

The background of the slide features a close-up, dark, and slightly blurred image of a film reel on the left and a clapperboard on the right. The clapperboard has labels for 'CAMERA', 'SCENE', and 'TAKE' visible at the bottom right.

# Film Industry Trends and Success Using IMDb Data Sets

Part 6: Project Presentation

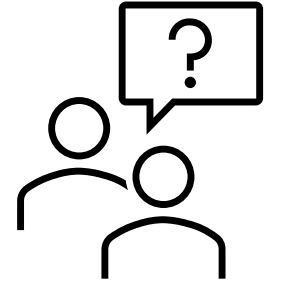
**Group 2:** Nathan Harris, Mayumi Shimobe

[https://github.com/CSPB4502-](https://github.com/CSPB4502-Group2/FilmIndustryTrendsAndSuccessUsingIMDbDataSet)

[Group2/FilmIndustryTrendsAndSuccessUsingIMDbDataSet](https://github.com/CSPB4502-Group2/FilmIndustryTrendsAndSuccessUsingIMDbDataSet)

CSPB 4502 Data Mining Spring 2025

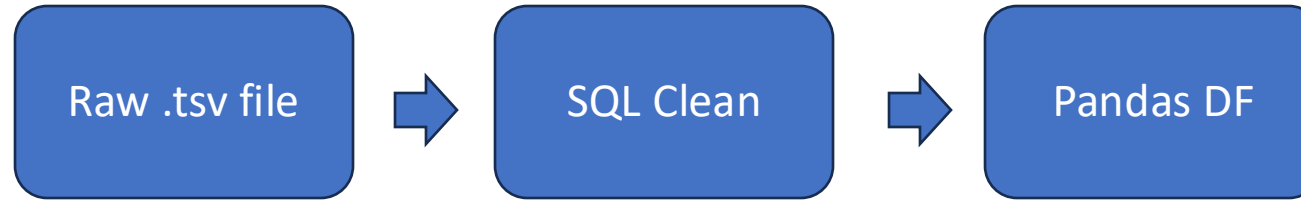
# Questions We Asked



- Which movie attributes drive audience ratings?
- Can we predict a film's rating or classify “success”?
- How do linear vs. ensemble models compare?



# Data Preparation



- Loaded 1.5 M IMDb titles & ratings via MySQL RDS
- Cleaned: convert “\N” → NULL, remove duplicates
- Flagged low-vote titles (< 10 votes), log-transformed votes
- One-hot encoded genres for 20+ categories



## Data Warehouse & Cubes

- Built imdb\_integrated view (basics+ratings)
- Star schema with fact table + genre, year dimensions
- Precomputed “yearly\_votes” & “genre\_performance” cubes

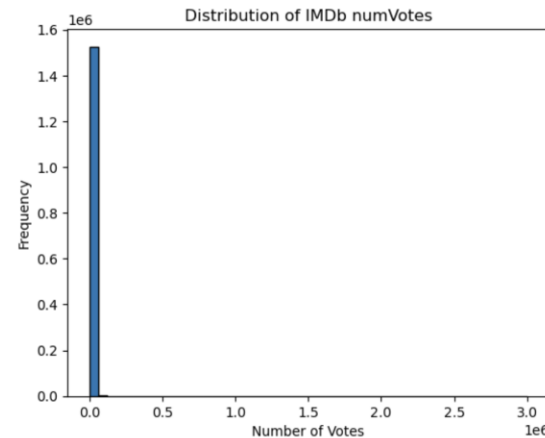
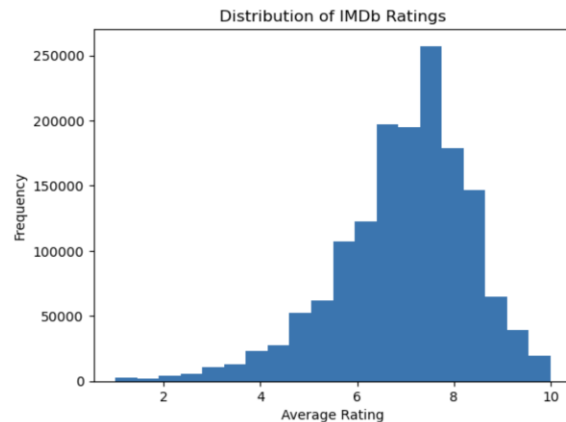
## Tools & Environment

- Database: MySQL on AWS RDS (+ SQLAlchemy, %sql magic)
- Analysis: Python / Jupyter, Pandas, NumPy, SciPy
- Modeling: scikit-learn (Linear, Logistic, RF, HGB)
- Visualization: Matplotlib and Seaborn

