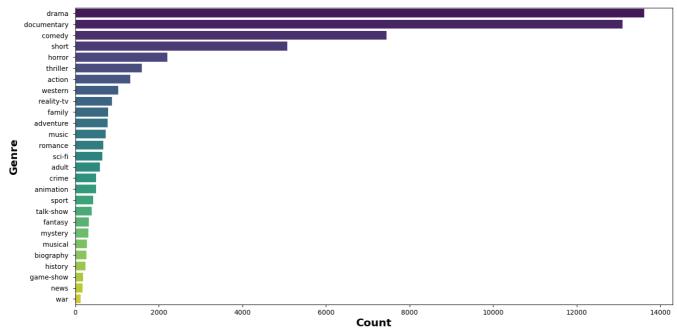
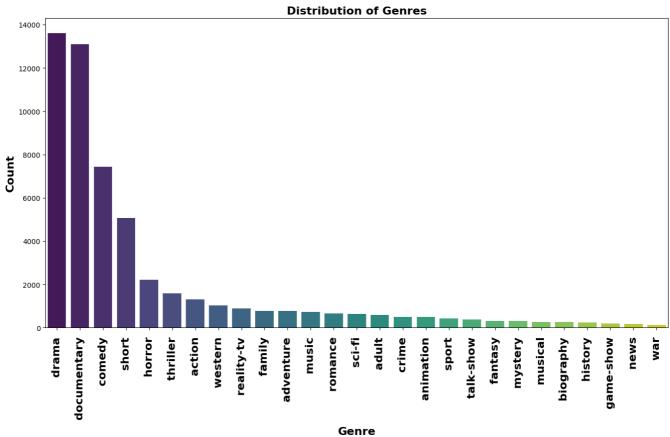
Notebook Release notes

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import re
import nltk
import string
from nltk.corpus import stopwords
from nltk.stem import LancasterStemmer
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.model_selection import train_test_split
from sklearn.naive bayes import MultinomialNB
from \ sklearn.metrics \ import \ accuracy\_score, \ classification\_report
nltk.download('stopwords')
nltk.download('punkt')
     [nltk\_data] \ \ Downloading \ package \ \ stopwords \ \ to \ \ /root/nltk\_data...
                   Package stopwords is already up-to-date!
     [nltk data]
     [nltk_data] Downloading package punkt to /root/nltk_data...
     [nltk_data]
                   Package punkt is already up-to-date!
     True
train_path = "/content/train_data.txt"
train_data = pd.read_csv(train_path, sep=':::', names=['Title', 'Genre', 'Description'], engine='python')
print(train_data.describe())
                                        Title
                                                 Genre
     count
                                        54214
                                                 54214
     unique
                                        54214
                                                    27
     top
              Oscar et la dame rose (2009)
                                                drama
     freq
                                                 13613
                                                     Description
                                                            54214
     count
                                                            54086
     unique
     top
              Grammy - music award of the American academy
     frea
print(train_data.info())
     <class 'pandas.core.frame.DataFrame'>
     Int64Index: 54214 entries, 1 to 54214
     Data columns (total 3 columns):
     # Column
                       Non-Null Count Dtype
     ---
          -----
                        -----
      9
         Title
                       54214 non-null object
          Genre
                        54214 non-null object
      2 Description 54214 non-null object
     dtypes: object(3)
     memory usage: 1.7+ MB
print(train_data.isnull().sum())
     Title
     Genre
                     0
     Description
                     a
     dtype: int64
test_path = "/content/test_data.txt"
test_data = pd.read_csv(test_path, sep=':::', names=['Id', 'Title', 'Description'], engine='python')
test_data.head()
                               Title
                                                                    Description
                                                                                    Ħ
         Ιd
      0 1
                  Edgar's Lunch (1998)
                                           L.R. Brane loves his life - his car, his apar...
               La guerra de papá (1977) Spain, March 1964: Quico is a very naughty ch...
      1
         2
      2
         3 Off the Beaten Track (2010)
                                           One year in the life of Albin and his family ...
      3
         4
               Meu Amigo Hindu (2015)
                                        His father has died, he hasn't spoken with hi...
      4
         5
                      Er nu zhai (1955) Before he was known internationally as a mart...
```

