```
# importing required primary packages
import numpy as np import pandas as
pd import seaborn as sns import
matplotlib.pyplot as plt
# reading the csv file df =
pd.read_csv('/content/Restaurant_Reviews.tsv',sep='\t')
df
                                   Review Liked
1
                            Crust is not good.
                    I think food should have flavor and texture an... 0
995
996
                    Appetite instantly gone. 0
997
                    Overall I was not impressed and would not go b...
                                                       0
                    The whole experience was underwhelming, and I \dots
998
999
                    Then, as if I hadn't wasted enough of my life ... 0
1000
                    rows x 2 columns
# getting info of the data set
df.info()
   <class 'pandas.core.frame.DataFrame'>
   RangeIndex: 1000 entries, 0 to 999
   Data columns (total 2 columns):
    # Column Non-Null Count Dtype
       -----
      Review 1000 non-null object
   1 Liked 1000 non-null int64
   dtypes: int64(1), object(1) memory
   usage: 15.8+ KB
# removing the repeated or duplicate values
df = df.drop_duplicates(keep = 'last') df
                                   Review Liked
     0
                        Wow... Loved this place.
                            Crust is not good.
     1
                                           0
               Not tasty and the texture was just nasty.
     2
         Stopped by during the late May bank holiday of...
```

995 I think food should have flavor and texture an... 0

The selection on the menu was great and so wer...

996 Appetite instantly gone. 0

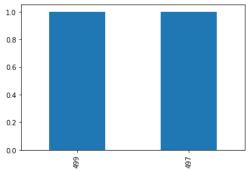
997 Overall I was not impressed and would not go b... 0
998 The whole experience was underwhelming, and I ... 0

999 Then, as if I hadn't wasted enough of my life ... 0

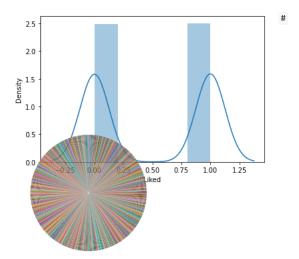
```
df.info()
     <class 'pandas.core.frame.DataFrame'>
     Int64Index: 996 entries, 0 to 999
     Data columns (total 2 columns):
     # Column Non-Null Count Dtype
     0 Review 996 non-null
1 Liked 996 non-null
                               object
                               int64
     dtypes: int64(1), object(1) memory
     usage: 23.3+ KB
df['Review'][900]
     'Spend your money elsewhere.'
df['Liked'][900]
     0
df['Review'][500]
     'I also had to taste my Mom's multi-grain pumpkin pancakes with pecan butter and they were
     amazing, fluffy, and delicious!'
df['Liked'][500]
     1
# finding number of unique values
data_frame=df['Liked'].value_counts().to_frame()
data_frame
        Liked
          499
      0
          497
```

pandas plotting data_frame['Liked'].value_counts().plot(kind='bar')

<matplotlib.axes._subplots.AxesSubplot at 0x7f44857cd2d0>



data visualization using seaborn distplot
sns.distplot(df['Liked']) /usr/local/lib/python3.7/distpackages/seaborn/distributions.py:2619: FutureWarning: `distpl
warnings.warn(msg, FutureWarning)
<matplotlib.axes._subplots.AxesSubplot at 0x7f4485713050>



data visualization using matplotlib
this piechart is showing the all 997 values with their
respective values (0&1) import matplotlib.pyplot as plt x
= df['Review'].values y = df['Liked'].values plt.pie(y)
plt.show()

[0, 0, 0, ..., 0, 0, 0]])

