Rating Prediction

Introduction

The problem which I have faced while doing this project is that to scrap the data from <u>Amazon</u> websites and to scrap the laptops reviews and Ratings as well with huge no of 38400 rows data .am face many time only one problem that is finding the Rating.

Data collection

- ► Firstly I have done the data extraction, for extracting I have used the amazon websites through that I have scraped the Rating and Reviews.
- ▶ And I got 38400 rows for Laptops

Analytical Problem Framing

```
#loading the Dataset
df_rating=pd.read_csv("D:/DATA_SCIENCE\I/intern Assignment/Review_rating/Review_Rating_final.csv")
```

1 df_rating.head()

Price is up a bit . In this price we are ge
t is come with pre loaded ms office nd 165 1.0
lelivery was well on time. The product was 5.0
zing product with proper delivery!!Perfect 5.0
ovo Legion 5i is a decent rig for gaming as 4.0

```
#cheking the Datatype
    df rating.dtypes
Unnamed: 0
                        int64
Product Review
                       object
Product Rating
                     float64
dtype: object
     #lets Drop the Unwanted data
     df rating.drop('Unnamed: 0',axis=1,inplace=True)
 1 df rating.head()
                              Product_Review Product_Rating
0
        1. Price is up a bit . In this price we are ge...
                                                          1.0
      If it is come with pre loaded ms office nd 165...
                                                          1.0
2 The delivery was well on time. The product was...
                                                          5.0
    Amazing product with proper delivery!!Perfect ...
                                                          5.0
    Lenovo Legion 5i is a decent rig for gaming as...
                                                          4.0
```

Data Pre-processing

```
1 #cheking the Datatype
  2 df rating.dtypes
Unnamed: 0
                        int64
                       object
Product Review
Product Rating
                     float64
dtype: object
    #lets Drop the Unwanted data
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                              Product_Review Product_Rating
        1. Price is up a bit . In this price we are ge...
                                                          1.0
 0
      If it is come with pre loaded ms office nd 165...
                                                           1.0
 2 The delivery was well on time. The product was...
                                                           5.0
    Amazing product with proper delivery!!Perfect ...
                                                           5.0
    Lenovo Legion 5i is a decent rig for gaming as...
                                                           4.0
```

```
1 stop_words = set(stopwords.words('english') + ['u', 'ü', 'ur', '4', '2', 'im', 'dont', 'doin', 'ure'])
  def Remove Stop Words(df,col):
      df[col]=df[col].apply(lambda x: ' '.join(term for term in x.split() if term not in stop words))
      return df
1 #steaming
2 lemmatizer = WordNetLemmatizer()
  def stemming(col):
      lema=lemmatizer.lemmatize(col,pos='a')
      return lema
1 #Tokenize and Lemmatize
  def preprocess(text):
      result=[]
      for token in text:
          if len(token)>=3:
              result.append(stemming(token))
       text=result
       return text
```

Data Pre-processing

Finally we are calling all the Data Pre-processing function on a single column

```
def data_cleaning(df,col):
    Text_cleaning(df,col)
    Remove_Punctuatuation(df,col)
    Remove_Stop_Words(df,col)
    stemming(col)
    preprocess(col)
    return df

final=data_cleaning(df_rating,'Product_Review')

df_new=final
```

