IMDB Movie Analysis

Project Description:

In this project we were provided with a CSV dataset having various columns of different IMDB movies. We are required to Frame the problem. For this task, we need to define a problem we want to shed some light on like top 250 movie, top director, highest grossing movies, top actors, etc. Once we have defined a problem, clean the data as necessary, and use Data Analysis skills to explore the data set and derive insights.

Approach:

My approach was first to study the dataset to better understand it then I cleaned the data set by removing blank spaces after that I used my excel knowledge which I learned in course video to draw out insights from the dataset to answer the questions.

Tech-Stack Used:

I have used MS Excel 2016 for data analysis and MS word to make the report.

Excel Dataset:

 $\frac{https://docs.google.com/spreadsheets/d/1tGN12pveejGIFPjs5zj8aDHxHICtjQMm/edit?usp=sharing\&ouid=115977276322304781744\&rtpof=true\&sd=trued$

Insights:

In this project I learned several new excel function and I also learned how to clean the data using whereas shortcut techniques i.e. Find & Select option. I also learned how to create table using charts to analyze the something which in my point very useful thing and save a lot of time.

Results:

A. Cleaning the data: This is one of the most important step to perform before moving forward with the analysis. Use your knowledge learned till now to do this. (Dropping columns, removing null values, etc.)

Insights: I clean the dataset by delete all the blank space using Find & Select Option, it reduce my dataset from 5044 rows to 3757 rows.

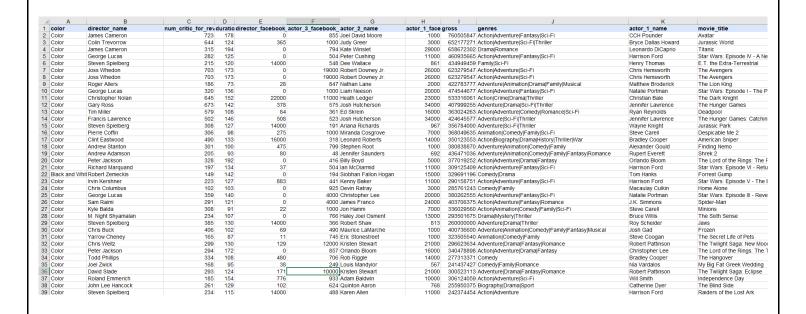


Fig: Cleaned Excel Data

B. Movies with highest profit: Create a new column called profit which contains the difference of the two columns: gross and budget. Sort the column using the profit column as reference. Plot profit (y-axis) vs budget (x- axis) and observe the outliers using the appropriate chart type. **Insights:**



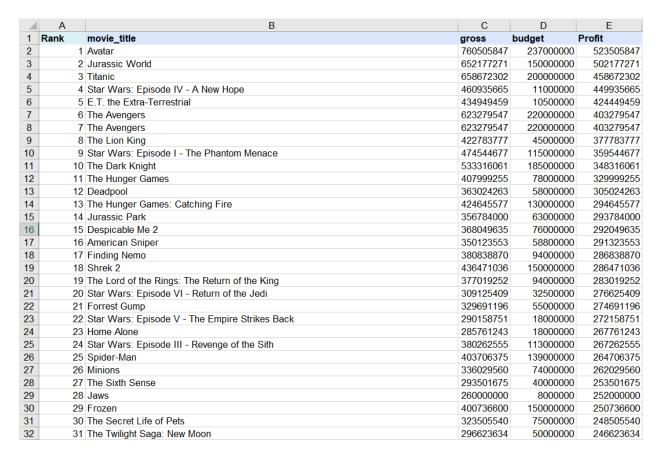


Fig: Movies with Most Profit

C. Top 250: Create a new column IMDb_Top_250 and store the top 250 movies with the highest IMDb Rating (corresponding to the column: imdb_score). Also make sure that for all of these movies, the num_voted_users is greater than 25,000. Also add a Rank column containing the values 1 to 250 indicating the ranks of the corresponding films.

Extract all the movies in the IMDb_Top_250 column which are not in the English language and store them in a new column named Top_Foreign_Lang_Film. You can use your own imagination also!

