

Introduction

IMDb, the Internet Movie Database, is a cornerstone for film enthusiasts and industry professionals alike. Established in 1990, it has evolved into a comprehensive online repository, offering a wealth of information about movies, TV shows, video games, and streaming content. From cast and crew details to release dates, plot summaries, and user ratings, IMDb provides a one-stop destination for exploring the world of entertainment. The platform's multimedia content, including trailers and interviews, enhances the user experience. With a robust social component, allowing users to rate, review, and engage in discussions, IMDb has become an essential tool for navigating the cinematic landscape.

1 import Necessary Library

In [305...

import pandas as pd
import numpy as np

2 import Dataset

In [306... df = pd.read_csv("/kaggle/input/imdb-movie/IMDB Movie.csv")
In [307... df.head()

Out[307...

o One of the other reviewers has mentioned that ... positive 1 A wonderful little production.

 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

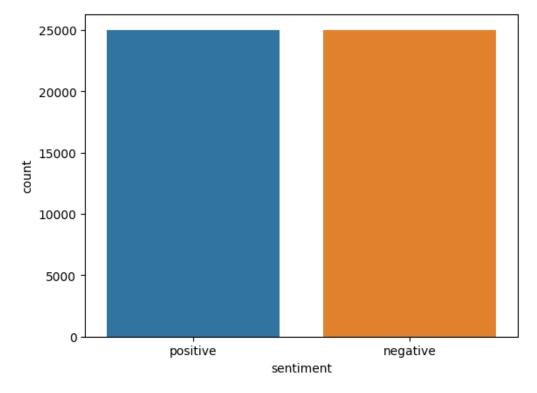
```
In [308... df.tail()
```

Out[308... review sentiment

49995 I thought this movie did a down right good job... positive
49996 Bad plot, bad dialogue, bad acting, idiotic di... negative
49997 I am a Catholic taught in parochial elementary... negative
49998 I'm going to have to disagree with the previou... negative
49999 No one expects the Star Trek movies to be high... negative

import seaborn as sns
sns.countplot(x='sentiment',data=df)

Out[311... <Axes: xlabel='sentiment', ylabel='count'>



```
positive_review=list(df[df['sentiment']=='positive']['review'])[:100]
            negative_review=list(df[df['sentiment']=='negative']['review'])[:100]
In [313...
            from wordcloud import WordCloud, STOPWORDS
            from matplotlib import pyplot as plt
            stopwords = set(STOPWORDS)
            stopwords
           {'a',
Out[313...
             'about',
             'above',
             'after',
             'again',
             'against',
             'all',
             'also',
             'am',
             'an',
             'and',
             'any',
             'are',
            "aren't",
             'as',
             'at',
             'be',
             'because',
             'been',
            'before',
             'being',
             'below',
             'between',
             'both',
             'but',
             'by',
             'can',
            "can't",
             'cannot',
             'com',
             'could',
             "couldn't",
             'did',
            "didn't",
             'do',
             'does',
            "doesn't",
             'doing',
            "don't",
             'down',
             'during',
             'each',
            'else',
             'ever',
             'few',
             'for',
             'from',
             'further',
             'get',
             'had',
            "hadn't",
             'has',
            "hasn't",
             'have',
             "haven't",
             'having',
             'he',
            "ה ה ה חו"
```

```
ne u ,
"he'll",
"he's",
'hence',
'her',
'here',
"here's",
'hers',
'herself',
'him',
'himself',
'his',
'how',
"how's",
'however',
'http',
'i',
"i'd",
"i'll",
"i'm",
"i've",
'if',
'in',
'into',
'is',
"isn't",
'it',
"it's",
'its',
'itself',
'just',
'k',
"let's",
'like',
'me',
'more',
'most',
"mustn't",
'my',
'myself',
'no',
'nor',
'not',
'of',
'off',
'on',
'once',
'only',
'or',
'other',
'otherwise',
'ought',
'our',
'ours',
'ourselves',
'out',
'over',
'own',
'r',
'same',
'shall'
"shan't",
'she',
"she'd"
"she'11",
"she's",
'should',
```

