DATA ANALYSIS REPORT

IMDb data from 2006

Arife Gül Yalçın

B1605.090054

A data set of 1000 popular movies on IMDb in the last years

First of all, I want to explain my aim in choosing this subject and dataset;

I personally like watching movies and doing research about them. Questions such as who directed the movie ,which movie is the most popular (user ranking) or what are movies genre, are very important. IMDb is a very advanced platform. By researching the movie you want to watch, you can reach very accurate results. On the IMDb platform, movies are rated by good people and appropriate ratings are revealed, which you will like almost. In addition, users can contribute to the scoring by voting.

In this part, I imported pandas to read our csv file. I also continue importing some things that will work with my future questions here.

In this section, we checked which columns we have in our csv file named 'IMDB-Movie-Data' and we read our file.

Which director has the most number of movies?

I wrote this way to see more than one director in this section. But as I mentioned; By writing this way, we can only see 1 director with the most movies.

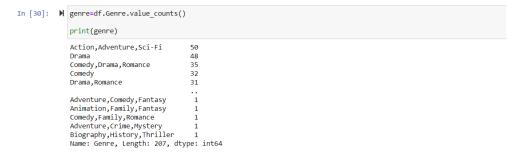


Here, Metascor is a value between 0 and 100, and by taking 100 Ms, the Boyhood movie in the Drama genre has the highest Ms.

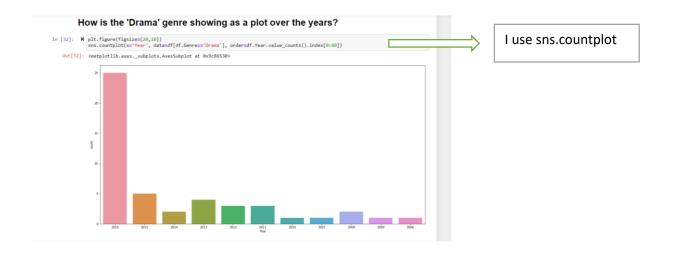


When we check out the least preferred movie genre, we have 1 movies in Action, Fantasy, Thriller.

What is the most preferred genre?



When we check out the most preferred movie genre, we have 50 movies in Action, Adventure, Sci-Fi. By writing this way, we can only see 1 genre .(genre.head(1))



In this section, we checked the plot representation of the 'Drama' type from our datasets between 2006-2016.



The 'Nine Lives' movie at the bottom of the list is our least watched movie between these years.



The most preferred year among these years in IMDb is 2016. This year there are 297 films.



In this part, the movie "The Dark Knight" in the genre of Action, Crime, Drama with the votes of 1791916 is the most popular movie.



Checked that the duration of the longest movie was 191 minutes



⇒ 'Star Wars: Episode VII - The Force Awakens' movie, Action, Adventure, Fantasy, has earned 936.63 million revenues.



I found the total movie count by using the for loop and increasing it by 1 each time



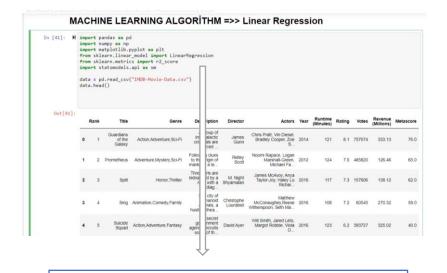
ightharpoonup This is how I checked the as 'quantile (0.80)' revenue from the 80th percentile, and revenue=134.52



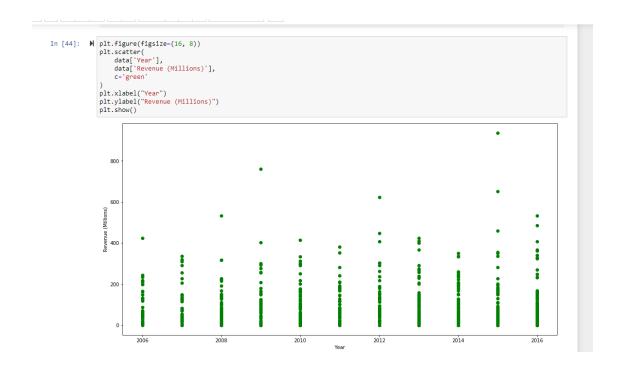
The Dark Knight movie with Rating 9 is the most popular movie(user ranking)

MACHINE LEARNING

- > Linear Regression
- > Logistic Regression
- Decision Tree
- > SVM
- Naive Bayes
- > kNN
- ➤ K-Means
- Random Forest
- Dimensionality Reduction Algorithms
- > Gradient Boosting algorithms
 - o GBM
 - XGBoost
 - o LightGBM
 - CatBoost



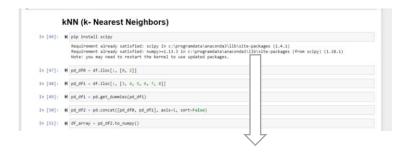
First of all;
import LinearRegression → from sklearn.linear_model
import r2_score → from sklearn.metrics



Run this cell of code and you should see this graph:

As you can see, there is a clear relationship between the Year and Revenue (Millions).

Looking at both coefficients, we have a p-value that is very low (although it is probably not exactly 0). This means that there is a strong correlation between these coefficients and the target.



SciPy is a <u>free and open-source Python</u> library used for <u>scientific computing</u> and technical computing.

SciPy contains modules for <u>optimization</u>, <u>linear</u> <u>algebra</u>, <u>integration</u>, <u>interpolation</u>, <u>special functions</u>, <u>FFT</u>, <u>signal</u> and <u>image</u> <u>processing</u>, <u>ODE</u> solvers and other tasks common in science and engineering.

```
In [52]: M IPDB - {}

for d in df_array:
    Rank = int(d[0])
    Title = d[1]
    year = d[2:]
    Revenue = map(int, year)

In [53]: M def getWeighbors(Rank, K):
    distances = []
    for imdb in IPDB:
        if (indb != Rank):
            distar = computeDistance(IPDB[Rank], IPDB[imdb])
        distances.sopend((imdb, dist)))
    distances.sopend((imdb, dist)))
    neighbors = []
    for x in range(K):
        neighbors.append(distances[x][0], distances[x][1]))
    return neighbors

In [54]: M def ComputeDistance(a, b):
    data* = a[1]
    data* = b[1]
    AttributeDistance = spatial.distance.cosine(dataA, dataB)
    return AttributeDistance
```