REGRESSION ANALYSIS "MOVIE INDUSTRY"

TEAM 5 - BAN 1 VINCENT | TOSHIAKI | ARJUN



ABOUT THIS ANALYSIS



- Movie industry
- Over 7000 records
- Information on:
 - Votes
 - Runtime
 - Gross
 - Budget

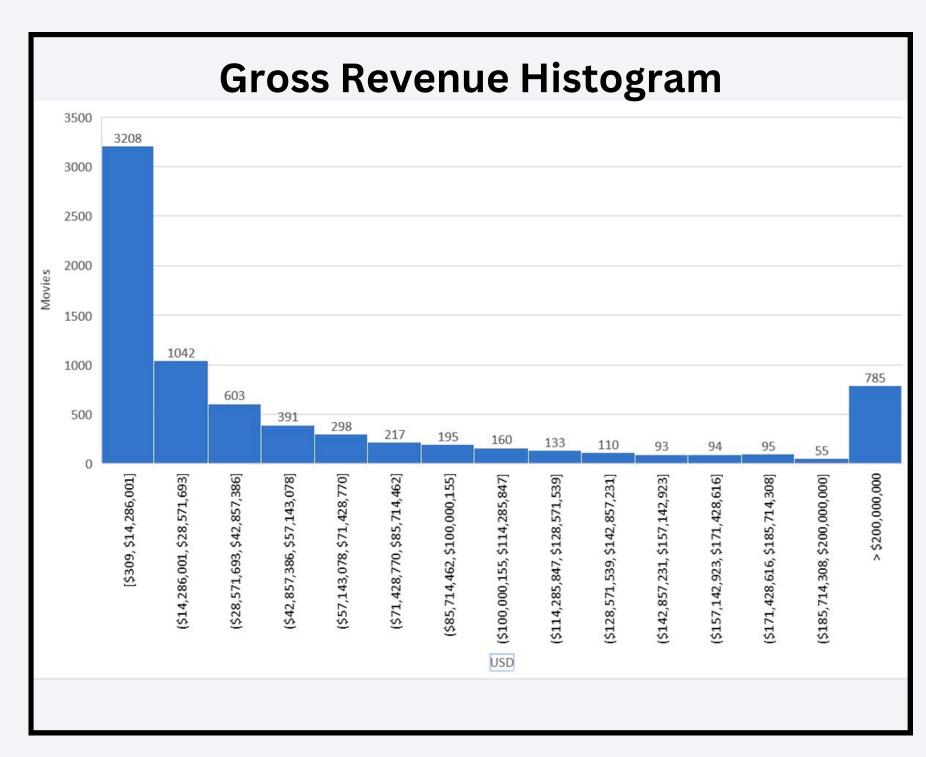


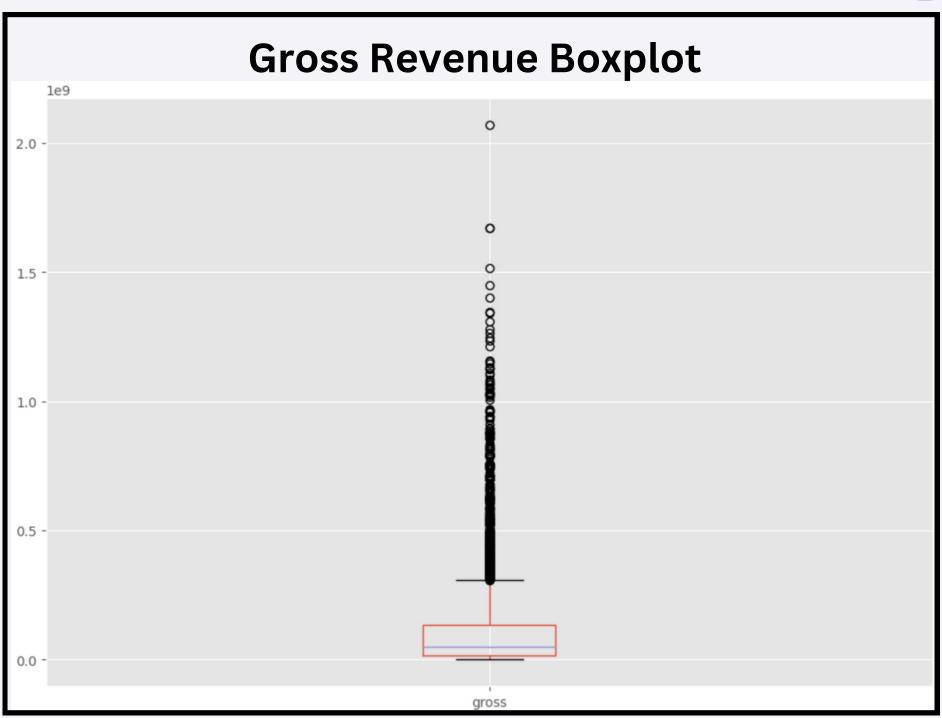
- Univariate and Bivariate Analysis.
- Outliers, missing values.
- Shape before: (7668, 6)
- Shape after: (4502, 52)
- Regression with scaler

