

Assignment

Task-1

Corpus

```
In [1]: ## SkLearn# Collection of string documents
corpus = ['this is the first document',
          'this document is the second document',
          'and this is the third one',
          'is this the first document' ]
```

SkLearn Implementation

```
In [2]: from sklearn.feature_extraction.text import TfidfVectorizer
vectorizer = TfidfVectorizer()
vectorizer.fit(corpus)
print(vectorizer.get_feature_names()) # sklearn feature names, they are sorted in alphabetic order by default
['and', 'document', 'first', 'is', 'one', 'second', 'the', 'third', 'this']
```

```
In [3]: # Here we will print the sklearn tfidf vectorizer idf values after applying the fit method
# After using the fit function on the corpus the vocab has 9 words in it, and each has its idf value.

print(vectorizer.idf_)
```

```
[1.91629073 1.22314355 1.51082562 1.          1.91629073 1.91629073
 1.          1.91629073 1.          ]
```

```
In [4]: # shape of sklearn tfidf vectorizer output after applying transform method.
skl_output = vectorizer.transform(corpus)
skl_output.shape
```

Out[4]: (4, 9)

```
In [5]: # sklearn tfidf values for first line of the above corpus.
# Here the output is a sparse matrix

print(skl_output[0])
```

```
(0, 8)      0.38408524091481483
(0, 6)      0.38408524091481483
(0, 3)      0.38408524091481483
(0, 2)      0.5802858236844359
(0, 1)      0.46979138557992045
```

```
In [6]: # sklearn tfidf values for first line of the above corpus.
# To understand the output better, here we are converting the sparse output matrix to dense matrix and printing it
# Notice that this output is normalized using L2 normalization. sklearn does this by default.

print(skl_output[0].toarray())
```

```
[[0.          0.46979139 0.58028582 0.38408524 0.          0.
  0.38408524 0.          0.38408524]]
```

My custom implementation

```
In [7]: # Required imports for my custom Implementation
from collections import Counter
from tqdm import tqdm
from scipy.sparse import csr_matrix
import math
import operator
from sklearn.preprocessing import normalize
import numpy as np
```

```
In [8]: # Fit function to define the features (unique words)
def fit(dataset):
    unique_words = set() # at first we will initialize an empty set
    # check if its list type or not
    if type(dataset) == list:
        for review in dataset: # for each review in the dataset
            for word in review.split(" "): # for each word in the review. #split method converts each string into list
                if len(word) >= 2:
                    unique_words.add(word)
        return sorted(list(unique_words))
    else:
        print("Invalid Input. Please give corpus as a list")
My_Custom_features = fit(corpus)
print(My_Custom_features)

['and', 'document', 'first', 'is', 'one', 'second', 'the', 'third', 'this']
```

```
In [9]: # function to calculate IDF value of any word in given corpus
def IDF(t, dataset):
    n = 0 #Number_of_documents_having_t = 0
    N = len(dataset) # total number of Docs in corpus
    for review in dataset: # for each review in the dataset
        for word in review.split(" "): # for each word in the review. #split method converts each string into list
            if word == t:
                n += 1
                break # goes to next sentence at first occurrence of the t in a document
    return 1+math.log((1+N)/(1+n))

idf = [IDF(word, corpus) for word in My_Custom_features] #List of IDF value in same order as vocab
idf
```

```
Out[9]: [1.916290731874155,
1.2231435513142097,
1.5108256237659907,
1.0,
1.916290731874155,
1.916290731874155,
1.0,
1.916290731874155,
1.0]
```

```
In [10]: # final custom tf-idf Vectorizer
def transform(Corpus, Dimensions):
    loc_r, loc_c, values = [], [], []

    if type(Corpus) == list:
        r = 0 # to store row index in sparse matrix
        for review in Corpus: # sets row
            word_freq = dict(Counter(review.split())) # counts occurrence of a word in a doc and stores as dict
            c = 0 # to store col index in sparse matrix
            for word in Dimensions: # sets column
                if word in review.split(): # checks if the word is in present row
                    loc_r.append(r) ; loc_c.append(c)
                    values.append(((word_freq[word])*IDF(word, corpus))/len(review.split()))
                    c += 1
            r += 1
        return normalize(csr_matrix((values, (loc_r, loc_c)), shape=(len(Corpus),len(Dimensions))), "l2")
    else:
        print("you need to pass list of strings")

My_Custom_Output = transform(corpus,My_Custom_features)
My_Custom_Output.shape
print(My_Custom_Output[0])

(0, 1)      0.4697913855799205
(0, 2)      0.5802858236844359
(0, 3)      0.3840852409148149
(0, 6)      0.3840852409148149
(0, 8)      0.3840852409148149
```

```
In [11]: print(My_Custom_Output[0].toarray())

[[0.         0.46979139 0.58028582 0.38408524 0.         0.
  0.38408524 0.         0.38408524]]
```