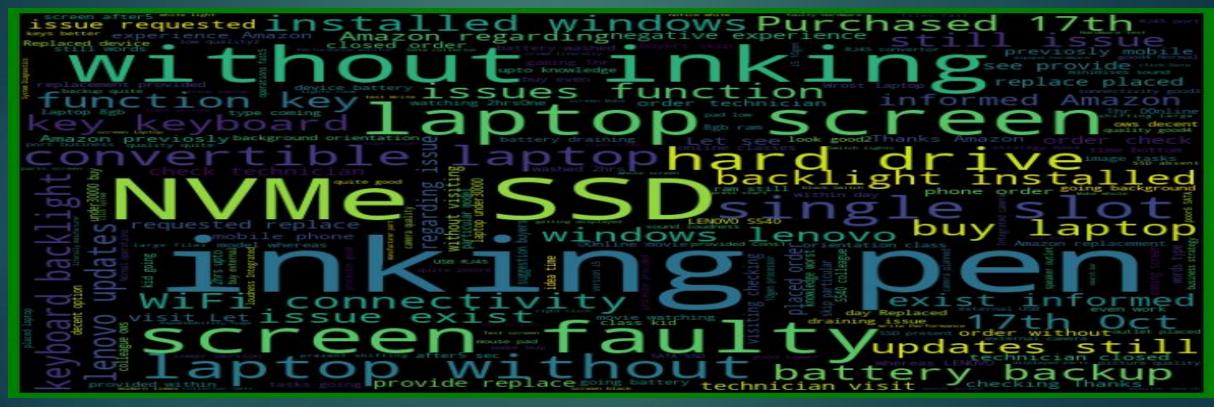
Now We have your cleaned Data

We can see here which words are heavy when its comes to releted to rating "1".



Analysis

▶ We can see here which words are heavy when its comes to releted to rating "2" \subseteq



Analysis

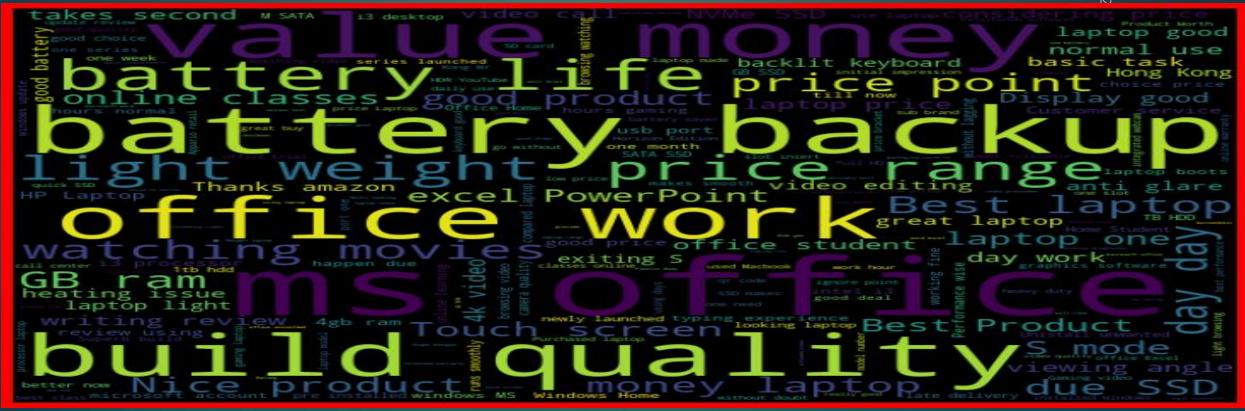


▶ We can see here which words are heavy when its comes to releted to rating "4".



Analysis

We can see here which words are heavy when its comes to releted to rating "5".



Feature Extraction

▶ TFIDF vectorizer.

```
# 1. Convert text into vectors using TF-IDF

from sklearn.feature_extraction.text import TfidfVectorizer

tf_vec = TfidfVectorizer()
features = tf_vec.fit_transform(df_new['Product_Review'])

x = features
y = df_new[['Product_Rating']]
```