"shouldn't"

```
J.11041411 C )
'since',
'so',
'some',
'such',
'than',
'that',
"that's",
'the',
'their',
'theirs',
'them',
'themselves',
'then',
'there',
"there's",
'therefore',
'these',
'they',
"they'd",
"they'll",
"they're",
"they've",
'this',
'those',
'through',
'to',
'too',
'under',
'until',
'up',
'very',
'was',
"wasn't",
'we',
"we'd",
"we'll",
"we're",
"we've",
'were',
"weren't",
'what',
"what's",
'when',
"when's",
'where',
"where's",
'which',
'while',
'who',
"who's",
'whom',
'why',
"why's",
'with',
"won't",
'would',
"wouldn't",
'www',
'you',
"you'd",
"you'11",
"you're",
"you've",
'your',
'yours',
'yourself',
'yourselves'}
```

In [315...

create_cloud(' '.join(positive_review).lower(), 'words is positive review')

words is positive review

small thinkgirl phamlet wonderfulbeautiful director true police comedy

family C true police truly police truly considered to the police truly considered truly considered to the police truly considered truly considered to the police truly considered truly considered to the police truly considered truly con

In [316...

create_cloud(' '.join(negative_review).lower(),'words in negative review')

words in negative review funny place book say role man idea löök stillseen fact Some 1 ette **O**watching go_{hard} many actually life eop. give muc worst chara 50 stupid kid looking Sign got must maybe cast Tgoing show actor every

```
TII [3T/...
             def text_processing(data):
                 from bs4 import BeautifulSoup
                 import re
                 def decontracted(phrase):
                      # specific
                      phrase= re.sub(r'<br /><br />',' ',phrase)
                      phrase = re.sub(r"won't", "will not", phrase)
                      phrase = re.sub(r"can\'t", "can not", phrase)
                      # general
                      phrase = re.sub(r"n\'t", " not", phrase)
phrase = re.sub(r"\'re", " are", phrase)
phrase = re.sub(r"\'s", " is", phrase)
phrase = re.sub(r"\'d", " would", phrase)
                      phrase = re.sub(r"\'ll", " will", phrase)
nhrase = re.sub(r"\'t", " not", phrase)
                      phrase = re.sub(r"\'ve", " have", phrase)
phrase = re.sub(r"\'m", " am", phrase)
                      phrase = re.sub(r'"', " ", phrase)
                      return phrase
                 stopwords=set(STOPWORDS)
                 # Combining all the above sentence
                 from tqdm import tqdm
                 preprocessed_reviews = []
                 # tqdm is for printing the status bar
                 for sentance in tqdm(data['review'].values):
                      sentance = re.sub(r"http\S+", "", sentance)
                      sentance = BeautifulSoup(sentance, 'lxml').get_text()
                      sentance = decontracted(sentance)
                      sentance = re.sub("\S*\d\S*", "", sentance).strip()
                      # https://gist.github.com/sebleier/554280
                      sentance = ' '.join(e.lower() for e in sentance.split() if e not in s
                      preprocessed_reviews.append(sentance.strip())
                 from nltk.stem import PorterStemmer
                 porter = PorterStemmer()
                 list_of_sentence=[]
                 for sentence in preprocessed_reviews:
                      words_in_sentence=[]
                      for words in sentence.split():
                          words in sentence.append(porter.stem(words))
                      list of sentence.append(' '.join(words in sentence))
                 return(list of sentence)
In [318...
             x=text processing(df[:1000])
                          818/1000 [00:00<00:00, 2073.64it/s]/tmp/ipykernel_43/22288752
          65.py:29: MarkupResemblesLocatorWarning: The input looks more like a filename t
          han markup. You may want to open this file and pass the filehandle into Beautif
          ul Soup.
            sentance = BeautifulSoup(sentance, 'lxml').get_text()
                        | 1000/1000 [00:00<00:00, 2049.99it/s]
In [319...
             df = df[:1000]
In [320...
             df.head()
Out[320...
                                                      review sentiment
            0 One of the other reviewers has mentioned that ...
                                                                 positive
```

```
1 A wonderful little production. <br /> <br /> 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
In [321...
             df['cleaned_review']=x
In [322...
             df.head()
Out[322...
                                         review
                                                  sentiment
                                                                                   cleaned_review
                   One of the other reviewers has
                                                                one review mention watch oz episod
            0
                                                     positive
                               mentioned that ...
                                                                                      will hooked...
                A wonderful little production. <br
                                                                   a wonder littl production. the film
            1
                                                     positive
                                  /> <br /> The...
                                                                                       techniqu u...
                   I thought this was a wonderful
                                                               i thought wonder way spend time hot
            2
                                                     positive
                                way to spend ti...
                                                                                     summer wee...
                 Basically there's a family where a
                                                               basic famili littl boy (jake) think zombi
            3
                                                    negative
                                     little boy ...
                                                                                             clos...
                 Petter Mattei's "Love in the Time
                                                                petter mattei love time money visual
            4
                                                    positive
                                   of Money" is...
                                                                                        stun film...
In [323...
             X = df['cleaned_review']
             Y = df['sentiment']
In [324...
             Х
Out[324...
            0
                    one review mention watch oz episod will hooked...
                    a wonder littl production. the film techniqu u...
            2
                    i thought wonder way spend time hot summer wee...
            3
                    basic famili littl boy (jake) think zombi clos...
            4
                    petter mattei love time money visual stun film...
            995
                    noth sacred. just ask erni fosselius. these da...
            996
                    i hate it. i hate self-awar pretenti inan masq...
            997
                    i usual tri profession construct i critic movi...
            998
                    if go see film histori class someth school, tr...
            999
                    thi zoolog textbook, given depict anim accurat...
            Name: cleaned_review, Length: 1000, dtype: object
In [325...
Out[325...
            0
                    positive
            1
                    positive
            2
                    positive
            3
                    negative
            4
                    positive
                      . . .
            995
                    positive
