In [9]:

from sklearn.datasets import fetch_20newsgroups

In [10]:

```
categories = ['alt.atheism', 'soc.religion.christian', 'comp.graphics', 'sci.med']
news_train = fetch_20newsgroups(subset = 'train', categories = categories, shuffle = True)
news_test = fetch_20newsgroups(subset = 'test', categories = categories, shuffle = True)
news_train.target_names
```

Out[10]:

['alt.atheism', 'comp.graphics', 'sci.med', 'soc.religion.christian']

In [11]:

```
from sklearn.feature extraction.text import CountVectorizer
count = CountVectorizer()
X_train = count.fit_transform(news_train.data)
X_train.shape
print(X_train)
  (0, 14887)
                 1
  (0, 29022)
                 1
  (0, 8696)
                 4
  (0, 4017)
                 2
  (0, 33256)
                 2
  (0, 21661)
                 3
  (0, 9031)
                 3
  (0, 31077)
                 1
  (0, 9805)
                 2
  (0, 17366)
                 1
  (0, 32493)
                 4
  (0, 16916)
                 2
  (0, 19780)
                 2
  (0, 17302)
                 2
  (0, 23122)
                 1
  (0, 25663)
                 1
  (0, 16881)
                 1
  (0, 16082)
                 1
  (0, 23915)
                 1
  (0, 32142)
                 5
  (0, 33597)
                 2
  (0, 20253)
                 1
  (0, 587)
                 1
  (0, 12051)
                 1
  (0, 5201)
  (2256, 13740) 1
  (2256, 14662) 1
  (2256, 20201) 1
  (2256, 12443) 6
  (2256, 30325) 3
  (2256, 4610) 1
  (2256, 33844) 1
  (2256, 17354) 1
  (2256, 26998) 1
  (2256, 20277) 1
  (2256, 20695) 1
  (2256, 20702) 1
  (2256, 9649)
  (2256, 9086)
  (2256, 26254) 1
  (2256, 17133) 2
  (2256, 4490)
  (2256, 13720) 1
  (2256, 5016)
                 1
  (2256, 9632)
  (2256, 11824) 1
  (2256, 29993) 1
```

1

1

(2256, 1298)

(2256, 2375)

(2256, 3921)

In [12]:

```
from sklearn.feature_extraction.text import TfidfTransformer

tfid = TfidfTransformer()

X_train_tfid = tfid.fit_transform(X_train)

X_train_tfid.shape

print(X_train_tfid)
```

```
(0, 35416)
              0.1348710554299733
(0, 35312)
              0.0312703097833574
(0, 34775)
              0.034481472140846715
(0, 34755)
              0.043341654399042764
(0, 33915)
              0.0999409997803694
(0, 33597)
              0.06567578043186388
(0, 33572)
              0.09313007554599557
(0, 33256)
              0.11819702490105698
(0, 32493)
              0.07283773941616518
(0, 32391)
              0.12806013119559947
(0, 32270)
              0.023871142738151236
(0, 32142)
              0.08865416253721688
(0, 32135)
              0.04910237380446671
(0, 32116)
              0.10218403421141944
(0, 31915)
              0.08631915131162177
(0, 31077)
              0.016797806021219684
(0, 30623)
              0.0686611288079694
(0, 29022)
              0.1348710554299733
(0, 28619)
              0.047271576160535234
(0, 27836)
              0.06899050810672397
(0, 26175)
              0.08497460943470851
(0, 25663)
              0.034290706362898604
(0, 25361)
              0.11947938145690981
(0, 25337)
              0.04935883383975408
(0, 24677)
              0.09796250319482307
(2256, 13720) 0.0969927054646086
(2256, 13521) 0.06264742916622883
(2256, 13498) 0.08574361554718753
(2256, 12626) 0.047531848081675473
(2256, 12443) 0.5533848656066114
(2256, 11824) 0.12028503503707108
(2256, 9649) 0.09916899209319778
(2256, 9632)
              0.11395377721095845
(2256, 9338)
              0.05248982587156077
(2256, 9086)
              0.1056108470261161
(2256, 9072)
              0.06784313233467505
(2256, 7766)
              0.030117682035074988
(2256, 7743)
              0.08847991007710146
(2256, 6507)
              0.26543973023130435
(2256, 6430)
              0.04825278674712813
(2256, 5698)
              0.032261754952242205
(2256, 5529)
              0.031207627784231296
(2256, 5016)
              0.12302132956698503
(2256, 4938)
              0.032031796462786304
(2256, 4610)
              0.0927607242624837
(2256, 4490)
              0.11395377721095845
(2256, 3921)
              0.13532523144219466
(2256, 2375)
              0.12625767908616808
(2256, 1298)
              0.12625767908616808
(2256, 587)
              0.06304633984640515
```

In [13]:

```
from sklearn.naive_bayes import MultinomialNB
a = MultinomialNB().fit(X_train_tfid, news_train.target)
```

In [14]:

```
X_test = count.transform(news_test.data)
X_test_tfid = tfid.transform(X_test)
predicted = a.predict(X_test_tfid)
print(predicted)
```

```
[2 2 3 ... 2 2 1]
```

In [15]:

```
doc = ['god is love', 'openGL on the GPU is fast']
X_new_counts = count.transform(doc)
X_new_tfid = tfid.transform(X_new_counts)
predicted = a.predict(X_new_tfid)
for x in predicted:
    print(x)
```

3 1

In [16]:

```
from sklearn.pipeline import Pipeline
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.naive_bayes import MultinomialNB
```

In [17]:

```
text_clf = Pipeline([('vect', TfidfVectorizer()), ('clf', MultinomialNB())])
text_clf.fit(news_train.data, news_train.target)
predicted = text_clf.predict(news_test.data)
```

In [18]:

```
from sklearn import metrics
from sklearn.metrics import accuracy_score
import numpy as np
```

In [19]:

```
print('Accuracy achieved is ' + str(np.mean(predicted == news_test.target)))
print(metrics.classification_report(news_test.target, predicted, target_names =
news_test.target_names)), metrics.confusion_matrix(news_test.target, predicted)
```

Accuracy achieved is 0.8348868175765646

	precision	recall	f1-score	support
alt.atheism	0.97	0.60	0.74	319
comp.graphics	0.96	0.89	0.92	389
sci.med	0.97	0.81	0.88	396
<pre>soc.religion.christian</pre>	0.65	0.99	0.78	398
accuracy			0.83	1502
macro avg	0.89	0.82	0.83	1502
weighted avg	0.88	0.83	0.84	1502

Out[19]:

In []: