

# **TASK**

# **Exploratory Data Analysis on the Movies Data Set**

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# Introduction

Summary of the data set

### **Data Cleaning**

During the data cleansing phase, several steps were taken to prepare the dataset for analysis.

# 1. Column Removal:

Redundant or unnecessary columns were identified and dropped from the dataset. These included 'homepage', 'keywords', 'original\_language', 'original\_title', 'overview', 'production\_companies', 'status', and 'tagline'.

## 2. Remove Duplicate Rows:

Duplicate rows were removed to maintain data integrity and ensure that each entry in the dataset is unique.

## 3. Remove Rows with Missing Data:

Rows with missing or zero values in crucial columns, such as budget or revenue, were filtered out. This step helps in focusing the analysis on complete and meaningful data.

#### 4. Change Data Types:

- Data types were adjusted to facilitate easier manipulation.
- 'release\_date' was converted to the DateTime format for better date handling.
- The release year was extracted from each release date.
- budget' and 'revenue' columns were converted to the integer data type (int64) using NumPy.

#### 5. Flatten JSON Columns:

- JSON-formatted columns ('genres', 'production\_countries', 'spoken\_languages') were flattened, making the data more accessible for analysis.

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# Data Stories and visualisations

# **Data Exploration:**

Identify Relationships Between Variables:

In the exploration phase, the goal is to identify interesting relationships or patterns within the dataset. Specific questions or relationships can be defined for further investigation.

## Top 5 Most Expensive Movies:

To understand the budget distribution, the top 5 most expensive movies were identified, providing insights into high-budget productions.

| Title  | Budget      |
|--|-------------|
| Pirates of the Caribbean: On Stranger<br>Tides | 380,000,000 |
| Pirates of the Caribbean: At World's End       | 300,000,000 |
| Avengers: Age of Ultron                        | 280,000,000 |
| Superman Returns                               | 270,000,000 |
| John Carter                                    | 260,000,000 |

## Top 5 Most Profitable Movies:

Profitability, calculated as the difference between revenue and budget, was used to determine the top 5 most profitable movies. This sheds light on successful ventures in terms of financial gains.

| Title          | Profit        |
|----------------|---------------|
| Avatar         | 2,550,965,087 |
| Titanic        | 1,645,034,188 |
| Jurassic World | 1,363,528,810 |
| Furious 7      | 1,316,249,360 |

| Title        | Profit        |
|--------------|---------------|
| The Avengers | 1,299,557,910 |

# Top 5 Most Popular Movies:

Sorting the dataset based on the 'popular' column revealed the top 5 most popular movies, giving an indication of audience interest and engagement.

| Title                   | Popularity |
|-------------------------|------------|
| Minions                 | 875.581305 |
| Interstellar            | 724.247784 |
| Deadpool                | 514.569956 |
| Guardians of the Galaxy | 481.098624 |
| Mad Max: Fury Road      | 434.278564 |

# Movies Rated Above 7:

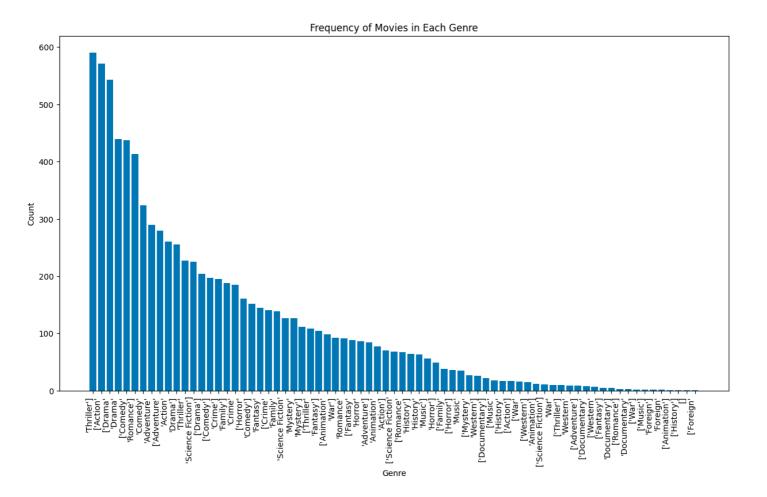
Movies with a rating above 7 were singled out, providing insights into critically acclaimed films based on user ratings.

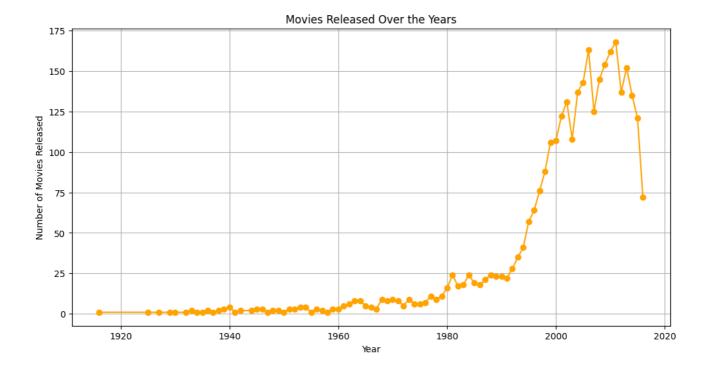
| Title                                     | Vote Average |
|---|--------------|
| Avatar                                    | 7.2          |
| The Dark Knight Rises                     | 7.6          |
| Tangled                                   | 7.4          |
| Avengers: Age of Ultron                   | 7.3          |
| Harry Potter and the Half-Blood<br>Prince | 7.4          |
| Roger & Me                                | 7.4          |
| Eraserhead                                | 7.5          |

| Title  | Vote Average |
|--------|--------------|
| Pi     | 7.1          |
| Clerks | 7.4          |
| Cure   | 7.4          |

#### Most Successful Genres:

A bar plot was created to visualise the frequency of movies in each genre. This analysis helps understand the distribution of genres in the dataset, potentially revealing trends or preferences.





This line graph depicts two notable trends: first, the upward trajectory in the production of movie releases over time, and second, a distinct decline in popularity. This decline could be attributed to either the impact of the 2020 pandemic or the surge in popularity of digital home movie platforms, which are less reliant on physical audience attendance.

This comprehensive data cleansing and exploration process set the foundation for further in-depth analysis and storytelling. The cleaned dataset now provides a solid basis for extracting meaningful insights from the world of movies.

THIS REPORT WAS WRITTEN BY: Elsy Theledi