

IMDB Movie Analysis

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Project Description

| The project is on the analysis of the dataset of the IMDB movies where I will study the data properly and ask various questions about it to draw attention to that problem.

After the problem is detected, I will do analysis and try to find the reason behind the problem and draw insights about it.

Approach

| The approach of this project is to first clean the data by removing duplicates, dropping null value, dropping columns etc. and then use the Root Cause Analysis, developed by Sakichi Toyoda which is also known as the Five 'Whys' Approach and do analysis according to it using Pivot table and various Excel function such as AVERAGE, COUNTIF, etc.

Tech-Stack Used

I have used Microsoft Excel to do the analysis of the given data.

I have used Microsoft Excel because it is a spreadsheet developed for various platforms. It has calculation or computation capabilities with graphing tools, pivot tables, and also has a macro programming language called Visual Basic for Applications which would make doing the analysis and visualisation effortless.

Data Cleaning

| First the data is imported to excel from the csv file after importing the data the duplicate data are checked and deleted using the data tools. After that we check for the gaps in the columns using the trim function, then we look for blank cells using the find special and delete them as it would be of no help for analysis. After that we check the columns and according to our basic understandings we delete the columns which will be of no use for the analysis and it would make analysis easy.

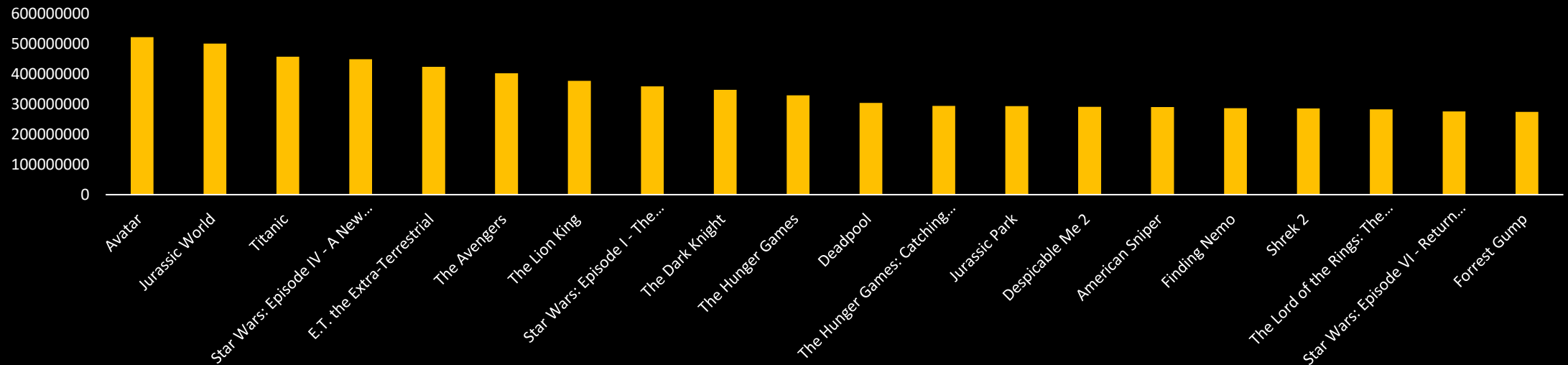
Movies with highest profit

The Top 20 movies with highest profit are given in the table

S.No	Movie	Profit	S.No	Movie	Profit
1	Avatar	\$ 523,505,847	11	Deadpool	\$ 305,024,263
2	Jurassic World	\$ 502,177,271	12	The Hunger Games: Catching Fire	\$ 294,645,577
3	Titanic	\$ 458,672,302	13	Jurassic Park	\$ 293,784,000
4	Star Wars: Episode IV - A New Hope	\$ 449,935,665	14	Despicable Me 2	\$ 292,049,635
5	E.T. the Extra-Terrestrial	\$ 424,449,459	15	American Sniper	\$ 291,323,553
6	The Avengers	\$ 403,279,547	16	Finding Nemo	\$ 286,838,870
7	The Lion King	\$ 377,783,777	17	Shrek 2	\$ 286,471,036
8	Star Wars: Episode I - The Phantom Menace	\$ 359,544,677	18	The Lord of the Rings: The Return of the King	\$ 283,019,252
9	The Dark Knight	\$ 348,316,061	19	Star Wars: Episode VI - Return of the Jedi	\$ 276,625,409
10	The Hunger Games	\$ 329,999,255	20	Forrest Gump	\$ 274,691,196

INSIGHTS

Avatar is the highest profitable movie having profit of \$523,505,847 followed by *Jurassic Park* and then *Titanic*. *The Hunger Games* is at 10th position and *Forest Gump* at 20th position.

