

Movies and their performance on the Box Office

TECHNICAL PROJECT REPORT

Submitted to:

Department of IT

Phoenix College of Management and IT



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Extra Course Internal Evaluation

Submitted by

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1. Introduction to Dataset

- The Movie Data Science Project dataset contains information on **10 popular movies**.
 - Variables include:
 - Movie (name),
 - Genre (category, e.g., Action, Sci-Fi, Romance, etc.),
 - BudgetMillion (budget in millions),
 - RevenueMillion (box office revenue in millions),
 - Rating (IMDb rating).
 - Used for exploratory analysis and insights into movie success factors.
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2. Data Overview

- **Shape:** 10 rows, 5 columns.
 - **Variable Types:**
 - Numerical: BudgetMillion, Revenue Million, Rating
 - Categorical: Movie, Genre
 - **Descriptive Statistics:**
 - Average budget: 144.7M
 - Average revenue: 1275.9M
 - Average rating: 8.288.28
 - No missing values detected.
 - **Genre Distribution:**
 - Action, Sci-Fi, Animation, Romance, Drama.
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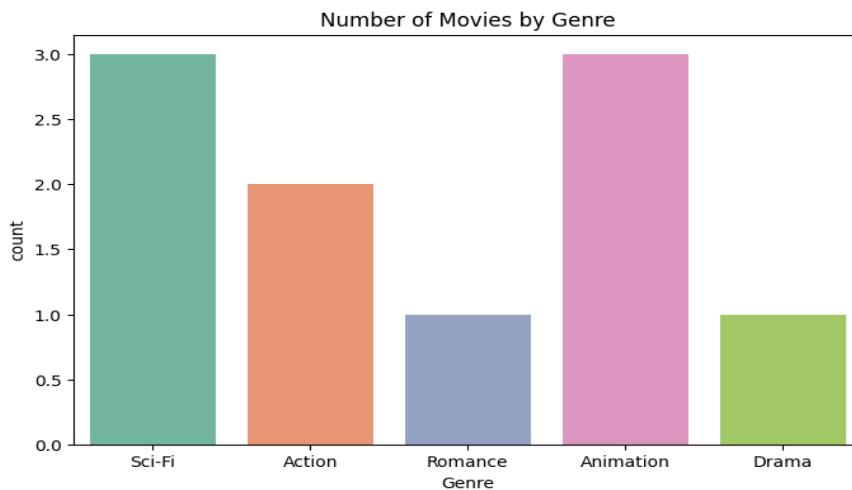
3. Data Quality

- **Completeness:** No missing values detected.
 - **Duplicates:** None present.
 - **Type:** Original.
 - **Outliers:**
 - Revenue: Extremely high (e.g., Avatar, Titanic).
 - Ratings: All above 7, The Dark Knight exception at 9.
 - No major inconsistencies, though sample size is small.
 - Creation: Created by Kunal Bhandari as it the list of his favorite movies.
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4. Visualization Overview

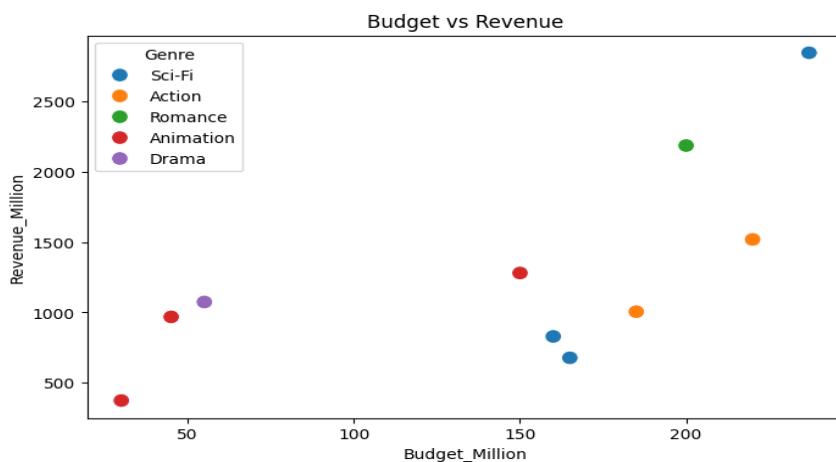
- **Genre Popularity:**

- Countplot: Highlights Animation and Sci-Fi as frequent genres in high-grossing movies.