                    negative
            996
            997
                    negative
            998
                    negative
```

```
999
                     negative
             Name: sentiment, Length: 1000, dtype: object
In [326...
              Y = list(Y)
              for i in range(len(Y)):
                   if Y[i]=='positive':
                        Y[i]=1
                   else:
                        Y[i]=0
              df['sentiment_score']=Y
              Y=df['sentiment score']
In [327...
              df
Out[327...
                                    review
                                             sentiment
                                                                      cleaned_review sentiment_score
                          One of the other
                                                            one review mention watch
               0
                             reviewers has
                                                                                                        1
                                                positive
                                                              oz episod will hooked...
                         mentioned that ...
                         A wonderful little
                                                            a wonder littl production.
                    production. <br/> <br/> <br/> <br/>
                                                                                                        1
               1
                                                positive
                                                                 the film techniqu u...
                                   />The...
                       I thought this was a
                                                                i thought wonder way
               2
                         wonderful way to
                                                              spend time hot summer
                                                                                                        1
                                                positive
                                 spend ti...
                                                                                wee...
                         Basically there's a
                                                            basic famili littl boy (jake)
               3
                       family where a little
                                               negative
                                                                                                        0
                                                                    think zombi clos...
                                     boy ...
                    Petter Mattei's "Love in
                                                               petter mattei love time
                                                positive
                                                                                                        1
                   the Time of Money" is...
                                                              money visual stun film...
                     Nothing is sacred. Just
                                                             noth sacred. just ask erni
             995
                                                positive
                                                                                                        1
                     ask Ernie Fosselius. T...
                                                                  fosselius. these da...
                      I hated it. I hate self-
                                                              i hate it. i hate self-awar
             996
                         aware pretentious
                                                                                                        0
                                               negative
                                                                 pretenti inan masq...
                                     inan...
                         I usually try to be
                                                                 i usual tri profession
             997
                          professional and
                                                                                                        0
                                               negative
                                                               construct i critic movi...
                                construct...
                     If you like me is going
                                                             if go see film histori class
             998
                                               negative
                                                                                                        0
                      to see this in a film ...
                                                                   someth school, tr...
                                                           thi zoolog textbook, given
                      This is like a zoology
             999
                                                                                                        0
                                               negative
                   textbook, given that it...
                                                                 depict anim accurat...
            1000 rows × 4 columns
In [328...
              from sklearn.model_selection import train_test_split
In [329...
              X_train, X_test, y_train, y_test = train_test_split(X[:1000], Y[:1000], test_
```

```
In [330...
            X train
Out[330...
          105
                  marion davi star remark comedi show peopl rele...
           68
                  i sure produc need trade name somewhat success...
           479
                  joseph bradi clarenc doolittl two sailors, fou...
           399
                  thi movi fairli entertain comedi murphi law ap...
           434
                  yes, indeed, good movie. a love biangle, (sorr...
          835
                  the stori told before. a deadli diseas spread ...
          192
                  nifti littl episod play mainli laughs, clever ...
          629
                  let keep simple: my two kid glu movie. it flaw...
           559
                  so i rent netflix somebodi gave roger ebert bo...
           684
                  the perfect murder foil wife(play mari ellen t...
           Name: cleaned_review, Length: 700, dtype: object
In [331...
           y_train
          105
Out[331...
                  1
                  0
           68
           479
                  1
           399
                  1
           434
                  0
           835
                  1
           192
                  1
           629
                  1
           559
                  0
           Name: sentiment_score, Length: 700, dtype: int64
In [332...
           list(y_test).count(0)
Out[332...
          162
In [333...
            list(y_test).count(1)
Out[333...
          138
In [334...
            from sklearn.feature_extraction.text import CountVectorizer
            vectorizer = CountVectorizer()
           X_train_bow = vectorizer.fit_transform(X_train)
           X_test_bow = vectorizer.transform(X_test)
In [335...
           X_train_bow.shape, X_test_bow.shape
           ((700, 12936), (300, 12936))
Out[335...
In [336...
           X_test_bow
           <300x12936 sparse matrix of type '<class 'numpy.int64'>'
Out[336...
                   with 28221 stored elements in Compressed Sparse Row format>
In [337...
           X_train.shape, X_test.shape
Out[337...
          ((700,),(300,))
```

```
In [338...
            y train.shape, y test.shape
Out[338...
           ((700,),(300,))
```

