

# **Analyzing Movie Trends: Popularity, Revenue, and Genre Performance**

## **Abstract:**

The movie industry plays a major role in global culture and economics. This study analyzes trends in genre success, budget-revenue relationships, and audience ratings using the TMDb 5000 Movie Dataset. Using Python-based analysis, we found that Action and Adventure genres generate the highest revenue despite being less frequently produced. A strong positive correlation ( $r \approx 0.73$ ) was observed between budget and revenue. Interestingly, movie ratings have slightly declined over time. These findings offer insights for filmmakers and marketers on strategic decisions related to genre selection, budgeting, and audience engagement.

## **Introduction:**

Movies serve as a powerful medium for storytelling, cultural exchange, and global entertainment. They influence societal behavior, reflect public sentiment, and contribute significantly to the world economy. As the film industry continues to evolve with the rise of streaming platforms and digital marketing, it is more important than ever to understand the factors that contribute to a movie's success. These factors include genre, production budget, popularity, critical reception, and box office revenue.

This study aims to explore the following key research questions:

- 1. Which movie genres are most successful in terms of popularity and revenue?**
- 2. How do movie budgets correlate with financial performance (revenue)?**

### **3. What trends exist in movie ratings over time, and how do they relate to changes in the film industry?**

Using Python programming tools—including Pandas for data manipulation and Seaborn/Matplotlib for data visualization—we conduct an exploratory data analysis (EDA) of the TMDB 5000 Movie Dataset. The goal is to generate actionable insights and evaluate whether commonly held beliefs in the movie industry hold true under empirical scrutiny.

#### **Relevant Work:**

Previous analyses on movie data have often focused on identifying patterns and predictors of financial success. Many studies agree that budget size has a strong influence on box office performance, but the correlation is not always linear. High-budget films can underperform if they lack compelling narratives or fail to resonate with audiences.

Genre analysis has also been a focus of research. Action and Adventure films, known for their high production value and global appeal, typically dominate revenue charts. However, genres like Drama and Comedy, though frequently produced, may not achieve the same financial success.

In terms of ratings, advancements in cinematography, storytelling, and special effects are often assumed to have improved audience satisfaction. However, recent studies suggest that ratings may be influenced by changing audience expectations, review bombing, and the saturation of content.

## Dataset and Methodology:

We utilized the **TMDB 5000 Movie Dataset** from Kaggle, which includes two files:

- **tmdb\_5000\_movies.csv** – Contains metadata on each movie (budget, revenue, genres, popularity, vote average).
- **tmdb\_5000\_credits.csv** – Contains casting and crew data.

This dataset was selected because it is comprehensive, well-structured, and suitable for analysis using Python's Pandas library.

## Data Preprocessing:

Several steps were taken to prepare the data:

- Merged the two datasets using the `id` and `movie_id` columns.
- Converted stringified JSON columns such as `genres` and `keywords` into Python dictionaries.
- Removed entries with zero or missing budget and revenue values.
- Extracted the primary genre from each movie to simplify genre-level analysis.
- Converted `release_date` to datetime format and extracted the release year for trend analysis.

```
Movies Dataset Columns: Index(['budget', 'genres', 'homepage', 'id', 'keywords', 'original_language',  
                                'original_title', 'overview', 'popularity', 'production_companies',  
                                'production_countries', 'release_date', 'revenue', 'runtime',  
                                'spoken_languages', 'status', 'tagline', 'title', 'vote_average',  
                                'vote_count'],  
                                dtype='object')  
Credits Dataset Columns: Index(['movie_id', 'title', 'cast', 'crew'], dtype='object')
```