Model creation

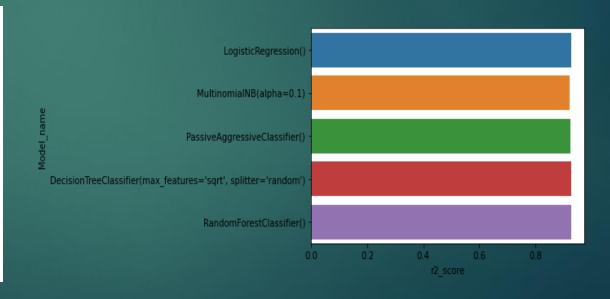
Here We are making a function for model Training.

```
from sklearn.metrics import classification report, confusion matrix, accuracy score
from sklearn.model selection import train test split
from sklearn.metrics import accuracy score, classification report, confusion matrix
from sklearn.metrics import roc auc score, roc curve, auc
final random state=[]
final r2score=[]
model=[]
def max acc(rgr,x,y):
   max acc=0
   model.append(rgr)
   for r in range(42,100):
       x train,x test,y train,y test=train test split(x,y,random state=r,test size=0.20)
       rgr.fit(x train,y train)
       y prd=rgr.predict(x test)
       rc=accuracy score(y test,y prd)
       if rc>max acc:
           max acc=rc
           final r=r
   final random state.append(final r)
   final r2score.append(max acc)
    print("max accuracy score coressponding to **+>", final r, "is*+", max acc*100)
```

Model Selection

▶ After trained our data in different model we got the result

	Model_name	r2_score	Random_State
0	LogisticRegression()	0.927604	73
1	MultinomiaINB(alpha=0.1)	0.922396	43
2	PassiveAggressiveClassifier()	0.925260	81
3	$Decision Tree Classifier (max_features = 'sqrt', sp$	0.927604	73
4	$(Decision Tree Classifier (max_features='auto', r$	0.927604	73



Model finalize

We got the result where Randomforest is working good so am finalize Randomforest classifier as my final model.

```
1 # Based on the Above Score am going to finalize the model as RandomForest Classifie
   RF=RandomForestClassifier()
    x train,x test,y train,y test=train test split(x,y,random state=73,test size=0.20)
    RF.fit(x train,y train)
    RF pred=RF.predict(x test)
    print(accuracy score(y test,RF pred))
    print(confusion matrix(y test,RF pred))
    print(classification report(y test,RF pred))
0.9276041666666667
[[2119
          8
                         38]
    33
        387
               0
                    1
                         16]
                   14
                        11
   49
               8 1451
                         60]
 133
                   38 2575]]
              precision
                            recall f1-score
                                                support
         1.0
                   0.89
                              0.96
                                        0.92
                                                   2205
                   0.94
                              0.89
                                        0.91
                                                    437
         3.0
                              0.87
                                        0.89
                   0.93
                                                    684
         4.0
                   0.95
                              0.93
                                        0.94
                                                   1568
         5.0
                   0.95
                              0.92
                                        0.94
                                                   2786
                                        0.93
                                                   7680
    accuracy
   macro avg
                   0.93
                              0.91
                                        0.92
                                                   7680
weighted avg
                   0.93
                              0.93
                                        0.93
                                                   7680
```

Actual vs Predict

```
#here we can check the predicted values
test=pd.DataFrame(data=y_test,)
test['Predicted values']=RF_pred

test.to_csv('Ratings_Predict.csv')

test
```

	Product_Rating	Predicted values
6379	5.0	5.0
17071	5.0	5.0
3402	4.0	4.0
24338	4.0	4.0
7646	2.0	2.0
4609	5.0	5.0
28224	3.0	3.0
15873	5.0	5.0
35155	5.0	5.0
22012	5.0	2.0

CONCLUSION

- Key Findings and Conclusions of the Study
- The key findings that I have find that I have scraped it from only one websites due to dead line I was able to scrap it .if I could scrap more websites we will get more better model prediction.
- By using 38400 data we for two best models Random Forest Classifier and Decision Tree Classifier. Because of limited data I haven't go for sampling only just used stratify method to balance the data.
- Limitations of this work and Scope for Future Work.
- In some algorithms where was taking to much time to execute but it was executed it in better way, because of that laptops where getting hang and as we accept we got better score in every