(1) KNN Algorithm

```
In [339...
           from sklearn.neighbors import KNeighborsClassifier
           from sklearn.metrics import accuracy_score
           from sklearn.metrics import f1 score
           for i in range(10,30):
               print('K',i)
               # initialization
               neigh = KNeighborsClassifier(n neighbors=i)
               # Training
               neigh.fit(X train bow, y train)
               # Test the training data
               y_pred_train = neigh.predict(X_train_bow)
               accuracy_train = accuracy_score(y_pred_train,y_train)
               f1_train = f1_score(y_pred_train,y_train)
               # Test the test data
               y pred test = neigh.predict(X test bow)
               accuracy_test = accuracy_score(y_pred_test,y_test)
               f1_test = f1_score(y_pred_test,y_test)
               print(accuracy_train,accuracy_test)
               print(f1 train,f1 test)
               print()
        0.69 0.5533333333333333
        0.7290886392009986 0.5838509316770186
        K 11
        0.66 0.5333333333333333
        0.7337807606263983 0.6089385474860335
        K 12
        0.6757142857142857 0.55
        0.7241798298906439 0.5896656534954408
        K 13
        0.6614285714285715 0.5066666666666667
        0.7363737486095662 0.6
        K 14
        0.6914285714285714 0.53
        0.7422434367541767 0.5936599423631124
        K 15
        0.6528571428571428 0.5166666666666667
        0.7338444687842279 0.6214099216710183
        K 16
        0.6742857142857143 0.536666666666666
        0.733644859813084 0.6084507042253522
```

0.6414285714285715 0.5233333333333333

0.7238723872387239 0.6285714285714286

```
K 18
         0.6571428571428571 0.56
         0.725400457665904 0.641304347826087
         K 19
         0.6257142857142857 0.5133333333333333
         0.7164502164502164 0.6256410256410257
         0.6428571428571429 0.5533333333333333
         0.7203579418344519 0.6417112299465241
         0.6071428571428571 0.526666666666666
         0.7089947089947091 0.6377551020408163
         K 22
         0.6271428571428571 0.5433333333333333
         0.7159956474428727 0.6422976501305483
         K 23
         0.6 0.53
         0.7089397089397088 0.6430379746835443
         K 24
         0.6171428571428571 0.5433333333333333
         0.7136752136752136 0.6422976501305483
         K 25
         0.6028571428571429 0.52
         0.7104166666666667 0.6363636363636364
        K 26
         0.6057142857142858 0.5333333333333333
         0.7044967880085653 0.6391752577319587
         K 27
         0.5957142857142858 0.51
         0.7055150884495316 0.631578947368421
         K 28
         0.6071428571428571 0.5166666666666667
         0.7083775185577942 0.6272493573264781
         K 29
         0.59 0.516666666666667
         0.7044284243048403 0.6401985111662531
In [340...
           X_train = X_train_bow.toarray()
           X_test = X_test_bow.toarray()
In [341...
           X_train[0]
          array([0, 0, 0, ..., 0, 0, 0])
Out[341...
In [342...
           X_test[0]
          array([0, 0, 0, ..., 0, 0, 0])
Out[342...
```

