

## **PROJECT DESCRIPTION**

IMDb stands for the Internet Movie Database. It is an online database that was founded in 1990 and is now one of the most comprehensive sources of information related to movies, TV shows, actors, directors, and other entertainment content.

I was given a dataset with information related to movies from 1916 to 2016. My task was to clean the data and analyze the data related to genre,

budget, movie duration, director, and language.



### **APPROACH**

STEP 1: I cleaned the data that was provided to me.

**STEP 2:** Analyzed the cleaned data.



**STEP 3:** Gained insights from the analysis.



**STEP 4:** Created charts and graphs for those insights.

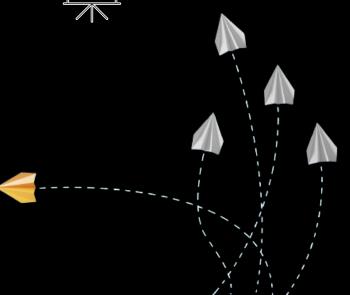


**STEP 5:** Created a presentation to outsource my information.



**STEP 6:** Presented my presentation.





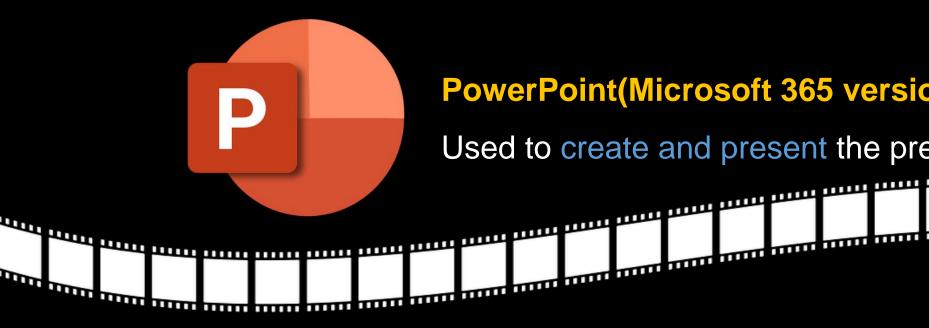
## **TECH-STACK USED**





### **Excel(Microsoft 365 version)**

Used to analyze the dataset provided to me.



### PowerPoint(Microsoft 365 version)

Used to create and present the presentation.

### **INSIGHTS**

### A. Movie genre analysis:

Determine the most common genres of movies in the dataset.

### **B.** Movie duration analysis:

Analyze the distribution of movie durations and its impact on the IMDb scores.

### C. Language analysis:

Determine the most common languages used in movies and analyze the impact on IMDb scores.

### D. Director analysis:

Identify the top directors based on their average IMDb scores.

### E. Budget analysis:

Analyze the correlation between movie budgets and gross earnings and identify the movies with the highest profit margin.

## A. MOVIE GENRE ANALYSIS

#### STEPS PERFORMED

- Firstly, I extracted movie\_title, imdb\_scores, and genres columns from the cleaned data.
- Then I separated genres into different columns using text to columns command and identified all the unique genres. There were a total of 24 genres.
- Then I created a table with having movie name and all 24 genres as my columns. For each movie, if the genre is present among 24 then it will be 1 else it will be blank.
- After that using the COUNTIFS function I created a table having information about how many movies were created for a genre and IMDb scores.
- This table helped me to gain the descriptive statistics i.e., mean, median, mode, max,
   min, variance, and standard deviation of the IMDb scores.





### **INSIGHTS**

- There are a total of 24 genres.
- Drama, Thriller, action, comedy, and romance are the top 5 genres that are performing well.
- Genre drama has the maximum count of movies (128 movies) for a 7.2 IMDb score.
- Drama, Comedy, and Thriller have an average count of 17-33 movies based on their IMDb scores.
- Drama has comparatively greater variance and standard deviation compared to any other genre.