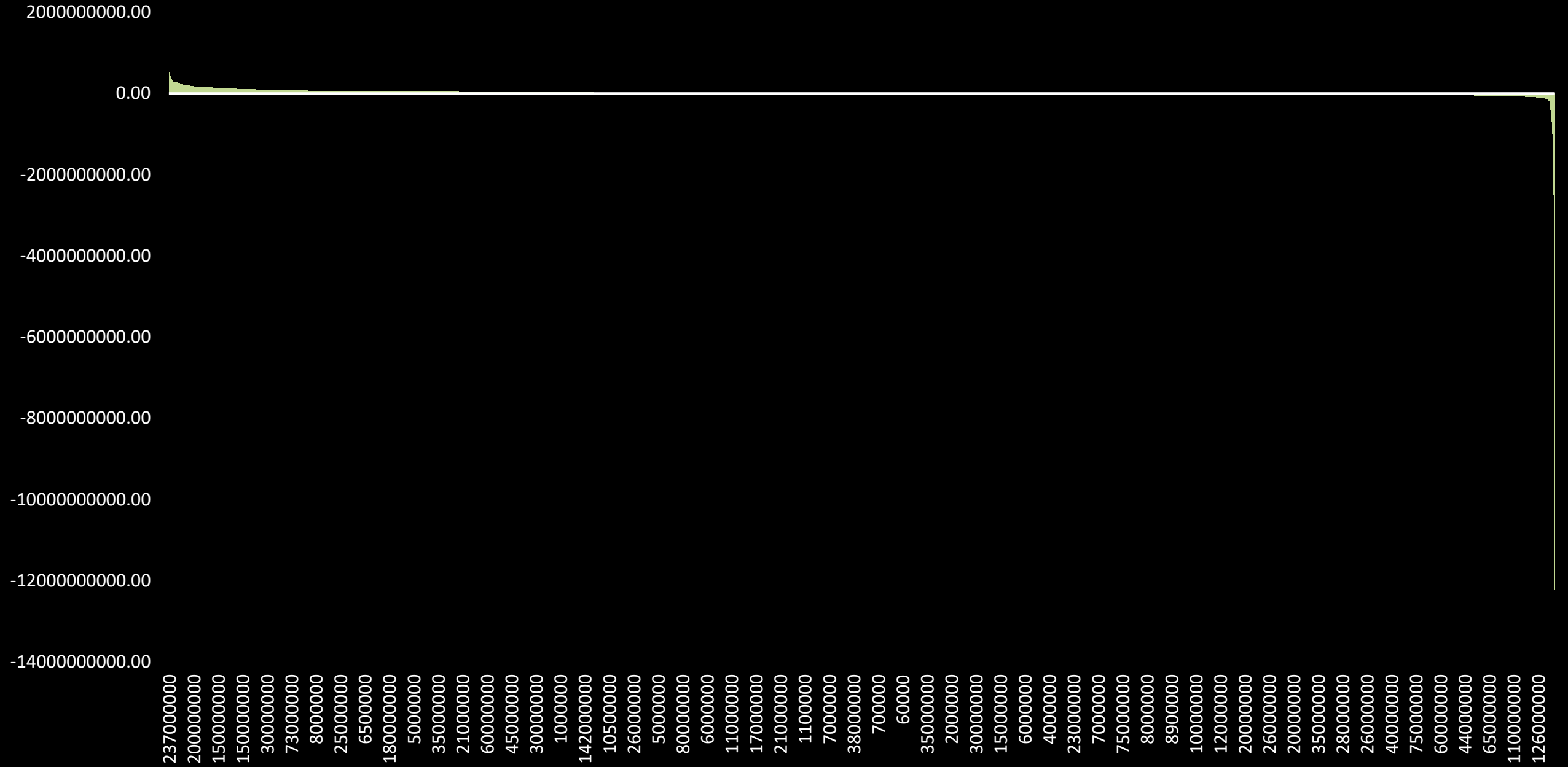


EXPLANATION

To Extract the desired data a new column named Profit is made which contains the difference of Gross and Budget of each movie. After that, the Profit column is sorted as highest to lowest and then the top movies are extracted for the analysis and various charts are plotted.

