IMDB Movie Analysis

Project Description –

Movie Analysis is one of the most important functions for a film distributing company in order to broadcast appropriate movies and gain huge number of profits and minimal loss.

In this project, I have used the 5 why approach to find solutions for various problems.

One of them (for eg) is:

A VOD platform company incurs a huge amount of loss due to broadcasting some of the movies.

5 why approach:

- Q. Why did the VOD company incur huge loss?
- A. Because the movies broadcasted were not seen / subscribed by many people.
- Q. Why were the movies not seen by many people?
- A. Because the movies did not meet the expectations of the public.
- Q. Why did the movies not meet the expectations of the public?
- A. Because the director and actor of the movies were not liked by much people.
- Q. Why did not company not know that those movies were not liked by much people?
- A. Because the ratings and score of those movies were miscalculated.
- Q. Why were the ratings and score miscalculated?
- A. Because of the ineffectiveness of the data analyst team.

As we can see, wrong assumptions made by the data analyst team can lead to huge loss for a company which is difficult to oversee and overcome.

Approach -

As a Data Analyst, it is my duty to perform analysis to find out the following which would be helpful for the IMDB Movie Analysis:

- 1. Movies with highest profit
- 2. Top movies (250)
- 3. Best directors
- 4. Popular genres
- 5. Critic favourite and audience favourite actors

For finding all these details, we must clean the collected dataset that is used for the analysis. Cleaning includes: deleting unwanted columns, removing rows with null values, removing duplicates etc.

Tech-Stack used -

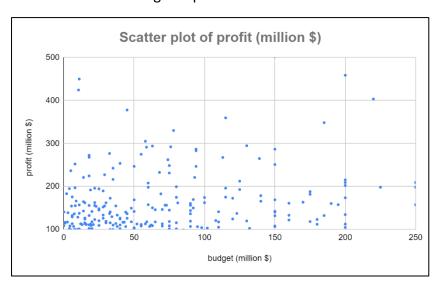
I have used **Google Spreadsheet** for this project.

Spreadsheet makes it easy to extract and visualize the data and interpret the results of the queries.

Insights -

After cleaning the provided dataset, I have derived the following information/insights using Excel:

1. Movies with highest profit:



Here is a scatter plot that plots budget in million \$ against profit in million \$.

The top 10 movies with highest profit are given in the table:

| movie_title | gross (million \$) | budget (million \$) | profit (million \$) |
|---|--------------------|---------------------|---------------------|
| Avatar | 760.5 | 237 | 523.5 |
| Jurassic World | 652.2 | 150 | 502.2 |
| Titanic | 658.7 | 200 | 458.7 |
| Star Wars: Episode IV - A New Hope | 460.9 | 11 | 449.9 |
| E.T. the Extra-Terrestrial | 434.9 | 10.5 | 424.4 |
| The Avengers | 623.3 | 220 | 403.3 |
| The Lion King | 422.8 | 45 | 377.8 |
| Star Wars: Episode I - The Phantom Menace | 474.5 | 115 | 359.5 |
| The Dark Knight | 533.3 | 185 | 348.3 |
| The Hunger Games | 408 | 78 | 330 |

2. IMDB Top 250 Movies:

The top 250 movies are given in the table below:

| IMDb_Top_250 | Rank |
|---|------|
| The Shawshank Redemption | 1 |
| The Godfather | 2 |
| The Dark Knight | 3 |
| The Godfather: Part II | 4 |
| The Lord of the Rings: The Return of the King | 5 |
| Schindler's List | 6 |
| Pulp Fiction | 7 |
| The Good, the Bad and the Ugly | 8 |
| Inception | 9 |
| The Lord of the Rings: The Fellowship of the Ring | 10 |
| Fight Club | 11 |
| Forrest Gump | 12 |
| Star Wars: Episode V - The Empire Strikes Back | 13 |
| The Lord of the Rings: The Two Towers | 14 |
| The Matrix | 15 |
| Goodfellas | 16 |
| Star Wars: Episode IV - A New Hope | 17 |
| One Flew Over the Cuckoo's Nest | 18 |
| City of God | 19 |
| Seven Samurai | 20 |
| Interstellar | 21 |
| Saving Private Ryan | 22 |
| Se7en | 23 |
| The Silence of the Lambs | 24 |
| Spirited Away | 25 |
| American History X | 26 |
| The Usual Suspects | 27 |
| Modern Times | 28 |
| The Dark Knight Rises | 29 |
| Gladiator | 30 |
| Terminator 2: Judgment Day | 31 |