Machine Learning Algorithm

(1) KNN 🖸

```
In [343... from sklearn.neighbors import KNeighborsClassifier
    from sklearn.metrics import accuracy_score
    from sklearn.metrics import f1_score

In [344... neigh = KNeighborsClassifier(n_neighbors=20)

In [345... neigh.fit(X_train_bow, y_train)
```

Out[345... KNeighborsClassifier(n_neighbors=20)

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

```
In [346...
# Test the training data
y_pred_train = neigh.predict(X_train_bow)
accuracy_train = accuracy_score(y_pred_train,y_train)
f1_train = f1_score(y_pred_train,y_train)

# Test the test data
y_pred_test = neigh.predict(X_test_bow)
accuracy_test = accuracy_score(y_pred_test,y_test)
f1_test = f1_score(y_pred_test,y_test)

print(accuracy_train,accuracy_test)
print("f1_train : ",f1_train)
print("f1_test : ",f1_test)
```

(2) Naive Bayes classifier

Out[350...

GaussianNB()

```
In a Jupyter environment, please rerun this cell to show the HTML representation
          or trust the notebook.
          On GitHub, the HTML representation is unable to render, please try loading this
          page with nbviewer.org.
In [351...
            train_predictions = G_classifier.predict(X_train)
            train_accuracy21 = accuracy_score(y_train, train_predictions)
In [352...
            test predictions = G classifier.predict(X test)
            test_accuracy21 = accuracy_score(y_test, test_predictions)
In [353...
            print(f"Training Accuracy: {train_accuracy21}")
            print(f"Testing Accuracy: {test_accuracy21}")
         Training Accuracy: 0.9985714285714286
         Testing Accuracy: 0.6233333333333333
In [354...
            # BernoulliNB
In [355...
            B_classifier = BernoulliNB()
In [356...
            B_classifier.fit(X_train, y_train)
          BernoulliNB()
Out[356...
          In a Jupyter environment, please rerun this cell to show the HTML representation
          or trust the notebook.
          On GitHub, the HTML representation is unable to render, please try loading this
          page with nbviewer.org.
In [357...
            train_predictions = B_classifier.predict(X_train)
            train_accuracy22 = accuracy_score(y_train, train_predictions)
In [358...
            test predictions = G classifier.predict(X test)
            test_accuracy22 = accuracy_score(y_test, test_predictions)
In [359...
            print(f"Training Accuracy: {train_accuracy22}")
            print(f"Testing Accuracy: {test_accuracy22}")
         Training Accuracy: 0.9928571428571429
         Testing Accuracy: 0.6233333333333333
In [360...
            # MultinomiaLNB
In [361...
            M_classifier = MultinomialNB()
```

```
In [362...
            M_classifier.fit(X_train, y_train)
          MultinomialNB()
Out[362...
          In a Jupyter environment, please rerun this cell to show the HTML representation
          or trust the notebook.
          On GitHub, the HTML representation is unable to render, please try loading this
          page with nbviewer.org.
In [363...
            train predictions = M classifier.predict(X train)
            train_accuracy23 = accuracy_score(y_train, train_predictions)
In [364...
           test_predictions = M_classifier.predict(X_test)
            test_accuracy23 = accuracy_score(y_test, test_predictions)
In [365...
            print(f"Training Accuracy: {train_accuracy23}")
            print(f"Testing Accuracy: {test_accuracy23}")
         Training Accuracy: 0.9914285714285714
         Testing Accuracy: 0.786666666666666
In [366...
           # GaussianNB
            # BernoulliNB
            # MultinomialNB
            # Being the best of them | BernoulliNB |
```

