

A_3

October 16, 2020

1 SK-Learn Implementation

```
[23]: corpus = [  
    'this is the first document',  
    'this document is the second document',  
    'and this is the third one',  
    'is this the first document',  
]  
  
from sklearn.feature_extraction.text import TfidfVectorizer  
vectorizer = TfidfVectorizer()  
vectorizer.fit(corpus)  
skl_output = vectorizer.transform(corpus)  
  
# sklearn feature names, they are sorted in alphabetic order by default.  
  
print(vectorizer.get_feature_names())  
  
# 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_)  
  
# shape of sklearn tfidf vectorizer output after applying transform method.  
  
skl_output.shape  
  
# sklearn tfidf values for first line of the above corpus.  
# Here the output is a sparse matrix  
  
print(skl_output[0])  
  
# sklearn tfidf values for first line of the above corpus.
```

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# 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.
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print(skl_output[0].toarray())
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```
['and', 'document', 'first', 'is', 'one', 'second', 'the', 'third', 'this']
[1.91629073 1.22314355 1.51082562 1.          1.91629073 1.91629073
 1.          1.91629073 1.          ]
(0, 8)          0.38408524091481483
(0, 6)          0.38408524091481483
(0, 3)          0.38408524091481483
(0, 2)          0.5802858236844359
(0, 1)          0.46979138557992045
[[0.          0.46979139 0.58028582 0.38408524 0.          0.
 0.38408524 0.          0.38408524]]
```

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[ ]:
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2 Your custom implementation:

3 Task-1

```
[32]: # Write your code here.
# Make sure its well documented and readable with appropriate comments.
# Compare your results with the above sklearn tfidf vectorizer
# You are not supposed to use any other library apart from the ones given below

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

## 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',
]
```

```
[57]: import warnings
warnings.filterwarnings("ignore")
import pandas as pd
from tqdm import tqdm
import os
# fit method is used to identify the unique words in the corpus and add
→dimension to it in the dictionary-format.
def fit(data):
    if type(data) == type(list()):
        s = set()
        for i in range(len(data)): # Iterating over every row in the corpus and
→finding the unique words of (length > 2)
            x = data[i].split()
            for j in range(len(x)):
                if len(x[j]) < 2:
                    continue
                s.add(x[j])
            d = {j:i for i,j in enumerate(sorted(s))} # d : ( keys = unique-words )
→and (values = dimension-number)
            return d
    else:
        print("you need to pass list of sentence")
```

```
[34]: def transform(corpus,vocab):
    idf_dict = dict() # keys = unique-words , values = number of documents
→contain the corresponding unique-word.

    idf_ = list() # used for printing the idf values

    for word in vocab.keys(): # this for-loop is used to find number of documents
→contain the corresponding unique-word.
        c=0
        for row in (corpus):
            if word in row:
                c+=1
        idf_dict[word] = c
        idf_.append(1+ math.log((1+len(corpus))/(c+1)))
    print('\n',idf_)
    rows = []
    cols = []
    vals = []
    print('\n','*'*50)
    for indx,row in enumerate(corpus):
        a = dict(Counter(row.split()))
        for word,freq in a.items():
            col_indx = vocab.get(word,-1)
            if col_indx != -1:
```

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        rows.append(indx)
        cols.append(col_indx)
        tf = freq / sum(a.values())
        idf = 1 + (math.log((1+len(corpus))/(1 + idf_dict[word])))
        res = (tf * idf)
        #print(tf, '*', idf, '=', res)
        vals.append(res)
    b = csr_matrix((vals, (rows,cols)), shape=(len(corpus),len(vocab)))
    b = normalize(b,norm='l2')
    return b

```

```

[35]: vocab = fit(corpus)
print(list(vocab.keys()))
print(transform(corpus,vocab).toarray())

```

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

```
[1.916290731874155, 1.2231435513142097, 1.5108256237659907, 1.0,
1.916290731874155, 1.916290731874155, 1.0, 1.916290731874155, 1.0]
```

```

*****
[[0.          0.46979139 0.58028582 0.38408524 0.          0.
  0.38408524 0.          0.38408524]
 [0.          0.6876236  0.          0.28108867 0.          0.53864762
  0.28108867 0.          0.28108867]
 [0.51184851 0.          0.          0.26710379 0.51184851 0.
  0.26710379 0.51184851 0.26710379]
 [0.          0.46979139 0.58028582 0.38408524 0.          0.
  0.38408524 0.          0.38408524]]

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4 Task-2

```

[36]: import pickle
import math
corpus = pickle.load(open("cleaned_strings", "rb"))

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```

[68]: def fit(data):
        if type(data) == type(list()):
            s = set()
            '''

```

```

Finding the unique words in the corpus
'''
for i in range(len(data)):
    x = data[i].split()
    for j in range(len(x)):
        if len(x[j]) < 2:
            continue
        s.add(x[j])

d1 = dict() # d1: is used to store the unique-words as keys and ( values =
→ number of documents contains this word )
fit.d2 = dict() # d2: ( keys = unique-words) and (values =
→ Inverse-document-freq values) in decending order of values

for word in s: # this for-loop is used to find number of documents
→ contain the corresponding unique-word.
    c=0
    for row in (data):
        if word in row:
            c+=1
    d1[word] = c
    fit.d2[word] = 1 + (math.log((1+len(data)) / (1 + d1[word])))
    s1 = (sorted(fit.d2.items(), key=lambda x:x[1],reverse=True)[:50]) # top
→ 50 words based on idf scores.
    # s1 = sorted(s1) # since the idf-scores for top-50 words are same, so
→ sorted these top-50 words in alphabetical order.
    #print(s1)
    s2 = [i[0] for i in s1]
    d = {j:i for i,j in enumerate((s2))}
    return d
else:
    print("you need to pass list of sentance")

```

```

[69]: def transform(corpus,vocab):
    rows = []
    cols = []
    vals = []
    for indx,row in enumerate(corpus):
        a = dict(Counter(row.split()))
        for word,freq in a.items():
            col_indx = vocab.get(word,-1)
            if col_indx != -1:
                rows.append(indx)
                cols.append(col_indx)
                tf = freq / sum(a.values())

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        idf = fit.d2[word] # using the same dictionary (fit.d2) from fit method
        →where ( keys = unique-words) and (values = Inverse-document-freq scores) in
        →decending order of values
        #print(idf)
        res = float(tf * idf)
        #print(tf, '*',idf, '=',res)
        vals.append(res)
    b = csr_matrix((vals, (rows,cols)), shape=(len(corpus),len(vocab)))
    b = normalize(b,norm='l2')
    return b

```

```

[70]: vocab = fit(corpus)
      print(list(vocab.keys()))
      print(transform(corpus,vocab).toarray())

```

```

['holding', 'shameful', 'landscapes', 'removing', 'secondary', 'revenge',
'politically', 'repeating', 'massive', 'cliff', 'kathy', 'rendering', 'virus',
'hayworth', 'fire', 'cutie', 'fanciful', 'reporter', 'boss', 'represents',
'sounded', 'regardless', 'portrayed', 'angelina', 'spy', 'quaid', 'applause',
'shell', 'drawn', 'angela', 'voyage', 'evidently', 'timing', 'truth',
'unlockable', 'smith', 'menacing', 'edward', 'murdering', 'merit', 'selections',
'females', 'recover', 'pledge', 'flicks', 'finest', 'washed', 'manages',
'colours', 'discovery']
[[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. 0. ... 0. 0. 0.]
 [0. 0. 0. ... 0. 0. 0.]]

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[72]:
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