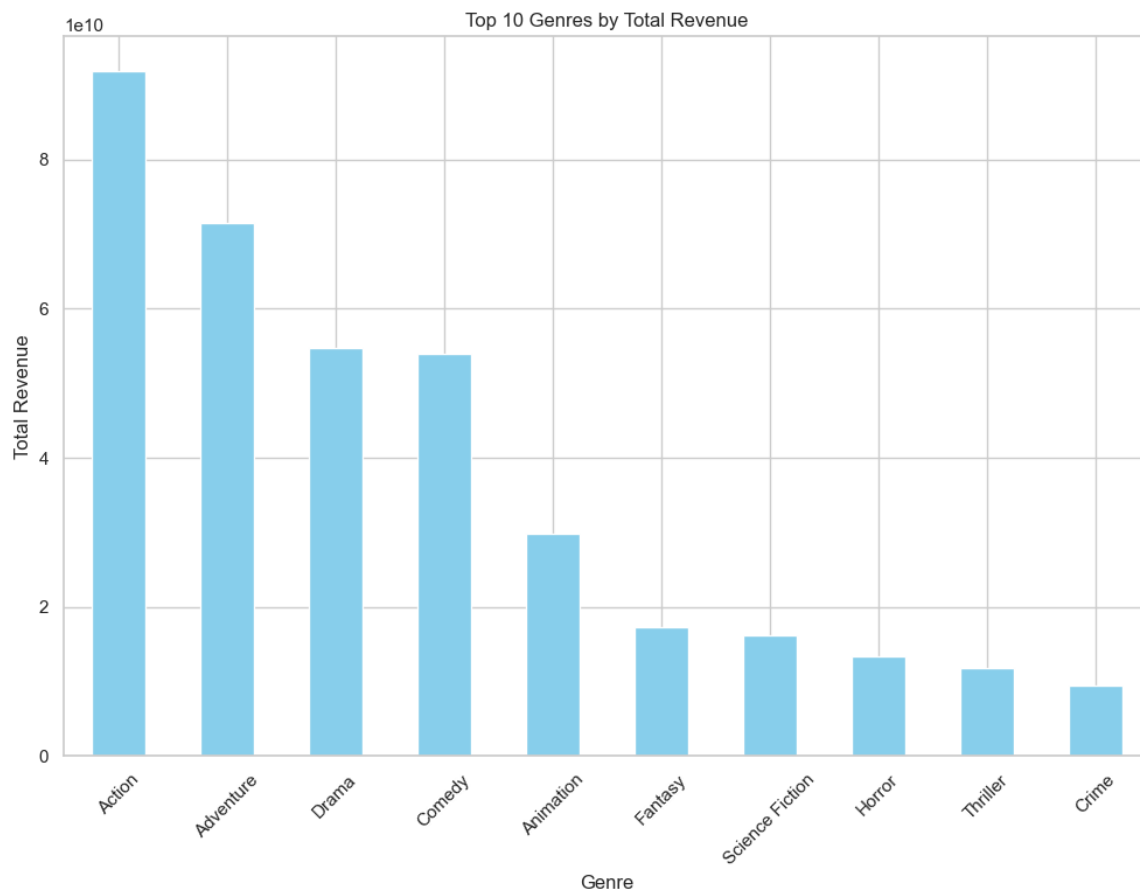
## Exploratory Analysis and Results:

