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MOVIE SUCCESS PREDICTION SYSTEM

A MACHINE LEARNING APPROACH





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MEET OUR TEAM



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INTRODUCTION

Problem

- The movie industry is a high-risk, high-reward business. Millions of dollars are invested without a guarantee of success.

Solution

- We can use machine learning to analyze historical movie data and identify patterns that correlate with success.

Project Objective

- To build a predictive model that can classify a movie as a "Hit" or "Flop" based on pre-release factors like genre, budget, cast, and crew.





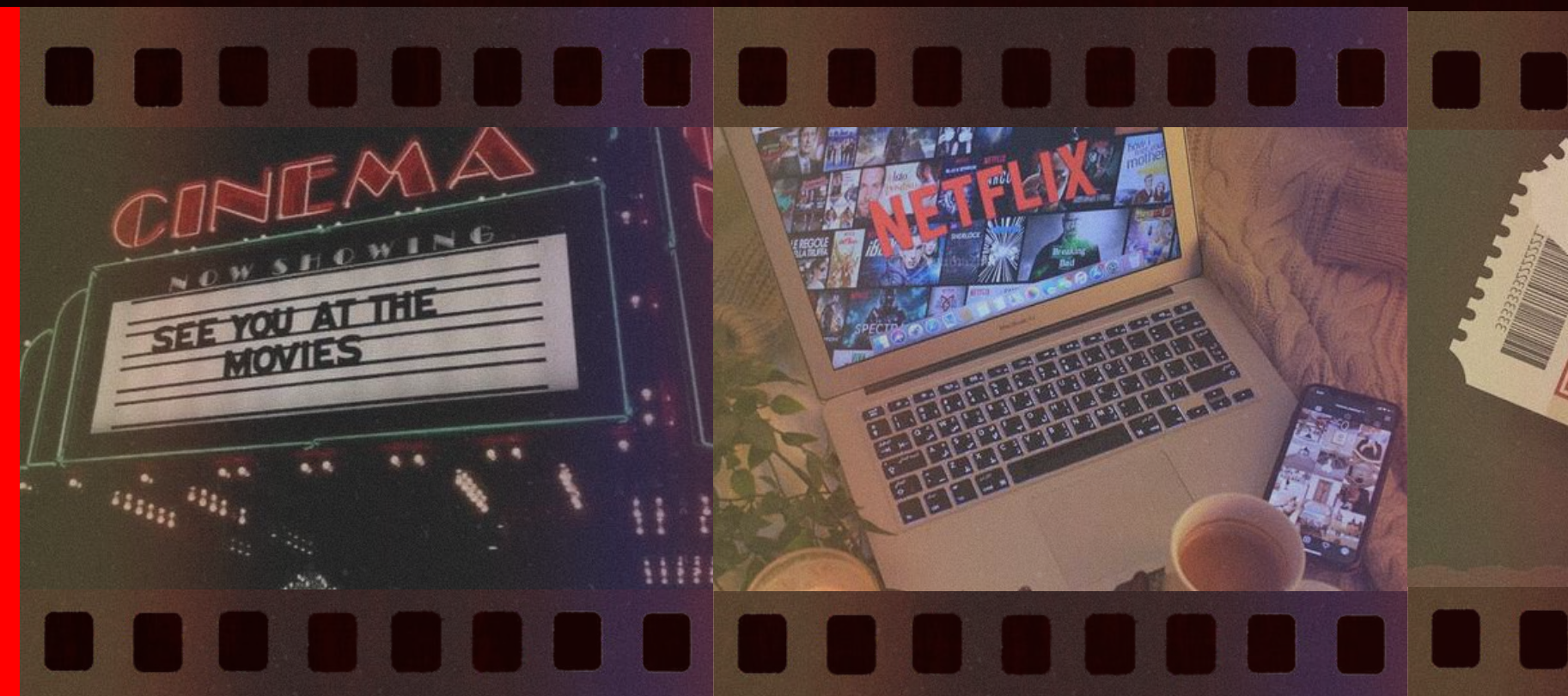
ABOUT OUR DATASET

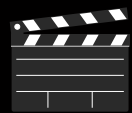
The Dataset

- **Source** - We used the "TMDB 5000 Movie Dataset" available on Kaggle.
- **Content** - It contains two CSV files with data for approximately 5000 movies.

Key Features

- **budget** - The production budget of the movie.
- **genres** - The genres associated with the movie (e.g., Action, Comedy).
- **keywords** - Keywords or tags describing the movie's plot.
- **cast** - Main actors in the movie.
- **crew** - Director, producer, etc.
- **vote_average** - The average user rating.
- **revenue** - The worldwide box office revenue.