#### **Unique genres**

Drama

Comedy

Thriller

Action

Romance

Adventure

Crime

Sci-Fi

Fantasy

Horror

Family

Mystery

Biography

Animation

Music

War

History

Sport

Musical

Documentary

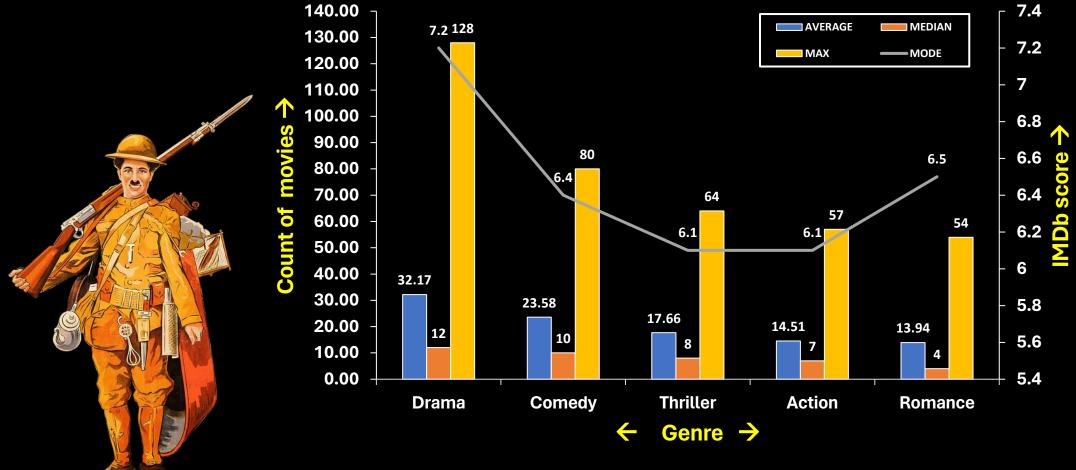
Western

Film-Noir

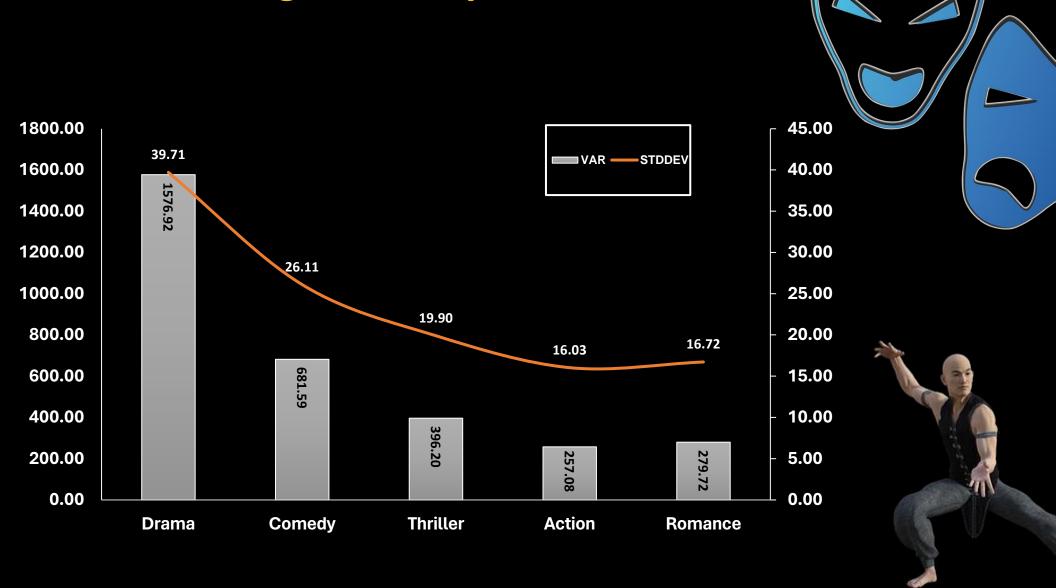
Short

News









## B. MOVIE DURATION ANALYSIS

## **INSIGHTS: Movie duration analysis**

#### **STEPS PERFORMED**

- At first, I extracted movie\_title, IMDb\_score and duration from the cleaned data.
- Then I created class intervals for the duration column which I used to show the relationship with IMDb scores.
- I created descriptive statistics for the duration column using the Data analysis command in the Analysis group which is in the data tab.
- Then I created a separate table using class interval and IMDb scores which denotes how many movies were created within that duration and what is their IMDb score.