BUDGET TO PROFIT CHART

■ Profit



IMDB Top 250

1	Towering Inferno	9.5	31	Queen of the Mountains	8.7	61	Raiders of the Lost Ark	8.5
2	The Shawshank Redemption	9.3	32	Seven Samurai	8.7	62	Psycho	8.5
3	The Godfather	9.2	33	Butterfly Girl	8.7	63	Back to the Future	8.5
4	Dekalog	9.1	34	Interstellar	8.6	64	Alien	8.5
4	Dekalog	9.1	35	Hannibal	8.6	65	Memento	8.5
6	The Dark Knight	9	36	Saving Private Ryan	8.6	66	Airlift	8.5
7	The Godfather: Part II	9	37	Luther	8.6	67	Samsara	8.5
8	Fargo	9	38	Se7en	8.6	68	Whiplash	8.5
9	The Lord of the Rings: The Return of the King	8.9	39	The Silence of the Lambs	8.6	69	The Lives of Others	8.5
10	Schindler's List	8.9	40	Spirited Away	8.6	70	Dr. Strangelove or: How I Learned to Stop Worrying and Love the Bomb	8.5
11	Pulp Fiction	8.9	41	Spartacus: War of the Damned	8.6	71	Happy Valley	8.5
12	The Good, the Bad and the Ugly	8.9	42	American History X	8.6	72	Children of Heaven	8.5
13	12 Angry Men	8.9	43	The Usual Suspects	8.6	73	WALL·E	8.4
14	Inception	8.8	44	Once Upon a Time in the West	8.6	74	Braveheart	8.4
15	The Lord of the Rings: The Fellowship of the Ring	8.8	45	Running Forever	8.6	75	Stargate SG-1	8.4
16	Daredevil	8.8	46	It's a Wonderful Life	8.6	76	Amélie	8.4
17	Fight Club	8.8	47	Modern Times	8.6	77	Baahubali: The Beginning	8.4
18	Forrest Gump	8.8	48	Casablanca	8.6	78	Lion of the Desert	8.4
19	It's Always Sunny in Philadelphia	8.8	49	The Dark Knight Rises	8.5	79	Star Wars: Episode VI - Return of the Jedi	8.4
20	Star Wars: Episode V - The Empire Strikes Back	8.8	50	Gladiator	8.5	80	Once Upon a Time in America	8.4
21	The Lord of the Rings: The Two Towers	8.7	51	Terminator 2: Judgment Day	8.5	81	Princess Mononoke	8.4
22	The Matrix	8.7	52	Django Unchained	8.5	82	The Shining	8.4
23	Friday Night Lights	8.7	53	The Departed	8.5	83	Aliens	8.4
24	The Honeymooners	8.7	54	The Lion King	8.5	84	American Beauty	8.4
25	Goodfellas	8.7	55	The Green Mile	8.5	85	Lawrence of Arabia	8.4
26	Star Wars: Episode IV - A New Hope	8.7	56	Outlander	8.5	86	U2 3D	8.4
27	Gomorrah	8.7	57	The Prestige	8.5	87	Das Boot	8.4
28	One Flew Over the Cuckoo's Nest	8.7	58	The Pianist	8.5	88	Veronica Mars	8.4
29	City of God	8.7	59	Apocalypse Now	8.5	89	The Inbetweeners	8.4
30	A Beginner's Guide to Snuff	8.7	60	Entourage	8.5	90	Rang De Basanti	8.4

91	Psych	8.4	121	Snatch	8.3	151	Blade Runner	8.2
92	Requiem for a Dream	8.4	122	2001: A Space Odyssey	8.3	152	Gran Torino	8.2
93	Anne of Green Gables	8.4	123	Good Will Hunting	8.3	153	Warrior	8.2
94	M*A*S*H	8.4	124	Preacher	8.3	154	Howl's Moving Castle	8.2
95	Batman: The Dark Knight Returns, Part 2	8.4	125	Lake of Fire	8.3	155	Into the Wild	8.2
96	Oldboy	8.4	126	The Sting	8.3	156	Pan's Labyrinth	8.2
97	To Kill a Mockingbird	8.4	127	The Great Escape	8.3	157	The Deer Hunter	8.2
98	Reservoir Dogs	8.4	128	The Hunt	8.3	158	The Big Lebowski	8.2
99	A Separation	8.4	129	The Apartment	8.3	159	Buffy the Vampire Slayer	8.2
100	The Other Dream Team	8.4	130	Judgment at Nuremberg	8.3	160	Incendies	8.2
101	A Charlie Brown Christmas	8.4	131	Some Like It Hot	8.3	161	War & Peace	8.2
102	Toy Story 3	8.3	132	Singin' in the Rain	8.3	162	Emma	8.2
103	Up	8.3	133	Inside Job	8.3	163	The Elephant Man	8.2
104	Inside Out	8.3	134	No End in Sight	8.3	164	Gone with the Wind	8.2
105	Batman Begins	8.3	135	Taxi Driver	8.3	165	The Secret in Their Eyes	8.2
106	Life	8.3	136	Hoop Dreams	8.3	166	Trainspotting	8.2
107	Inglourious Basterds	8.3	137	The Sound and the Shadow	8.3	167	The Bridge on the River Kwai	8.2
108	Indiana Jones and the Last Crusade	8.3	138	Monty Python and the Holy Grail	8.3	168	Lage Raho Munna Bhai	8.2
109	The Returned	8.3	139	Peace, Propaganda & the Promised Land	8.3	169	Mr. Smith Goes to Washington	8.2
110	L.A. Confidential	8.3	140	Captain America: Civil War	8.2	170	Growing Up Smith	8.2
111	Toy Story	8.3	141	How to Train Your Dragon	8.2	171	Lock, Stock and Two Smoking Barrels	8.2
112	Scarface	8.3	142	Godzilla Resurgence	8.2	172	Rebecca	8.2
113	Shaun the Sheep	8.3	143	The Wolf of Wall Street	8.2	173	The Act of Killing	8.2
114	Eternal Sunshine of the Spotless Mind	8.3	144	Finding Nemo	8.2	174	On the Waterfront	8.2
115	Amadeus	8.3	145	A Beautiful Mind	8.2	175	Rise of the Entrepreneur: The Search for a Better Way	8.2
116	Raging Bull	8.3	146	Casino	8.2	176	Home Movies	8.2
117	Metropolis	8.3	147	V for Vendetta	8.2	177	Cries & Whispers	8.2
118	Unforgiven	8.3	148	The Thing	8.2	178	It Happened One Night	8.2
119	Downfall	8.3	149	Trapped	8.2	179	The Brain That Sings	8.2
120	Room	8.3	150	Die Hard	8.2	180	The Last Waltz	8.2