| Django Unchained | 32 |
|--|----|
| The Departed | 33 |
| The Lion King | 34 |
| The Green Mile | 35 |
| The Prestige | 36 |
| The Pianist | 37 |
| Apocalypse Now | 38 |
| Raiders of the Lost Ark | 39 |
| Psycho | 40 |
| Back to the Future | 41 |
| Alien | 42 |
| Memento | 43 |
| Whiplash | 44 |
| The Lives of Others | 45 |
| Children of Heaven | 46 |
| WALL-E | 47 |
| Braveheart | 48 |
| Amélie | 49 |
| Baahubali: The Beginning | 50 |
| Star Wars: Episode VI - Return of the Jedi | 51 |
| Once Upon a Time in America | 52 |
| Princess Mononoke | 53 |
| Aliens | 54 |
| American Beauty | 55 |
| Lawrence of Arabia | 56 |
| Das Boot | 57 |
| Requiem for a Dream | 58 |
| Oldboy | 59 |
| Reservoir Dogs | 60 |
| A Separation | 61 |
| Toy Story 3 | 62 |
| Up | 63 |
| Inside Out | 64 |
| Batman Begins | 65 |
| Inglourious Basterds | 66 |

| Indiana Jones and the Last Crusade | 67 |
|---------------------------------------|-----|
| L.A. Confidential | 68 |
| Toy Story | 69 |
| Scarface | 70 |
| Eternal Sunshine of the Spotless Mind | 71 |
| Amadeus | 72 |
| Raging Bull | 73 |
| Metropolis | 74 |
| Unforgiven | 75 |
| Downfall | 76 |
| Room | 77 |
| Snatch | 78 |
| 2001: A Space Odyssey | 79 |
| Good Will Hunting | 80 |
| Snatch | 81 |
| The Sting | 82 |
| The Hunt | 83 |
| Some Like It Hot | 84 |
| Monty Python and the Holy Grail | 85 |
| Captain America: Civil War | 86 |
| How to Train Your Dragon | 87 |
| The Wolf of Wall Street | 88 |
| Finding Nemo | 89 |
| A Beautiful Mind | 90 |
| Casino | 91 |
| V for Vendetta | 92 |
| The Thing | 93 |
| Die Hard | 94 |
| Blade Runner | 95 |
| Gran Torino | 96 |
| Warrior | 97 |
| Howl's Moving Castle | 98 |
| Into the Wild | 99 |
| Pan's Labyrinth | 100 |
| The Big Lebowski | 101 |

| Incendies | 102 |
|--|-----|
| Gone with the Wind | 103 |
| The Secret in Their Eyes | 104 |
| Trainspotting | 105 |
| The Bridge on the River Kwai | 106 |
| Lock, Stock and Two Smoking Barrels | 107 |
| On the Waterfront | 108 |
| The Avengers | 109 |
| Guardians of the Galaxy | 110 |
| Mad Max: Fury Road | 111 |
| The Revenant | 112 |
| The Bourne Ultimatum | 113 |
| Pirates of the Caribbean: The Curse of the Black Pearl | 114 |
| Monsters, Inc. | 115 |
| The Martian | 116 |
| Shutter Island | 117 |
| Jurassic Park | 118 |
| Gone Girl | 119 |
| The Truman Show | 120 |
| The Avengers | 121 |
| Deadpool | 122 |
| Kill Bill: Vol. 1 | 123 |
| Prisoners | 124 |
| The Sixth Sense | 125 |
| Sin City | 126 |
| Rush | 127 |
| The Grand Budapest Hotel | 128 |
| Million Dollar Baby | 129 |
| The Help | 130 |
| No Country for Old Men | 131 |
| There Will Be Blood | 132 |
| 12 Years a Slave | 133 |
| Spotlight | 134 |
| Hotel Rwanda | 135 |
| The Imitation Game | 136 |

| The Princess Bride | 137 |
|------------------------------------|-----|
| Groundhog Day | 138 |
| The Sea Inside | 139 |
| Tae Guk Gi: The Brotherhood of War | 140 |
| Stand by Me | 141 |
| Akira | 142 |
| Elite Squad | 143 |
| The Terminator | 144 |
| Platoon | 145 |
| Butch Cassidy and the Sundance Kid | 146 |
| Donnie Darko | 147 |
| Annie Hall | 148 |
| The Wizard of Oz | 149 |
| Before Sunrise | 150 |
| The Best Years of Our Lives | 151 |
| Amores Perros | 152 |
| The Celebration | 153 |
| Rocky | 154 |
| X-Men: Days of Future Past | 155 |
| Ratatouille | 156 |
| Star Trek | 157 |
| Life of Pi | 158 |
| Casino Royale | 159 |
| Blood Diamond | 160 |
| The Incredibles | 161 |
| Cinderella Man | 162 |
| Big Fish | 163 |
| The Pursuit of Happyness | 164 |
| Kill Bill: Vol. 2 | 165 |
| Catch Me If You Can | 166 |
| The Iron Giant | 167 |
| JFK | 168 |
| Serenity | 169 |
| Magnolia | 170 |
| District 9 | 171 |
| | |

