IMDB MOVIE ANALYSIS - FINAL REPORT

1. Introduction

- The IMDB 5000 Movie Dataset contains details of over 5,000 movies, including attributes such as title, director, genre, budget, gross revenue, duration, IMDb rating, and number of user votes.
- The objective of this project is to perform a complete data analytics workflow by cleaning the dataset, analyzing it through multiple tools, and deriving meaningful insights.
- This project demonstrates the use of Python, Excel, SQL, and Power BI to explore trends, identify top-performing movies/directors, and visualize patterns in the film industry.

2. Methodology

Python (Data Cleaning & EDA)

- Imported dataset using Pandas.
- Cleaned missing values and duplicates.
- Performed Univariate Analysis (IMDb scores, duration, budgets).
- Performed Bivariate Analysis (Budget vs Gross, Genre vs IMDb score).
- Created plots: histograms, scatter plots, boxplots, heatmaps.

♦ Excel (Pivot & Dashboard)

- Built PivotTables to analyze:
- Top 10 highest-grossing movies.
- Average IMDb rating per genre.
- Movie production trend per year.
- Applied Conditional Formatting:
- Highlighted Top 5 Directors (by movie count).
- Highlighted movies with IMDb score > 8.
- Designed an interactive dashboard with slicers for genre, director, and year.

♦ *SQL* (*Structured Queries*)

- > Imported dataset into MySQL.
- > Executed queries:
- Find top 5 highest-rated movies.
- Get average IMDb score per director.
- List movies with gross > 100M.

- Count movies per genre.
- Find director with most movies.

Power BI (Interactive Dashboard)

- > Imported dataset into Power BI.
- > Built dashboard with visuals:
- Total gross revenue and average IMDb score (KPIs).
- Genre-wise movie distribution.
- Yearly production trend.
- Scatter plot (Budget vs Gross).
- Added slicers for filtering by genre, director, year.

3. Key Findings & Insights Correlation:

- Higher budgets generally lead to higher gross revenue.
- Genre Popularity: Drama, Comedy, and Action dominate the dataset.
- Directors: Steven Spielberg directed the most movies.
- IMDb Ratings: Most movies fall between 5.5 7.5.
- Trends: After 2000, movies show larger budgets and revenues.

4.GitHub Repository:

URL:

https://github.com/swe912mads/IMDB-Movie-Analysis

STRUCTURE:

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IMDB-Movie-Analysis
```

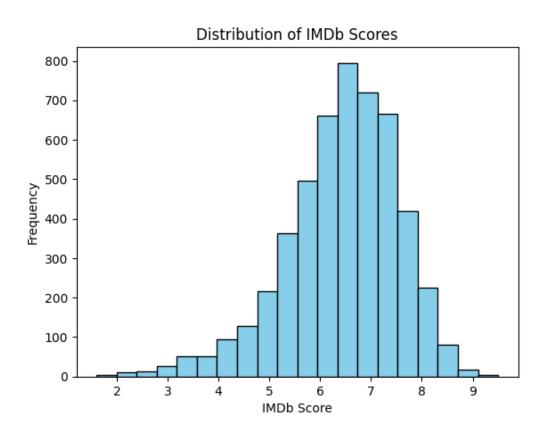
- **-** Data
- imdb cleaned.csv
- Python
- EDA.ipynb
- **-** Excel
- **-** sql
- SQL Queries.sql
- ☐ SQL Results.csv

- **-** PowerBI
- **F** Report
- **│ └** Final_Report.pdf
- **∟** README.md

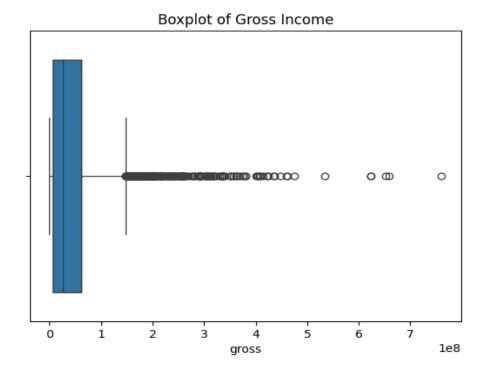
5. Visuals:

Charts &Plots:

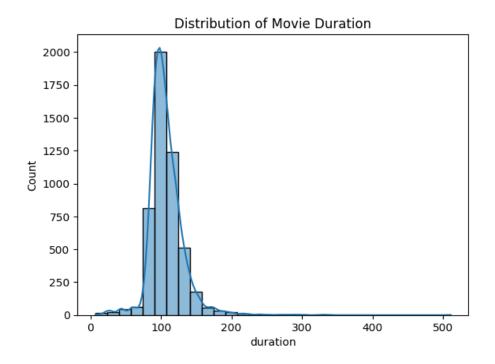
Univariate analysis:



IMDb score distribution

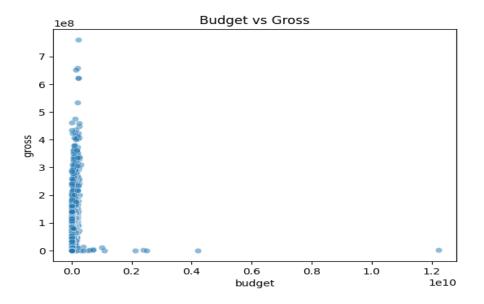


Boxplot of Gross

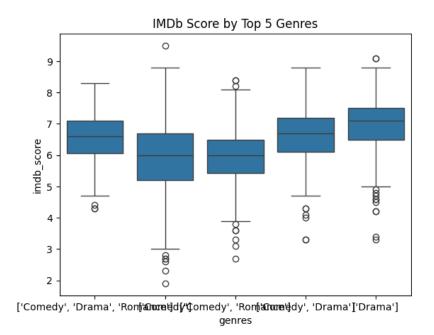


Duration distribution

Bivariate analysis:

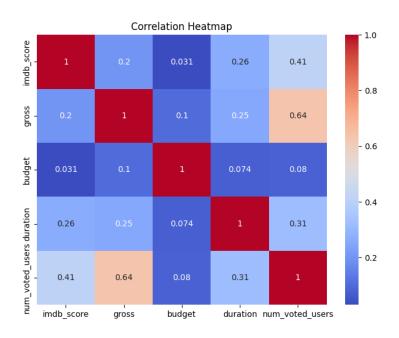


Budget vs Gross



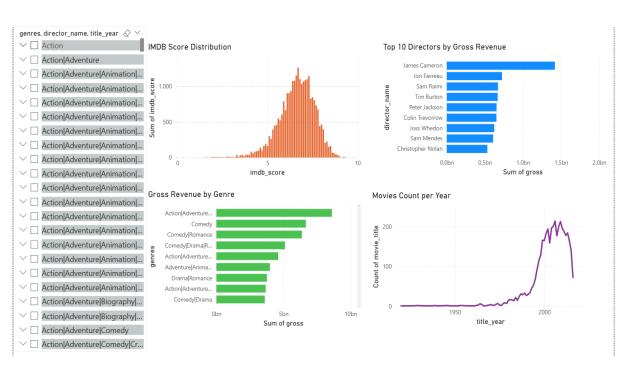
Genre vs IMDb Score

Visualizations using Matplotlib & Seaborn:



Heatmap

Dashboard:



5. Conclusion

This project highlights how multiple tools complement each other in analytics:

- Python for data cleaning and exploratory analysis.
- Excel for pivot analysis and dashboards.
- SQL for structured querying.
- Power BI for advanced interactive visualizations.

The analysis provided meaningful insights into movie trends, revenue patterns, genre performance, and director contributions.