



# MOVIE ANALYSIS

OPEN DATA 1916-2017

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## EXECUTIVE SUMMARY

### CONTEXT:

This data contains about 5000 movies from year 1916 to 2017 and was generated from the movie database API by using TMDb API.

There are total of 4803 observations in this dataset and includes at least three continues columns which are popularity, release\_date and runtime. Also, it includes at least 3 classification columns like country, status, and language.

### CONTENT:

All the data comes from real movie data, and it doesn't contain any columns that need to be anonymized due any ethic concerns in future.

### OBJECTIVE:

Would track the most popular movies through the world and what caused those movies are famous based on the profit gained.

Also, find out any correlations between famous movies and unfamous movies.

**DATA SOURCE:** [Movies Dataset](#) | [Kaggle](#)

### LIMITATIONS AND ETHICS:

- This dataset only contains 5000 movies may cause rank error due to the limitation.
- This dataset was considered as not never update frequency would be caused timeliness problem.

### DATA WRANGLIN AND CONSISTENCY CHECKING

- Data wrangling check:
  - ✓ Dropped 8 columns won't be used in analysis and keeps another 14 columns.
  - ✓ Columns names are all easy understanding.
- Data Consistency Check:
  - ✓ Mixed-type columns check, column country, genres, overview, and language contain nan values considered as mixed-type column.
  - ✓ Nan values checked, found 181 nan values in country column and 28 nan values in genres column and 132 nan values in language column.
  - ✓ Didn't do anything with those nan values as they will be ok for my analysis
  - ✓ No duplicates found in this dataset.

## DATA PROFILE

| Column Name  | Description                                 | Data type                | Time Variant |
|--------------|---|--------------------------|--------------|
| budget       | The total spending to cast the movie        | Quantitative, continuous | No           |
| genres       | What kind movie is defined as.              | Qualitative, nominal     | No           |
| id           | Movie unique number                         | Quantitative, discrete   | No           |
| overview     | Customers' experiencing about the movie     | Qualitative, nominal     | No           |
| popularity   | The index of how famous bout the movie      | Quantitative, continuous | No           |
| country      | Stay days in weekend                        | Quantitative, continuous | No           |
| Release_date | The date that the movie was released.       | Quantitative, continuous | Yes          |
| revenue      | Total gains from the movie                  | Quantitative, continuous | No           |
| runtime      | The length of the movie                     | Quantitative, continuous | No           |
| language     | Movie's spoken language                     | Qualitative, nominal     | No           |
| status       | Whether the movie was released or not       | Qualitative, binary      | No           |
| title        | Movie's name                                | Qualitative, nominal     | No           |
| Vote_average | The average number of movies that was voted | Quantitative, discrete   | No           |
| Vote_count   | The total number of movies that was voted   | Quantitative, discrete   | No           |

### QUESTIONS TO DIG OUT:

- How can we define a movie as popularity? Are there any correlation actors among the observations?
- Which region contributes the most revenues and what are the profits? Does the top revenue indicate the top profit?
- Does budget can be connected to a movie length, like high budget means long length?
- Which year contributes the most revenue and which year contain the most popularity movie? Where are these movies come from?