## **Project Summary**

**Design :** This project was designed to develop a linear regression model that predicts revenue of movies using Budget, running\_time, genre,rating, genre.

**Data:** Data used for this project was scrapped from Boxoffice mojo.com and IMDB.com. The data is taken for movies released between 2000 to 2019.

## Algorithm:

1. Data scraping

From Boxoffice mojo - movie-title, domestic\_gross, Total\_gross, budget, runing\_time\_minutes, genres

From IMBD.com - Director, producer and cinematographer

- 2. EDA
- 2.1 Data cleaning got rid of nan-values
- 2.2 Analyze the relationship between variables and the target using correlation map
- Features selection and tuning After checking on the frequency of occurrences of the values in the director, cinematographer and producers, they don't have significant predictive value for the target therefore, they were dropped.
- -Adjust the categorical variables distributor, ratings and genre by putting the unpopular or low count values into one category i.e others for each feature.
- -Converted the categorical variables Distributor, Rating and genre into dummy variables.
  - 3. Made a baseline model linear regression model, fit and evaluate
  - 4. Adjust the data further by getting rid of outliers
    - Movies that made more than 1,000,000,000 and less than 10,000,000 were dropped
    - The target variable was adjusted using log transformation to get rid off the high positive skewness it had
    - More variables were used to develop the model further and to increase performance
    - The model was overfitting, regularization was applied to improve it
  - 5. Regularization using lasso and ridge models- the lasso regression model gave the best r2 and smallest MAE than the linear and ridge model

## Tools:

For webscarpping - Beautiful soup For analysis and model making - python