Insights:

	Α	В	С	D
1	Ranking	movie_title	num_voted_users	imdb_score
2	1	The Shawshank Redemption	1689764	9.3
3	2	The Godfather	1155770	9.2
4	4	The Dark Knight	1676169	9
5	3	The Godfather: Part II	790926	9
6	8	Pulp Fiction	1324680	8.9
7	7	The Lord of the Rings: The Return of the King	1215718	8.9
8	6	Schindler's List	865020	8.9
9	5	The Good, the Bad and the Ugly	503509	8.9
10	13	Inception	1468200	8.8
11	12	Fight Club	1347461	8.8
12	11	Forrest Gump	1251222	8.8
13	10	The Lord of the Rings: The Fellowship of the Ring	1238746	8.8
14	9	Star Wars: Episode V - The Empire Strikes Back	837759	8.8
15	20	The Matrix	1217752	8.7
16	19	The Lord of the Rings: The Two Towers	1100446	8.7
17	18	Star Wars: Episode IV - A New Hope	911097	8.7
18	17	Goodfellas	728685	8.7
19	16	One Flew Over the Cuckoo's Nest	680041	8.7
20	15	City of God	533200	8.7
21	14	Seven Samurai	229012	8.7
22	28	Se7en	1023511	8.6
23	27	Interstellar	928227	8.6
24	26	The Silence of the Lambs	887467	8.6
25	25	Saving Private Ryan	881236	8.6
26	24	American History X	782437	8.6
27	23	The Usual Suspects	740918	8.6
28	22	Spirited Away	417971	8.6
29	21	Modern Times	143086	8.6
30	46	The Dark Knight Rises	1144337	8.5
31	45	Gladiator	982637	8.5
32	44	Django Unchained	955174	8.5

Fig: Top 250 movies based on IMDB Score

Top_Foreign_Lang_film	num_voted_users	imdb_score	language
The Good, the Bad and the Ugly	503509	8.9	Italian
Seven Samurai	229012	8.7	Japanese
City of God	533200	8.7	Portuguese
Spirited Away	417971	8.6	Japanese
Children of Heaven	27882	8.5	Persian
The Lives of Others	259379	8.5	German
Samsara	22457	8.5	None
A Separation	151812	8.4	Persian
Das Boot	168203	8.4	German
Princess Mononoke	221552	8.4	Japanese
Oldboy	356181	8.4	Korean
Amélie	534262	8.4	French
Metropolis	111841	8.3	German
The Hunt	170155	8.3	Danish
Downfall	248354	8.3	German
Incendies	80429	8.2	French
The Secret in Their Eyes	131831	8.2	Spanish
Howl's Moving Castle	214091	8.2	Japanese
Pan's Labyrinth	467234	8.2	Spanish
The Act of Killing	23836	8.2	Indonesian
Tae Guk Gi: The Brotherhood of War	31943	8.1	Korean
The Sea Inside	64556	8.1	Spanish
The Celebration	65951	8.1	Danish
Elite Squad	81644	8.1	Portuguese
Akira	106160	8.1	Japanese
Amores Perros	173551	8.1	Spanish
Central Station	28951	8	Portuguese
Waltz with Bashir	46107	8	Hebrew
My Name Is Khan	69759	8	Hindi
Persepolis	70194	8	French
A Fistful of Dollars	147566	8	Italian

Fig: Top Foreign Language Movie

D. Best Directors: Group the column using the director_name column.

Find out the top 10 directors for whom the mean of imdb_score is the highest and store them in a new column top10director. In case of a tie in IMDb score between two directors, sort them alphabetically.

Insights:

1	Director	Mean imdb_score
2	Akira Kurosawa	8.7
3	Charles Chaplin	8.6
4	Michael Curtiz	8.6
5	Tony Kaye	8.6
6	Damien Chazelle	8.5
7	Majid Majidi	8.5
8	Ron Fricke	8.5
9	Sergio Leone	8.433333333
10	Christopher Nolan	8.425
11	Asghar Farhadi	8.4

Fig: Top 10 Director Based on Mean IMDB Score

E. Popular Genres: Perform this step using the knowledge gained while performing previous steps.

Insights:

Fig: Top Genre Based on their count

1	Genre	count
2	Drama	1893
3	Comedy	1461
4	Thriller	1115
5	Action	959
6	Romance	858
7	Adventure	781
8	Crime	709
9	Fantasy	507
10	Sci-Fi	496
11	Family	442
12	Horror	392
13	Mystery	384
14	Biography	239
15	Animation	196
16	War	152
17	Music	151
18	History	149
19	Sport	148
20	Musical	96
21	Western	59
22	Documenta	45
23	Film-Noir	1

F. Charts: Create three new columns namely, Meryl_Streep, Leo_Caprio, and Brad_Pitt which contain the movies in which the actors: 'Meryl Streep', 'Leonardo DiCaprio', and 'Brad Pitt' are the lead actors. Use only the actor_1_name column for extraction. Also, make sure that you use the names 'Meryl Streep', 'Leonardo DiCaprio', and 'Brad Pitt' for the said extraction.

