Custom TFIDF

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0.1 Task - 1

```
[2]: from collections import Counter
     from scipy.sparse import csr_matrix
     import math
     import operator
     from sklearn.preprocessing import normalize
     import numpy as np
     from sklearn.feature_extraction.text import TfidfVectorizer
     def IDF(corpus, unique_words):
         HHHH
         This function calculates and return the IDF values.
         idf_dict={}
         N=len(corpus)
         for i in unique_words:
             count=0
             for sen in corpus:
                 if i in sen.split(): # checks if the word from unique_word present ⊔
      \rightarrow in the sentence
                                         # and increment if it is there
                      count=count+1
                  idf_dict[i]=1+(math.log((1+N)/(count+1))) #IDF formula
         return idf_dict
     def fit(dataset):
         11 11 11
         This function creates dict of unique words in the corpus(vocab) and returns \Box
      \hookrightarrow vocab and Idf values
         11 11 11
                                 # at first we will initialize an empty set
         unique_words = set()
         if isinstance(dataset, (list,)): # proceed if it is list
             for row in dataset:
                 for word in row.split(" "):
                      if len(word) < 2: # removes if the word is only one letter</pre>
                          continue
                      unique_words.add(word) # Add the word to the set
```

```
unique_words = sorted(list(unique_words)) # convert set to list and_
 \hookrightarrowsort it
        vocab = {j:i for i, j in enumerate(unique_words)} # dictionary of_
→unique words of the corpus
    Idfs = IDF(dataset,unique_words)
    return vocab, Idfs
def transform(dataset, vocabulary, idf_values):
    This function calculates TF-IDF values and returns normalized sparse matrix,
\hookrightarrow of those values.
    11 11 11
    sparse_matrix = csr_matrix( (len(dataset), len(vocabulary)), dtype=np.
 →float64)
    for row in range(0,len(dataset)):
        sen_words=Counter(dataset[row].split()) # words are stored as keys and
→ their counts are stored as values.
        for word in dataset[row].split():
            if word in list(vocabulary.keys()):
                tf_idf_value=(sen_words[word]/len(dataset[row].
→split()))*(idf_values[word]) #Finding TF-IDF value
                sparse_matrix[row,vocabulary[word]]=tf_idf_value # converting_
\hookrightarrow to sparse matrix
    output = normalize(sparse_matrix, norm='12', axis=1, copy=False,__
→return_norm=False) #Normalizing the final output
    return output
corpus = [
     'this is the first document',
     'this document is the second document',
     'and this is the third one',
     'is this the first document',
]
#using SKLearn
vectorizer = TfidfVectorizer()
vectorizer.fit(corpus)
skl_output = vectorizer.transform(corpus)
#Custom Code
```

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Vocabulary, idf_of_vocabulary = fit(corpus)
print("Features from SKL :-", vectorizer.get_feature_names())
print("Features from custom code:-",list(Vocabulary.keys()))
print("\n\nIDF values from SKL :-",vectorizer.idf_)
print("IDF values from custom code :-",list(idf_of_vocabulary.values()))
final_output=transform(corpus, Vocabulary, idf_of_vocabulary)
print("\n\nshape of sklearn tfidf vectorizer output: ",skl_output.shape)
print("shape of custom code final output: ",final_output.shape)
print("\n\nfinal output using SKL(First Row) :-",skl_output[0].sorted_indices())
print("\nfinal output using custom code(First Row) :-",final_output[0])
print("\n\nfinal output using SKL(First Row, Dense Matrix) :-", skl_output[0].
 →toarray())
print("\nfinal output using custom code(First Row, Dense Matrix) :
 →-",final output[0].toarray())
Features from SKL :- ['and', 'document', 'first', 'is', 'one', 'second', 'the',
'third', 'this']
Features from custom code:- ['and', 'document', 'first', 'is', 'one', 'second',
'the', 'third', 'this']
IDF values from SKL :- \[ \int 1.91629073 \] 1.22314355 1.51082562 1. \[ 1.91629073 \]
1.91629073
            1.91629073 1.
