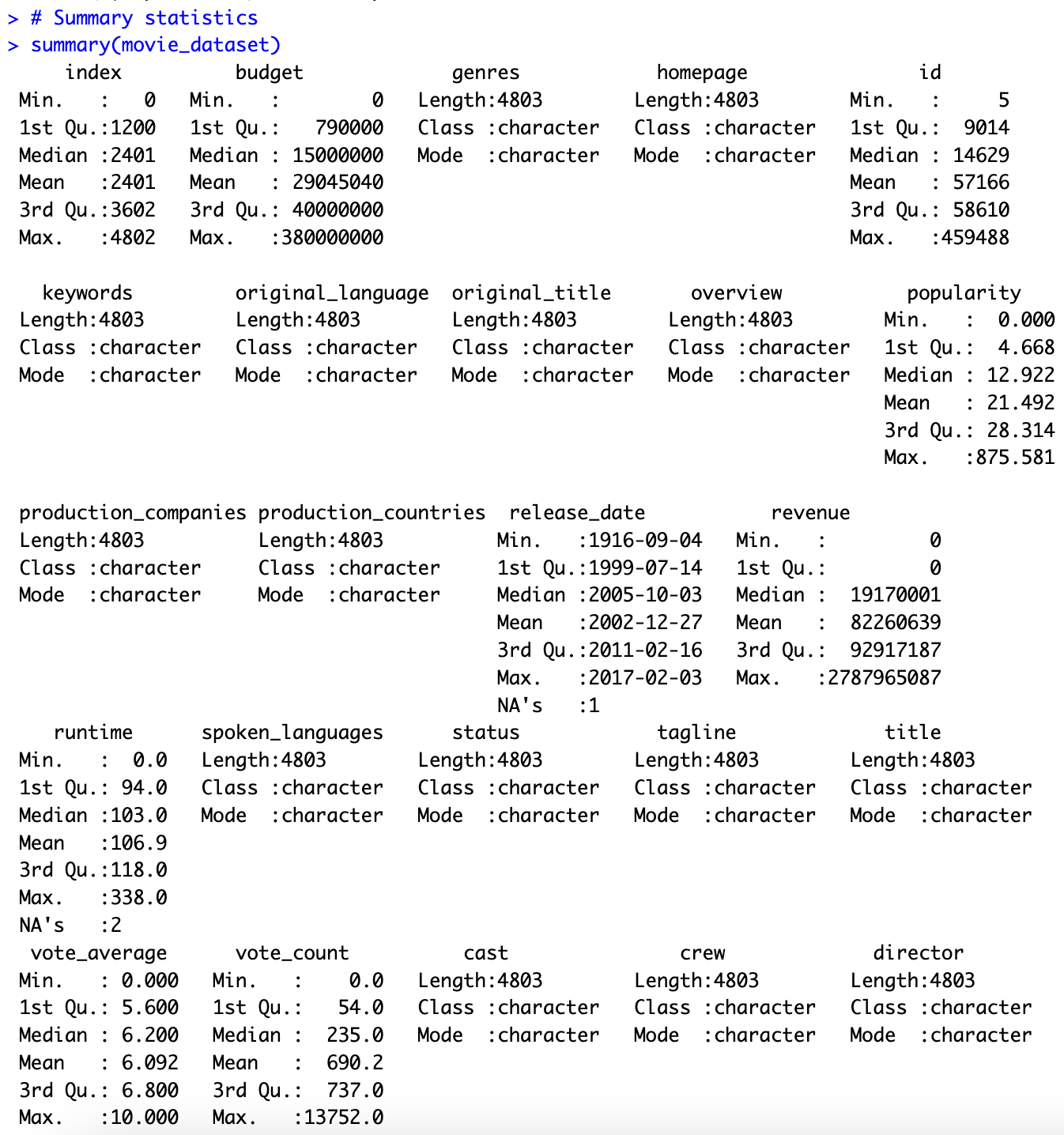
Final Project: Proposal/Dataset Selection   
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ALY6015: Intermediate Analytics  
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INTRODUCTION

The entertainment industry, particularly the movie sector, is characterized by the considerable financial investments and uncertainties over return on those investments. This preliminary analysis looks at the dataset of movies to understand the relationship between the budgets and revenue. Utilizing statistical techniques such as descriptive statistics, correlation analysis and regression modeling, we hope to derive the fundamental aspects of what makes a successful movie.

