```
Name:Mahfuzatul Bushra, ID:22-92354-1
```

```
Project: Feature engineering to represent text data using TF-IDF Model.
....."Import libraries and dependencies and settings".....
import warnings
warnings.filterwarnings('ignore') #supress warning in python
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
pd.options.display.max colwidth = 100
%matplotlib inline
import nltk
import re
....."Corpus Sample".....
corpus sample=['I would say football is my favourite sport.',
                'The dog next door kept barking all night.',
               'A youth was stuffing a bag full of medical supplies.',
               'Billy turned on a radio to get the sports news.',
               'His job requires him to travel frequently.',
               'She excels at sport.',
               'We can expect rainy weather tomorrow.',
               'My wife works in a travel agency.',
               'Korean food is generally very spicy.',
               'We cannot exist without food or water.',
                'The weather is unpredictable around here.',
               'I spent the day at the medical facility.'
labels=['sport','animal','medicine','sport','travel','sport','weather'
,'travel','food','food','weather','medicine']
corpus array=np.array(corpus sample)
corpus_df = pd.DataFrame({'Document': corpus_sample,
                           'Category': labels})
corpus df = corpus df[['Document', 'Category']]
corpus df
                                                  Document
                                                            Category
0
             I would say football is my favourite sport.
                                                               sport
1
               The dog next door kept barking all night.
                                                              animal
2
    A youth was stuffing a bag full of medical supplies.
                                                            medicine
3
         Billy turned on a radio to get the sports news.
                                                               sport
4
              His job requires him to travel frequently.
                                                              travel
5
                                     She excels at sport.
                                                               sport
6
                   We can expect rainy weather tomorrow.
                                                             weather
7
                        My wife works in a travel agency.
                                                              travel
```

```
8
                     Korean food is generally very spicy.
                                                                   food
9
                   We cannot exist without food or water.
                                                                    food
10
                The weather is unpredictable around here.
                                                               weather
11
                 I spent the day at the medical facility.
                                                               medicine
..... "Corpus Pre_Processing".....
nltk.download('stopwords')
word per text=nltk.WordPunctTokenizer()
#Removing stop words with NLTK in Python(such as "the", "a", "an",
"in")
stop words = nltk.corpus.stopwords.words('english')
[nltk data] Downloading package stopwords to /root/nltk data...
[nltk data] Unzipping corpora/stopwords.zip.
def norm doc(Document):
  Document = re.sub(r'[^a-zA-Z\s]', '', Document, re.I|re.A)#convert
to lower case and remove special characters\whitespaces
  Document= Document.lower()
  Document= Document.strip()
  Tokens = word per text.tokenize(Document) # Tokenize the Document
  filtered tokens = [token for token in Tokens if token not in
stop words]
  Document=' '.join(filtered tokens)# re-create document from filtered
tokens
  return Document
normalize corpus = np.vectorize(norm doc)
normalize_corpus = normalize_corpus(corpus sample)
normalize_corpus
array(['would say football favourite sport',
        'dog next door kept barking night',
       'youth stuffing bag full medical supplies',
       'billy turned radio get sports news',
       'job requires travel frequently', 'excels sport',
'expect rainy weather tomorrow', 'wife works travel agency',
'korean food generally spicy', 'cannot exist without food
water',
        'weather unpredictable around', 'spent day medical facility'],
      dtype='<U40')
....."TF-IDF Model".....
from sklearn.feature extraction.text import CountVectorizer
#Get Features in Sparse Format
Count Vectorizer=CountVectorizer(min df=0., max df=1.)
#Fit and transform
CV BOW= Count Vectorizer.fit transform(normalize corpus)
CV BOW
```

```
<12x48 sparse matrix of type '<class 'numpy.int64'>'
   with 53 stored elements in Compressed Sparse Row format>
CV BOW = CV BOW.toarray() # Convert to an array
CV BOW
array([[0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 0,
    0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0,
0,
    0, 0, 1, 0],
    [0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
1,
    0,
    0, 0, 0, 0],
    0,
    0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0,
0,
    0, 0, 0, 1],
    0,
    0, 0, 1, 0, 0, 1, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0,
0,
    0, 0, 0, 0],
    0,
    0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0,
0,
    0, 0, 0, 0],
    0,
    0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0,
0,
    0, 0, 0, 0],
    0,
    0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1,
0,
    0, 0, 0, 0],
    0,
    1,
    0, 1, 0, 0],
    0,
    1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
0,
```