(3) Decision Tree

```
In [367... from sklearn.tree import DecisionTreeClassifier
In [368... clf = DecisionTreeClassifier()
In [369... clf.fit(X_train, y_train)
Out[369... DecisionTreeClassifier()
In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
    On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.
In [370... train_predictions = clf.predict(X_train)
    train_accuracy3 = accuracy_score(y_train, train_predictions)
```

test_predictions = clf.predict(X_test)

test_accuracy3 = accuracy_score(y_test, test_predictions)

In [371...

```
In [372...
           print(f"Training Accuracy: {train_accuracy3}")
           print(f"Testing Accuracy: {test_accuracy3}")
```

Training Accuracy: 1.0

Testing Accuracy: 0.686666666666666

(4) Random Forest

RandomForestClassifier(random_state=42)

```
In [373...
           from sklearn.ensemble import RandomForestClassifier
```

In [374... rf_classifier = RandomForestClassifier(n_estimators=100, random_state=42)

In [375... rf_classifier.fit(X_train, y_train)

Out[375... In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

> On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

```
In [376...
           train_predictions = rf_classifier.predict(X_train)
           train_accuracy4 = accuracy_score(y_train, train_predictions)
```

```
In [377...
           test_predictions = rf_classifier.predict(X_test)
           test accuracy4 = accuracy score(y test, test predictions)
```

```
In [378...
           print(f"Training Accuracy: {train accuracy4}")
           print(f"Testing Accuracy: {test accuracy4}")
```

Training Accuracy: 1.0

Testing Accuracy: 0.816666666666667

(5) Boosting Algorithm 😂



```
In [379...
           from sklearn.ensemble import AdaBoostClassifier
In [380...
           base classifier = DecisionTreeClassifier(max depth=1)
In [381...
           adaboost_classifier = AdaBoostClassifier(base_classifier, n_estimators=50, ra
In [382...
           adaboost_classifier.fit(X_train, y_train)
         AdaBoostClassifier(estimator=DecisionTreeClassifier(max_depth=1),
Out[382...
                               random state=42)
```

In a Jupyter environment, please rerun this cell to show the HTML representation

or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

Training Accuracy: 0.8871428571428571 Testing Accuracy: 0.7566666666666667

(6).SVM 🖸

```
In [386... from sklearn.preprocessing import StandardScaler from sklearn.svm import SVC

In [387... scaler = StandardScaler()
    X_train = scaler.fit_transform(X_train)
    X_test = scaler.transform(X_test)

In [388... svm_classifier = SVC(kernel='linear', C=1.0)

In [389... svm_classifier.fit(X_train, y_train)

Out[389... SVC(kernel='linear')
    In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.
```

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

Training Accuracy: 1.0

Testing Accuracy: 0.756666666666667

(7). Logistic Regression

```
In [393... | from sklearn import linear_model

In [394... | lrg = linear_model.LogisticRegression()

In [395... | lrg.fit(X_train, y_train)

Out[395... | LogisticRegression()
In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.
```

Training Accuracy: 1.0 Testing Accuracy: 0.77

(8).Linear Regression

```
In [399... from sklearn.linear_model import LinearRegression from sklearn.metrics import mean_squared_error

In [400... model = LinearRegression()

In [401... model.fit(X_train, y_train)

Out[401... LinearRegression()

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this page with nbviewer.org.

In [402... train_predictions = clf.predict(X_train)
```

train_accuracy8 = accuracy_score(y_train, train_predictions)

Training Accuracy: 0.9514285714285714 Testing Accuracy: 0.7033333333333334

(9). Gradient Boosting Machines (GBM)

```
In [405... from sklearn.ensemble import GradientBoostingClassifier

In [406... model = GradientBoostingClassifier(n_estimators=100, learning_rate=0.1, max_d

In [407... model.fit(X_train, y_train)

Out[407... GradientBoostingClassifier(random_state=42)

In a Jupyter environment, please rerun this cell to show the HTML representation or trust the notebook.

On GitHub, the HTML representation is unable to render, please try loading this
```

Training Accuracy: 0.9742857142857143 Testing Accuracy: 0.8033333333333333

page with nbviewer.org.

Random Forest, Decision Tree, Gradient Boosting Machines (GBM), Algorithm is the best accuracy

(GradientBoostingClassifier)

Training Accuracy mean: 1.0

Testing Accuracy mean: 0.95