## **INSIGHTS: Movie duration analysis**

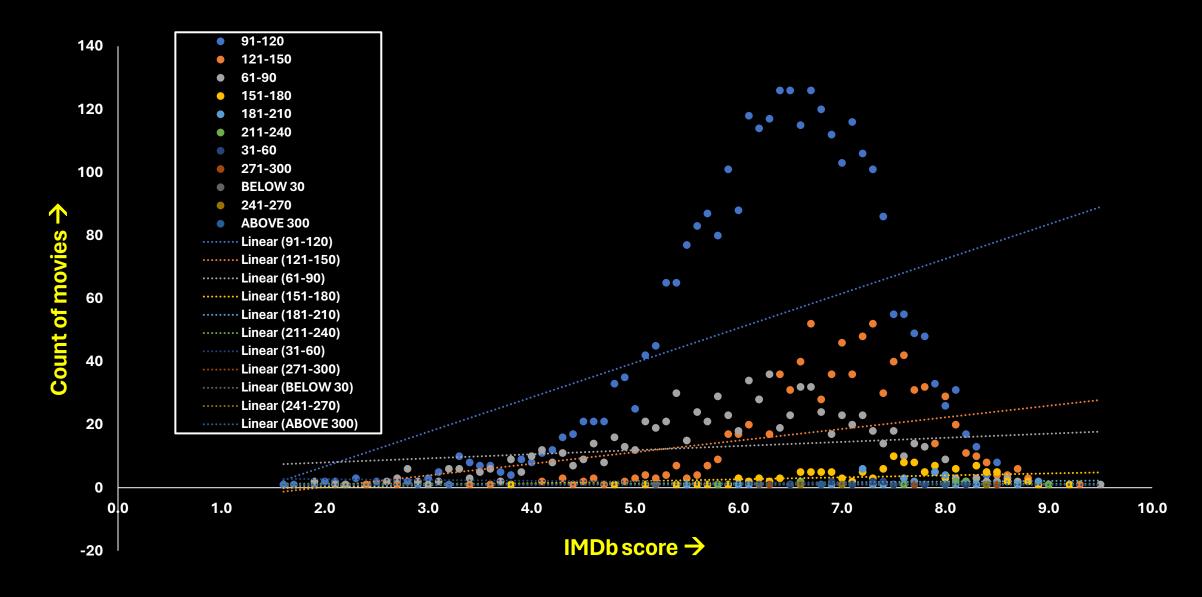
### **INSIGHTS**

- Movies that run for 91-120 minutes have the highest number of films and tend to receive an IMDb score between 5.5 and 7.5.
- Most of the movies are created with a duration of 61-120 minutes.
- Shortest duration of a movie is 7 mins, and the longest duration is 330 mins.





## **INSIGHTS: Movie duration analysis**

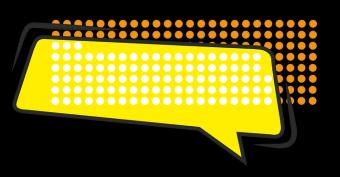


## C. LANGUAGE ANALYSIS

### **STEPS PERFORMED**

- At first, I extracted movie\_title, Language, and IMDb score columns from the cleaned data.
- Using the data from the table, I created a pivot table to analyze the number of movies produced in each language and their corresponding IMDb scores.
- I created one more table that denotes the number of movies produced in each language.
- Then using the pivot table, I created a table that gives the mean, mode, median, max, min, variance, and standard deviation for each language.