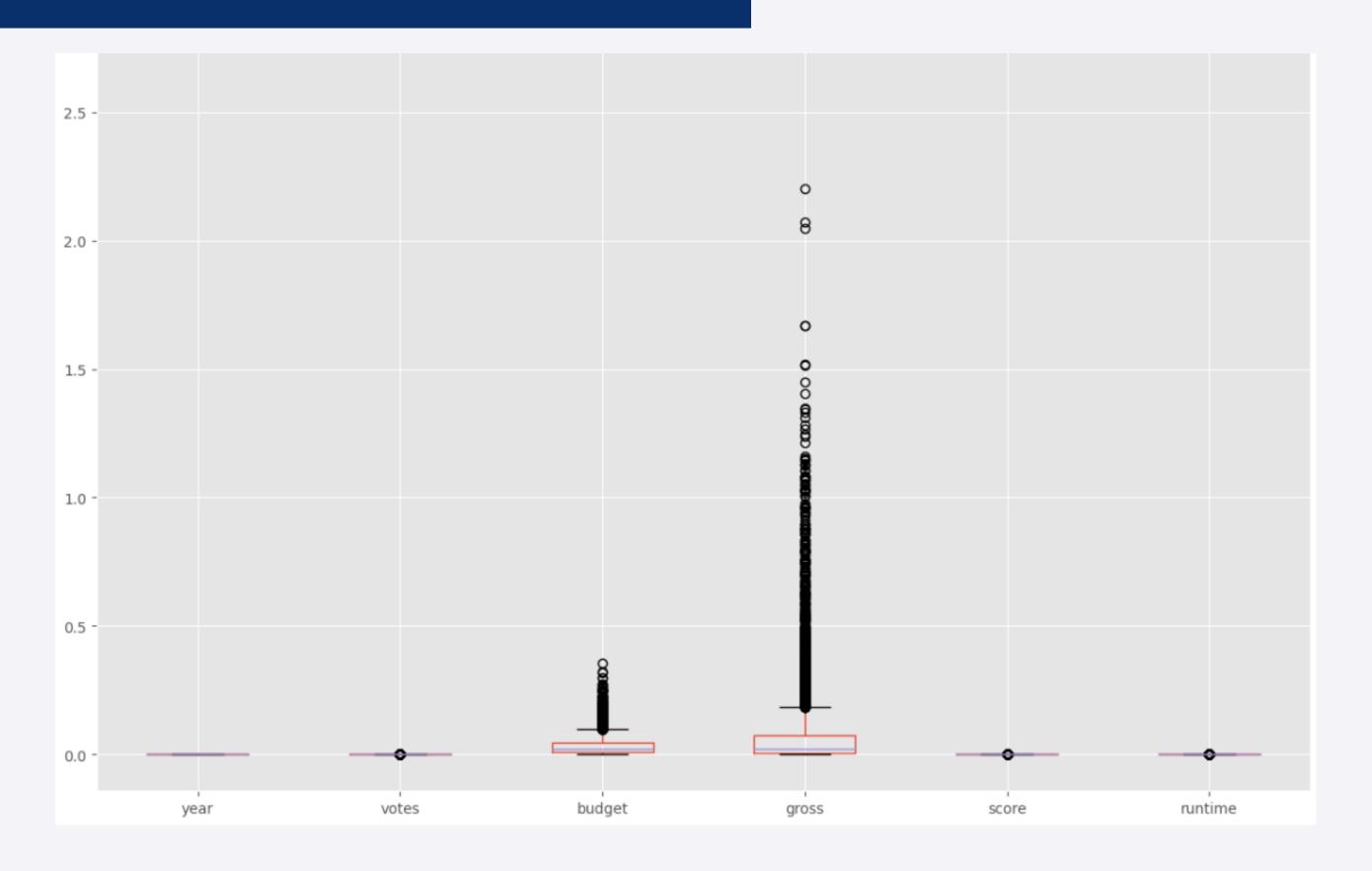
BUSINESS CASE



What are the driving factors of a good-scoring movie?