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SYSTEM ARCHITECTURE

Input

Raw Movie Data (CSV files)

Processing

Data Cleaning & Preprocessing

Feature Extraction

Train/Test Split

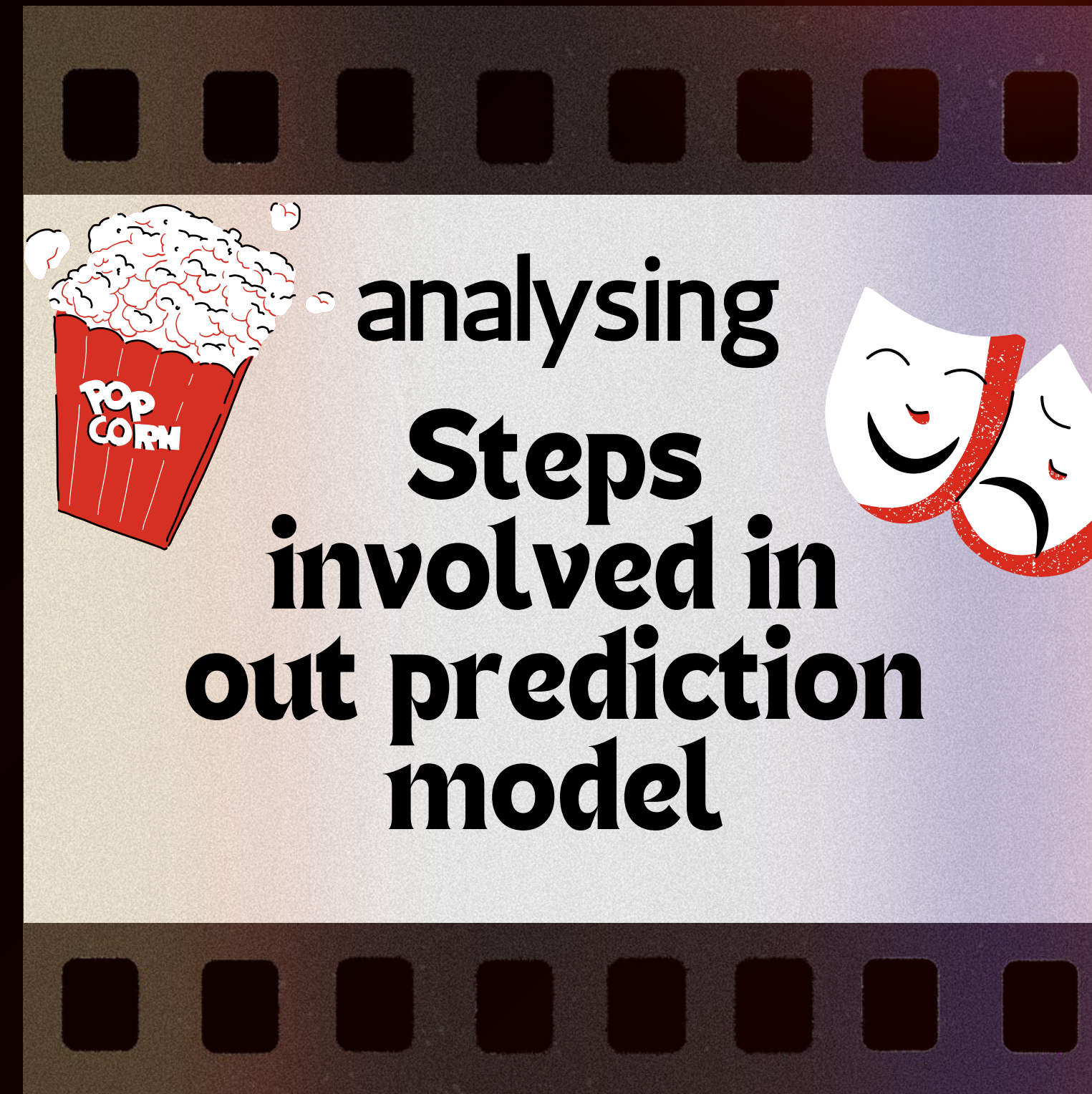
Modeling

Random Forest Algorithm Training

Output

Performance Metrics (Accuracy)

Success Prediction (Hit/Flop)