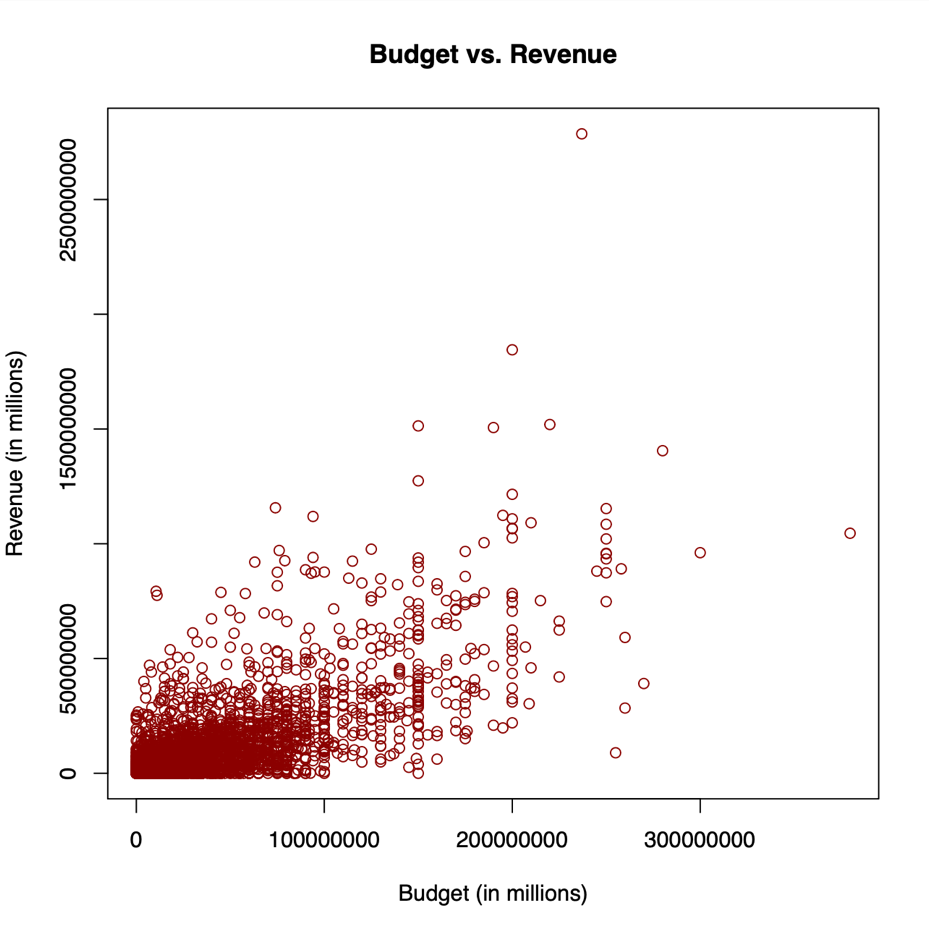