UNIVARIATE ANALYSIS GROSS

MIN

\$2,554

MAX

\$2,069,521,700

MEAN

\$115,114,071.48

MEDIAN

\$49,843,011

MODE

\$140,000,000

VOTES

MEAN 115,066

MEDIAN 52,000

MODE NaN

MIN 195

MAX 2,400,000

BUDGET

MEAN 39,945,720

MEDIAN 250,000,000

MODE 30,000,000

MIN 6,000

MAX 356,000,000

RUNTIME

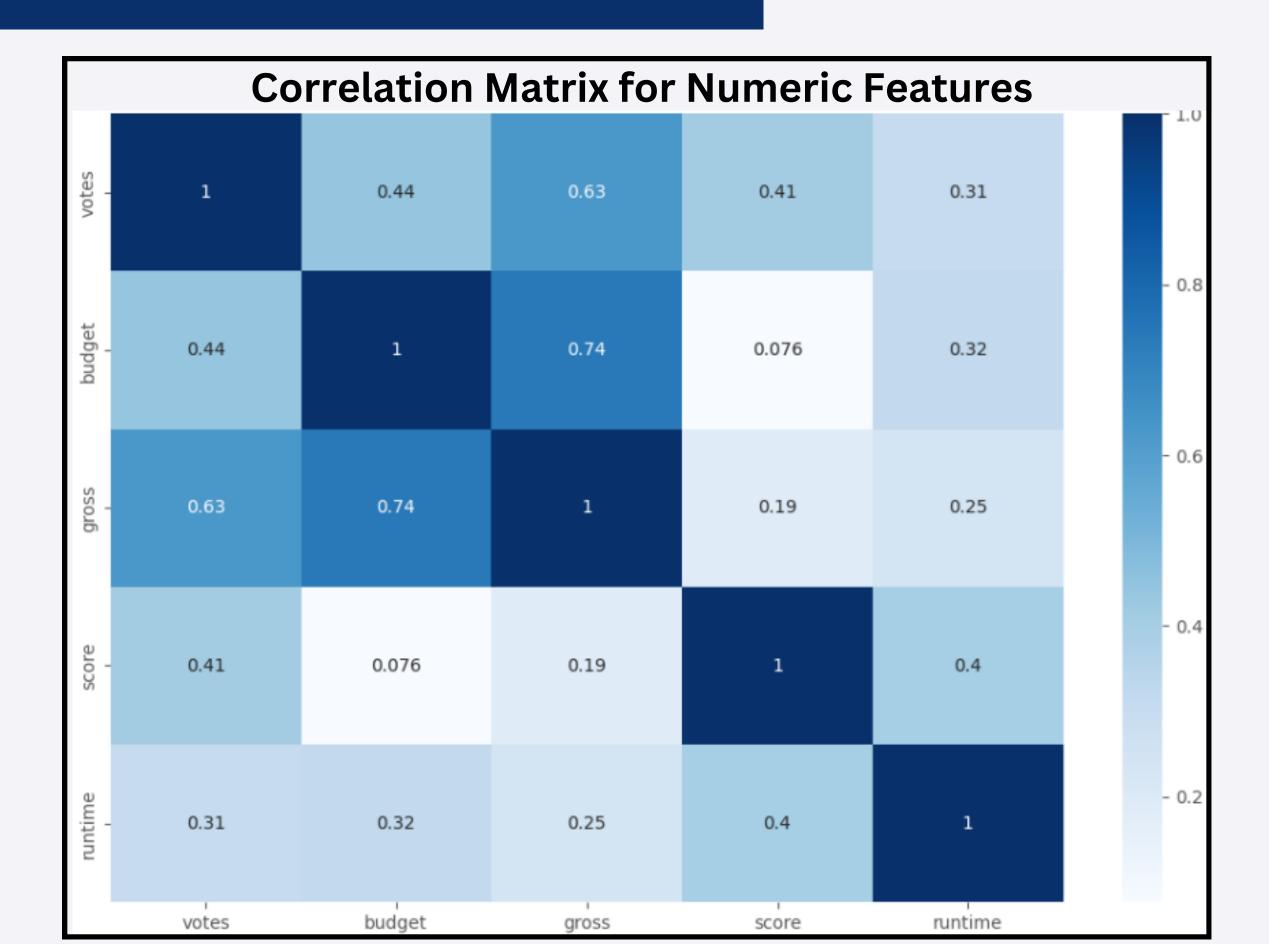
MEAN 109

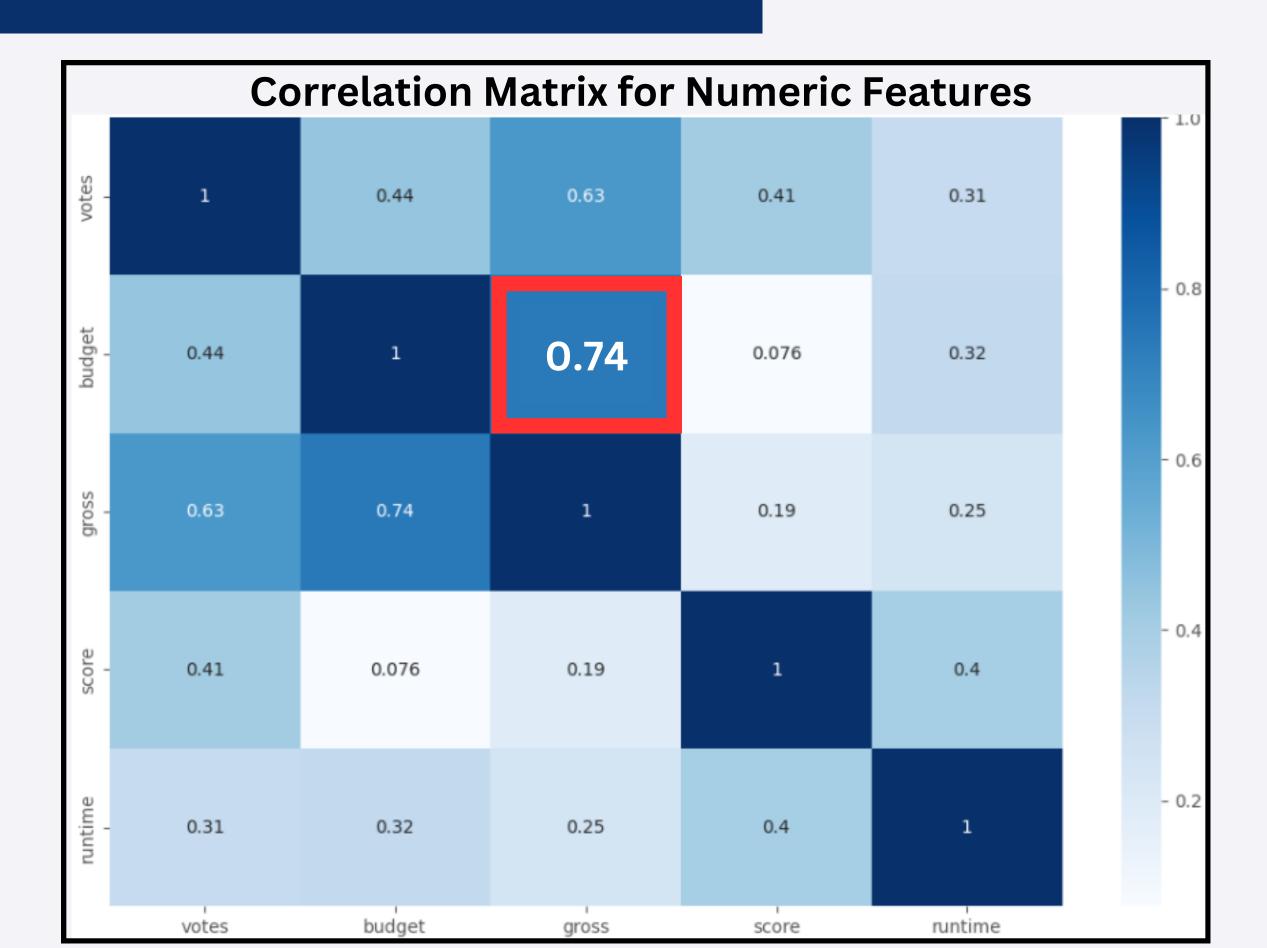
MEDIAN 106

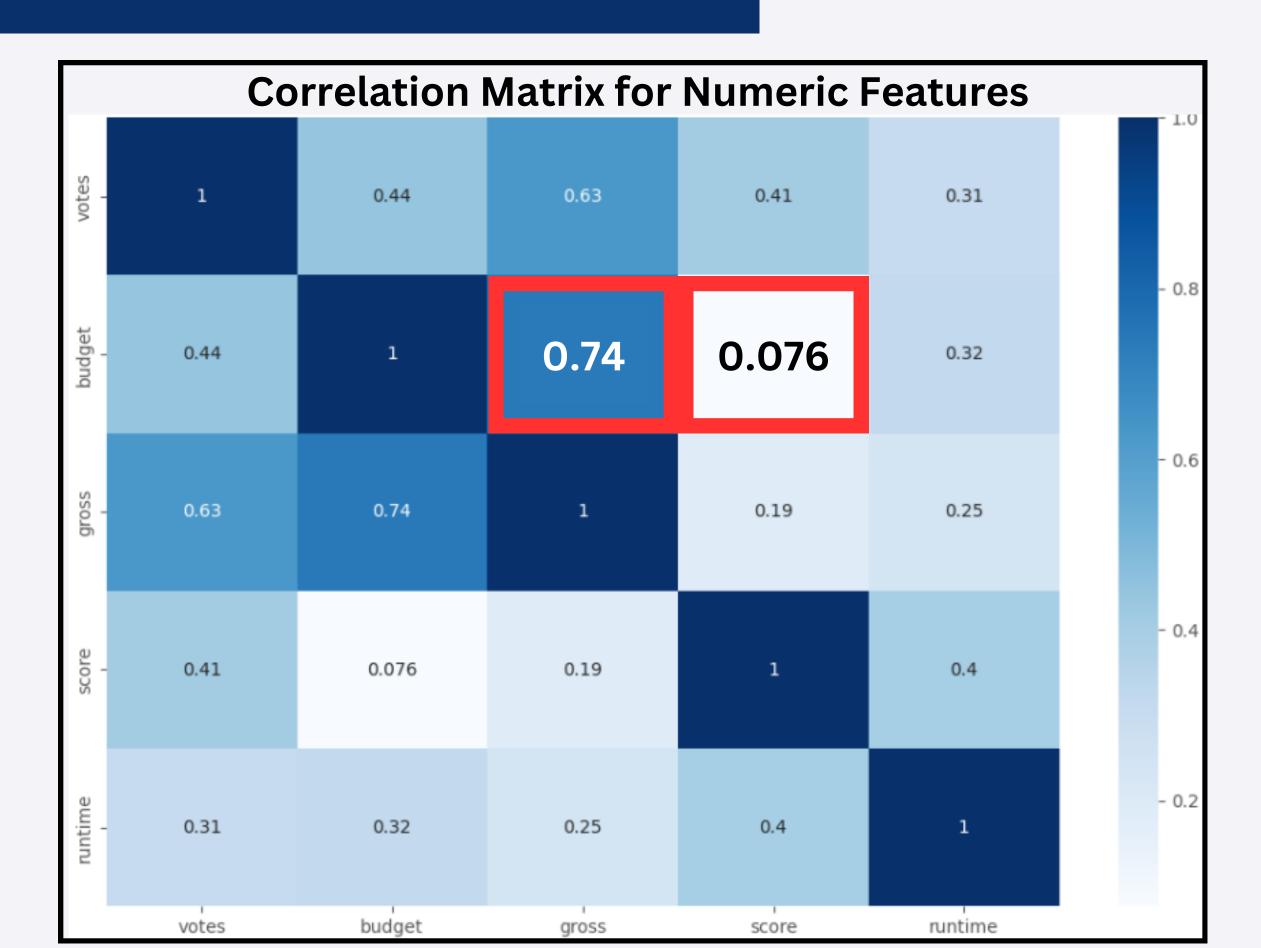
MODE 101

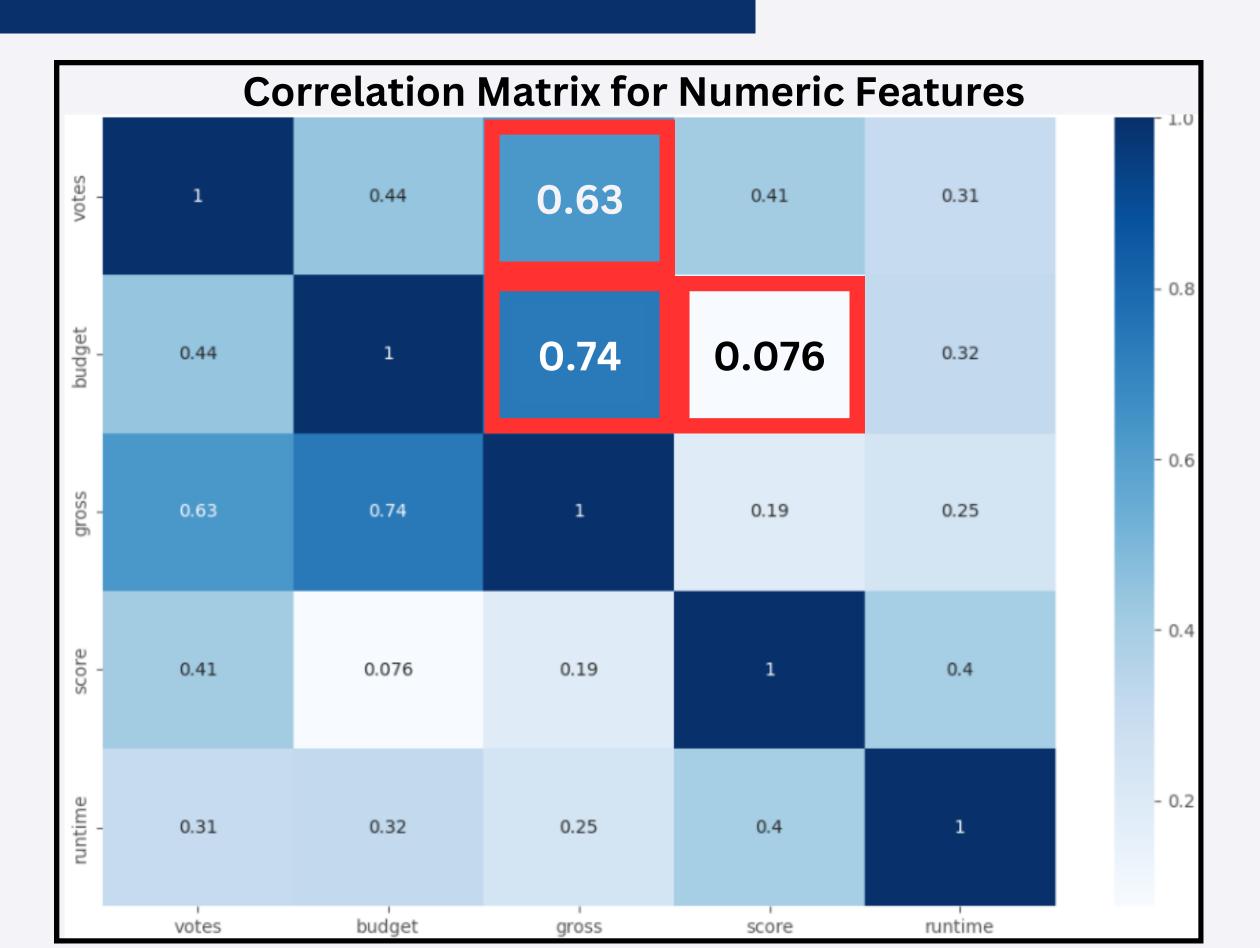