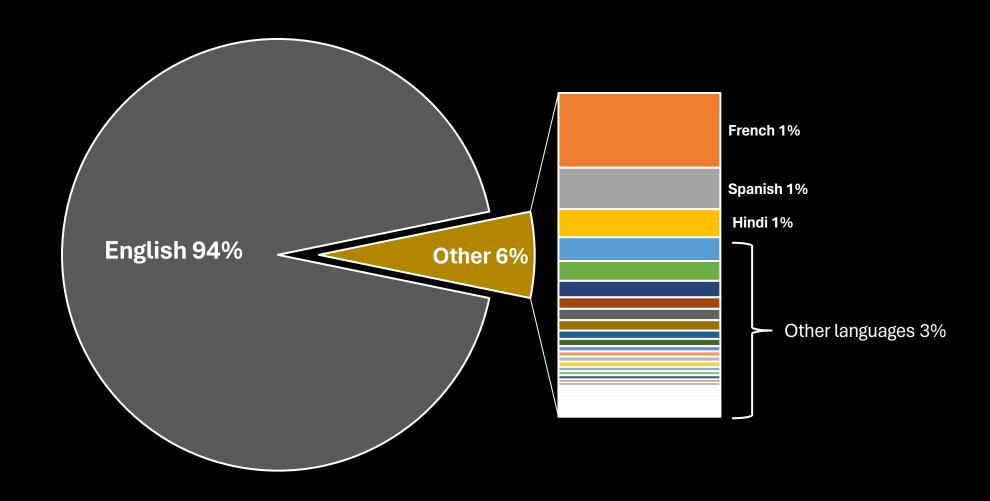
### **INSIGHTS**

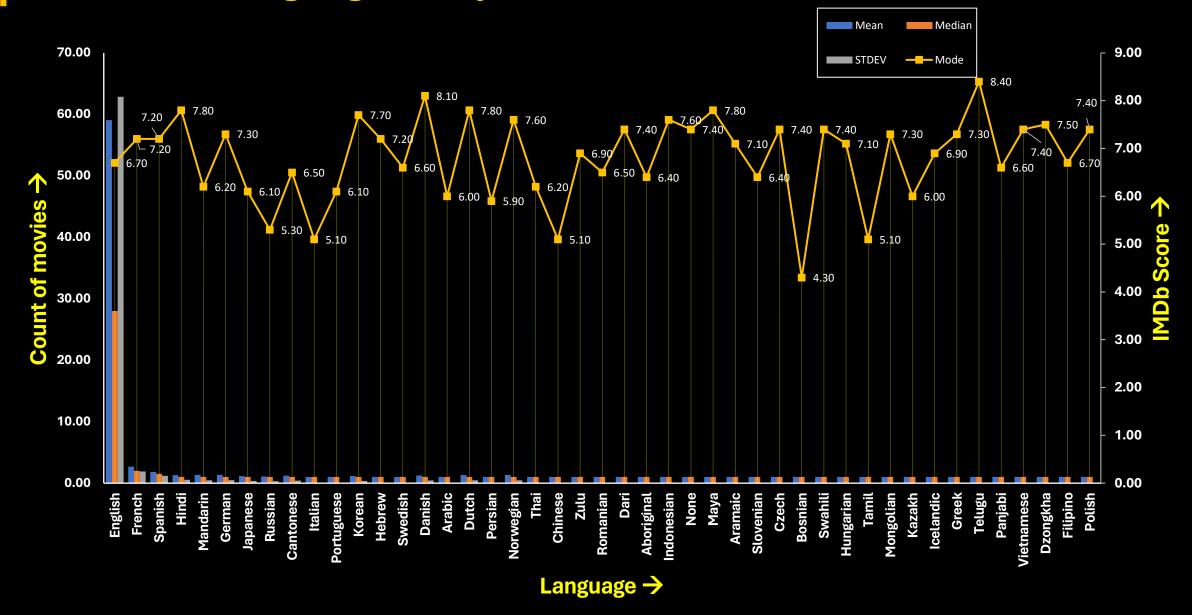
- 94% are English movies.
- Mean, median, and standard deviation of English movies are much greater than other movies.
- 207 English movies have an average IMDb score of 6.7.