Append the rows of all these columns and store them in a new column named Combined.

Group the combined column using the actor_1_name column.

Find the mean of the num_critic_for_reviews and num_users_for_review and identify the actors which have the highest mean.

Observe the change in number of voted users over decades using a bar chart. Create a column called decade which represents the decade to which every movie belongs to. For example, the title_year year 1923, 1925 should be stored as 1920s. Sort the column based on the column decade, group it by decade and find the sum of users voted in each decade. Store this in a new data frame called df by decade.

Insights:

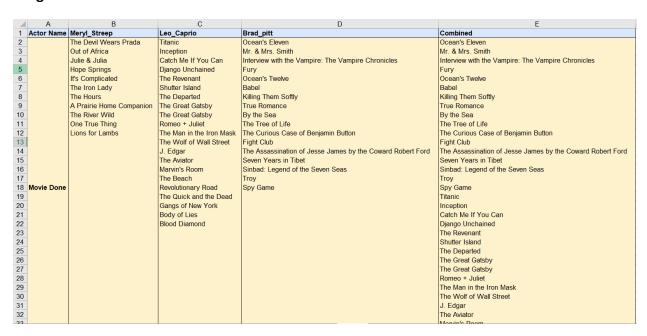


Fig: Movie Done by Meryl Streep, Leonardo DiCaprio, Brad pitt

	Α	В	С	D	E	F
1	Actor Name	Average of num_critic_for_review	Average of num_user_for_reviews			
2	Albert Finney	750	1498			
3	Phaldut Sharma	738	1885			
4	Peter Capaldi	654	995			
5	Craig Stark	596	1018			
6	Bérénice Bejo	576	583			
7	Suraj Sharma	552	755			
8	Ellar Coltrane	548	836		Critic Favourite	Heather Donahue
9	Mike Howard	546	405		Audience Favorit	Albert Finney
10	Lou Taylor Pucci	543	789			
11	Joel Courtney	539	849			
12	Maika Monroe	533	631			
13	Tim Holmes	525	511			
14	Elina Alminas	489	611			
15	Kurt Fuller	487	509			
16	Iko Uwais	481	316			
17	Quvenzhané Wallis	478.6666667	392.3333333			
18	Edgar Arreola	478	461			
19	Sharlto Copley	472	1262			
20	Cory Hardrict	452	326			
21	Matt Frewer	451	1229			
22	Elizabeth McGovern	447	801			
23	Isabelle Huppert	447	190			
24	Michael Fassbender	434.3333333	837.5			
25	Wood Harris	432	588			
26	Jennifer Lawrence	418.9285714	583.2142857			
27	Anil Kapoor	418	1017			
28	Jessica Barden	417	438			
29	Sally Hawkins	415	384			
30	Chris Hemsworth	411.7333333	702.3333333			
31	Danielle Kotch	411	524			
32	Micah Sloat	409	1189			

Fig: Critic Favorite and Audience Favorite

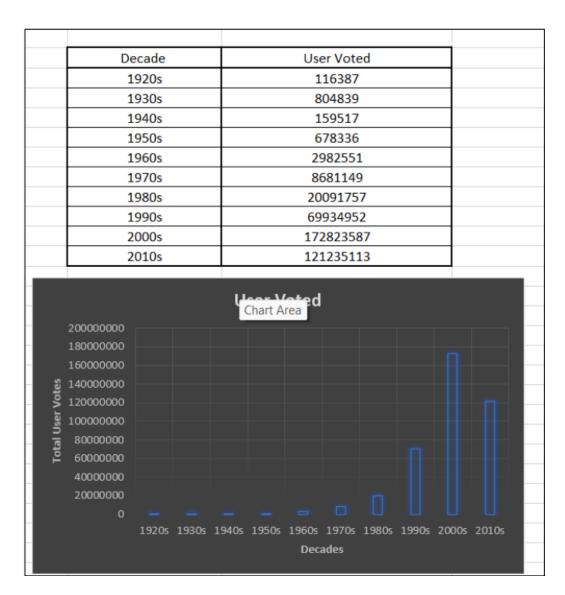


Fig: Number of voted User over Decade