[67]: {'fire': 0.0,
      'walk': 0.14285714285714285,
      'for': 0.14285714285714285,
      'man': 0.14285714285714285,
      'went': 0.14285714285714285,
      'children': 0.0,
      'a': 0.14285714285714285,
      'sat': 0.0,
      'the': 0.14285714285714285,
      'around': 0.0,
      'out': 0.14285714285714285}
[68]: tfB
'walk': 0.0,
      'for': 0.0,
      'man': 0.0,
      'went': 0.0,
      'a': 0.0,
      'out': 0.0}
[71]: #Calculate the IDF values
[82]: def computeIDF(documents):
        import math
        N=len(documents)
        idfDict=dict.fromkeys(documents[0].keys(),0)
        #calculating the df
        for document in documents:
           for word, val in document.items():
               if val>0:
                  idfDict[word]+=1
        #calculating the idf
        for word,val in idfDict.items():
           idfDict[word] = math.log(N/float(val))
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```
return idfDict
[83]: idfs=computeIDF([numofwordsA,numofwordsB])
      idfs
[83]: {'fire': 0.6931471805599453,
       'walk': 0.6931471805599453,
       'for': 0.6931471805599453,
       'man': 0.6931471805599453,
       'went': 0.6931471805599453,
       'children': 0.6931471805599453,
       'a': 0.6931471805599453,
       'sat': 0.6931471805599453,
       'the': 0.0,
       'around': 0.6931471805599453,
       'out': 0.6931471805599453}
[84]: #calculatin the TfIdf values of each term
[87]: def computeTFIDF(tfbagofwords,idfs):
          tfidf={}
          for word,val in tfbagofwords.items():
              tfidf[word]=val*idfs[word]
          return tfidf
[90]: tfidfA=computeTFIDF(tfA,idfs)
      tfidfB=computeTFIDF(tfB,idfs)
[91]: tfidfA
[91]: {'fire': 0.0,
       'walk': 0.09902102579427789,
       'for': 0.09902102579427789,
       'man': 0.09902102579427789,
       'went': 0.09902102579427789,
       'children': 0.0,
       'a': 0.09902102579427789,
       'sat': 0.0,
       'the': 0.0,
       'around': 0.0,
       'out': 0.09902102579427789}
[92]: tfidfB
[92]: {'fire': 0.11552453009332421,
       'walk': 0.0,
       'for': 0.0,
```

```
'man': 0.0,
       'went': 0.0,
       'children': 0.11552453009332421,
       'a': 0.0,
       'sat': 0.11552453009332421,
       'the': 0.0,
       'around': 0.11552453009332421,
       'out': 0.0}
[97]: df=pd.DataFrame([tfidfA,tfidfB])
      df
      #0 - doc1
      #1 - doc2
[97]:
             fire
                       walk
                                   for
                                                      went children
                                             man
      0.000000 \ 0.099021 \ 0.099021 \ 0.099021 \ 0.099021 \ 0.000000 \ 0.099021
      1 \quad 0.115525 \quad 0.000000 \quad 0.000000 \quad 0.000000 \quad 0.115525 \quad 0.000000
              sat the
                          around
                                        out
      0 0.000000 0.0 0.000000 0.099021
      1 0.115525 0.0 0.115525 0.000000
 []:
 []:
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