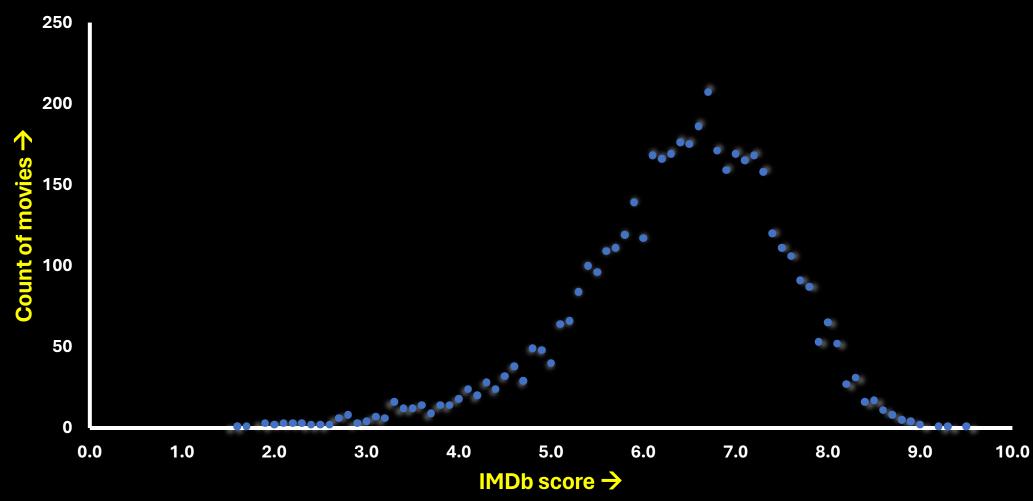


#### Distribution of movies based on language









# D. DIRECTOR ANALYSIS

### STEPS PERFORMED

- At first, I extracted movie\_title, director\_name, and imdb\_score columns from the cleaned data.
- Using the data from the table, I created a pivot table that tells us how many movies were directed by each director and what is their average IMDb score.
- From that table I was able to know the percentile of each director based on their IMDb score.
- Using the data, I extracted the directors whose percentile is equal to or greater than 99%.





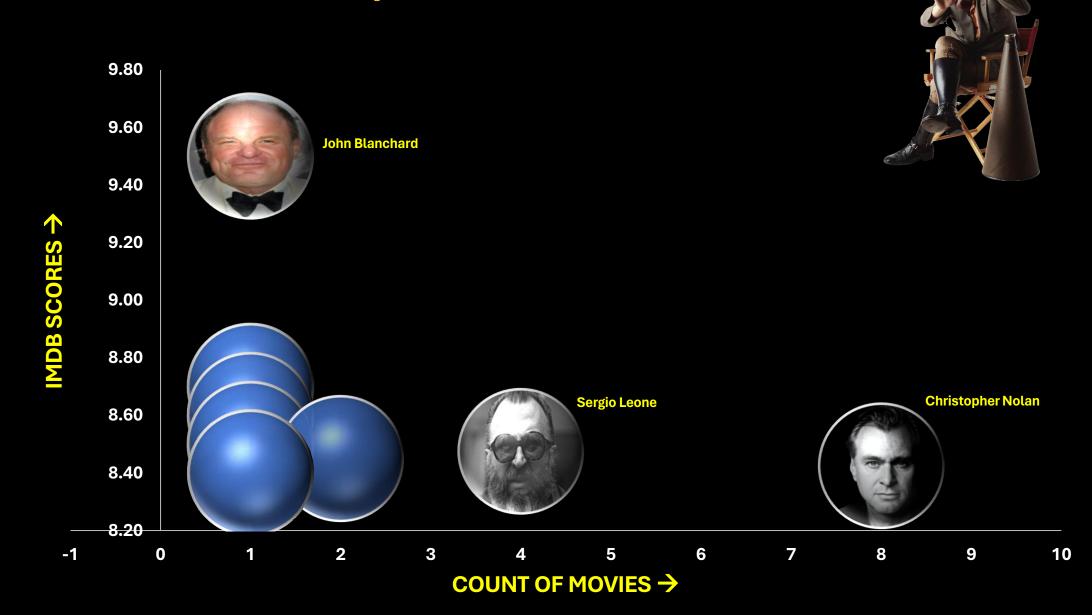
### **INSIGHTS**

- John Blanchard who has directed Towering Inferno which has an IMDb rating of 9.5 has a percentile of 100%.
- Christoper Nolan's has directed a total of 8 movies and its average IMDb score is 8.43 and has a percentile of 99.40%.
- There are a total of 23 directors whose percentile is equal to or greater than the 99% percentile.



Row Labels	Count of movie_title	Average of imdb_score	Percentile
John Blanchard	1	9.50	100.00%
Sadyk Sher-Niyaz	1	8.70	99.80%
Mitchell Altieri	1	8.70	99.80%
Cary Bell	1	8.70	99.80%
Mike Mayhall	1	8.60	99.70%
Charles Chaplin	1	8.60	99.70%
Ron Fricke	1	8.50	99.60%
Raja Menon	1	8.50	99.60%
Majid Majidi	1	8.50	99.60%
Damien Chazelle	1	8.50	99.60%
Sergio Leone	4	8.48	99.50%
Tony Kaye	2	8.45	99.50%
Christopher Nolan	8	8.43	99.40%
S.S. Rajamouli	1	8.40	99.00%
Rakeysh Omprakash Mehra	1	8.40	99.00%
Richard Marquand	1	8.40	99.00%
Robert Mulligan	1	8.40	99.00%
Moustapha Akkad	1	8.40	99.00%
Marius A. Markevicius	1	8.40	99.00%
Jay Oliva	1	8.40	99.00%
Catherine Owens	1	8.40	99.00%
Asghar Farhadi	1	8.40	99.00%
Bill Melendez	1	8.40	99.00%





