```
In [1]:
         import pandas as pd
In [2]: imdb df = pd.read csv('IMDB Dataset.csv')
In [3]: | imdb df.info() #shows 19999 records with no null values
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 19999 entries, 0 to 19998
         Data columns (total 2 columns):
             Column
                         Non-Null Count Dtype
                          -----
         ____
          \cap
             review
                         19999 non-null object
            sentiment 19999 non-null object
         dtypes: object(2)
         memory usage: 312.6+ KB
In [4]: imdb sample df = imdb df.sample(n= 15000, random state= 42) #randomly picking 15000 recor
         imdb sample df.info()
         <class 'pandas.core.frame.DataFrame'>
         Int64Index: 15000 entries, 10650 to 13861
         Data columns (total 2 columns):
             Column
                         Non-Null Count Dtype
            review 15000 non-null object
          1 sentiment 15000 non-null object
         dtypes: object(2)
         memory usage: 351.6+ KB
In [5]:
         imdb sample df = imdb sample df.reset index(drop= True)
In [8]:
         imdb sample df
Out[8]:
                                                   review sentiment
              0
                   I don't want to bore everyone by reiterating w...
                                                            positive
                  I like Wes Studi & especially Adam Beach, but ...
                                                            negative
              2
                 I'm sure this was one of those "WOAH!" attract...
                                                            negative
              3
                    As a fan of Henriksen (I liked him in the "Mil...
                                                            negative
                 Best animated movie ever made. This film explo...
                                                            positive
         14995
                Harlan Knowles (Lance Henriksen) brings a grou...
                                                            negative
         14996
                     I can't help thinking that this is Franco's 'h...
                                                            negative
         14997
                  The movie starts in spring 2001. A soldier nam...
                                                            negative
         14998
                    I found the first bit of stop motion animation...
                                                            negative
         14999
                  I haven't read the source, Richard Brooks' nov...
                                                            positive
        15000 rows × 2 columns
In [9]: imdb sample df['sentiment'] = imdb sample df['sentiment'].map({'positive': 1, 'negative'
```

review sentiment

In [10]:

Out[10]:

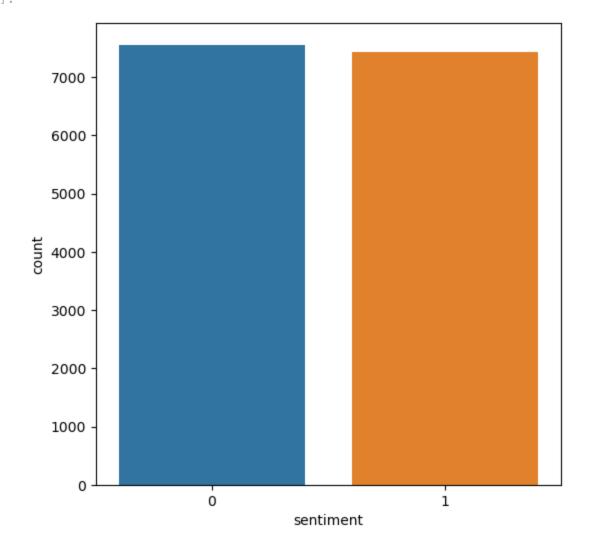
imdb sample df.head()

```
I don't want to bore everyone by reiterating w...
I like Wes Studi & especially Adam Beach, but ...
I'm sure this was one of those "WOAH!" attract...
As a fan of Henriksen (I liked him in the "Mil...
Best animated movie ever made. This film explo...
```

```
In [12]: import matplotlib.pyplot as plt
import seaborn as sn
%matplotlib inline
```

```
In [21]: plt.figure(figsize=(6,6))
sn.countplot(x= 'sentiment', data= imdb_sample_df)#checking if the positive and negative
#imbalanced
```

Out[21]: <AxesSubplot:xlabel='sentiment', ylabel='count'>



```
In [23]: imdb_sample_df['sentiment'].value_counts(normalize= True) #data is balanced with sentimen
#under-sampling, over-sampling is not required
```

```
Out[23]: 0     0.503933
1     0.496067
Name: sentiment, dtype: float64
```

```
In [24]: import re import nltk
```

performing text cleaning and preprocessing

```
In [25]:
         from nltk.corpus import stopwords
         from nltk.stem.porter import PorterStemmer
         from nltk.stem import WordNetLemmatizer
In [26]: lemmatizer = WordNetLemmatizer()
In [29]: corpus = []
         for i in range(0,len(imdb sample df)):
             review = re.sub('[^a-zA-Z]', ' ', imdb sample df['review'][i])
             review = review.lower()
             review = review.split() #extracting words out of a sentence that is tokens
             review = [lemmatizer.lemmatize(word) for word in review if not word in stopwords.wor
             review = ' '.join(review) #re joining words or tokens to create a sentence
             corpus.append(review)
In [30]: from sklearn.feature_extraction.text import CountVectorizer
         cv = CountVectorizer(stop words= 'english')
         X = cv.fit transform(corpus).toarray()
In [31]: Y = imdb sample df['sentiment']
Out[31]:
                  0
         14995
         14996
         14997
         14998
         14999
         Name: sentiment, Length: 15000, dtype: int64
```

finding max_features required

```
In [32]:
          import numpy as np
          featurecount = np.sum(X, axis= 0)
In [33]:
In [34]:
          features = cv.get feature names out()
In [35]:
          featurecountdf = pd.DataFrame(dict(features = features, counts = featurecount))
In [36]:
          featurecountdf
Out[36]:
                                                 features counts
              0
                                                              5
                                                     aaa
                                                aaaaahhhh
                                                 aaaarrgh
                                           aaaggghhhhhhhh
                                                              1
```