MIN 63

MAX 271

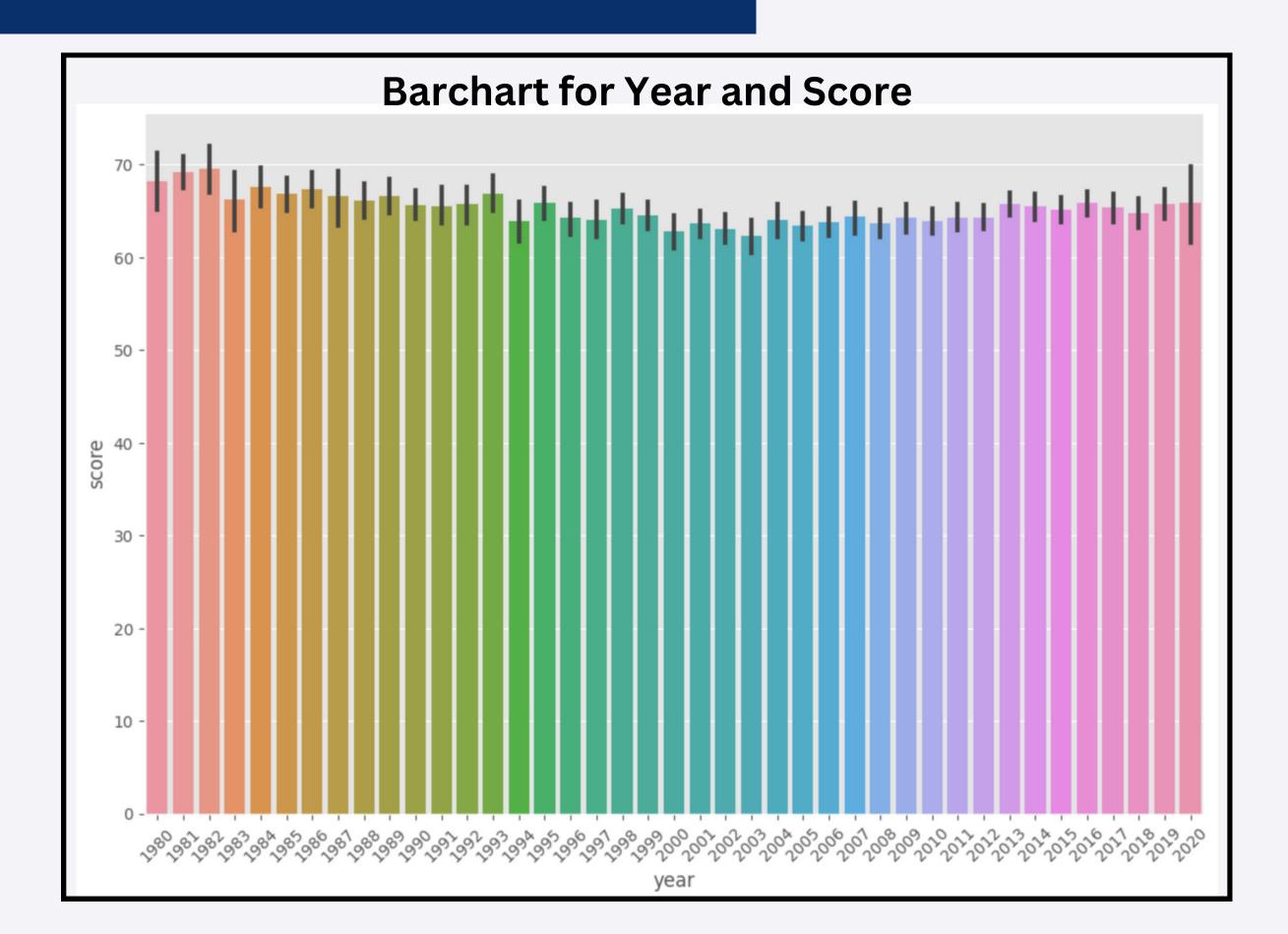




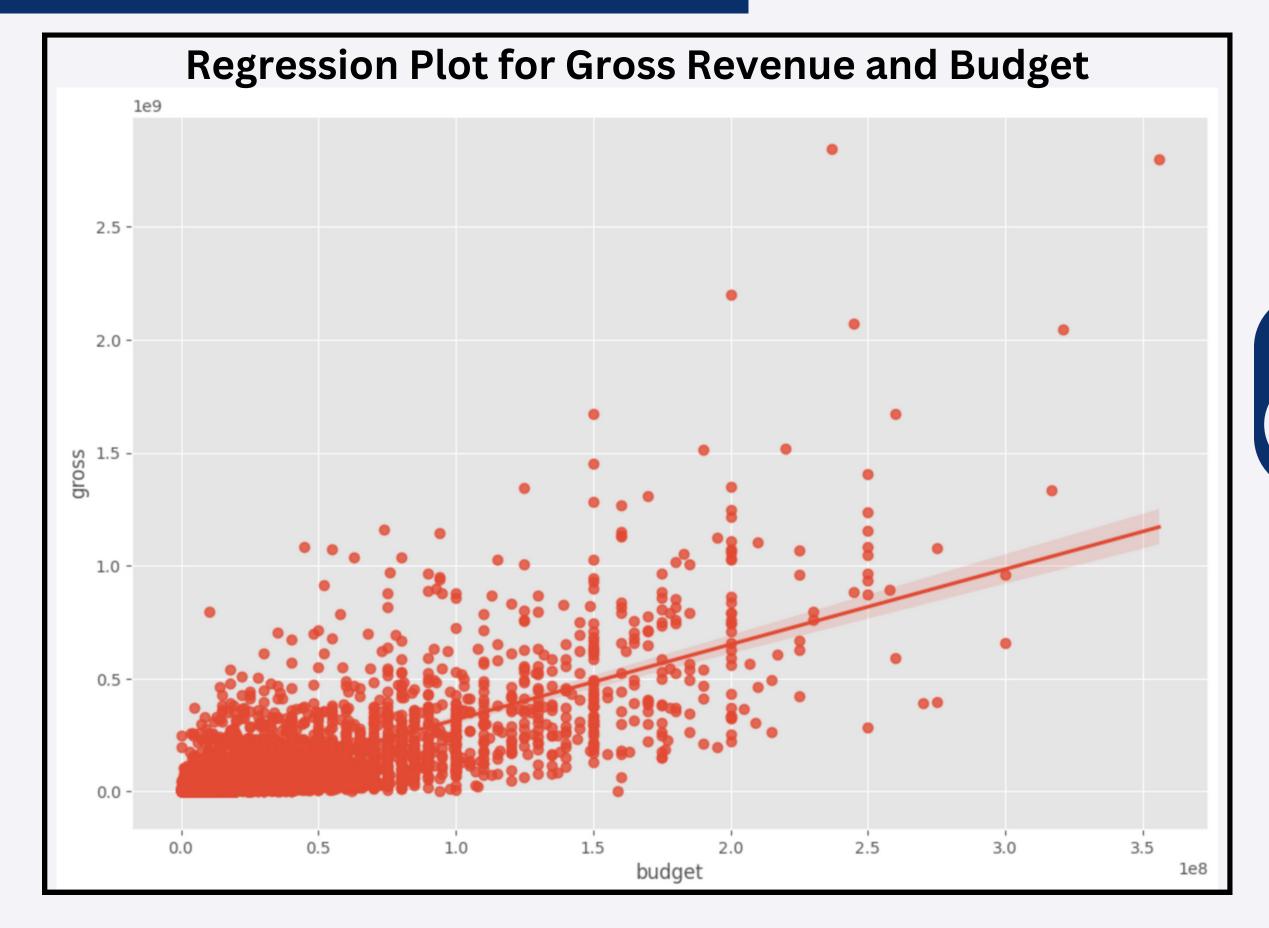




YEAR V/S SCORE



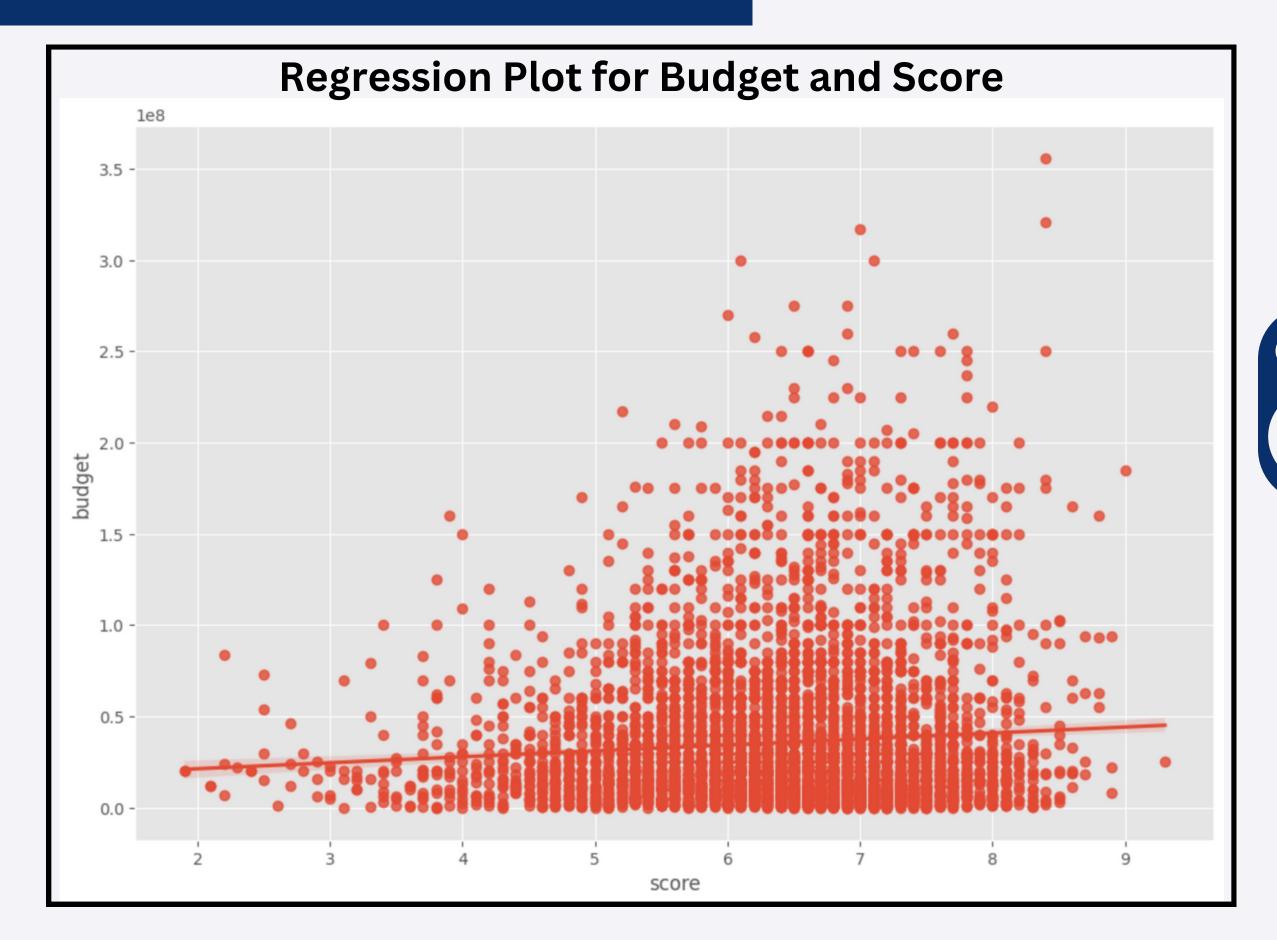
GROSS V/S BUDGET



CORRELATION

0.74

BUDGET V/S SCORE



