IMDB Movie Analysis

1.1 PROJECT DESCRIPTION

The primary goal of the project is to analyze the factors that help in the success of a movie. Various features related to the making, release, and production of movies are considered to understand the underlying patterns between them. Multiple valuable and data-driven insights are generated and visualized to further understand the dynamics of this field.

1.2 APPROACH

The dataset is available in .csv and .xlsx format. It is first cleaned, rows with missing values are deleted, and duplicates are removed. All unnecessary features that would not be required for analysis are deleted for simplification of the database. Out of **28 features**, **19 features** are **deleted**. The analysis is done based on a cleaned dataset with **9 features** and **3786 rows**.

1.3 TECH STACK USED

MS Excel is used for data storage, data processing, manipulation, and visualization. MS Word is used to display actionable insights in an easy-to-understand format, that is, the following report.

1.4 INSIGHTS

1.4.1 Movie Genre Analysis

 Determine the most common genres of movies in the dataset. Then, for each genre, calculate descriptive statistics (mean, median, mode, range, variance, standard deviation) of the IMDB scores.

Approach:

- A column called **Genres** is created using the following formula:
 - =UNIQUE('Cleaned Db'!D1:D3786)
- A column called Count is created using the following:
 - =COUNTIF('Cleaned Db'!12:13786,'Cleaned Db'!12)
- Both columns are sorted based on Count from Largest to Smallest value.
- Create columns called Mean, Medians, Mode, Max, Min, Variance, Standard Deviation using the following formulae:
 - =AVERAGE(IF('Cleaned Db'!\$D\$2:\$D\$3786='Movie Genre Analysis'!\$K2,'Cleaned Db'!\$1\$2:\$1\$3786))
 - =MEDIAN(IF('Cleaned Db'!\$D\$2:\$D\$3786='Movie Genre Analysis'!\$K2,'Cleaned Db'!\$|\$2:\$|\$3786))
 - =MODE(IF('Cleaned Db'!\$D\$2:\$D\$3786='Movie Genre Analysis'!\$K2,'Cleaned Db'!\$I\$2:\$I\$3786))
 - =MAX(IF('Cleaned Db'!\$D\$2:\$D\$3786='Movie Genre Analysis'!\$K2,'Cleaned Db'!\$I\$2:\$I\$3786))
 - =MIN(IF('Cleaned Db'!\$D\$2:\$D\$3786='Movie Genre Analysis'!\$K2,'Cleaned Db'!\$1\$2:\$1\$3786))
 - =VAR.S(IF('Cleaned Db'!\$D\$2:\$D\$3786='Movie Genre Analysis'!\$K2,'Cleaned Db'!\$1\$2:\$1\$3786))
 - =STDEV.S(IF('Cleaned Db'!\$D\$2:\$D\$3786='Movie Genre Analysis'!\$K2,'Cleaned Db'!\$|\$2:\$|\$3786))

Output:

| Top 5 Movie Genres are as follows: | | | | | | | | |
|------------------------------------|-------|----------|--------|------|-----|-----|----------|--------------------|
| Genres | Count | Mean | Median | Mode | Max | Min | Variance | Standard Deviation |
| Drama | 154 | 7.040132 | 7.15 | 7.3 | 8.8 | 3.4 | 0.69116 | 0.83136048 |
| Comedy Drama Romance | 151 | 6.495302 | 6.5 | 6.5 | 8 | 4.3 | 0.555451 | 0.745285685 |
| Comedy Drama | 148 | 6.583673 | 6.7 | 6.7 | 8.8 | 3.3 | 0.7348 | 0.857204825 |
| Comedy | 147 | 5.84069 | 6 | 6.5 | 8 | 1.9 | 1.481875 | 1.217322686 |
| Comedy Romance | 136 | 5.896296 | 6 | 6.1 | 8.4 | 2.7 | 0.76827 | 0.87650999 |

Inference:

 This analysis gives an understanding of the top movie genres and it's statistical analysis based on their imdb_score.