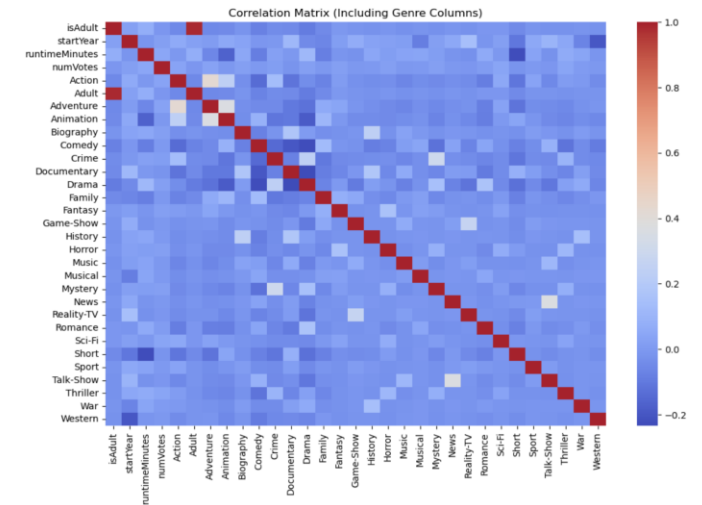


# Exploratory Data Analysis

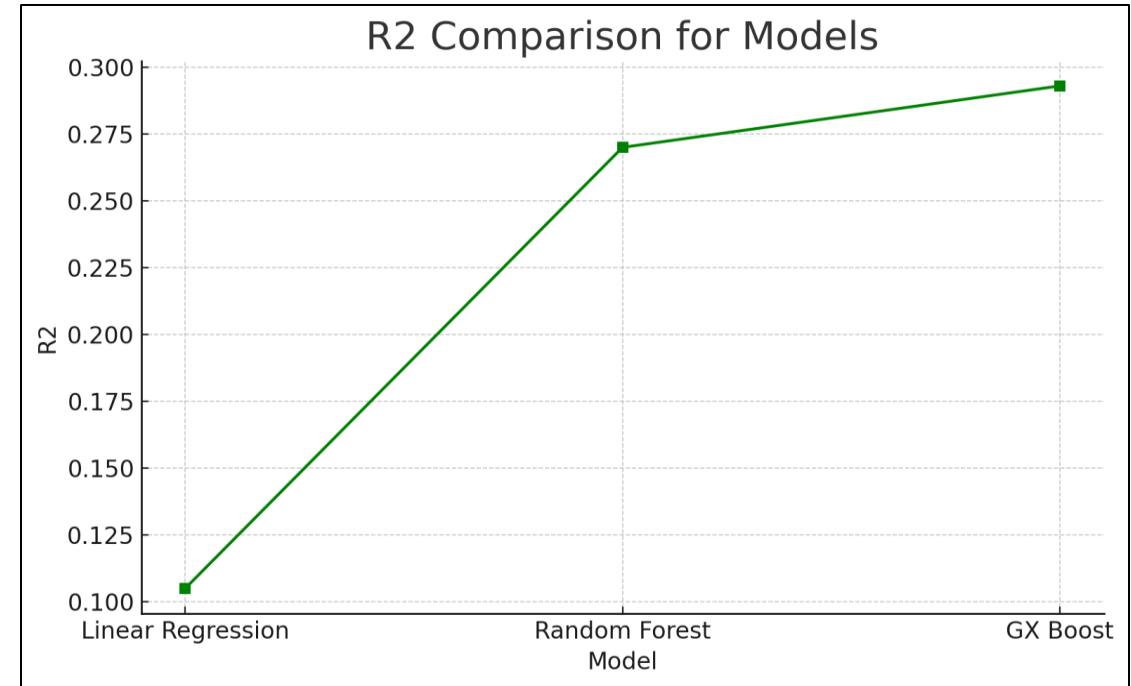
- Ratings histogram peaked at 6–7 (Gaussian-like)
- Votes histogram heavy-tailed (skew  $\approx 57$ )  $\rightarrow \log^{1+}$  transform
- Correlation heatmap: nearly zero pairwise  $\rho \rightarrow$  non-linear focus



Skewness of numVotes: 57.31708778938491



# Regression Models



Model	RMSE	R <sup>2</sup>
Linear Regression	0.946	0.105
Random Forest	1.135	0.270
GX Boost	1.117	0.293



# Classification Models

Model	Acc	Recall	Precision
Logistic Regression	0.79	0.00	0.50
Random Forest	0.56	0.87	0.30
GX Boost	0.80	0.07	0.67



# Knowledge Gained

- Vote volume is critical but skewed → log transform
- Weak linear correlations justify non-linear / ensemble methods
- Class imbalance dominates classification performance

# Applications

- **Studios** forecast ratings pre-release to guide budgets
- **Marketing** target campaigns to high-potential titles
- **Platforms** integrate scores into recommendation engines





Thank you!



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