CORRELATION
0.076

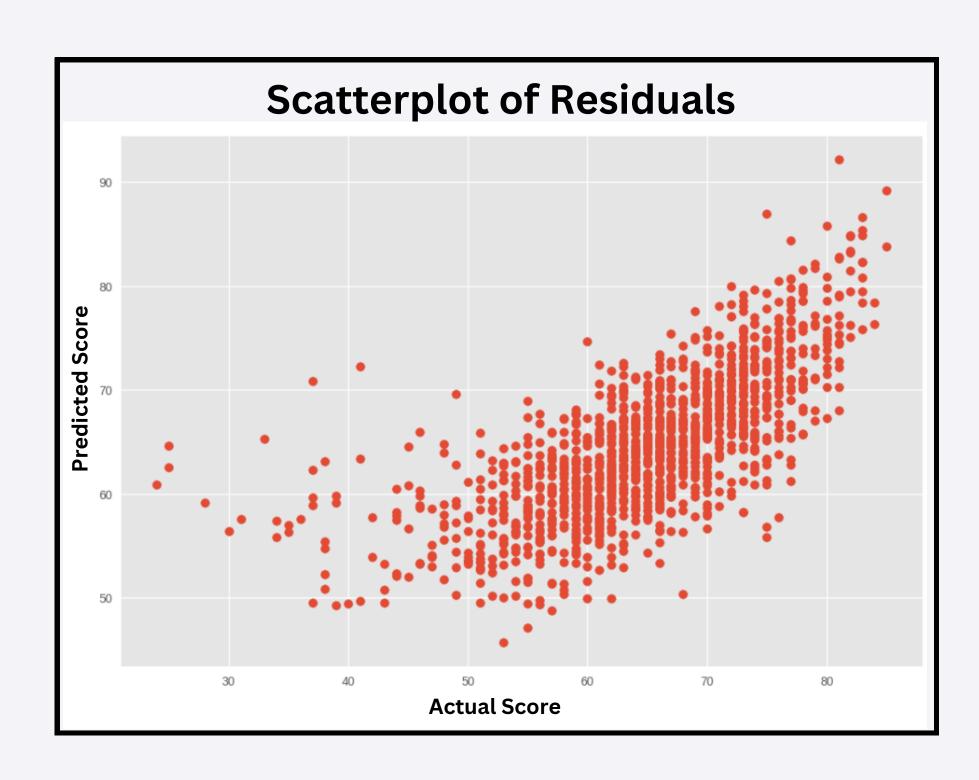
REGRESSION ANALYSIS

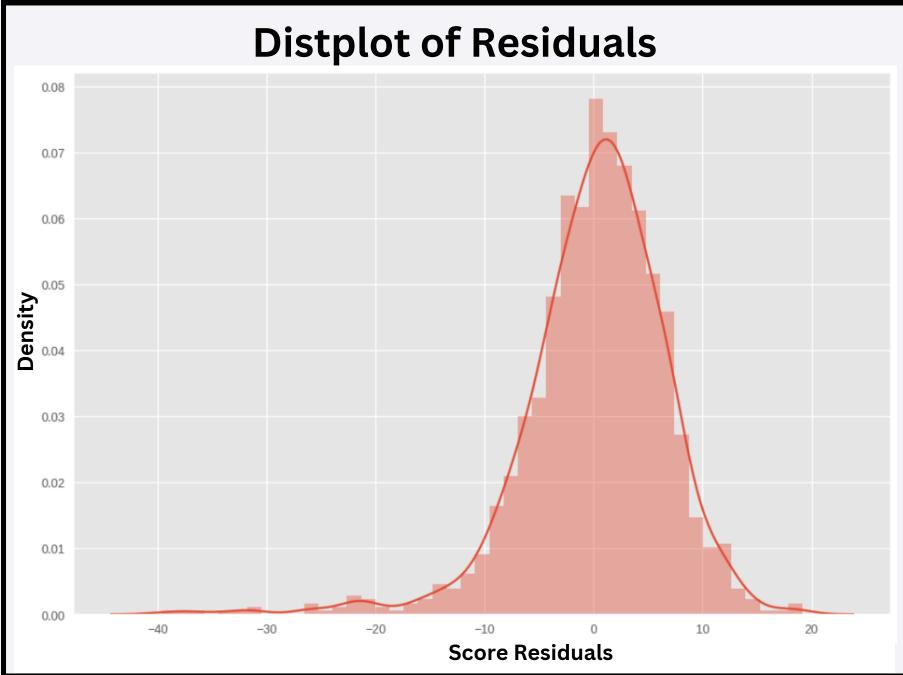
	Model	Score
2	Linear Regression 3	57.68
1	Linear Regression 2	42.91
0	Linear Regression 1	38.59

- Country Dummies
- Scaled Data
- Holidays
- Famous Writer, Director, Star
- Genre Dummies
- Numeric Variables

p-values were all under 0.005

RESIDUAL PLOTS





FINDINGS & RECOMMENDATIONS

FINDINGS

- 1 Budget ≠ Score
- 2 Marvel Studios

RECOMMENDATIONS

- 1 USA or Not?
- 2 Feedback is Key
- **Best Genres**

LEARNINGS

Bias is closer to a project than we realize.

As for us with the year example.

Collaboration is a must know skill.

Data can be normalized which makes for better regression analysis.