x_test_vect.toarray()

```
# rows - 996 and columns - 2
df.shape
     (996, 2)
\ensuremath{\text{\#}} splitting the date into training data and testing data from
sklearn.model selection import train test split
x\_train, x\_test, y\_train, y\_test=train\_test\_split(x, y, random\_state=0)
x_train.shape
     (747,)
x\_test.shape
     (249,)
# using countvectorizer coverting the text data to numerical values
# that is it involves removing of stop words and counting the occurence of each useful word and storing them
from sklearn.feature_extraction.text import CountVectorizer
vect = CountVectorizer(stop_words='english') x_train_vect =
vect.fit_transform(x_train) x_test_vect =
vect.transform(x_test)
x_train_vect.toarray()
     array([[0, 0, 0, ..., 0, 0, 0],
     [0, 0, 0, ..., 0, 0, 0],
            [0, 0, 0, ..., 0, 0, 0],
            [0, 0, 0, ..., 0, 0, 0],
[0, 0, 0, ..., 0, 0, 0],
```

```
array([[0, 0, 0, ..., 0, 0, 0],
    [0, 0, 0, ..., 0, 0, 0],
           [0, 0, 0, \ldots, 0, 0, 0],
           [0, 0, 0, ..., 0, 0, 0],
           [0, 0, 0, \ldots, 0, 0, 0],
           [0, 0, 0, ..., 0, 0, 0]])
x_train_vect.shape
    (747, 1531)
x_test_vect.shape
    (249, 1531)
#METHOD 1 # using support vector
classifier from sklearn.svm
import SVC model1 = SVC()
model1.fit(x_train_vect,y_train)
    SVC()
# predicted values y_pred1 =
model1.predict(x_test_vect) y_pred1
     1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 1, 0, 1, 1,
           1, 0, 0, 1, 0, 0, 1, 1, 1, 1, 0, 1, 0, 1, 0, 1, 1, 0, 0, 1, 1, 1,
           0,\ 0,\ 0,\ 0,\ 1,\ 1,\ 0,\ 0,\ 0,\ 1,\ 0,\ 1,\ 1,\ 1,\ 0,\ 0,\ 0,\ 1,\ 0,\ 0,\ 1,
           0, 0, 1, 1, 0, 1, 1, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0,
           0, 1, 1, 0, 0, 1, 1, 1, 0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 0, 0, 1, 1,
           1, 0, 0, 1, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 1, 1, 1, 0, 0,
           1, 1, 0, 0, 1, 0, 0, 0, 0, 0, 1, 1, 1, 1, 0, 1, 0, 0, 0, 1, 0, 1,
           0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0, 1, 0, 1, 1, 0, 0, 0, 1, 0,
           1, 1, 0, 0, 1, 0, 0, 1, 1, 0, 1, 1, 0, 0, 1, 1, 0, 1, 0, 1, 0, 1,
           0, 0, 1, 1, 0, 1, 1, 1, 0, 0, 1, 0, 1, 1, 0, 1, 0, 0, 0, 0, 0, 1,
    1, 0, 1, 0, 1, 0, 1])
# finding accuracy of the model from
sklearn.metrics import accuracy_score
accuracy_score(y_pred1,y_test)
    0.7269076305220884
# MFTHOD 2
# combining countvectorizer and support vector
classifier using pipeline from sklearn.pipeline import
make_pipeline model2 =
make_pipeline(CountVectorizer(),SVC())
model2.fit(x_train,y_train)
    Pipeline(steps=[('countvectorizer', CountVectorizer()), ('svc', SVC())])
# predicted values y_pred2 =
model2.predict(x_test) y_pred2
     array([0, 1, 0, 0, 1, 0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 1, 1, 1, 1, 1, 0, 1, 1,
            1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 0, 1,
           1, 0, 1, 1, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 0, 1, 1, 1,
           0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 1, 0, 1, 1,
           1, 0, 0, 1, 0, 0, 1, 0, 1, 1, 0, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0,
           0, 1, 0, 0, 0, 1, 1, 1, 0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 0, 1, 1,
           1, 1, 1, 1, 1, 0, 0, 0, 0, 0, 1, 1, 0, 0, 1, 0, 1, 1, 1, 0, 0,
           1, 1, 0, 0, 1, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 0, 0, 0, 1, 0,
           0, 1, 0, 0, 0, 1, 1, 0, 1, 1, 0, 0, 1, 1, 0, 1, 1, 0, 0, 0, 1, 1,
           1, 1, 1, 0, 1, 0, 0, 1, 1, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 1, 1, 0,
           1, 1, 0, 1, 0, 1, 1, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 0, 0, 0, 1, 1,
    1, 0, 1, 1, 0, 0, 1])
# finding accuracy of the model
accuracy_score(y_pred2,y_test)
```