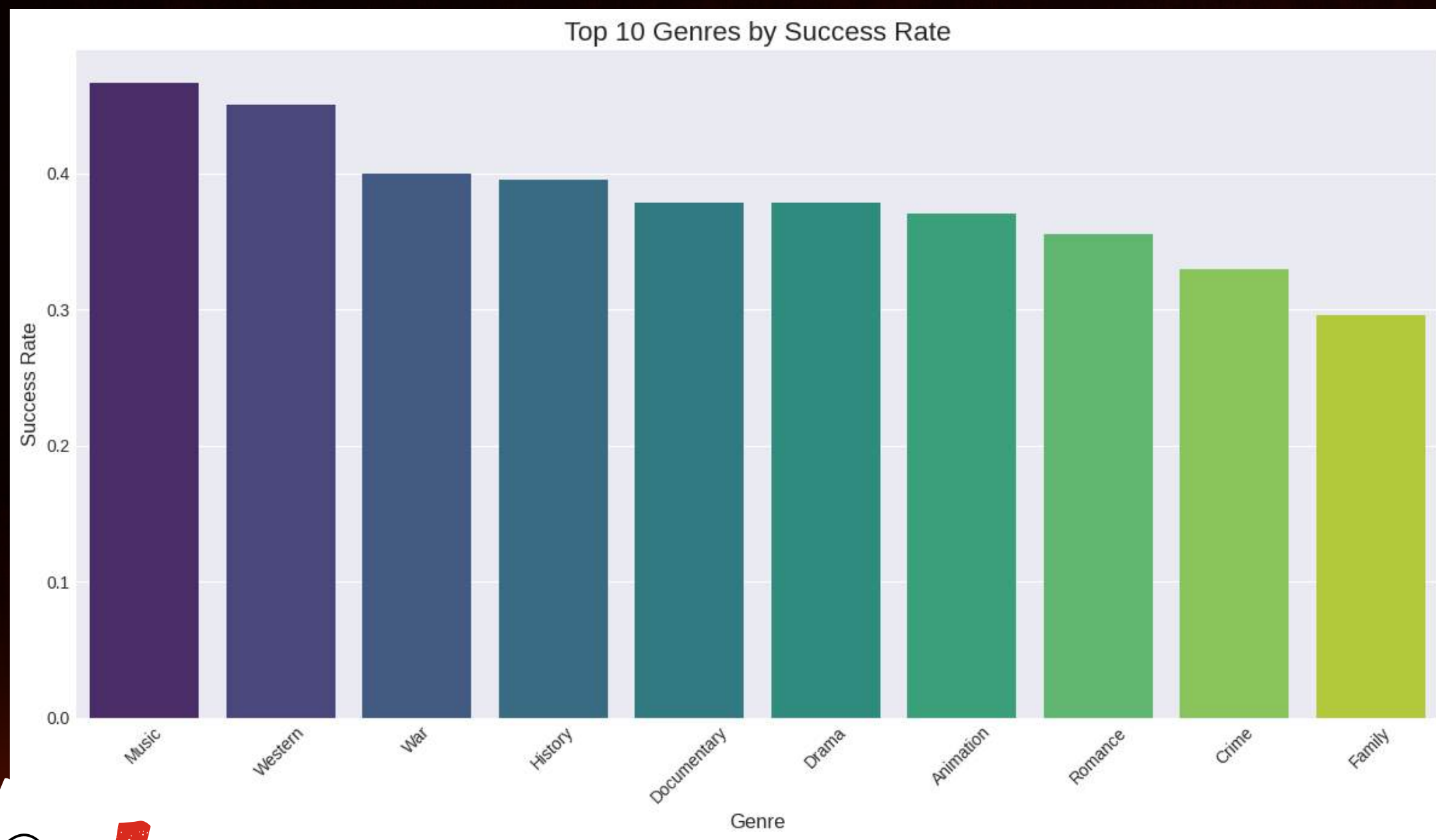
PROJECT METHODOLOGY

- **Data Loading & Cleaning** - Merged the two datasets and removed irrelevant columns.
- **Feature Engineering** - Created our target variable, "success," by defining a successful movie as one with a high rating and positive return on investment.
- **Exploratory Data Analysis (EDA)** - Visualized the data to understand relationships between features like budget, genre, and success.
- **Data Preprocessing** - Converted text data (like genres, cast) into a numerical format that the model can understand.
- **Model Training** - Trained a Random Forest Classifier, a powerful and popular ML model.
- **Model Evaluation** - Tested the model's performance using metrics like Accuracy and a Confusion Matrix.