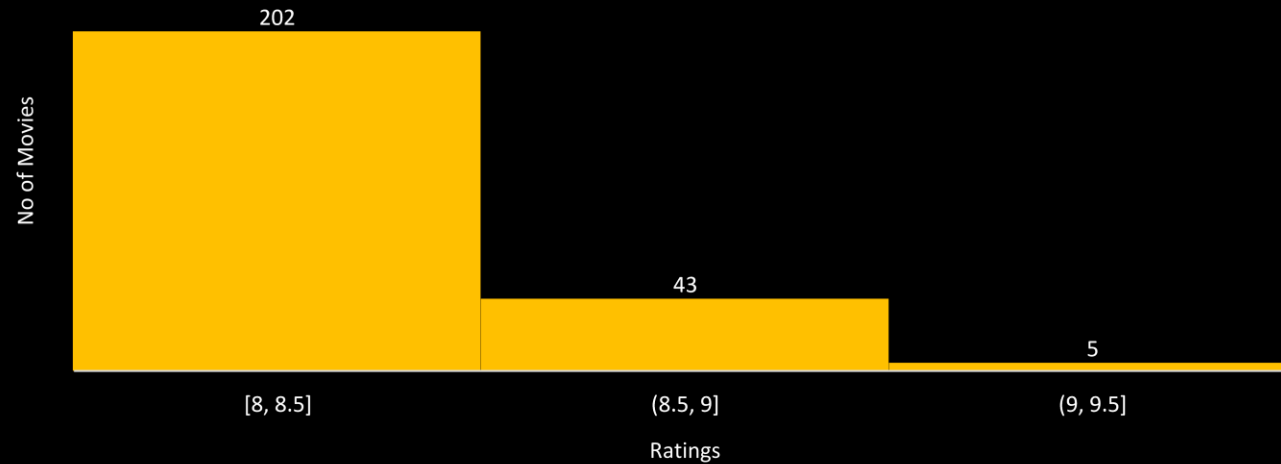
181	The Avengers	8.1
182	Guardians of the Galaxy	8.1
183	Mad Max: Fury Road	8.1
184	The Revenant	8.1
185	The Bourne Ultimatum	8.1
186	Pirates of the Caribbean: The Curse of the Black Pearl	8.1
187	Monsters, Inc.	8.1
188	The Martian	8.1
189	Shutter Island	8.1
190	The Missing	8.1
191	Jurassic Park	8.1
192	Gone Girl	8.1
193	The Truman Show	8.1
194	Deadpool	8.1
195	Kill Bill: Vol. 1	8.1
196	Solaris	8.1
197	Prisoners	8.1
198	The Sixth Sense	8.1
199	Sin City	8.1
200	Rush	8.1
201	The Grand Budapest Hotel	8.1
202	Million Dollar Baby	8.1
203	The Help	8.1
204	No Country for Old Men	8.1

