```
import numpy as np # NumPy: Numerical computing library for arrays and matrices.
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
# Import necessary modules
from sklearn.preprocessing import OneHotEncoder
from sklearn.model_selection import train_test_split
from sklearn.naive_bayes import MultinomialNB
from sklearn.metrics import accuracy_score, classification_report
# Import the necessary module
from sklearn.feature_extraction.text import TfidfVectorizer
# Load the training data
train_path = "train_data.txt"
train_data = pd.read_csv(train_path, sep=':::', names=['Title', 'Genre', 'Description'], engine='python')
print(train data.describe())
print(train data.info())
print(train_data.isnull().sum())
# Load the test data
test_path = "test_data.txt"
test_data = pd.read_csv(test_path, sep=':::', names=['Id', 'Title', 'Description'], engine='python')
test_data.head()
#Count each genre value
train_data.Genre.value_counts()
# Plot count plot
plt.figure(figsize=(12,8))
counts = train_data.Genre.value_counts()
sns.barplot(x=counts.index, y=counts, color='blue')
plt.xlabel('Genre' ,fontsize=14, fontweight='bold')
plt.ylabel('Count', fontsize=14, fontweight='bold')
plt.title('Distribution of Genres', fontsize=16, fontweight='bold')
plt.xticks(rotation=90, fontsize=14, fontweight='bold');
# Plot the distribution of genres in the training data
plt.figure(figsize=(14, 7))
sns.countplot(data=train_data, y='Genre', order=train_data['Genre'].value_counts().index, palette='viridis')
plt.xlabel('Count', fontsize=14, fontweight='bold')
plt.ylabel('Genre', fontsize=14, fontweight='bold')
# Plot the distribution of genres using a bar plot
plt.figure(figsize=(14, 7))
counts = train_data['Genre'].value_counts()
\verb|sns.barplot(x=counts.index, y=counts, palette='viridis')|\\
plt.xlabel('Genre', fontsize=14, fontweight='bold')
plt.ylabel('Count', fontsize=14, fontweight='bold')
plt.title('Distribution of Genres', fontsize=16, fontweight='bold')
plt.xticks(rotation=90, fontsize=14, fontweight='bold')
plt.show()
train_data.info()
#Finda any null value
train_data.isnull().sum()
# 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
    text = text.lower()  # Lowercase data considerate

text = re.sub(r'@\S+', '', 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
    text = "".join([i for i in words if i not in stopwords and len(i) > 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)
# Droping the redundant data
print("shape before drop nulls",train_data.shape)
train_data = train_data.drop_duplicates()
print("shape after drop nulls",train_data.shape)
# 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()
# Set up the figure with two subplots
plt.figure(figsize=(12, 6))
# Subplot 1: Original text length distribution
plt.subplot(1, 2, 1)
original_lengths = train_data['Description'].apply(len)
plt.hist(original lengths, bins=range(0, max(original lengths) + 100, 100), color='blue', alpha=0.7)
plt.title('Original Text Length')
plt.xlabel('Text Length')
plt.ylabel('Frequency')
# Subplot 2: Cleaned text length distribution
plt.subplot(1, 2, 2)
cleaned_lengths = train_data['Text_cleaning'].apply(len)
plt.hist(cleaned_lengths, bins=range(0, max(cleaned_lengths) + 100, 100), color='green', alpha=0.7)
plt.title('Cleaned Text Length')
plt.xlabel('Text Length')
plt.ylabel('Frequency')
# Adjust layout and display the plots
plt.tight_layout()
plt.show()
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'])
X = X \text{ 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)
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))
X_test_predictions = classifier.predict(X_test)
test_data['Predicted_Genre'] = X_test_predictions
test_data.to_csv('predicted_genres.csv', index=False)
# Display the 'test_data' DataFrame with predicted genres
print(test data)
```

```
Title
                                           Genre
count
                                  54214
                                           54214
unique
                                  54214
                                              27
top
         Oscar et la dame rose (2009)
                                          drama
                                           13613
freq
                                               Description
count
                                                      54214
                                                      54086
unique
top
         Grammy - music award of the American academy \dots
freq
                                                        12
<class 'pandas.core.frame.DataFrame'>
Index: 54214 entries, 1 to 54214
Data columns (total 3 columns):
#
    Column
                  Non-Null Count
                                   Dtype
0
     Title
                  54214 non-null
                                   object
1
                  54214 non-null object
    Genre
2
    Description 54214 non-null object
dtypes: object(3)
memory usage: 1.7+ MB
None
Title
               0
Genre
               0
Description
dtype: int64
<ipython-input-50-3206035593e8>:46: FutureWarning:
```

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue` and set `le

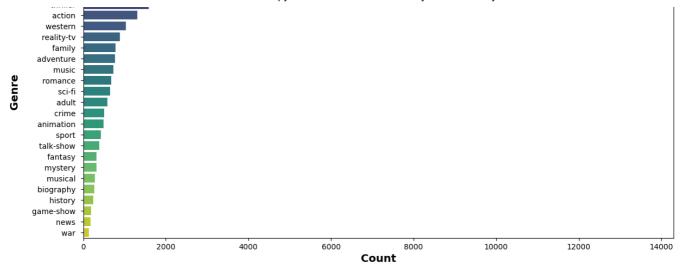
 $sns.countplot(data=train_data, y='Genre', order=train_data['Genre'].value_counts().index, palette='viridis') < ipython-input-50-3206035593e8>:53: FutureWarning:$