0.8152610441767069

```
# METHOD 3 # using Naive Bayes from
sklearn.naive_bayes import MultinomialNB
model3 = MultinomialNB()
model3.fit(x_train_vect,y_train)
     MultinomialNB()
# predicted values y_pred3 =
model3.predict(x_test_vect) y_pred3
     \mathsf{array}([\,0,\ 1,\ 0,\ 0,\ 1,\ 0,\ 1,\ 0,\ 1,\ 1,\ 1,\ 1,\ 0,\ 1,\ 0,\ 0,\ 1,\ 0,\ 1,\ 0,\ 1,\ 1,
            1, 0, 1, 0, 0, 1, 1, 0, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 0, 1,
            1, 0, 0, 1, 0, 0, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 1, 0, 0, 1, 1, 1,
            1, 0, 0, 0, 1, 1, 0, 0, 0, 0, 0, 1, 1, 1, 0, 1, 0, 1, 0, 0, 1,
            1, 0, 1, 1, 1, 1, 0, 1, 1, 0, 1, 1, 0, 0, 0, 0, 0, 1, 0, 1, 0,
            0, 1, 0, 1, 0, 1, 1, 1, 0, 0, 1, 0, 0, 0, 1, 0, 1, 0, 0, 0, 1, 1,
            1, 1, 0, 1, 1, 1, 0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 0, 0, 0,
            1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 0, 1, 1, 1, 0, 1, 1, 1, 0, 1, 1, 1,
            0, 0, 0, 1, 0, 0, 1, 0, 1, 0, 1, 0, 1, 1, 1, 1, 1, 1, 0, 1, 1, 1,
            1, 1, 1, 0, 1, 0, 0, 1, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 0, 1, 0, 1,
            1, 0, 1, 1, 1, 1, 0, 1, 0, 0, 0, 1, 1, 0, 1, 1, 1, 0, 0, 1, 1,
     1, 0, 1, 1, 0, 0, 1])
# finding accuracy of the model
accuracy_score(y_pred3,y_test)
     0.7469879518072289
# MFTHOD 4
# combining count vectorizer and Naive Bayes
using pipeline from sklearn.pipeline import
make_pipeline model4 =
make_pipeline(CountVectorizer(),MultinomialNB())
model4.fit(x_train,y_train)
     Pipeline(steps=[('countvectorizer', CountVectorizer()),
                     ('multinomialnb', MultinomialNB())])
# predicted values y_pred4 =
model4.predict(x_test) y_pred4
     array([0, 1, 0, 0, 1, 0, 1, 0, 1, 1, 0, 1, 0, 1, 0, 0, 1, 0, 1, 0, 1, 1,
            1, 0, 1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 0, 1, 1, 0, 0, 1, 1, 0, 1,
            1, 0, 1, 1, 0, 0, 1, 1, 0, 0, 0, 1, 0, 0, 0, 1, 1, 0, 0, 1, 1, 1,
            0, 0, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 1, 0, 1, 0, 0, 0, 0, 1,
            1, 0, 0, 1, 1, 0, 1, 0, 1, 1, 0, 1, 1, 0, 0, 0, 1, 0, 1, 0, 1, 0,
            0, 0, 0, 0, 0, 1, 1, 1, 0, 0, 0, 0, 1, 0, 1, 0, 1, 0, 0, 1, 1, 1,
            1, 1, 0, 1, 0, 1, 0, 0, 0, 1, 0, 1, 0, 0, 1, 0, 1, 1, 0, 0, 0,
            1, 1, 0, 1, 1, 0, 0, 0, 1, 0, 1, 1, 0, 1, 0, 1, 1, 0, 0, 1, 1, 1,
            0, 0, 0, 0, 0, 1, 0, 0, 1, 1, 0, 0, 1, 1, 1, 1, 1, 0, 0, 1, 1, 1,
            1, 1, 1, 0, 1, 0, 0, 1, 1, 0, 1, 1, 1, 0, 0, 0, 0, 0, 0, 1, 0, 0,
            1, 1, 1, 1, 1, 1, 0, 1, 0, 0, 0, 1, 1, 0, 1, 1, 0, 0, 0, 1, 1,
     1, 0, 1, 1, 0, 0, 1])
# finding accuracy of the model
accuracy_score(y_pred4,y_test)
     0.7791164658634538
# THE ACCURACIES OF ALL MODELS CREATED import
pandas as pd
accuracy = [['SVC',0.7269076305220884],['SVC Pipeline',0.8152610441767069]
               ,['MultinomialNB',0.7469879518072289],['MultinomialNB Pipeline',0.7791164658634538]]
accuracy_df = pd.DataFrame(accuracy,columns = ['Model','Accuracy']) accuracy_df =
accuracy_df.style.set_properties(**{'text-align':'left'}) accuracy_df =
accuracy_df.set_table_styles([dict(selector = 'th',props=[('text-align','center')])]) accuracy_df
                Mode1
                              Accuracy
      0 SVC
                  0.726908
                            0.815261
      1 SVC Pipeline
```