205	There Will Be Blood	8.1
206	Gandhi	8.1
207	12 Years a Slave	8.1
208	Spotlight	8.1
209	Hotel Rwanda	8.1
210	Hachi: A Dog's Tale	8.1
211	Lilyhammer	8.1
212	The Imitation Game	8.1
213	The Princess Bride	8.1
214	Groundhog Day	8.1
215	The Sea Inside	8.1
216	Del 1 - Män som hatar kvinnor	8.1
217	Tae Guk Gi: The Brotherhood of War	8.1
218	Barry Lyndon	8.1
219	Stand by Me	8.1
220	Akira	8.1
221	Elite Squad	8.1
222	The Terminator	8.1
223	Platoon	8.1
224	Butch Cassidy and the Sundance Kid	8.1
225	Donnie Darko	8.1
226	Animal Kingdom	8.1
227	Annie Hall	8.1
228	Network	8.1

229	A Christmas Story	8.1
230	The Man Who Shot Liberty Valance	8.1
231	Cat on a Hot Tin Roof	8.1
232	The Wizard of Oz	8.1
233	Before Sunrise	8.1
234	The Best Years of Our Lives	8.1
235	Amores Perros	8.1
236	Touching the Void	8.1
237	Strangers with Candy	8.1
238	In the Shadow of the Moon	8.1
239	The Square	8.1
240	The Celebration	8.1
241	Rocky	8.1
242	The Conformist	8.1
243	High Noon	8.1
244	Destiny	8.1
245	Woodstock	8.1
246	Nothing But a Man	8.1
247	Ordet	8.1
248	Archaeology of a Woman	8.1
249	X-Men: Days of Future Past	8
250	Ratatouille	8

INSIGHTS

'Towering Inferno' is the highest rated movie on IMDB with 9.5 rating. A total of 8 movies are rated above 9, 240 movies are rated between 8-9 and 2 movies are rated 8.



EXPLANATION

To get the desired results, sort the 'imdb_score' column from highest to lowest and then extract the data accordingly for Non-English movies, sort the 'language' column by deselecting the English language.

MORE INSIGHTS

Non-English movies in IMDB top 250

Movie	IMDB Rating	Movie	IMDB Rating	Movie	IMDB Rating
A Separation	8.4	Godzilla Resurgence	8.2	Spirited Away	8.6
Airlift	8.5	Gomorrah	8.7	Tae Guk Gi: The Brotherhood of War	8.1
Akira	8.1	Howl's Moving Castle	8.2	The Act of Killing	8.2
Amélie	8.4	Incendies	8.2	The Brain That Sings	8.2
Amores Perros	8.1	Lage Raho Munna Bhai	8.2	The Celebration	8.1
Baahubali: The Beginning	8.4	Metropolis	8.3	The Conformist	8.1
Children of Heaven	8.5	Oldboy	8.4	The Good, the Bad and the Ugly	8.9
City of God	8.7	Ordet	8.1	The Hunt	8.3
Cries & Whispers	8.2	Pan's Labyrinth	8.2	The Lives of Others	8.5
Das Boot	8.4	Princess Mononoke	8.4	The Returned	8.3
Dekalog	18.2	Rang De Basanti	8.4	The Sea Inside	8.1
Del 1 - Män som hatar kvinnor	8.1	Samsara	8.5	The Secret in Their Eyes	8.2
Downfall	8.3	Seven Samurai	8.7	The Square	8.1
Elite Squad	8.1	Solaris	8.1	Trapped	8.2

Top 10 language with highest avg rating

Rank	Language	Avg of IMDB Rating
1	Telugu	8.40
2	Polish	8.25
3	None	7.95
4	Indonesian	7.90
5	Maya	7.80
6	Hebrew	7.58
7	Persian	7.58
8	Icelandic	7.55
9	Danish	7.50
10	Dari	7.50