## E. BUDGET ANALYSIS

## **INSIGHTS: Budget analysis**

### **STEPS PERFORMED**

- At first, I extracted movie\_title, gross, and budget columns from the cleaned data.
- Then I created a column for the profit margin on each movie.
- I also created a profit margin % column which tells the percentage increase of gross with respect to its budget.
- Then by using the CORREL function I found the correlation coefficient between gross and budget.



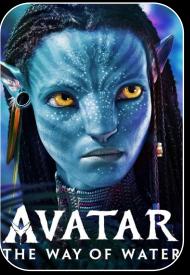
## **INSIGHTS: Budget analysis**

### **INSIGHTS**

- The correlation coefficient of gross and budget is 0.100969202. It means that they have a
  weak positive correlation which implies that if the budget increases it does not mean that the
  gross will also increase.
- Avatar movie has the highest profit margin of \$523 Million(approx.).
- Paranormal Activity movie has the highest profit margin percentage of 719349% because it has

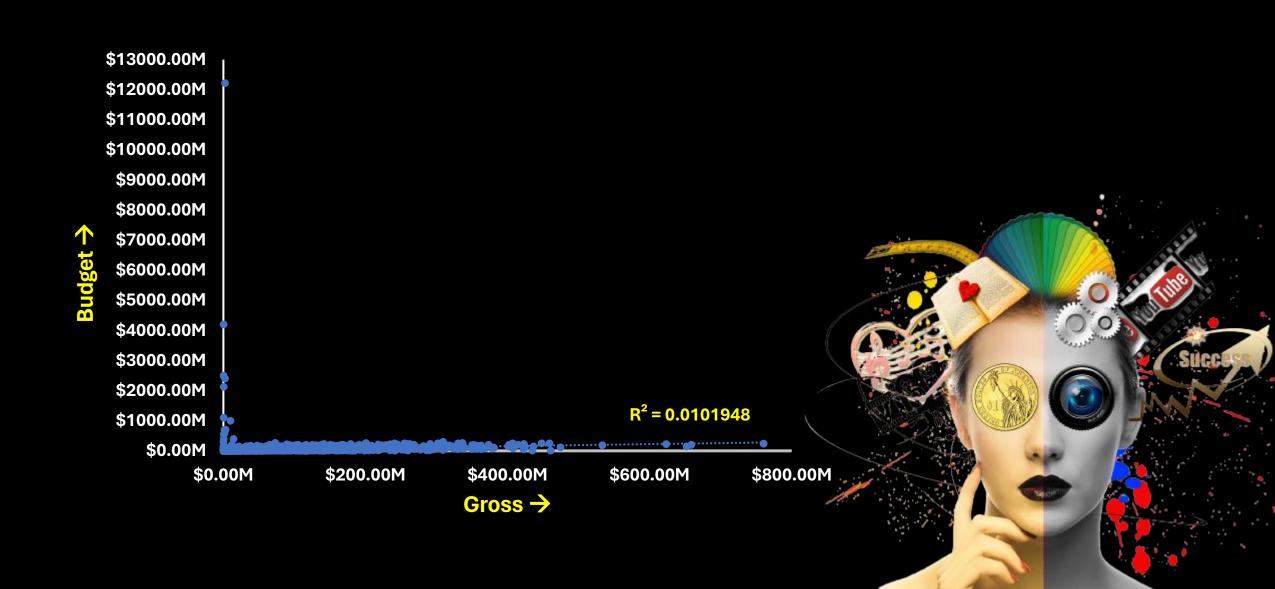
a budget of \$15000 but the gross reached \$107 Million(approx.).







## **INSIGHTS: Budget analysis**





**Excel Link:** <u>click here</u>