Task-2

```
In [5]: # Required imports for my custom Implementation
from collections import Counter
from tqdm import tqdm
from scipy.sparse import csr_matrix
import math
import operator
from sklearn.preprocessing import normalize
import numpy as np
```

```
In [6]: import pickle
with open('/content/drive/My Drive/Colab Notebooks/3_CountVectorizer/cleaned_strings', 'rb') as f:
    corpus_2 = pickle.load(f)

# printing the length of the corpus loaded
print("Number of documents in corpus = ",len(corpus_2))

Number of documents in corpus = 746
```

```
In [7]: # fit function to return vocab as unique values
def fit(Corpus):
    unique_words = set() # at first we will initialize an empty set
    # check if its list type or not
    if type(Corpus) == list:
        for review in Corpus: # for each review in the Corpus
            for word in review.split(" "): # for each word in the review. #split method converts each string into list
                if len(word) >= 2:
                    unique_words.add(word)
        return sorted(list(unique_words))
    else:
        print("Invalid Input. Please give corpus as a list")
vocab = fit(corpus_2)
print(len(vocab))

2886
```

```
In [8]: def IDF(t, Corpus):
    n = 0 #Number_of_documents_having_t = 0
    N = len(Corpus) # total number of Docs in corpus
    for doc in Corpus: # for each doc in the Corpus
        for word in doc.split(" "): # for each word in the doc. #split method converts each string into list
            if word == t:
                n +=1 ;
                break # goes to next sentence at first occurrence of the t in a document
    return 1+math.log((1+N)/(1+n))

list_idf = [IDF(word, corpus_2) for word in vocab] #List of IDF value in same order as vocab
print(len(list_idf))
print(list_idf[:5])

2886
[6.922918004572872, 6.922918004572872, 6.229770824012927, 6.922918004572872, 5.3134800921387715]
```

```
In [22]: # ordered idf values for alphabetical words in the vocab
idf_vocab = tuple(zip(vocab, list_idf)) # tuple of word along with its idf value as (word, idf)
list_idf.sort(reverse = True) # sorting idf values in descending order
Rank_list = []
for idf in list_idf:
    for entry in idf_vocab:
        if entry[0] in Rank_list:
            continue
        if entry[1] == idf:
            Rank_list.append(entry[0]) ; break
Dimensions = Rank_list[:50] # Top 50 idf words as dimensions
print(Dimensions)

['aailiyah', 'abandoned', 'ability', 'abroad', 'absolutely', 'abstruse', 'abysmal', 'academy', 'accent', 's', 'accessible', 'acclaimed', 'accolades', 'accurate', 'accurately', 'accused', 'achievement', 'achille', 'ackerman', 'act', 'acted', 'acting', 'action', 'actions', 'actor', 'actors', 'actress', 'actresse', 's', 'actually', 'adams', 'adaptation', 'add', 'added', 'addition', 'admins', 'admiration', 'admitted', 'adorable', 'adrift', 'adventure', 'advise', 'aerial', 'aesthetically', 'affected', 'affleck', 'afraid', 'africa', 'afternoon', 'age', 'aged', 'ages']
```

```
In [25]: # final custom tf-idf Vectorizer
def transform(Corpus, Dimensions):
    loc_r, loc_c, values = [], [], []
    if type(Corpus) == list:
        r = 0
        for doc in Corpus: # for each document in the Corpus
            word_freq = dict(Counter(doc.split())) ; c = 0
            for word in Dimensions:
                if word in doc.split():
                    loc_r.append(r) ; loc_c.append(c)
                    values.append(((word_freq[word])*IDF(word, Corpus))/len(doc.split()))
                    c += 1
            r += 1
        normalized_sparse_matrix = normalize(csr_matrix((values, (loc_r, loc_c)), shape=(len(Corpus),len(Dimensions))))
        return normalized_sparse_matrix
    else:
        print("you need to pass list of strings")
Custom_TF_IDF = transform(corpus_2,Dimensions)
print(Custom_TF_IDF[10].toarray())
Custom_TF_IDF.shape
```

Out[25]: (746, 50)