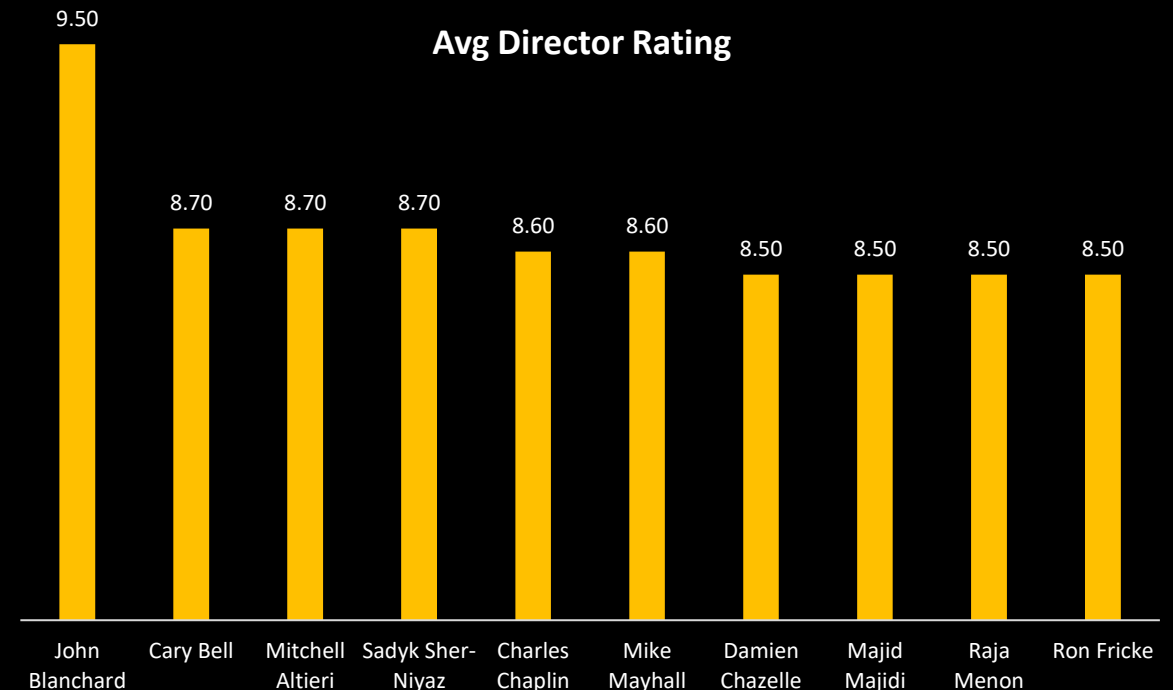
Hindi movies in IMDB top 250 movies

Movie	IMDB Rating
Airlift	8.5
Lage Raho Munna Bhai	8.2
Rang De Basanti	8.4

Best Directors

Best 10 directors with highest average IMDB ratings are:

Rank	Director	Average IMDB score
1	John Blanchard	9.50
2	Cary Bell	8.70
3	Mitchell Altieri	8.70
4	Sadyk Sher-Niyaz	8.70
5	Charles Chaplin	8.60
6	Mike Mayhall	8.60
7	Damien Chazelle	8.50
8	Majid Majidi	8.50
9	Raja Menon	8.50
10	Ron Fricke	8.50



INSIGHTS

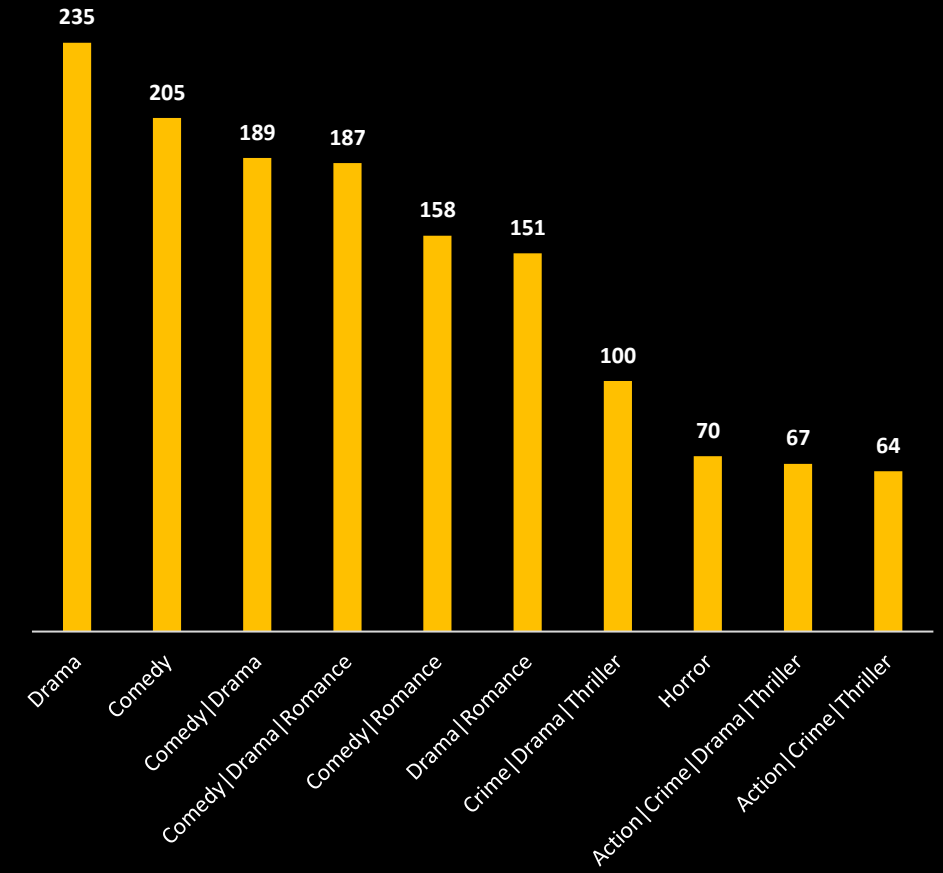
'John Blanchard' is the best Director in the IMDB list with average rating of 9.5, *'Ron Fricke'* is at the 10th position with Avg. IMDB rating of 8.5. The Average rating of Top 10 directors is 8.68.