0, 0, 0, 0],

```
[0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0,
0,
       0,
       1, 0, 0, 0],
      0,
       0,
       0, 0, 0, 0],
      [0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0,
0,
       0, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
0,
       0, 0, 0, 0]])
from sklearn.feature extraction.text import TfidfTransformer
#Compute the IDF values
TD IDF= TfidfTransformer(norm='l2', use idf=True, smooth_idf=True)
#fit and transform
TDIDF matrix = TD IDF.fit transform(CV BOW)
#to array
TDIDF matrix=TDIDF matrix.toarray()
#Create DataFrame
features=Count Vectorizer.get feature names()
pd.DataFrame(np.round(TDIDF matrix, 2), columns=features)
   agency around
                  bag
                       barking
                               billy
                                     cannot
                                              day
                                                   dog
                                                       door
excels
     0.00
                 0.00
                          0.00
                                0.00
             0.0
                                       0.00
                                             0.00
                                                  0.00
                                                       0.00
0.00
     0.00
             0.0
                 0.00
                          0.41
                                0.00
                                       0.00
                                             0.00
                                                  0.41
                                                       0.41
1
0.00
     0.00
             0.0
                                0.00
                                       0.00
2
                 0.42
                          0.00
                                             0.00
                                                  0.00
                                                       0.00
0.00
     0.00
             0.0
                 0.00
                          0.00
                                0.41
                                       0.00
                                             0.00
                                                  0.00
                                                       0.00
3
0.00
4
     0.00
             0.0
                 0.00
                          0.00
                                0.00
                                       0.00
                                             0.00
                                                  0.00
                                                       0.00
0.00
5
     0.00
             0.0
                 0.00
                          0.00
                                0.00
                                       0.00
                                             0.00
                                                  0.00
                                                       0.00
0.76
     0.00
             0.0
                 0.00
                          0.00
                                0.00
                                       0.00
                                             0.00
                                                  0.00
                                                       0.00
6
0.00
     0.52
             0.0
                 0.00
                          0.00
                                0.00
                                       0.00
                                            0.00
                                                  0.00
                                                       0.00
7
0.00
     0.00
                 0.00
                          0.00
                                0.00
                                       0.00
8
             0.0
                                             0.00
                                                  0.00
                                                       0.00
0.00
     0.00
             0.0
                 0.00
                          0.00
                                0.00
                                       0.46
                                             0.00
                                                  0.00
                                                       0.00
0.00
```

10 0.00	0.00	6	0.6	0.00	(0.00	0.00	9	0.00	0.00	0.00	0.00
11 0.00	0.00	6	0.0	0.00	(9.00	0.00	9	0.00	0.52	0.00	0.00
		ravel	tι	ırned	unpre	edicta	able	wate	er we	ather	wife	without
0.00 1 . 0.00 2 . 0.00 3 . 0.00 4 .		0.00		0.00			0.0	0.0	90	0.00	0.00	0.00
		0.00		0.00			0.0	0.0	00	0.00	0.00	0.00
		0.00		0.00			0.0	0.0	90	0.00	0.00	0.00
		0.00		0.41			0.0	0.0	00	0.00	0.00	0.00
		0.44		0.00			0.0	0.0	00	0.00	0.00	0.00
		0.00		0.00			0.0	0.0	90	0.00	0.00	0.00
_		0.00		0.00			0.0	0.0	90	0.44	0.00	0.00
_		0.44		0.00			0.0	0.0	90	0.00	0.52	0.00
0		0.00		0.00			0.0	0.0	90	0.00	0.00	0.00
^		0.00		0.00			0.0	0.4	16	0.00	0.00	0.46
10	• •	0.00		0.00			0.6	0.0	00	0.52	0.00	0.00
		0.00		0.00			0.0	0.0	00	0.00	0.00	0.00
0 1 2 3 4 5 6 7 8 9	ould 0.46 0.00 0.00 0.00 0.00 0.00 0.00 0.00	youth 0.00 0.00 0.00 0.00 0.00 0.00 0.00))))))))									

[12 rows x 48 columns]