pickling

```
2 MultinomialNB
                            0.746988
      3 MultinomialNB Pipeline 0.779116
# pickling
# using joblib for saving the model having highest accuracy
import joblib joblib.dump(model2,'restaurant_reviews')
     ['restaurant_reviews']
# loading the model import joblib reload_model =
joblib.load('restaurant_reviews')
# predicting the values with highest accuracy model after pickling
reload_model.predict(['good'])
     array([1])
reload_model.predict(['not so good'])
     array([0])
reload_model.predict(['thankyou'])
     array([1])
reload_model.predict(['not worth'])
     array([0])
# creating the webapp using streamlit
# installing streamlit
!pip install streamlit --quiet
                                             10.1 MB 27.1 MB/s
                                             111 kB 64.3 MB/s
                                             181 kB 66.5 MB/s
                                             77 kB 5.6 MB/s
                                             4.3 MB 68.2 MB/s
                                             164 kB 66.6 MB/s
                                             63 kB 1.7 MB/s
                                             131 kB 60.7 MB/s
                                             130 kB 61.3 MB/s
                                             793 kB 56.2 MB/s
                                             428 kB 60.3 MB/s
                                           | 381 kB 61.2 MB/s
       Building wheel for blinker (setup.py) ... done
       Building wheel for validators (setup.py) ... done
     ERROR: pip's dependency resolver does not currently take into account all the packages that are installed. This behaviour is the source
     jupyter-console 5.2.0 requires prompt-toolkit<2.0.0,>=1.0.0, but you have prompt-toolkit 3.0.29 which is incompatible.
     google-colab 1.0.0 requires ipykernel~=4.10, but you have ipykernel 6.13.0 which is incompatible. google-colab 1.0.0
     requires ipython~=5.5.0, but you have ipython 7.33.0 which is incompatible. google-colab 1.0.0 requires
```

tornado~=5.1.0; python_version >= "3.0", but you have tornado 6.1 which is incompatible.

```
# using file handiling in python
# write mode for writing the reviews
\# button - on click displays the output as 0 or 1 \%\!\! writefile
reviews.py import streamlit as st import joblib from PIL import Image
image = Image.open('/content/photo.jpg') st.image(image,
caption='Artificial Intelligence and Machine learning') print()
st.title('SENTIMENT ANALYSIS') reload_model =
joblib.load('restaurant_reviews')
ip = st.text_input("Give your expensive review : ")
op = reload_model.predict([ip])
if st.button('PREDICT'):
 st.text('OUTPUT ')
st.text('1 - POSITIVE REVIEW')
st.text('0 - NEGATIVE REVIEW')
 st.title(op[0])
     Overwriting reviews.py
\ensuremath{\text{\#}}\xspace using local tunnel tool for creating a URL to we
bapp
!streamlit run reviews.py & npx localtunnel --port 8501
     2022-05-28 12:32:11.460 INFO
                                          numexpr.utils: NumExpr defaulting to 2 threads.
     npx: installed 22 in 2.593s
        You can now view your Streamlit app in your browser.
        Network URL: <a href="http://172.28.0.2:8501">http://172.28.0.2:8501</a>
        External URL: <a href="http://34.85.226.213:8501">http://34.85.226.213:8501</a>
     your url is: <a href="https://fair-turtles-stop-34-85-226-213.loca.lt">https://fair-turtles-stop-34-85-226-213.loca.lt</a>
        Stopping...
     ^C
```



Actificial in belignment and Machine beaming.

SENTIMENT ANALYSIS

Give your expensive reviews:

That's right...the red velvet case....ohih this stuff is so good.

PREDICT

OUTPUT

1 - POISTIVE REVIEW

6 - REGATIVE REVIEW

1