```
# Plot the distribution of genres in the training data
plt.figure(figsize=(16, 8))
sns.countplot(data=train_data, y='Genre', order=train_data['Genre'].value_counts().index, palette='viridis')
plt.xlabel('Count', fontsize=16, fontweight='bold')
plt.ylabel('Genre', fontsize=16, fontweight='bold')

# Plot the distribution of genres using a bar plot
plt.figure(figsize=(16, 8))
counts = train_data['Genre'].value_counts()
sns.barplot(x=counts.index, y=counts, palette='viridis')
plt.xlabel('Genre', fontsize=16, fontweight='bold')
plt.ylabel('Count', fontsize=16, fontweight='bold')
plt.title('Distribution of Genres', fontsize=16, fontweight='bold')
plt.xticks(rotation=90, fontsize=16, fontweight='bold')
plt.show()
```





```
# Initialize the stemmer and stop words
stemmer = LancasterStemmer()
stop_words = set(stopwords.words('english'))
# Define the clean_text function
def clean text(text):
    text = text.lower() # Lowercase all characters
    \label{eq:text} \texttt{text} = \texttt{re.sub}(\texttt{r'@\S+'}, \texttt{''}, \texttt{text}) \quad \text{\# Remove Twitter handles}
    text = re.sub(r'http\S+', '', text) # Remove URLs
text = re.sub(r'pic.\S+', '', text)
    text = re.sub(r"[^a-zA-Z+']", ' ', text) # Keep only characters
text = re.sub(r'\s+[a-zA-Z]\s+', ' ', text + ' ') # Keep words with length > 1 only
    text = "".join([i for i in text if i not in string.punctuation])
    words = nltk.word_tokenize(text)
    stopwords = nltk.corpus.stopwords.words('english') # Remove stopwords
    \texttt{text} = \texttt{"".join}([\texttt{i for i in words if i not in stopwords and len(i)} \ \gt 2])
    text = re.sub("\s[\s]+", " ", text).strip() # Remove repeated/leading/trailing spaces
    return text
# Apply the clean_text function to the 'Description' column in the training and test data
train_data['Text_cleaning'] = train_data['Description'].apply(clean_text)
test_data['Text_cleaning'] = test_data['Description'].apply(clean_text)
# Calculate the length of cleaned text
train_data['length_Text_cleaning'] = train_data['Text_cleaning'].apply(len)
# Visualize the distribution of text lengths
plt.figure(figsize=(8, 7))
sns.histplot(data=train_data, x='length_Text_cleaning', bins=20, kde=True, color='blue')
plt.xlabel('Length', fontsize=14, fontweight='bold')
plt.ylabel('Frequency', fontsize=14, fontweight='bold')
plt.title('Distribution of Lengths', fontsize=16, fontweight='bold')
plt.show()
# Initialize the TF-IDF vectorizer
tfidf vectorizer = TfidfVectorizer()
# Fit and transform the training data
X_train = tfidf_vectorizer.fit_transform(train_data['Text_cleaning'])
# Transform the test data
X_test = tfidf_vectorizer.transform(test_data['Text_cleaning'])
# Split the data into training and validation sets
X = X train
y = train_data['Genre']
X_train, X_val, y_train, y_val = train_test_split(X, y, test_size=0.2, random_state=42)
# Initialize and train a Multinomial Naive Bayes classifier
classifier = MultinomialNB()
classifier.fit(X_train, y_train)
# Make predictions on the validation set
y_pred = classifier.predict(X_val)
# Evaluate the performance of the model
accuracy = accuracy_score(y_val, y_pred)
print("Validation Accuracy:", accuracy)
print(classification_report(y_val, y_pred))
     Validation Accuracy: 0.44526422576777647
                     precision recall f1-score
                                                       support
            action
                           9.99
                                     9.99
                                                9.99
                                                            263
             adult
                           0.00
                                     0.00
                                                 0.00
                                                            112
         adventure
                           0.00
                                      0.00
                                                 0.00
                                                            139
         animation
                           0.00
                                      0.00
                                                0.00
                                                            104
        biography
                           0.00
                                      0.00
                                                 0.00
                                                           1443
            comedy
                           0.61
                                      0.04
            crime
                           0.00
                                      0.00
                                                 0.00
                                                            107
                           0.54
                                      0.90
                                                0.67
                                                           2659
      documentary
                                                           2697
                           0.38
                                      0.88
            drama
                                                 0.53
            family
                           0.00
                                     0.00
                                                0.00
                                                            150
          fantasy
                           0.00
                                     9.99
                                                9.99
                                                             74
        game-show
                           0.00
                                      0.00
                                                 0.00
                                                             40
          history
                           0.00
                                      0.00
                                                0.00
                                                             45
            horror
                           0.00
                                      0.00
                                                0.00
                                                            431
             music
                           0.00
                                      0.00
                                                 0.00
                                                            144
                           0.00
                                      0.00
          musical
                                                 0.00
                                                             50
          mystery
                           0.00
                                      0.00
                                                 0.00
                                                             56
                           0.00
                                      0.00
                                                 0.00
             news
                           0.00
                                     0.00
                                                0.00
                                                            192
       reality-ty
```