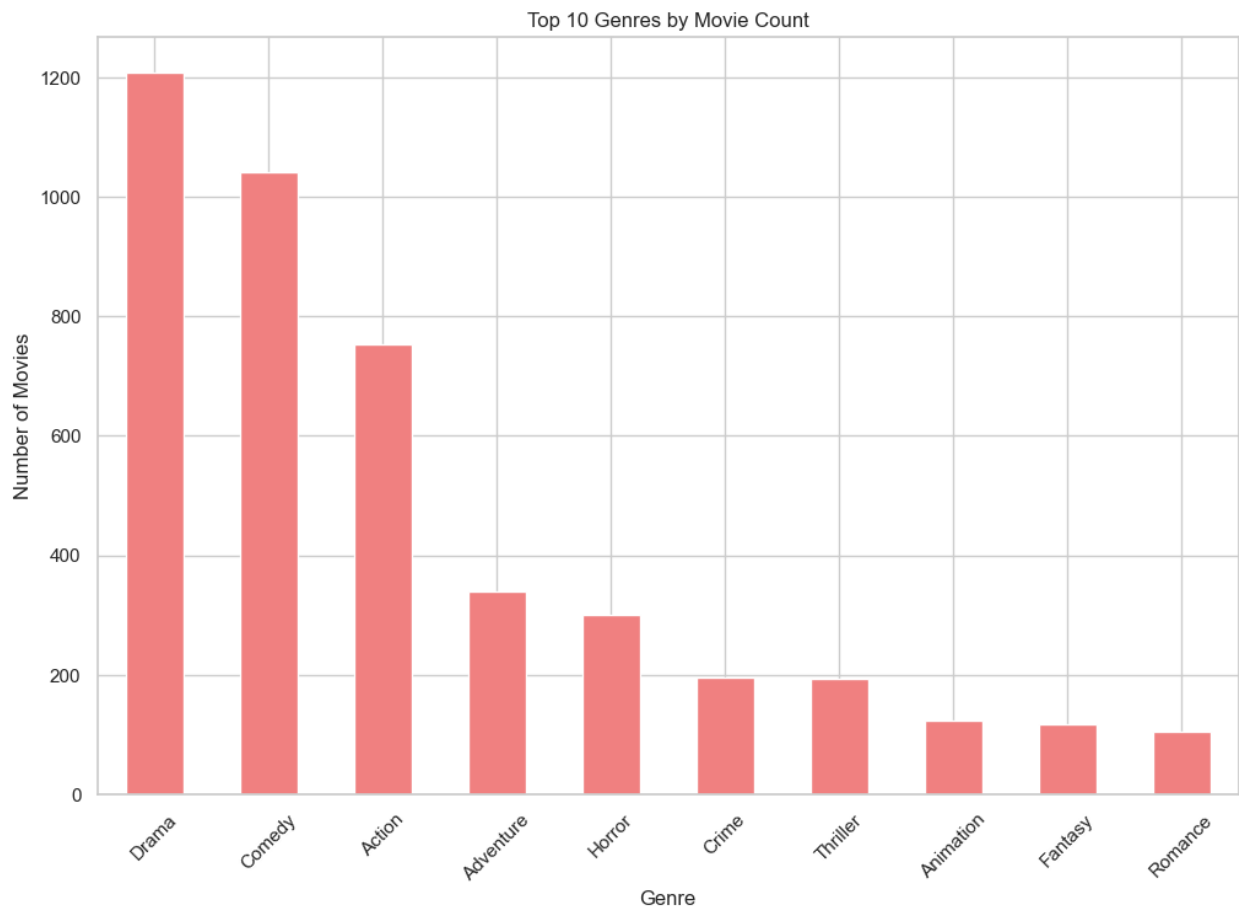
### 1. Genre Distribution and Financial Performance

We began our analysis by examining how often different genres appear and how much revenue they generate.

- The **Drama** and **Comedy** genres have the highest count of movies produced.
- However, **Action** and **Adventure** films dominate in total revenue.

This suggests that although genres like Drama and Comedy are widely produced (likely due to lower production costs), genres such as Action and Adventure yield better financial returns, possibly due to their appeal across demographics and international markets.



