Vectorizing Raw Data: TF-IDF

TF-IDF

Creates a document-term matrix where the columns represent single unique terms (unigrams) but the cell represents a weighting meant to represent how important a word is to a document.

```
image.png
```

▼ Read in text

```
import pandas as pd
import re
import string
import nltk
pd.set_option('display.max_colwidth', 100)

stopwords = nltk.corpus.stopwords.words('english')
ps = nltk.PorterStemmer()

data = pd.read_csv("SMSSpamCollection.tsv", sep='\t')
data.columns = ['label', 'body_text']
```

▼ Create function to remove punctuation, tokenize, remove stopwords, and stem

```
def clean_text(text):
    text = "".join([word.lower() for word in text if word not in string.punctuation])
    tokens = re.split('\W+', text)
    text = [ps.stem(word) for word in tokens if word not in stopwords]
    return text
```

▼ Apply TfidfVectorizer

```
from sklearn.feature_extraction.text import TfidfVectorizer

tfidf_vect = TfidfVectorizer(analyzer=clean_text)
X_tfidf = tfidf_vect.fit_transform(data['body_text'])
print(X_tfidf.shape)
print(tfidf_vect.get_feature_names())

(5567, 8104)
['', '0', '008704050406', '0089mi', '0121', '01223585236', '01223585334', '0125698789', '02', '020603', '0207', '02070836089', '0
```

Apply TfidfVectorizer to smaller sample

```
data_sample = data[0:20]

tfidf_vect_sample = TfidfVectorizer(analyzer=clean_text)

X_tfidf_sample = tfidf_vect_sample.fit_transform(data_sample['body_text'])

print(X_tfidf_sample.shape)

print(tfidf_vect_sample.get_feature_names())

(20, 192)
   ['08002986030', '08452810075over18', '09061701461', '1', '100', '100000', '11', '12', '150pday', '16', '2', '20000', '2005', '21s
```

Vectorizers output sparse matrices

Sparse Matrix: A matrix in which most entries are 0. In the interest of efficient storage, a sparse matrix will be stored by only storing the locations of the non-zero elements.

```
X_tfidf_df = pd.DataFrame(X_tfidf_sample.toarray())
X_tfidf_df.columns = tfidf_vect_sample.get_feature_names()
X_tfidf_df
```

	08002986030	08452810075over18	09061701461	1	100	100000	11	12	150pd
0	0.000000	0.198986	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.0000
1	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.0000
2	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.0000
3	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.0000
4	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.0000
5	0.000000	0.000000	0.231645	0.000000	0.000000	0.000000	0.000000	0.231645	0.0000
6	0.197682	0.000000	0.000000	0.000000	0.000000	0.000000	0.197682	0.000000	0.0000
7	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.0000
8	0.000000	0.000000	0.000000	0.000000	0.224905	0.000000	0.000000	0.000000	0.2249
9	0.000000	0.000000	0.000000	0.252972	0.000000	0.252972	0.000000	0.000000	0.0000
10	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.0000
11	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.0000
12	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.0000
13	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.0000
14	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.0000
15	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.0000
16	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.0000
17	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.0000
18	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.0000
19	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.000000	0.0000