 $from \ sklearn.feature_extraction.text \ import \ TfidfVectorizer$

```
#Compute the IDF values
TD IDF V= TfidfVectorizer(min df=0.,max df=1.,norm='l2', use idf=True,
smooth idf=True)
#Fit and transform
TDIDF V matrix = TD IDF V.fit transform(normalize corpus)
#to array
TDIDF V matrix=TDIDF V matrix.toarray()
#Create DataFrame
features_V=Count_Vectorizer.get_feature_names()
pd.DataFrame(np.round(TDIDF V matrix, 2), columns=features V)
    agency around
                           barking
                                     billy cannot
                      bag
                                                      day
                                                             dog
                                                                  door
excels \
      0.00
                     0.00
                               0.00
                                      0.00
                                                            0.00
                0.0
                                               0.00
                                                     0.00
                                                                  0.00
0.00
1
      0.00
                0.0
                     0.00
                               0.41
                                      0.00
                                               0.00
                                                     0.00
                                                            0.41
                                                                  0.41
0.00
      0.00
                0.0
                                      0.00
                                                     0.00
2
                     0.42
                               0.00
                                               0.00
                                                            0.00
                                                                  0.00
0.00
3
      0.00
                0.0
                     0.00
                               0.00
                                      0.41
                                               0.00
                                                     0.00
                                                            0.00
                                                                  0.00
0.00
4
      0.00
                0.0
                     0.00
                               0.00
                                      0.00
                                               0.00
                                                     0.00
                                                            0.00
                                                                  0.00
0.00
      0.00
                0.0
                     0.00
5
                               0.00
                                      0.00
                                               0.00
                                                     0.00
                                                            0.00
                                                                  0.00
0.76
      0.00
                               0.00
6
                0.0
                     0.00
                                      0.00
                                               0.00
                                                     0.00
                                                            0.00
                                                                  0.00
0.00
7
      0.52
                0.0
                     0.00
                               0.00
                                      0.00
                                               0.00
                                                     0.00
                                                            0.00
                                                                  0.00
0.00
      0.00
                0.0
                     0.00
                               0.00
                                      0.00
                                               0.00
                                                     0.00
                                                            0.00
                                                                  0.00
8
0.00
      0.00
9
                0.0
                     0.00
                               0.00
                                      0.00
                                               0.46
                                                     0.00
                                                            0.00
                                                                  0.00
0.00
      0.00
10
                0.6
                     0.00
                               0.00
                                      0.00
                                               0.00
                                                     0.00
                                                            0.00
                                                                  0.00
0.00
      0.00
                0.0
                               0.00
                                      0.00
                                               0.00
11
                     0.00
                                                     0.52
                                                            0.00
                                                                  0.00
0.00
         travel
                  turned
                          unpredictable water
                                                  weather
                                                            wife
                                                                  without
works \
0
           0.00
                    0.00
                                     0.0
                                            0.00
                                                     0.00
                                                            0.00
                                                                      0.00
    . . .
0.00
                                                                      0.00
           0.00
                    0.00
                                     0.0
                                            0.00
                                                     0.00
                                                            0.00
1
    . . .
0.00
2
           0.00
                    0.00
                                     0.0
                                            0.00
                                                     0.00
                                                            0.00
                                                                      0.00
0.00
3
           0.00
                    0.41
                                     0.0
                                            0.00
                                                     0.00
                                                            0.00
                                                                      0.00
    . . .
0.00
4
           0.44
                    0.00
                                     0.0
                                            0.00
                                                     0.00
                                                            0.00
                                                                      0.00
    . . .
```

0.00							
5 0.00	0.00	0.00	0.0	0.00	0.00	0.00	0.00
6 0.00	0.00	0.00	0.0	0.00	0.44	0.00	0.00
7 0.52	0.44	0.00	0.0	0.00	0.00	0.52	0.00
8 0.00	0.00	0.00	0.0	0.00	0.00	0.00	0.00
9 0.00	0.00	0.00	0.0	0.46	0.00	0.00	0.46
10 0.00	0.00	0.00	0.6	0.00	0.52	0.00	0.00
11	0.00	0.00	0.0	0.00	0.00	0.00	0.00
would 0 0.46 1 0.00 2 0.00 3 0.00 4 0.00 5 0.00 6 0.00 7 0.00 8 0.00 9 0.00 10 0.00 11 0.00	youth 0.00 0.00 0.42 0.00 0.00 0.00 0.00 0.00						

[12 rows x 48 columns]