IDF values from custom code :- [1.916290731874155, 1.2231435513142097,
1.5108256237659907, 1.0, 1.916290731874155, 1.916290731874155, 1.0,
1.916290731874155, 1.0]
shape of sklearn tfidf vectorizer output: (4, 9)
shape of custom code final output: (4, 9)
                                       (0, 1) 0.46979138557992045
final output using SKL(First Row) :-
  (0, 2)
                0.5802858236844359
  (0, 3)
                0.38408524091481483
  (0, 6)
               0.38408524091481483
  (0, 8)
               0.38408524091481483
final output using custom code(First Row) :- (0, 1) 0.4697913855799205
  (0, 2)
               0.580285823684436
```

0.2 Task - 2

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[4]: def fit_50(dataset):
         This function creates dict of unique words in the corpus(vocab) and returns \Box
      \hookrightarrow vocab and Idf values
         n n n
         top50_vocab = []
         unique_words = set()  # at first we will initialize an empty set
         if isinstance(dataset, (list,)): # proceed if it is list
             for row in dataset:
                 for word in row.split(" "):
                     if len(word) < 2: # removes if the word is only one letter</pre>
                         continue
                     unique_words.add(word) # Add the word to the set
             unique_words = sorted(list(unique_words)) # convert set to list and_
      \rightarrowsort it
         Idfs = IDF(dataset,unique_words)
         #extracting the top 50 words based on IDF values
         sorted_idfs = {k: v for k, v in sorted(Idfs.items(), key=lambda item:__
      →item[1],reverse=True)} # sorting in descending order
         top50_Idfs = {k: sorted_idfs[k] for k in list(sorted_idfs)[:50]} #_U
      ⇒selecting first 50 elements from sorted idfs
         for i in unique_words:
                                           # getting the words with highest idf values
             if i in top50_Idfs.keys():
                 top50_vocab.append(i)
         top50_vocab = {j:i for i,j in enumerate(top50_vocab)} # list to dictionary
         return top50_vocab, top50_Idfs
     import pickle
     with open('cleaned_strings', 'rb') as f:
         corpus1 = pickle.load(f)
```

```
Vocabulary_50, idf_of_vocabulary_50 = fit_50(corpus1)
print("Top 50 Features:-",list(Vocabulary_50.keys()))
print("\nTop 50 IDF values:-",list(idf_of_vocabulary_50.values()))
final_output_50 = transform(corpus1, Vocabulary_50, idf_of_vocabulary_50)
print("\nshape of final output: ",final output 50.shape)
print("\nfinal output(First Row) :-",final_output_50[0])
print("\nfinal output(First Row, Dense Matrix) :-",final_output_50[0].toarray())
Top 50 Features:- ['aailiyah', 'abandoned', 'abroad', 'abstruse', 'academy',
'accents', 'accessible', 'acclaimed', 'accolades', 'accurate', 'accurately',
'achille', 'ackerman', 'actions', 'adams', 'add', 'added', 'admins',
'admiration', 'admitted', 'adrift', 'adventure', 'aesthetically', 'affected',
'affleck', 'afternoon', 'aged', 'ages', 'agree', 'agreed', 'aimless', 'aired',
'akasha', 'akin', 'alert', 'alike', 'allison', 'allow', 'allowing', 'alongside',
'amateurish', 'amaze', 'amazed', 'amazingly', 'amusing', 'amust', 'anatomist',
'angel', 'angela', 'angelina']
Top 50 IDF values:- [6.922918004572872, 6.922918004572872, 6.922918004572872,
6.922918004572872, 6.922918004572872, 6.922918004572872, 6.922918004572872,
6.922918004572872, 6.922918004572872, 6.922918004572872, 6.922918004572872,
6.922918004572872, 6.922918004572872, 6.922918004572872, 6.922918004572872,
6.922918004572872, 6.922918004572872, 6.922918004572872, 6.922918004572872,
6.922918004572872, 6.922918004572872, 6.922918004572872, 6.922918004572872,
6.922918004572872, 6.922918004572872, 6.922918004572872, 6.922918004572872,
6.922918004572872, 6.922918004572872, 6.922918004572872, 6.922918004572872,
6.922918004572872, 6.922918004572872, 6.922918004572872, 6.922918004572872,
6.922918004572872, 6.922918004572872, 6.922918004572872, 6.922918004572872,
6.922918004572872, 6.922918004572872, 6.922918004572872, 6.922918004572872,
6.922918004572872, 6.922918004572872, 6.922918004572872, 6.922918004572872,
6.922918004572872, 6.922918004572872, 6.922918004572872]
shape of final output: (746, 50)
final output(First Row) :- (0, 30)
                                      1.0
final output(First Row, Dense Matrix) :- [[0. 0. 0. 0. 0. 0. 0. 0. 0. 0. 0. 0. 0.
0. 0. 0. 0. 0. 0. 0. 0. 0. 0. 0.
 0. 0.]]
```