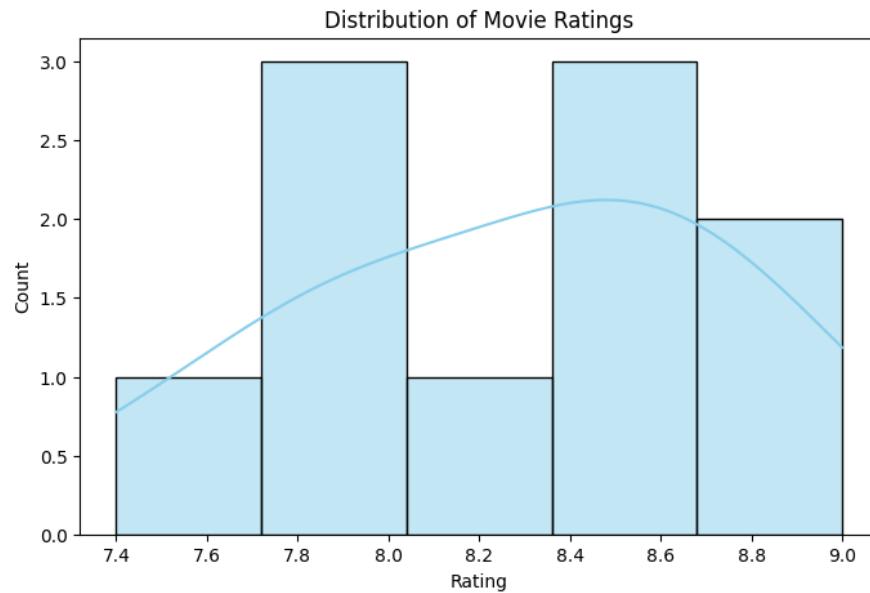
- **Budget vs Revenue:**

- Scatterplot: Shows that higher budgets often correlate with higher revenue, but not always with higher ratings.

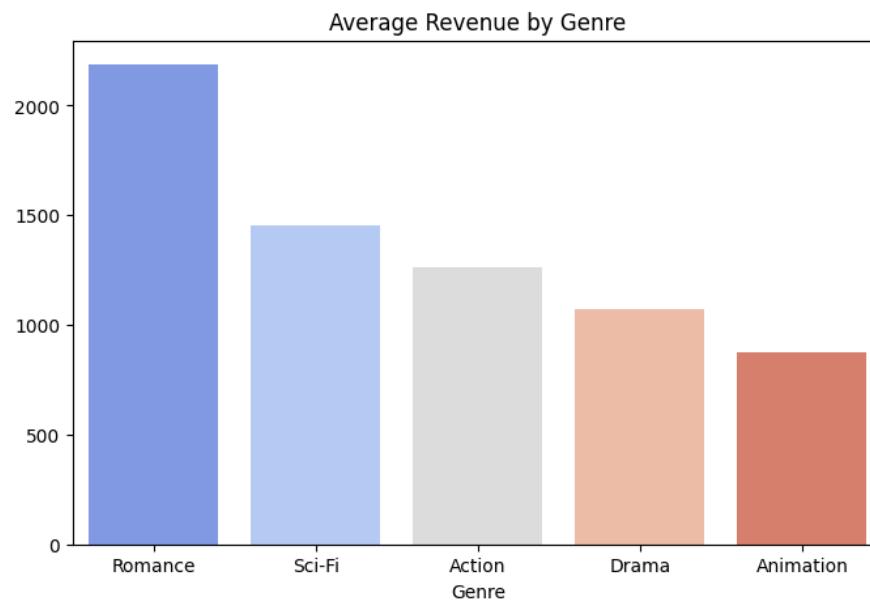


- **Rating Distribution:**

- Histogram: Most movies rated above 8, indicating strong critical performance.



- **Average Revenue by Genre:**
 - Barplot: Action and Sci-Fi genres show highest average revenues.



5. Feature Engineering

- **Encoding:**
 - One-hot encoding suitable for Genre if used for further machine learning.
 - **Binning:**
 - Rating can be grouped into categories (High, Medium, Low) for interpretability.
 - **Redundancy:**
 - No constant or unused columns detected.
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6. Key Findings and Conclusion

- **Insights:**
 - Action and Sci-Fi movies tend to generate higher revenues.
 - Animation movies also perform strongly, particularly in revenue relative to budget.
 - High budget does not guarantee higher ratings.
 - All movies analyzed are rated well (above 7), with notable outlier performances (The Dark Knight).
 - **Conclusion:**
 - The dataset provides an overview of popular, high-performing movies.
 - Further analysis with larger datasets may yield deeper insights into genre trends and success factors.
 - Feature engineering and visual exploration offer actionable perspectives for understanding movie industry patterns.
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GitHub Repo link: https://github.com/Anyonomous1/Python-Data-Science/blob/main/Movie_DataScience_Project.ipynb