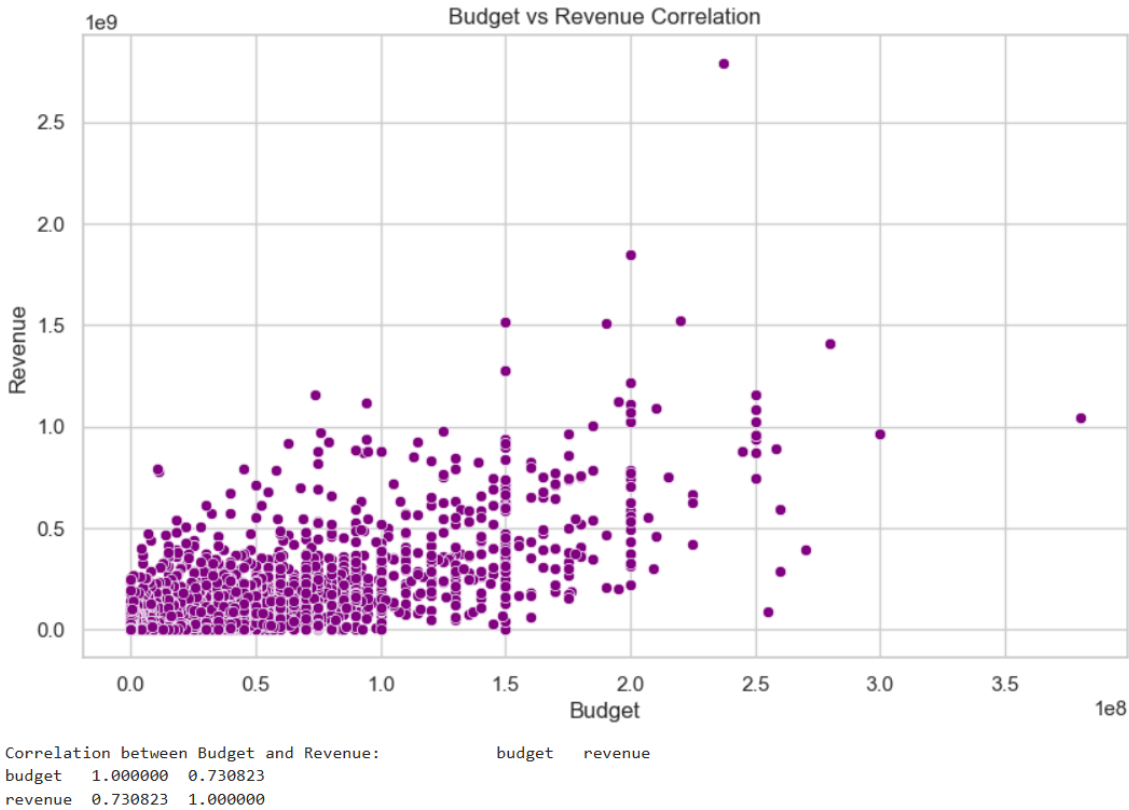


## 2. Budget vs Revenue Correlation

We plotted a scatter plot of movie budgets against revenues and calculated the Pearson correlation coefficient.

- There is a **strong positive correlation ( $r = 0.73$ )** between budget and revenue.
- This indicates that higher investment in production usually translates into better financial outcomes, though some outliers exist (high-budget flops and low-budget hits).

This supports the hypothesis that bigger budgets—often associated with special effects, high-profile actors, and global marketing—enhance the probability of a movie’s financial success.