54136 rows × 2 columns

```
In [37]: len(featurecountdf[featurecountdf.counts == 1]) #checking count of words that occured onl
Out[37]: 20628
```

performing bag of words

```
In [38]: cv = CountVectorizer(stop_words= 'english', max_features= 5000)
In [39]: X = cv.fit_transform(corpus).toarray()
```

splitting the dataset into train and test

```
In [40]: from sklearn.model_selection import train_test_split
In [41]: X_train, X_test, y_train, y_test = train_test_split(X, Y, test_size= 0.3, random_state=
```

performing modeling

accuracy

```
In [43]: from sklearn.naive bayes import MultinomialNB
In [44]: mnb = MultinomialNB()
         sentiment detection model = mnb.fit(X train, y train)
In [45]: #prediciton
         y pred = sentiment detection model.predict(X test)
In [46]: from sklearn.metrics import accuracy score, classification report
In [47]: score = accuracy score(y test, y pred)
         print(score)
         0.83955555555556
In [48]: print(classification report(y pred, y test))
                       precision recall f1-score
                                                        support
                            0.85 0.83
0.83 0.85
                    0
                                                0.84
                                                          2293
                                                0.84
                                                          2207
```

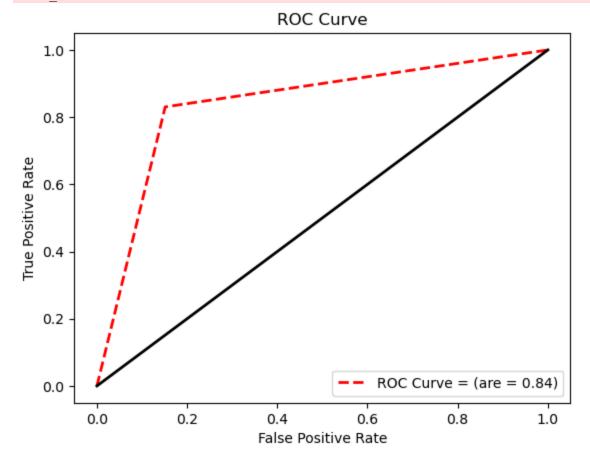
0.84

4500

```
0.84
            macro avg
                             0.84
                                       0.84
                                                            4500
                            0.84
                                                 0.84
                                                            4500
         weighted avg
                                       0.84
In [50]:
         from sklearn.metrics import confusion matrix
In [51]:
         cm = confusion matrix(y test, y pred)
In [52]: sn.heatmap(cm, annot= True, fmt= '.2f')
         plt.ylabel('True Label')
         plt.xlabel('Predicted Label')
          #confusion matrix shows the true and predicted values are having more accurate numbers
         Text(0.5, 23.5222222222222, 'Predicted Label')
Out[52]:
                                                                          - 1800
                                                                          - 1600
                         1912.00
                                                    341.00
            0 -
                                                                          - 1400
          True Label
                                                                          - 1200
                                                                          - 1000
                                                                          - 800
                          381.00
                                                   1866.00
                                                                          - 600
                            0
                                                       1
                                   Predicted Label
In [57]: from sklearn.metrics import roc curve, auc
In [55]:
         import matplotlib.pyplot as plt
         import seaborn as sn
         %matplotlib inline
In [58]: fpr, tpr, threshold = roc curve(y test, y pred)
         roc auc = auc(fpr, tpr)
         plt.figure()
         plt.plot(fpr, tpr, 'k--', color = 'red', label = 'ROC Curve = (are = {0:.2f})'.format(ro
         plt.plot([0, 1], [0, 1], color='black', lw=2, linestyle='-')
         plt.xlim([-0.05, 1.05])
         plt.ylim([-0.05, 1.05])
         plt.xlabel('False Positive Rate')
         plt.ylabel('True Positive Rate')
         plt.title('ROC Curve')
         plt.legend(loc='lower right')
         plt.show()
         /var/folders/3g/2bh9znv97b109382p2y xj9m0000gn/T/ipykernel 3905/1878658684.py:4: UserWar
         ning: color is redundantly defined by the 'color' keyword argument and the fmt string "k
```

```
--" (-> color='k'). The keyword argument will take precedence.

plt.plot(fpr, tpr, 'k--', color = 'red', label = 'ROC Curve = (are = {0:.2f})'.format
(roc auc), lw = 2)
```



testing with reviews

```
In [65]:
         def split words(words):
             review = re.sub('[^a-zA-Z]', '', words)
             review = review.lower()
             review = review.split() #extracting words out of a sentence that is tokens
             review = [lemmatizer.lemmatize(word) for word in review if not word in stopwords.wor
             review = ' '.join(review) #re joining words or tokens to create a sentence
             return review
In [78]:
         testreview = {"It is this combination of maximalism, nationalism, fatalism, and two-dime
                       "Disappointingly, the film's female characters are given precious little to
                       "In sci-fi terms, The Wandering Earth II feels less obviously derivative of
         clean review = []
         for i in testreview.keys():
             clean review.append(split words(i))
         vector comment = cv.transform(clean review)
         vector comment = vector comment.toarray()
In [79]: predict = sentiment detection model.predict(vector comment)
         result = {1: 'positive', 0: 'negative'}
         values = list(testreview.values())
         print("comment num \tpredict \treal")
In [86]:
         for i, val in enumerate(predict):
             print(f"comment {i+1} \t{result[val]} \t{values[i]}")
         comment num
                         predict
         comment 1
                         negative
                                          positive
```

comment 2negativenegativecomment 3positivepositive