EXPLANATION

To get the desired result a pivot table is made in which 'director_name' column is put into rows and 'imdb_score' column is put into sum where the value is changed to average. Datas from the Pivot Table is extracted and then new tables are made from the data and then the table with average imdb score is sorted from high to low. Then we get the desired result and can use it for further visual analysis.

Popular Genre

Rank	Genre	No of movies
1	Drama	235.00
2	Comedy	205.00
3	Comedy Drama	189.00
4	Comedy Drama Romance	187.00
5	Comedy Romance	158.00
6	Drama Romance	151.00
7	Crime Drama Thriller	100.00
8	Horror	70.00
9	Action Crime Drama Thriller	67.00
10	Action Crime Thriller	64.00



INSIGHTS

'Drama' is the most popular genre on IMDB with most numbers of movies i.e. 235 followed by 'Comedy' with 205 movies.

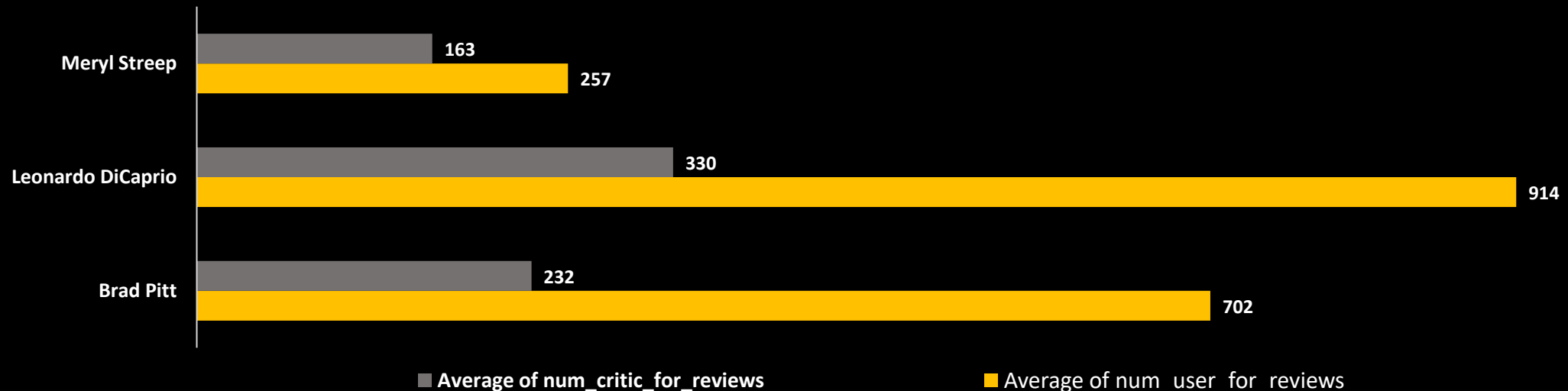
'Action|Crime|Thriller' is the 10th most popular genre with 64 movies.

EXPLANATION

To get the desired result a Pivot Table is made from the data in which 'genre' column is set into rows and 'imdb_score' is set into values changed to count to get the count of the genres or movies. After that the data is extracted and new table is made and the count column is sorted from largest to smallest thus getting the desired results for analysis.

Favorite Actors

Actor	Avg no of user for reviews	Avg no of critic for reviews
Brad Pitt	702.44	231.94
Leonardo DiCaprio	914.48	330.19
Meryl Streep	257.31	163.15