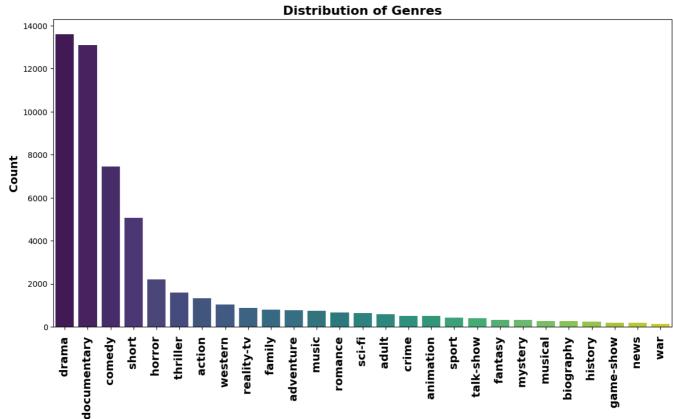
Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set `le sns.barplot(x=counts.index, y=counts, palette='viridis')

Distribution of Genres 14000 12000 10000 8000 Count 6000 4000 2000 short sci-fi sport romance adult news drama horror thriller action western reality-tv family adventure music crime animation talk-show history war documentary comedy fantasy mystery game-show musical biography

Genre

drama documentary comedy short horror thriller





Genre

<class 'pandas.core.frame.DataFrame'>
Index: 54214 entries, 1 to 54214
Data columns (total 3 columns):
Column Non-Null Count Dtype
-----0 Title 54214 non-null object
1 Genre 54214 non-null object

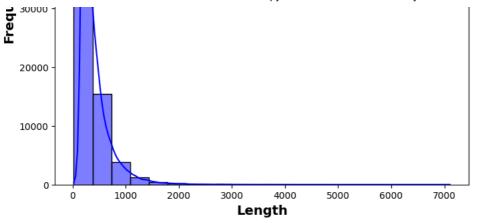
54214 non-null

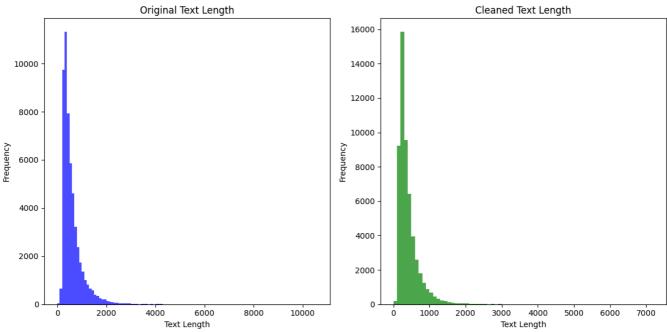
object

2 Description 5421 dtypes: object(3) memory usage: 1.7+ MB

shape before drop nulls (54214, 4) shape after drop nulls (54214, 4)

50000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 400000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 4000000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 40000 - 400





Validation Accuracy: 0.44526422576777647

/usr/local/lib/python3.10/dist-packages/sklearn/metrics/_classification.py:1344: UndefinedMetricWarning: Precision and F-score are i _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 i _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 i _warn_prf(average, modifier, msg_start, len(result))

_warn_prr(average, moul	rrier, msg_	_Start, Tem	(Lesurr))	
	precision	recall	f1-score	support	
action	0.00	0.00	0.00	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	61	
comedy	0.61	0.04	0.07	1443	
crime	0.00	0.00	0.00	107	
documentary	0.54	0.90	0.67	2659	
drama	0.38	0.88	0.53	2697	
family	0.00	0.00	0.00	150	
fantasy	0.00	0.00	0.00	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	
musical	0.00	0.00	0.00	50	
mystery	0.00	0.00	0.00	56	
news	0.00	0.00	0.00	34	
reality-tv	0.00	0.00	0.00	192	
romance	0.00	0.00	0.00	151	
sci-fi	0.00	0.00	0.00	143	
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	
war	0.00	0.00	0.00	20	
western	0.00	0.00	0.00	200	
accuracy			0.45	10843	
macro av	-	0.07	0.05	10843	
weighted av	9 0.36	0.45	0.31	10843	
Id			Title	e /	
0 1		Edgar's Lunch (1998)			

```
2
                     La guerra de papá (1977)
           3
                  Off the Beaten Track (2010)
2
3
           4
                       Meu Amigo Hindu (2015)
4
           5
                            Er nu zhai (1955)
54195
       54196
               "Tales of Light & Dark" (2013)
       54197
54196
                  Der letzte Mohikaner (1965)
                          Oliver Twink (2007)
54197
       54198
54198
       54199
                            Slipstream (1973)
54199
       54200
                    Curitiba Zero Grau (2010)
                                             Description \
        L.R. Brane loves his life - his car, his apar...
0
1
        Spain, March 1964: Quico is a very naughty ch...
2
        One year in the life of Albin and his family \dots
3
        His father has died, he hasn't spoken with hi...
        Before he was known internationally as a mart...
4
        Covering multiple genres, Tales of Light & Da...
54195
54196
        As Alice and Cora Munro attempt to find their...
        A movie 169 years in the making. Oliver Twist...
54197
54198
        Popular, but mysterious rock D.J Mike Mallard...
54199
        Curitiba is a city in movement, with rhythms \dots
                                           Text_cleaning Predicted_Genre
0
       brane loves life car apartment job especially ...
                                                                   drama
1
       spain march quico naughty child three belongin...
                                                                   drama
2
       one year life albin family shepherds north tra...
                                                            documentary
       father died hasnt spoken brother years serious...
3
                                                                  drama
       known internationally martial arts superstar b...
4
                                                                  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 ...
                                                                  drama
54198
       popular mysterious rock mike mallard askew bro...
                                                                  drama
54199
       curitiba city movement rhythms different pulsa...
                                                            documentary
```

[54200 rows x 5 columns]