

## TASK-1 (MOVIE GENRE CLASSIFICATION)

### PROGRAM:

```
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score, classification_report,
confusion_matrix

# Step 1: Load the dataset from the local path
file_path = r"/content/train_data.txt"
data = pd.read_csv(file_path, sep=':::', engine='python',
header=None) # Adjusted for your specific file format

# Inspect the data to understand its structure
print(data.head())
print(data.columns)

# Step 2: Data Preprocessing
# Manually specify the columns if they are not automatically parsed
correctly
data.columns = ['Index', 'Title', 'Genre', 'Plot'] # Example
columns based on the data sample you provided

# Now, let's focus on the 'Plot' for text and 'Genre' for labels
X_text = data['Plot'] # Features: movie plot summaries
y = data['Genre'] # Labels: movie genres

# Split the dataset into training and test sets
X_train, X_test, y_train, y_test = train_test_split(X_text, y,
test_size=0.2, random_state=42)

# Step 3: Convert text data to numerical features using TF-IDF
```

```

tfidf = TfidfVectorizer(stop_words='english', max_df=0.7)
X_train_tfidf = tfidf.fit_transform(X_train)
X_test_tfidf = tfidf.transform(X_test)

# Step 4: Train the model (Logistic Regression)
model = LogisticRegression(max_iter=1000)
model.fit(X_train_tfidf, y_train)

# Step 5: Make predictions and evaluate the model
y_pred = model.predict(X_test_tfidf)

```

OUTPUT:

Confusion Matrix:

```

[[ 53  0  0  0  0 23  1 38 111  0  0  0  0  9
   0  0  0  0  0  0  4  9  4  0 10  0  1]
 [ 0 20  9  0  0 37  0  5 30  0  0  0  0  2
   0  0  0  0  0  0  0  8  0  0  0  0  1]
 [ 3  1 14  0  0 17  0 31 48  1  0  0  0  9
   0  0  0  0  1  0  2  8  0  0  2  0  2]
 [ 1  0  0  3  0 27  0 19 27  4  0  0  0  2
   1  0  0  0  0  0  6 14  0  0  0  0  0]
 [ 0  0  0  0  0  2  0 38 19  0  0  0  0  0
   0  0  0  0  0  0  0  2  0  0  0  0  0]
 [ 3  0  0  0  0 846  0 87 447  2  0  0  0 11
   2  0  0  0  3  0  1 36  0  0  5  0  0]
 [ 6  0  0  0  0 15  1 14 59  0  0  0  0  3
   0  0  0  0  0  0  0  2  0  0  6  0  1]
 [ 3  0  0  0  0 49  0 2300 220  0  0  0  0  7
  11  0  0  0  3  0  1 64  0  0  1  0  0]
 [ 7  0  0  0  0 194  0 228 2192  0  0  0  0 11
   0  0  0  0  1  0  0 55  1  0  7  0  1]
 [ 0  0  3  0  0 37  0 34 49  8  0  0  0  1
   4  0  0  0  2  0  0 11  0  1  0  0  0]
 [ 0  1  4  1  0 12  0 11 30  0  0  0  0  6
   0  0  0  0  0  0  2  7  0  0  0  0  0]
 [ 0  0  0  0  0 15  0  9  1  0  0 12  0  0
   0  0  0  0  2  0  0  0  1  0  0  0  0]
 [ 0  0  0  0  0  0  0 27 18  0  0  0  0  0

```

```

    0 0 0 0 0 0 0 0 0 0 0 0 0 0]
[ 3 0 0 0 0 27 0 29 100 1 0 0 0 245
  0 0 0 0 1 0 2 15 0 0 8 0 0]
[ 0 0 0 0 0 11 0 61 9 0 0 0 0 0
 56 0 0 0 1 0 0 6 0 0 0 0 0]
[ 0 0 0 0 0 9 0 7 26 0 0 0 0 0
 3 0 0 0 0 0 0 4 0 0 0 0 1]
[ 0 0 0 0 0 7 1 5 29 0 0 0 0 5
 0 0 0 0 0 0 0 1 0 0 8 0 0]
[ 0 0 0 0 0 3 0 30 0 0 0 0 0 0
 0 0 0 0 0 0 0 0 0 1 0 0 0]
[ 0 0 0 0 0 51 0 98 15 1 0 0 0 0
 1 0 0 0 22 0 0 4 0 0 0 0 0]
[ 0 1 0 0 0 26 0 1 119 0 0 0 0 0
 0 0 0 0 0 0 0 4 0 0 0 0 0]
[ 2 0 0 0 0 14 0 37 42 0 0 0 0 10
 0 0 0 0 1 0 24 10 0 0 3 0 0]
[ 2 0 0 0 0 117 0 299 300 0 0 0 0 7
 1 0 0 0 1 0 0 317 0 0 1 0 0]
[ 2 0 0 0 0 7 0 59 3 0 0 0 0 0
 0 0 0 0 0 0 0 6 15 1 0 0 0]
[ 0 0 0 0 0 10 0 48 3 0 0 0 0 0
 2 0 0 0 4 0 0 7 0 7 0 0 0]
[ 8 0 0 0 0 28 0 19 171 0 0 1 0 32
 0 0 0 0 0 0 0 15 0 0 35 0 0]
[ 1 0 0 0 0 0 0 6 10 0 0 0 0 0
 0 0 0 0 1 0 0 2 0 0 0 0 0]
[ 1 0 0 0 0 14 0 4 47 0 0 0 0 0
 0 0 0 0 0 0 0 0 0 0 1 0 133]]

```