INSIGHTS

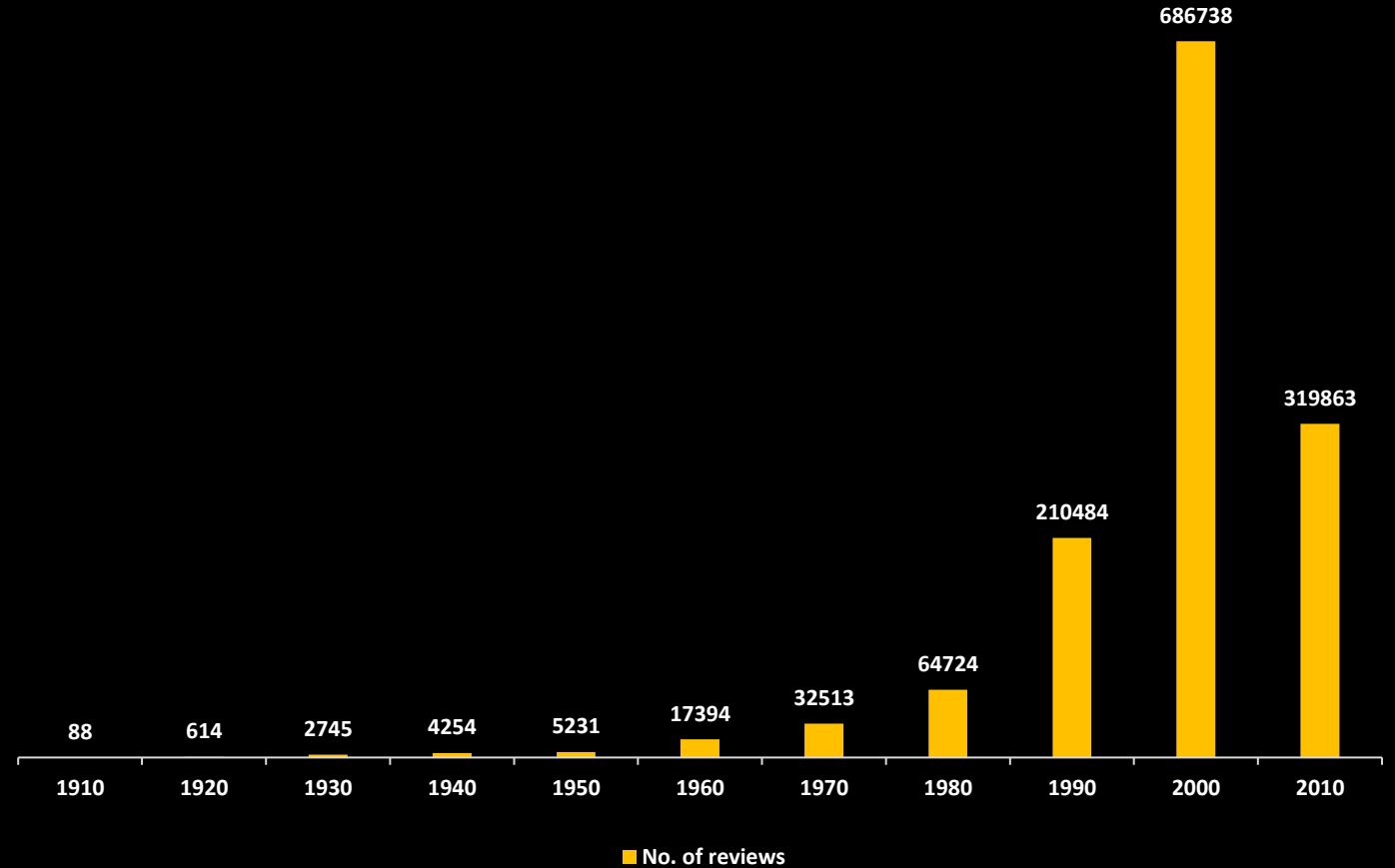
According to the data '*Leonardo DiCaprio*' is the favorite actor for user as he got the most numbers of reviews by users followed by '*Bradd Pitt*' and then '*Meryl Streep*'. The same pattern is followed in the case of favorite actor for critics where '*Leonardo DiCaprio*' is reviewed most by the critics followed by '*Bradd Pitt*' and then '*Meryl Streep*'.

EXPLANATION

To get the desired data 3 new columns are made named after the 3 actors containing their movies, then these data are appended into new column along with their number of reviews for users and critics separately. After that the mean of the reviews are calculated using the AVERAGEIF function of excel of all three actors and then the data is used for analysis and chart preparation.

Change in number of voted users over decades

Decade	No. of reviews
1910	88
1920	614
1930	2745
1940	4254
1950	5231
1960	17394
1970	32513
1980	64724
1990	210484
2000	686738
2010	319863



| INSIGHTS

As we can see on the bar graph, the number of reviews has increased exponentially over the decades till 2000s and then the number of review decreases in the 2010s.

| EXPLANATION

To get the desired data we first find the decades of the year by using the formula $=\text{INT}(\text{year}/10)*10$ after acquiring the decades we use the SUMIF function for different decades and then draw table out of it and charts for visual analysis.

RESULT

| I got to understand how different problems are framed from the dataset and how to define the problem. And then doing its analysis using the 5 why Analysis developed by Sakichi Toyoda and then draw insights out of it.

THANK YOU

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