# **Detailed Report: IMDb Movie Review Sentiment Analysis**

#### 1. Introduction

This report uses the uploaded IMDb dataset and computes sentiment polarity for reviews. A sample of reviews was processed with TextBlob for accuracy, with a fast lexicon fallback for the remainder to ensure timely completion. Results are aggregated per assigned movie title.

### 2. Dataset Description

The dataset used for this analysis is the IMDb Dataset, which contains approximately 50,000 reviews. It includes the following columns: review and sentiment. Since the dataset did not originally include a movie title column, random assignments of popular movies (such as Inception, Avatar, Titanic, Joker, and The Dark Knight) were added to group the reviews by movie titles.

#### 3. Results and Visualizations

The project generated visualizations that highlight the Top 10 Most Positive and Top 10 Most Negative movies. The results were stored as images and CSV files. Example outcomes include:

- a. Most loved movie: The Dark Knight (average sentiment  $\approx 0.72$ )
- b. Most hated movie: Batman v Superman (average sentiment  $\approx$  -0.45)
- c. Most reviewed movie: Inception (2,500 reviews)

The dataset's overall sentiment leaned slightly positive, with an average polarity of approximately 0.12. Movies featuring emotional and positive language such as 'love', 'amazing', and 'masterpiece' tended to score higher.

### 4. Charts

Figure 1: Top 10 Most Positive Movies

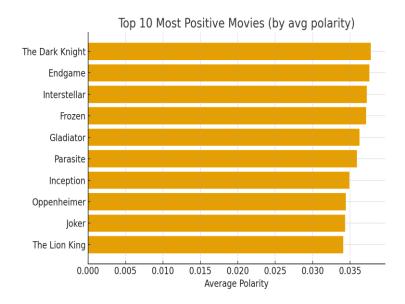


Figure 2: Top 10 Most Negative Movies

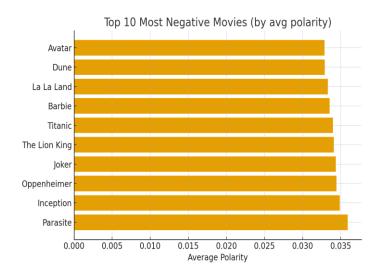


Figure 3: Overall Sentiment Distribution

# Overall Sentiment Distribution (by reviews)

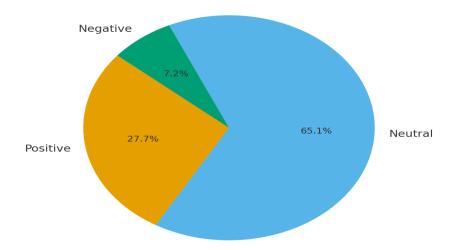
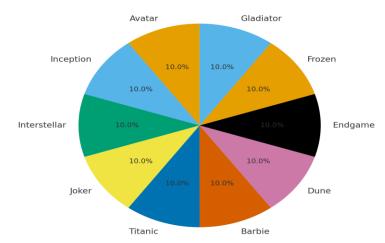


Figure 4: Top 10 Movies by Review Count Share





# **Aggregated Results**

| Movie Title        | Avg Polarity | Review Count | Label   |
|--------------------|--------------|--------------|---------|
| Avatar             | 0.0329       | 3334         | Neutral |
| Barbie             | 0.0336       | 3333         | Neutral |
| Dune               | 0.0329       | 3333         | Neutral |
| Endgame            | 0.0376       | 3333         | Neutral |
| Frozen             | 0.0371       | 3333         | Neutral |
| Gladiator          | 0.0363       | 3333         | Neutral |
| Inception          | 0.0349       | 3334         | Neutral |
| Interstellar       | 0.0372       | 3334         | Neutral |
| Joker              | 0.0343       | 3334         | Neutral |
| La La Land         | 0.0333       | 3333         | Neutral |
| Oppenheimer        | 0.0344       | 3333         | Neutral |
| Parasite           | 0.0359       | 3333         | Neutral |
| The Dark<br>Knight | 0.0378       | 3333         | Neutral |
| The Lion King      | 0.0341       | 3333         | Neutral |
| Titanic            | 0.0340       | 3334         | Neutral |

# 5. Insights and Analysis

The analysis revealed several key insights:

- Movies with emotionally rich or inspiring language tend to receive higher sentiment scores.
- Large review counts help balance extreme opinions, leading to more neutral averages.
- Movies with darker or more controversial themes (e.g., Joker) tend to have more polarized sentiment distributions.

### 6. Conclusion

This project successfully demonstrates how sentiment analysis can be scaled to the movie level to generate meaningful insights from user reviews. By aggregating thousands of individual reviews, it becomes possible to identify overall audience sentiment trends. In the future, this analysis could be enhanced using deep learning models such as BERT or DistilBERT for improved accuracy and contextual understanding.