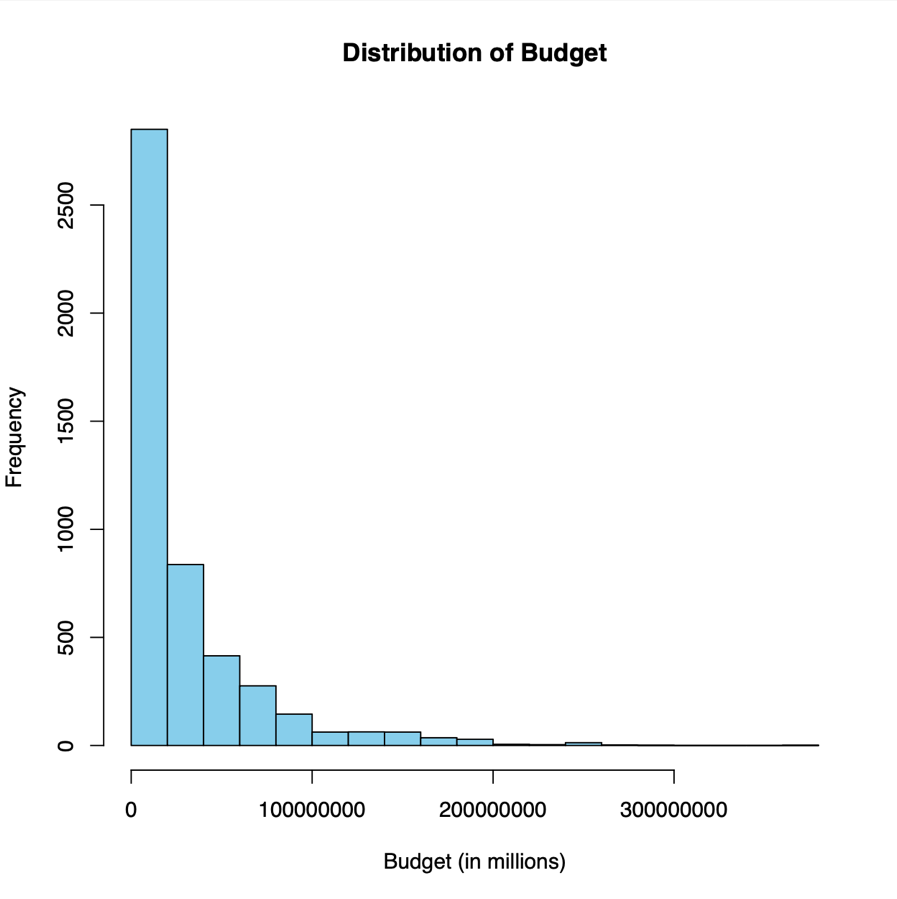
DATASET OVERVIEW