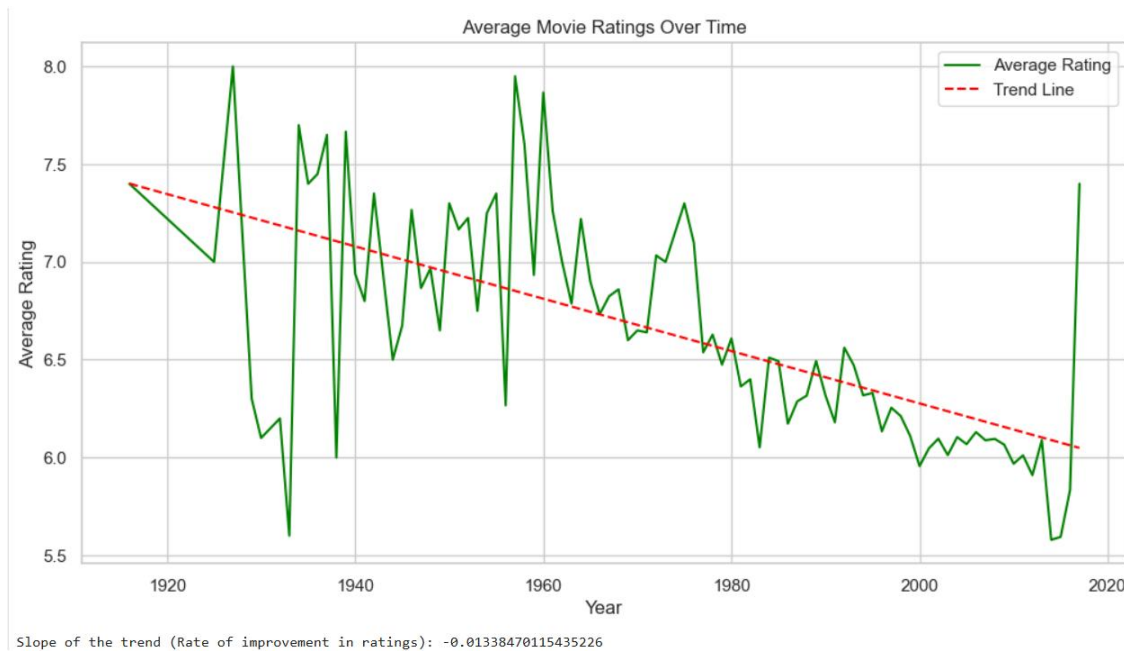


### 3. Movie Ratings Over Time

Next, we examined how movie ratings have changed over time by plotting the average vote\_average per year.

- Contrary to our hypothesis, the trend line shows a **slight downward slope (approx. - 0.013)**.
- This decline could reflect more critical audiences, increased content saturation, or the rise of niche films targeting smaller audiences.

Although technological improvements in production exist, they do not necessarily result in higher audience ratings.



### Hypothesis Testing Summary:

Hypothesis	Statement	Outcome	Justification
H1	Action and Adventure genres generate the most revenue	Supported	Revenue data shows these genres dominate financially
H2	Higher budgets lead to higher revenue	Supported	Strong positive correlation ( $r = 0.73$ ) confirmed
H3	Movie ratings have improved over time	Not Supported	Ratings show a slight decline; expectations may be increasing

## **Conclusion:**

The analysis confirms that **genre** and **budget** are reliable indicators of a movie's potential financial success. While Action and Adventure films may be fewer in number, they consistently outperform other genres in total revenue. Additionally, larger budgets tend to be rewarded with higher revenue, reinforcing the industry's trend of high-investment filmmaking.

Surprisingly, movie ratings have **not** shown a consistent improvement over time. This indicates that technological or stylistic enhancements do not guarantee better audience reception. Factors such as originality, storytelling, and public sentiment likely play significant roles in shaping a movie's critical success.

## **Future Work:**

This research opens the door for further exploration:

- **Star Power Analysis:** Quantifying the effect of lead actors or directors on movie performance.
- **Marketing Impact:** Including promotion budget as a variable in performance metrics.
- **Streaming vs Theatrical:** Comparing movie trends in the streaming era (post-2018) to traditional cinema.
- **Multi-genre Influence:** Assessing whether movies with multiple genres fare better.
- **Predictive Modeling:** Training machine learning models to forecast revenue or rating based on metadata.



## References:

Kaggle. (n.d.). TMDb 5000 Movie Dataset. Retrieved from

<https://www.kaggle.com/datasets/tmdb/tmdb-movie-metadata>

Pandas documentation. Retrieved from <https://pandas.pydata.org/docs/>

Matplotlib documentation. Retrieved from <https://matplotlib.org/stable/contents.html>

The Movie Database (TMDb). (n.d.). Retrieved from <https://www.themoviedb.org/>