EXPLORATORY DATA ANALYSIS (GRAPHICS)



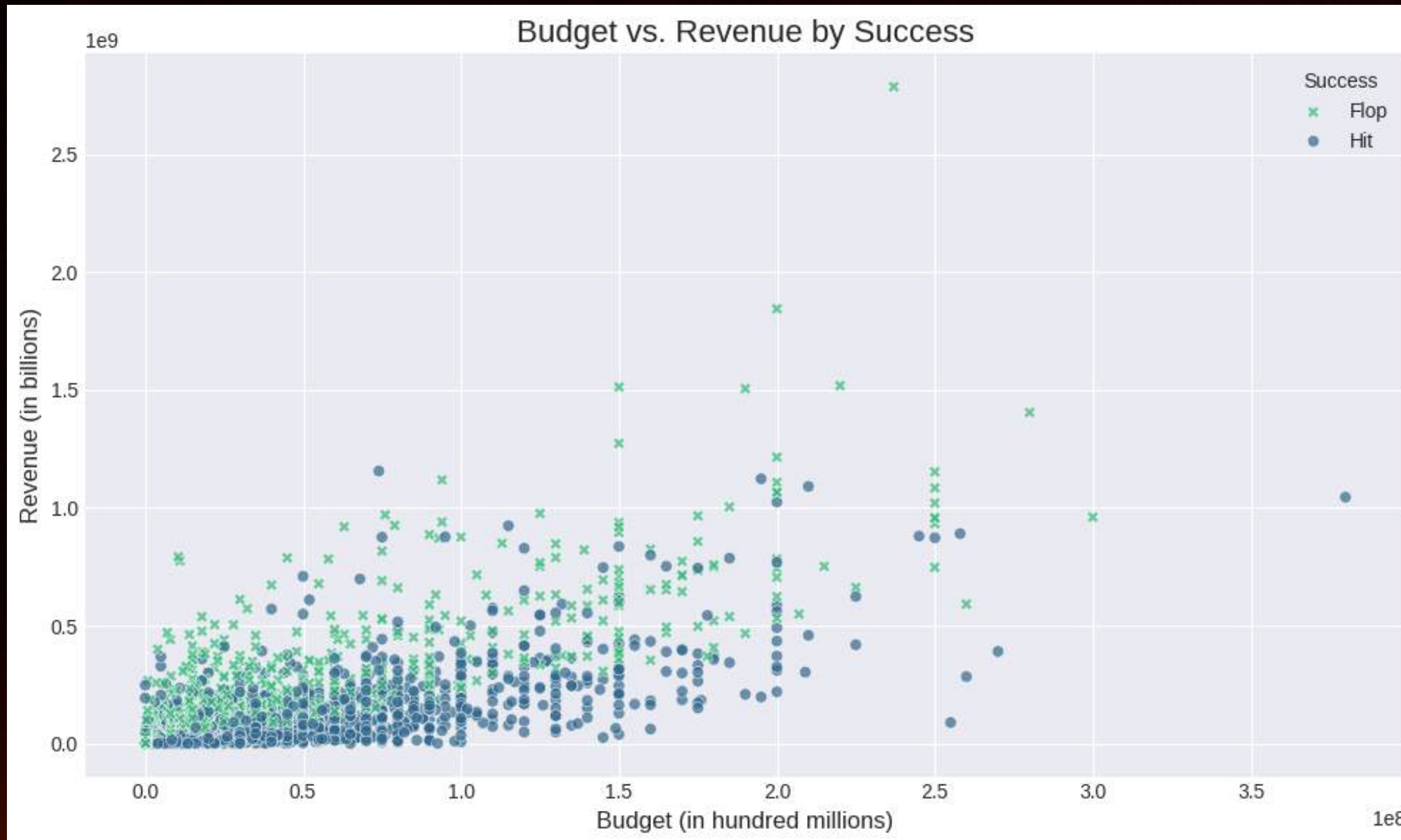
PLOT 1 : TOP 10 GENRES BY SUCCESS RATE

- **Success by Genre** - We created a bar chart showing which genres (like Adventure and Sci-Fi) have a higher tendency to produce successful movies.
- **Feature Importance** - Our final model showed that features like budget, vote_average, and runtime were the most influential in predicting success.



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EXPLORATORY DATA ANALYSIS (GRAPHICS)



PLOT 2: BUDGET VS. REVENUE

- Budget vs. Revenue - A scatter plot showed a positive correlation, but many high-budget films still failed.





