

Bag of Words Model with Naive Bayes:

```
In [1]: import pandas as pd
from bs4 import BeautifulSoup
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
from nltk.stem import WordNetLemmatizer
from sklearn.feature_extraction.text import CountVectorizer, TfidfVectorizer
from sklearn.naive_bayes import MultinomialNB
from sklearn.model_selection import GridSearchCV
from sklearn.metrics import accuracy_score
from sklearn.metrics import classification_report
```

```
In [2]: data = pd.read_csv('IMDB Dataset.csv')
data.head()
```

Out[2]:

	review	sentiment
0	One of the other reviewers has mentioned that ...	positive
1	A wonderful little production. The...	positive
2	I thought this was a wonderful way to spend ti...	positive
3	Basically there's a family where a little boy ...	negative
4	Petter Mattei's "Love in the Time of Money" is...	positive

Basic Statistics

```
In [3]: print("Number of rows: ", data.shape[0])
print("Number of columns: ", data.shape[1])
```

```
Number of rows: 50000
Number of columns: 2
```

```
In [4]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 50000 entries, 0 to 49999
Data columns (total 2 columns):
 #   Column      Non-Null Count  Dtype
---  -
 0   review     50000 non-null  object
 1   sentiment  50000 non-null  object
dtypes: object(2)
memory usage: 781.4+ KB
```

```
In [5]: data.sentiment.value_counts()
```

```
Out[5]: positive    25000
        negative    25000
        Name: sentiment, dtype: int64
```

from the above, we can confirm that the data is equally partitioned.

Data Cleaning and preprocessing

```
In [6]: data['review'][1]
```

```
Out[6]: 'A wonderful little production. <br /><br />The filming technique is very unassuming- very old-time-BBC fashion and
gives a comforting, and sometimes discomforting, sense of realism to the entire piece. <br /><br />The actors are ex
tremely well chosen- Michael Sheen not only "has got all the polari" but he has all the voices down pat too! You can
truly see the seamless editing guided by the references to Williams\' diary entries, not only is it well worth the w
atching but it is a terrificly written and performed piece. A masterful production about one of the great master\'s
of comedy and his life. <br /><br />The realism really comes home with the little things: the fantasy of the guard w
hich, rather than use the traditional \'dream\' techniques remains solid then disappears. It plays on our knowledge
and our senses, particularly with the scenes concerning Orton and Halliwell and the sets (particularly of their flat
with Halliwell\'s murals decorating every surface) are terribly well done.'
```

In the above data we can see
\ break tags. We need to remove them before using this data.

```
In [7]: cleantext = BeautifulSoup(data["review"][1], 'lxml').text
```

We need to remove the slash

```
In [8]: import re
cleantext = re.sub(r'^\w\s', '', cleantext)
cleantext
```

```
Out[8]: 'A wonderful little production The filming technique is very unassuming very oldtimeBBC fashion and gives a comforti
ng and sometimes discomfoting sense of realism to the entire piece The actors are extremely well chosen Michael She
en not only has got all the polari but he has all the voices down pat too You can truly see the seamless editing gui
ded by the references to Williams diary entries not only is it well worth the watching but it is a terrificly writte
n and performed piece A masterful production about one of the great masters of comedy and his life The realism reall
y comes home with the little things the fantasy of the guard which rather than use the traditional dream techniques
remains solid then disappears It plays on our knowledge and our senses particularly with the scenes concerning Orton
and Halliwell and the sets particularly of their flat with Halliwells murals decorating every surface are terribly w
ell done'
```

```
In [9]: import nltk
from nltk.corpus import stopwords
```

```
In [10]: nltk.download('stopwords')
stopwords.words('english')
```

```
'not',
'only',
'own',
'same',
'so',
'than',
'too',
'very',
's',
't',
'can',
'will',
'just',
'don',
"don't",
'should',
"should've",
'now',
'd',
'll',
```

```
In [11]: token = cleantext.lower().split()
stopword = set(stopwords.words('english'))
token_list = [ word for word in token if word.lower() not in stopword ]
```

```
In [12]: " ".join(token_list)
```

```
Out[12]: 'wonderful little production filming technique unassuming oldtimebbc fashion gives comforting sometimes discomfortin
g sense realism entire piece actors extremely well chosen michael sheen got polari voices pat truly see seamless edi
ting guided references williams diary entries well worth watching terrificly written performed piece masterful produ
ction one great masters comedy life realism really comes home little things fantasy guard rather use traditional dre
am techniques remains solid disappears plays knowledge senses particularly scenes concerning orton halliwell sets pa
rticularly flat halliwells murals decorating every surface terribly well done'
```



```
In [18]: clean_data[0]
```

```
Out[18]: 'one reviewers mentioned watching 1 oz episode youll hooked right exactly happened methe first thing struck oz bruta  
lity unflinching scenes violence set right word go trust show faint hearted timid show pulls punches regards drugs s  
ex violence hardcore classic use wordit called oz nickname given oswald maximum security state penitentiary focuses m  
ainly emerald city experimental section prison cells glass fronts face inwards privacy high agenda em city home many  
aryans muslims gangstas latinosaurs christians italians irish moreso scuffles death stares dodgy dealings shady agreemen  
ts never far awayi would say main appeal show due fact goes shows wouldnt dare forget pretty pictures painted mainst  
ream audiences forget charm forget romanceoz doesnt mess around first episode ever saw struck nasty surreal couldnt  
say ready watched developed taste oz got accustomed high levels graphic violence violence injustice crooked guards w  
holl sold nickel inmates wholl kill order get away well mannered middle class inmates turned prison bitches due lack  
street skills prison experience watching oz may become comfortable uncomfortable viewingthats get touch darker side'
```