1.4.2 Movie Duration Analysis

Analyze the distribution of movie durations and identify the relationship between movie duration and IMDB score.

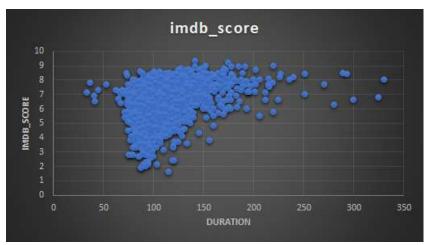
Approach:

- The columns of duration and imdb_score are considered.
- Average, Median, Standard Deviation are calculated for duration based on imdb_score using the following formulae:

- =AVERAGE(A2:A3786)
- =MEDIAN(A2:A3786)
- =STDEV.S(A2:A3786)

Output:

| Average | 109.8103 |
|--------------------|----------|
| Median | 105 |
| Standard Deviation | 22.76594 |



Inference:

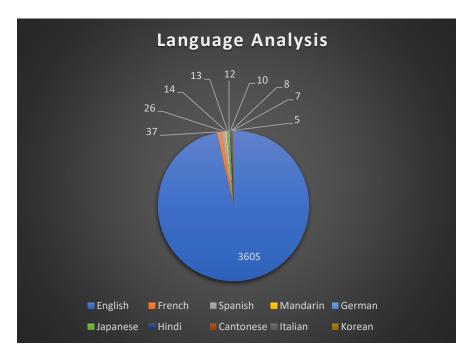
- A scatter plot is used to visualize the impact of duration on imdb_score.
- It is observed that the duration between 50 to 150 seems to be the sweet spot. A dense cluster in that area indicates that this range of duration has a wide variety of movies covering almost the entire range of available imdb_scores.

1.4.3 Language Analysis

- Determine the most common languages used in movies and analyze their impact on the IMDB score using descriptive statistics.
- Approach:
 - A column called **Language Distribution** is created using the following formula:
 - =UNIQUE('Cleaned Db'!F2:F3786)
 - A column called **Count** is created that has the count of each Language in the dataset using the following formula:
 - =COUNTIF('Cleaned Db'!F2:F3786,'Language Analysis'!I2)
 - Both columns are sorted based on Count from Largest to Smallest value.
 - The columns Mean, Median and Mode are created using the following formulae:
 - =AVERAGE(IF('Cleaned Db'!\$F\$2:\$F\$3786='Language Analysis'!\$12,'Cleaned Db'!\$I\$2:\$I\$3786))
 - =MEDIAN(IF('Cleaned Db'!\$F\$2:\$F\$3786='Language Analysis'!\$12,'Cleaned Db'!\$I\$2:\$I\$3786))
 - =STDEV.P(IF('Cleaned Db'!\$F\$2:\$F\$3786='Language Analysis'!\$12,'Cleaned Db'!\$I\$2:\$I\$3786))

Output:

| Top 10 Most Used Languages: | | | | |
|-----------------------------|-------|----------|--------|--------------------|
| Language Distribution | Count | Mean | Median | Standard Deviation |
| English | 3605 | 6.421664 | 6.5 | 1.052409957 |
| French | 37 | 7.286486 | 7.2 | 0.553691378 |
| Spanish | 26 | 7.05 | 7.15 | 0.810151933 |
| Mandarin | 14 | 7.021429 | 7.25 | 0.737930089 |
| German | 13 | 7.692308 | 7.7 | 0.615769111 |
| Japanese | 12 | 7.625 | 7.8 | 0.861321659 |
| Hindi | 10 | 6.76 | 7.05 | 1.05470375 |
| Cantonese | 8 | 7.2375 | 7.3 | 0.412121038 |
| Italian | 7 | 7.185714 | 7 | 1.069617517 |
| Korean | 5 | 7.7 | 7.7 | 0.509901951 |



o Inference:

- A pie chart is created to visualize the analysis generated.
- English is the most common language and Korean is the least common language when top 10 languages are considered.