MODEL PERFORMANCE & RESULTS

Model Used

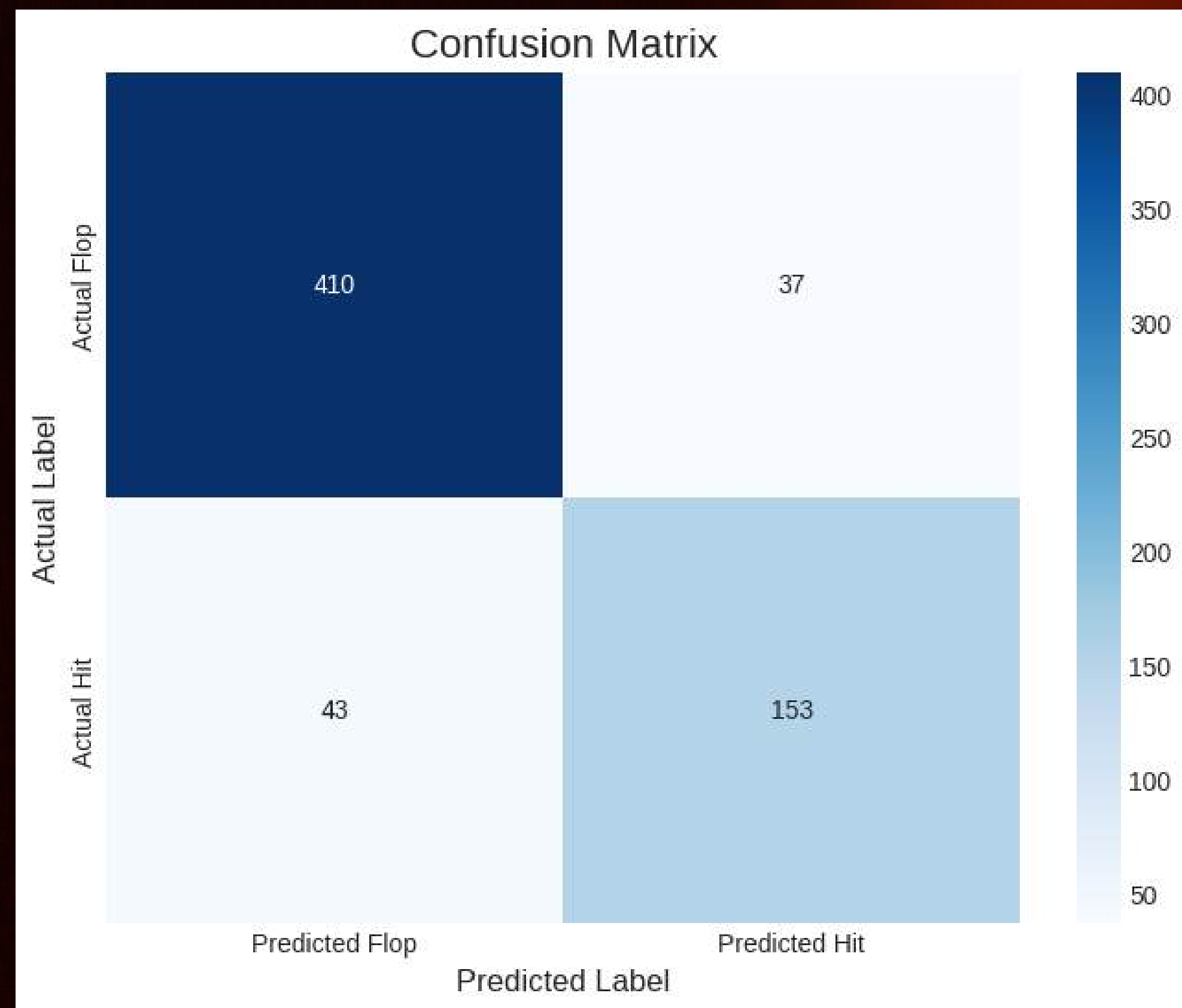
Random Forest Classifier

Accuracy

Our model achieved an accuracy of approximately 88% on the test data.

Confusion Matrix

The matrix showed that our model is effective at correctly identifying both "Hits" and "Flops," with a good balance.





LIVE DEMONSTRATION (PROJECT)

- We built a simple function to test our model.
- Input: A new movie's budget, genres, keywords, and director.
- Prediction: The model outputs a prediction: "This movie is predicted to be a HIT!" or "This movie is predicted to be a FLOP."