```
0.00
                               0.00
                                                    151
                                         0.00
    romance
                    0.00
                              0.00
                                         0.00
                                                    143
     sci-fi
      short
                    0.50
                              0.00
                                         0.00
                                                    1045
      sport
                    0.00
                               0.00
                                         0.00
                                                      93
  talk-show
                    0.00
                              0.00
                                         0.00
                                                      81
   thriller
                    0.00
                               0.00
                                         0.00
                                                     309
                    0.00
                              0.00
                                         0.00
                                                      20
        war
    western
                    0.00
                              0.00
                                         0.00
                                                     200
                                         0.45
                                                   10843
    accuracy
                    0.08
                               0.07
                                                   10843
   macro avg
                                         0.05
weighted avg
                                                  10843
                    0.36
                              0.45
                                         0.31
```

/usr/local/lib/python3.10/dist-packages/sklearn/metrics/\_classification.py:1344: UndefinedMetricWarning: Precision and F-score are \_warn\_prf(average, modifier, msg\_start, len(result))

/usr/local/lib/python3.10/dist-packages/sklearn/metrics/\_classification.py:1344: UndefinedMetricWarning: Precision and F-score are \_warn\_prf(average, modifier, msg\_start, len(result))

/usr/local/lib/python3.10/dist-packages/sklearn/metrics/\_classification.py:1344: UndefinedMetricWarning: Precision and F-score are \_\_warn\_prf(average, modifier, msg\_start, len(result))

# Use the trained model to make predictions on the test data
X\_test\_predictions = classifier.predict(X\_test)
test\_data['Predicted\_Genre'] = X\_test\_predictions

# Save the test\_data DataFrame with predicted genres to a CSV file test\_data.to\_csv('predicted\_genres.csv', index=False)

# Display the 'test\_data' DataFrame with predicted genres
print(test data)

```
Ιd
                                         Title \
a
          1
                         Edgar's Lunch (1998)
1
           2
                     La guerra de papá (1977)
                  Off the Beaten Track (2010)
2
           3
                      Meu Amigo Hindu (2015)
3
           4
4
                            Er nu zhai (1955)
           5
54195 54196
               "Tales of Light & Dark" (2013)
                  Der letzte Mohikaner (1965)
54196
       54197
54197
                          Oliver Twink (2007)
       54198
54198
       54199
                            Slipstream (1973)
54199 54200
                    Curitiba Zero Grau (2010)
```

Description \ 0 L.R. Brane loves his life - his car, his apar... Spain, March 1964: Quico is a very naughty ch... 1 2 One year in the life of Albin and his family ... 3 His father has died, he hasn't spoken with hi... Before he was known internationally as a mart... 4 54195 Covering multiple genres, Tales of Light & Da... 54196 As Alice and Cora Munro attempt to find their... 54197 A movie 169 years in the making. Oliver Twist... 54198 Popular, but mysterious rock D.J Mike Mallard... 54199 Curitiba is a city in movement, with rhythms ...

Text\_cleaning Predicted\_Genre drama 0 brane loves life car apartment job especially ... spain march quico naughty child three belongin... drama 1 one year life albin family shepherds north tra... 2 documentary father died hasnt spoken brother years serious... 3 drama 4 known internationally martial arts superstar b...  $\,$ drama 54195 covering multiple genres tales light dark anth... drama 54196 alice cora munro attempt find father british o... drama 54197 movie years making oliver twist artful dodger  $\dots$ drama 54198 popular mysterious rock mike mallard askew bro... drama curitiba city movement rhythms different pulsa... documentary

[54200 rows x 5 columns]