



VIT[®]

Vellore Institute of Technology

(Deemed to be University under section 3 of UGC Act, 1956)

School of Computer Science and Engineering

Register Number: 18BCE0745

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Code:

```
import pandas as pd

from sklearn.feature_extraction.text import TfidfTransformer
from sklearn.feature_extraction.text import CountVectorizer

docs=["plot: two teen couples go to a church party, drink and then drive.",
"films adapted from comic books have had plenty of success , whether they're about superheroes (
batman , superman , spawn ) , or geared toward kids ( casper ) or the arthouse crowd ( ghost world ) ,
but there's never really been a comic book like from hell before .",
"every now and then a movie comes along from a suspect studio , with every indication that it will be a
stinker , and to everybody's surprise ( perhaps even the studio ) the film becomes a critical darling . ",
"damn that y2k bug ."
]
```

#####

```
print("TFIDF")
```

```
from sklearn.feature_extraction.text import TfidfVectorizer
```

```
vectorizer = TfidfVectorizer()
```

```
vectors = vectorizer.fit_transform(docs)
```

```
feature_names = vectorizer.get_feature_names()
```

```
dense = vectors.todense()
```

```
denselist = dense.tolist()
```

```
df = pd.DataFrame(denselist, columns=feature_names)
```

```
print(df)
```

```
print("BOW")
```

```
from sklearn.feature_extraction.text import CountVectorizer
```

```
cv=CountVectorizer()
```

```
word_count_vector=cv.fit_transform(docs)
```

```
word_count_vector.shape
```

```
print(word_count_vector.toarray())
```

```
print("IDF")
```

```
tfidf_transformer=TfidfTransformer(smooth_idf=True,use_idf=True)
```

```
tfidf_transformer.fit(word_count_vector)
```

```
# print idf values
```

```
df_idf = pd.DataFrame(tfidf_transformer.idf_, index=cv.get_feature_names(),columns=["idf_weights"])
```

```
# sort ascending
```

```
df_idf.sort_values(by=['idf_weights'])
```

```
print(df_idf)
```

```
print("TF")

#instantiate CountVectorizer()
cv=CountVectorizer()

# this steps generates word counts for the words in your docs
word_count_vector=cv.fit_transform(docs)

word_count_vector.shape

pd = pd.DataFrame(word_count_vector.toarray(), columns = cv.get_feature_names())
print(pd)


count_vector=cv.transform(docs)
print(count_vector.toarray())


tf_idf_vector=tfidf_transformer.transform(count_vector)
print(tf_idf_vector.toarray())


feature_names = cv.get_feature_names()

print(feature_names)
```

```
Help Variable explorer Plots Files Find
Console 1/A x
index.py ; will = C:/Users/3/Streamline/OneDrive/Desktop/Machine Learning/Week 1/TFIDF/Lab 4
TFIDF
      about    adapted    along    ...    with    world    y2k
0  0.000000  0.000000  0.000000  ...  0.000000  0.000000  0.000000
1  0.150571  0.150571  0.000000  ...  0.000000  0.150571  0.000000
2  0.000000  0.000000  0.170352  ...  0.170352  0.000000  0.000000
3  0.000000  0.000000  0.000000  ...  0.000000  0.000000  0.525473

[4 rows x 74 columns]
BOW
[[0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0 1 0 0 0 0 1 1 0 0 0 0 0 0 0 0 1 0 0
  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 1 0 0 1 0 0 1 0 1 0 0 0
  0 0]
[1 1 0 0 1 1 0 0 1 1 1 1 0 1 1 0 0 2 0 0 1 0 0 0 0 0 0 0 0 0 0 1 2 1 1 0 1 1
  1 0 0 1 1 0 1 0 1 2 0 0 1 0 1 1 1 0 0 1 1 1 0 0 0 0 1 0 1 1 0 1 0 1 0 0
  1 0]
[0 0 1 2 0 0 1 1 0 0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 0 0 1 2 1 1 0 1 0 0 0 0 0
  0 1 1 0 0 1 0 1 0 0 0 1 0 0 0 0 0 1 2 0 0 0 1 1 0 1 2 1 0 0 1 0 0 0 0 1 1
  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 1 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 0 0 0 0 0 0 0 0 0 0 0 0
  0 1]]
IDF
      idf_weights
about      1.916291
adapted    1.916291
along      1.916291
and        1.510826
arthouse   1.916291
...        ...
whether      1.916291
will       1.916291
```

```
lab5.ipynb 8/20/2020 3:26 PM
WMLab5.py 8/20/2020 4:25 PM

Help Variable explorer Plots Files Find

Console 1/A x
wheener 1.916291
will 1.916291
with 1.916291
world 1.916291
y2k 1.916291

[74 rows x 1 columns]
TF
  about adapted along and arthouse ... whether will with world y2k
0      0         0     0  1         0 ...      0     0     0     0     0
1      1         1     0  0         1 ...      1     0     0     1     0
2      0         0     1  2         0 ...      0     1     1     0     0
3      0         0     0  0         0 ...      0     0     0     0     1

[4 rows x 74 columns]
[[0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 1 0 0 1 0 0 0 0 0 1 1 0 0 0 0 0 0 0 0 1 0 0
  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 1 0 0 1 0 1 0 0 0
  0 0]
[1 1 0 0 1 1 0 0 1 1 1 1 0 1 1 0 0 2 0 0 1 0 0 0 0 0 0 0 0 0 0 1 2 1 1 0 1 1
  1 0 0 1 1 0 1 0 1 2 0 0 1 0 1 1 1 0 0 1 1 1 0 0 0 0 1 0 1 1 0 1 0 1 0 0
  1 0]
[0 0 1 2 0 0 1 1 0 0 0 0 0 0 0 0 0 1 0 0 1 0 0 1 0 0 1 2 1 1 0 1 0 0 0 0 0
  0 1 1 0 0 1 0 1 0 0 0 1 0 0 0 0 0 1 2 0 0 0 1 1 0 1 2 1 0 0 1 0 0 0 1 1
  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 1 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 0 0 0 0 0 0 0 0 0 0
  0 1]]
[[0.      0.      0.      0.23918972 0.      0.
  0.      0.      0.      0.      0.      0.
  0.      0.      0.      0.30338183 0.      0.
  0.30338183 0.      0.      0.      0.      0.30338183
  0.30338183 0.      0.      0.      0.      0.      0.]
```