Train test split

```
In [19]: X_train, X_test, y_train, y_test = train_test_split(data, data.sentiment, test_size=0.2, random_state=42, stratify=da
```

```
In [20]: le = LabelEncoder()  
y_train = le.fit_transform(y_train)  
le_test = LabelEncoder()  
y_test = le_test.fit_transform(y_test)
```

```
In [21]: print(X_train.shape, y_train.shape)  
print(X_test.shape, y_test.shape)
```

```
(40000, 2) (40000,)  
(10000, 2) (10000,)
```



```
In [24]: clean_data_test_data = data_cleaner(X_test.review.values)
X_test['cleaned_text'] = clean_data_test_data
X_test.head()
```

```
0%|          | 0/10000 [00:00<?, ?i
t/s]C:\Users\santh\anaconda3\lib\site-packages\bs4\__init__.py:435: MarkupResemblesLocatorWarning: The input looks m
ore like a filename than markup. You may want to open this file and pass the filehandle into BeautifulSoup.
  warnings.warn(
100%|██████████| 10000/10000 [00:02<00:00, 3435.80i
t/s]
```

Out[24]:

	review	sentiment	cleaned_text
18870	Yes, MTV there really is a way to market Daria...	negative	yes mtv really way market daria started clever...
39791	The story of the bride fair is an amusing and ...	negative	story bride fair amusing engaging one filmmake...
30381	A team varied between Scully and Mulder, two o...	positive	team varied scully mulder two scientists pilot...
42294	This was a popular movie probably because of t...	negative	popular movie probably humor fastmoving story ...
33480	This movie made me so angry!! Here I am thinki...	negative	movie made angry thinking heres new horror mov...

Vectorizer

```
In [25]: vec = CountVectorizer()
vec = vec.fit(X_train.cleaned_text)
train_x_bow = vec.transform(X_train.cleaned_text)
test_x_bow = vec.transform(X_test.cleaned_text)
```

```
In [26]: print(train_x_bow.shape)
print(test_x_bow.shape)
```

```
(40000, 192139)
(10000, 192139)
```


Naive Bayes with Hyperparameter Tuning

```
In [27]: classifier = MultinomialNB()
```

```
In [28]: alpha_ranges = {"alpha": [0.001, 0.01, 0.1, 1, 10.0, 100]}
```

```
In [29]: grid_search = GridSearchCV(classifier, param_grid=alpha_ranges, scoring='accuracy', cv=3, return_train_score=True)
grid_search.fit(train_x_bow, y_train)
```

```
Out[29]:
```

```
└─ GridSearchCV
  └─ estimator: MultinomialNB
    └─ MultinomialNB
```

```
In [30]: alpha = [0.001, 0.01, 0.1, 1, 10.0, 100]
train_acc = grid_search.cv_results_['mean_train_score']
train_std = grid_search.cv_results_['std_train_score']

test_acc = grid_search.cv_results_['mean_test_score']
test_std = grid_search.cv_results_['std_test_score']
```

```
In [31]: grid_search.best_estimator_
```

```
Out[31]:
```

```
└─ MultinomialNB
  └─ MultinomialNB(alpha=1)
```

```
In [32]: classifier = MultinomialNB(alpha=1)
classifier.fit(train_x_bow, y_train)
```

```
Out[32]:
┆ MultinomialNB
┆ MultinomialNB(alpha=1)
```

```
In [33]: predict = classifier.predict(test_x_bow)
```

```
In [34]: print("Accuracy is ", accuracy_score(y_test, predict))
```

Accuracy is 0.8599

```
In [35]: print("Accuracy is ", classification_report(y_test, predict))
```

Accuracy is		precision	recall	f1-score	support
	0	0.85	0.88	0.86	5000
	1	0.87	0.84	0.86	5000
	accuracy		0.86	10000	
	macro avg	0.86	0.86	0.86	10000
	weighted avg	0.86	0.86	0.86	10000

TF-IDF Model with Naive Bayes:

```
In [36]: # Vectorize the text using TF-IDF model
tfidf_vectorizer = TfidfVectorizer()
X_train_tfidf = tfidf_vectorizer.fit_transform(X_train.cleaned_text)
X_test_tfidf = tfidf_vectorizer.transform(X_test.cleaned_text)

# Train a Naive Bayes classifier on the TF-IDF features
nb_classifier_tfidf = MultinomialNB()
nb_classifier_tfidf.fit(X_train_tfidf, y_train)

# Predict and calculate accuracy
predictions_tfidf = nb_classifier_tfidf.predict(X_test_tfidf)
accuracy_tfidf = accuracy_score(y_test, predictions_tfidf)
print("Accuracy using TF-IDF model:", accuracy_tfidf)
```

Accuracy using TF-IDF model: 0.867