Initial exploration of the dataset revealed a dataset with a variety of movie budgets, popularity, revenue, and other characteristics. The summary statistics provided insight that the budget and revenues of movies range widely and as a result there will be that much variability within our dataset. For instance, the median budget was at $15 million, and the median revenue was approximately at $19.17 million. This sort of variability underscores the need for further analysis to understand the key drivers of movie success.

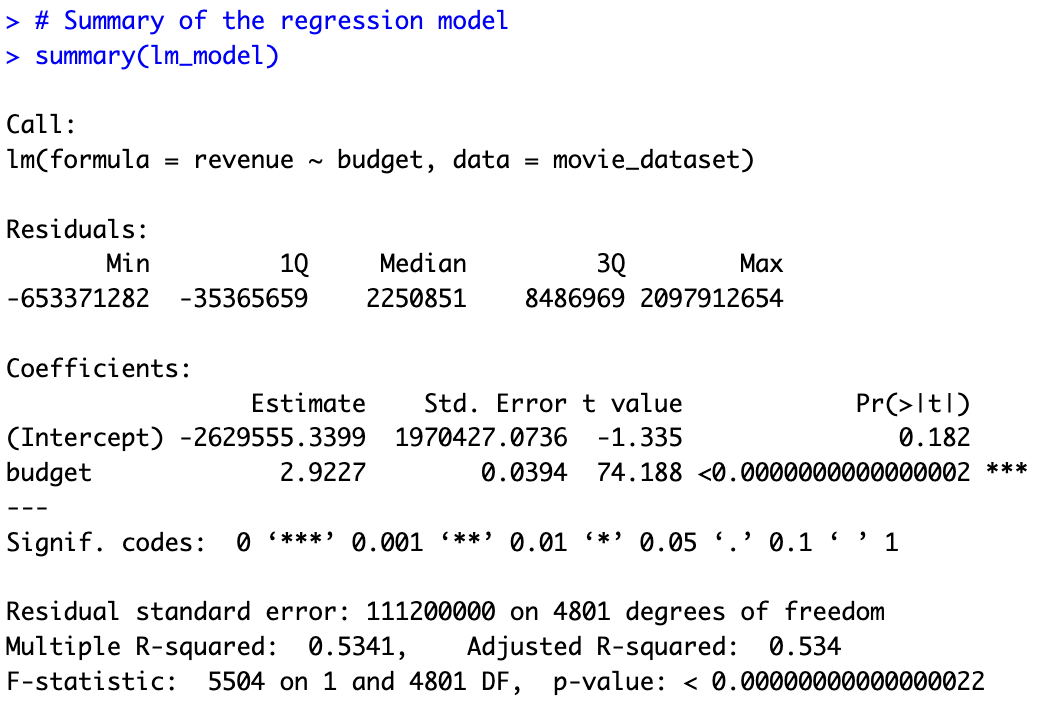


PRELIMINARY ANALYSIS

We started our preliminary analysis of the relationship between movie budget and revenue by examining the Pearson correlation coefficient. Visualization, through histograms and scatter plots, to understand the distribution and relationship between budgets and revenues, helped us get a feel for the data.



A linear regression model was fitted to assess the relationship between budget and revenue. The regression analysis indicated a statistically significant positive association between the two variables. The positive correlation between budget and revenue indicates that higher investments in production typically result in higher returns. For each additional unit of currency spent on the budget, revenue increased by approximately 2.9227 units of currency.



It is important to note, however, that this study naturally has its limitations, and one of them might be the inclusion of factors such as genre, release date, and marketing strategies which could have a significant influence on a movie's success.

For the final submission, we will expand our investigation by including additional variables, such as genre, release date, and marketing expenditure. These will enable us to build out our regression model and account more fully for multiple factors that influence a movie's revenue. Furthermore, an in-depth analysis would include the chi-square test, ANOVA, and other advanced statistical methodologies, which will allow us to gain insights into relationships present within the dataset.

CONCLUSION

Our preliminary analysis indicates that budget plays a significant role in determining movie revenue. While it’s true that higher investments are generally aimed at higher returns, there are several other success factors for filmmakers and industry stakeholders to consider. By conducting further analysis and utilizing more advanced statistical techniques, we can identify actionable insights that can help decision-making in the dynamic and hyper-competitive movie business.

REFERNCES

Bluman, A. (2018). Elementary statistics: A step by step approach (10th ed.). McGraw Hill.

Kabacoff, R. I. (2022). R in action: Data analysis and graphics with R and tidyverse (3rd ed.).

Manning Publications.

Northeastern University – Canvas – (Panopto) videos by Prof. Thomas Goulding

APPENDIX

# Load necessary libraries

library(readr)

library(ggplot2)

# Load the dataset

movie\_dataset <- read\_csv("/Users/m.joubert/Documents/Final Project: Initial Analysis Report - Joubert/movie\_dataset.csv")

# Explore the structure of the dataset

str(movie\_dataset)

# Summary statistics

summary(movie\_dataset)

# Histogram of budget

hist(movie\_dataset$budget, breaks = 20, col = "skyblue", main = "Distribution of Budget", xlab = "Budget (in millions)")

# Scatter plot of budget vs. revenue

plot(movie\_dataset$budget, movie\_dataset$revenue,

main = "Budget vs. Revenue",

xlab = "Budget (in millions)",

ylab = "Revenue (in millions)",

col = "darkred")

# Fit linear regression model

lm\_model <- lm(revenue ~ budget, data = movie\_dataset)

# Summary of the regression model

summary(lm\_model)