1.4.4 Director Analysis

o Identify the top directors based on their average IMDB score and analyze their contribution to the success of movies using percentile calculations.

Approach:

- A column called **Director** is created using the following formula:
 - =UNIQUE('Cleaned Db'!A2:A3786)
- Average imdb_score is calculated and put into a column called Average using the following formula:
 - =AVERAGE(IF('Cleaned Db'!\$!\$2:\$!\$3786))

Db'!\$A\$2:\$A\$3786=A2,'Cleaned

- Calculate Percentile using the following formula:
 - =PERCENTILE.INC(F4:F13,90%)

o Output:

| Top 10 Directors: | | | |
|-------------------|----------|--|--|
| Director | Average | | |
| Vicente Amorim | 8.6 | | |
| Ronan Chapalain | 8.6 | | |
| Matthew Vaughn | 8.5 | | |
| Richard Curtis | 8.5 | | |
| Nick Cassavetes | 8.5 | | |
| Claude Miller | 8.5 | | |
| Fred Walton | 8.433333 | | |
| Andrew Adamson | 8.425 | | |
| Gil Junger | 8.4 | | |
| Niels Arden Oplev | 8.4 | | |

| Percentile | 8.6 |
|------------|-----|
|------------|-----|

Inference:

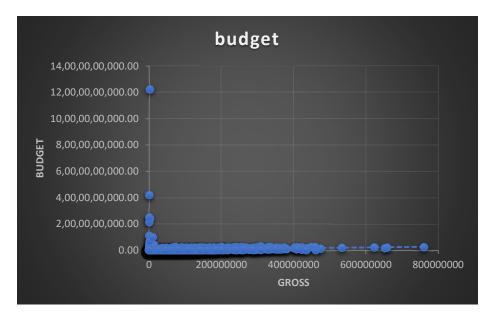
The **Percentile** gives the average score which is in the top 10% from the entire range of average **imdb_score**. Here, **8.6** is the score with percentile over 90% when compared to the overall distribution in the dataset.

1.4.5 Budget Analysis

- Analyze the correlation between movie budgets and gross earnings, and identify the movies with the highest profit margin.
- Approach:
 - The columns gross, movie_title, budget are considered.
 - **Profit Margin** is calculated using the following formulae:
 - =\$A2-\$C2
 - Correlation is calculated using the following formula:
 - =CORREL(A2:A3786,C2:C3786)
 - Maximum profit and movie that has that profit are calculated using the following formulae:
 - =MAX(D2:D3786)
 - =INDEX(B2:B3786, MATCH(1,IF(D2:D3786=F7, 1),0))
 - A Scatter Plot is used to visualize gross and budget as well as gross and profit margin.

Output:

| Max Profit | | Movie Name | | |
|------------|-----------|------------|--|--|
| | 523505847 | AvatarÂ | | |





o Inference:

Since the correlation constant when calculated for gross and budget is closer to 1, therefore the relationship is considered to be strong and positively linear. The budget graph compares budget with gross. It can be concluded that budget linearly and positively affects the gross income generated. In the next graph, profit margin is divided in bins for visualization.

1.5 RESULTS

The analysis of the current dataset helps in understanding the movie industry and how various factors affect how a movie does in the market. The results are as follows:

- Drama, comedy, romance and combinations of these genres are very popular among the audience.
- Movies with the duration of 110 minutes on an average are preferred.
- Maximum movies are in English, French and Spanish.
- The score is considered to be more than that of 90 percentage of movies or in other words imdb score is above 90 percentile when it is 8.6.
- Budget linearly affects the gross income in the positive direction.

1.6 HYPERLINK OF THE EXCEL SHEET

 $\frac{https://docs.google.com/spreadsheets/d/19HRj4SkBIDOs9IjDdU5jWASNumdj0Yl0/edit?usp=sharing \\ \& ouid=105545149670713438068 \\ \& rtpof=true \\ \& sd=true$