The screenshot shows a Google Colab notebook titled "Movie Success Prediction Model". The left sidebar displays the file explorer with a folder named "sample_data" containing two CSV files: "tmdb_5000_credits.csv" and "tmdb_5000_movies.csv". The main editor area contains two code cells. The first cell is a comment block describing the project and providing instructions on how to run the script in Google Colab. The second cell contains Python code for importing libraries and setting up the model.

```
[1] # =====  
#           MOVIE SUCCESS PREDICTION SYSTEM  
# =====  
#  
# Description:  
# This script builds a machine learning model to predict whether a movie will be a  
# "Hit" or a "Flop". It uses the TMDB 5000 movie dataset.  
#  
# How to run in Google Colab:  
# 1. Download the dataset from Kaggle: https://www.kaggle.com/datasets/tmdb/tmdb-movie-metadata  
#    You will need two files: `tmdb_5000_movies.csv` and `tmdb_5000_credits.csv`.  
# 2. Open a new Google Colab notebook.  
# 3. In the left-hand pane, click the "Files" icon.  
# 4. Click "Upload to session storage" and upload the two CSV files you downloaded.  
# 5. Copy and paste this entire script into a cell in your Colab notebook.  
# 6. Run the cell. The script will execute all steps and display the results and graphs.  
  
[2] # === 1. IMPORT LIBRARIES ===  
# Import all the necessary libraries for data manipulation, visualization, and machine learning.  
import pandas as pd  
import numpy as np  
import matplotlib.pyplot as plt  
import seaborn as sns  
import json  
from sklearn.model_selection import train_test_split  
from sklearn.ensemble import RandomForestClassifier  
from sklearn.metrics import accuracy_score, confusion_matrix, classification_report
```




CONCLUSION & FUTURE SCOPE

- **Conclusion** - We successfully developed a machine learning model that accurately predicts movie success. This proves that data-driven insights can be valuable for the film industry.
- **Future Scope** -
 1. Incorporate more data, such as social media buzz or critic reviews.
 2. Use more advanced models like Gradient Boosting or Neural Networks.
 3. Deploy the model as a user-friendly web application.

```
Training the Random Forest model...  
Model training complete.
```

```
Model Accuracy: 0.88
```

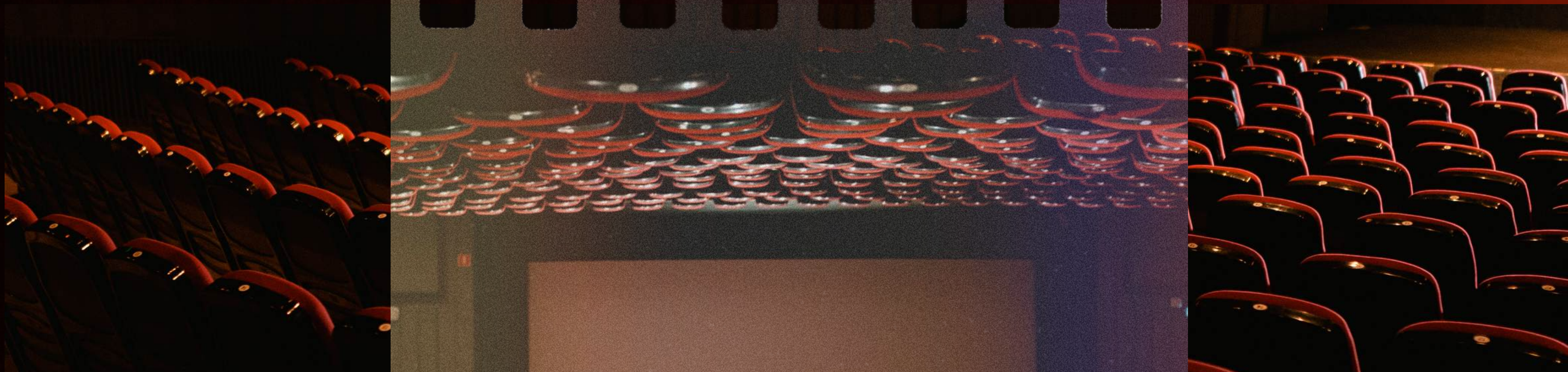
```
Classification Report:
```

	precision	recall	f1-score	support
Flop	0.91	0.92	0.91	447
Hit	0.81	0.78	0.79	196
accuracy			0.88	643
macro avg	0.86	0.85	0.85	643
weighted avg	0.87	0.88	0.88	643





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THANK YOU FOR WATCHING