| Mystic River | 172 |
|---------------------------------|-----|
| Aladdin | 173 |
| Rain Man | 174 |
| Her | 175 |
| Dances with Wolves | 176 |
| Dead Poets Society | 177 |
| The Artist | 178 |
| The King's Speech | 179 |
| Brazil | 180 |
| In Bruges | 181 |
| Mulholland Drive | 182 |
| Slumdog Millionaire | 183 |
| Black Swan | 184 |
| The Perks of Being a Wallflower | 185 |
| True Romance | 186 |
| Dancer in the Dark | 187 |
| The Exorcist | 188 |
| Jaws | 189 |
| Casino Royale | 190 |
| Doctor Zhivago | 191 |
| The Straight Story | 192 |
| Fiddler on the Roof | 193 |
| Sicko | 194 |
| My Name Is Khan | 195 |
| The Sound of Music | 196 |
| Persepolis | 197 |
| Dallas Buyers Club | 198 |
| Shaun of the Dead | 199 |
| Sling Blade | 200 |
| Boyhood | 201 |
| Bowling for Columbine | 202 |
| Central Station | 203 |
| Young Frankenstein | 204 |
| Before Sunset | 205 |
| Waltz with Bashir | 206 |
| | |

| A Fistful of Dollars | 207 |
|-------------------------------------|-----|
| Avatar | 208 |
| The Hobbit: The Desolation of Smaug | 209 |
| Iron Man | 210 |
| Edge of Tomorrow | 211 |
| Big Hero 6 | 212 |
| The Hobbit: An Unexpected Journey | 213 |
| How to Train Your Dragon 2 | 214 |
| Toy Story 2 | 215 |
| Children of Men | 216 |
| The Insider | 217 |
| The Hateful Eight | 218 |
| The Bourne Identity | 219 |
| Almost Famous | 220 |
| Captain Phillips | 221 |
| Shrek | 222 |
| Hero | 223 |
| The Notebook | 224 |
| Glory | 225 |
| Walk the Line | 226 |
| Straight Outta Compton | 227 |
| The Blues Brothers | 228 |
| The Right Stuff | 229 |
| Taken | 230 |
| The Untouchables | 231 |
| The World's Fastest Indian | 232 |
| Edward Scissorhands | 233 |
| Glory | 234 |
| Ed Wood | 235 |
| My Fair Lady | 236 |
| Halloween | 237 |
| Hero | 238 |
| Hot Fuzz | 239 |
| Crouching Tiger, Hidden Dragon | 240 |
| Halloween | 241 |

| The Remains of the Day | 242 |
|----------------------------|-----|
| Boogie Nights | 243 |
| Letters from Iwo Jima | 244 |
| The Fighter | 245 |
| E.T. the Extra-Terrestrial | 246 |
| Crash | 247 |
| Amour | 248 |
| Nightcrawler | 249 |
| Little Miss Sunshine | 250 |

The movies under top 250 that are not in English are:

| Top_Foreign_Lang_Film | language |
|------------------------------------|------------|
| The Good, the Bad and the Ugly | Italian |
| City of God | Portuguese |
| Seven Samurai | Japanese |
| Spirited Away | Japanese |
| The Lives of Others | German |
| Children of Heaven | Persian |
| Amélie | French |
| Baahubali: The Beginning | Telugu |
| Princess Mononoke | Japanese |
| Das Boot | German |
| Oldboy | Korean |
| A Separation | Persian |
| Metropolis | German |
| Downfall | German |
| The Hunt | Danish |
| Howl's Moving Castle | Japanese |
| Pan's Labyrinth | Spanish |
| Incendies | French |
| The Secret in Their Eyes | Spanish |
| The Sea Inside | Spanish |
| Tae Guk Gi: The Brotherhood of War | Korean |
| Akira | Japanese |

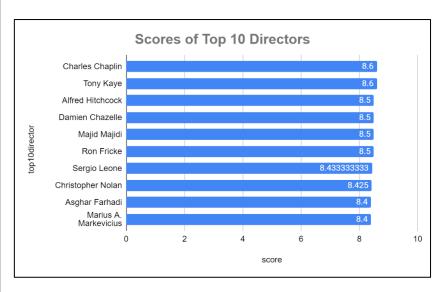
| Elite Squad | Portuguese |
|--------------------------------|------------|
| Amores Perros | Spanish |
| The Celebration | Danish |
| My Name Is Khan | Hindi |
| Persepolis | French |
| Central Station | Portuguese |
| Waltz with Bashir | Hebrew |
| A Fistful of Dollars | Italian |
| Hero | Mandarin |
| Hero | Mandarin |
| Crouching Tiger, Hidden Dragon | Mandarin |
| Letters from Iwo Jima | Japanese |
| Amour | French |
| Veer-Zaara | Hindi |
| The Chorus | French |
| Nine Queens | Spanish |
| 4 Months, 3 Weeks and 2 Days | Romanian |
| Apocalypto | Maya |
| Black Book | Dutch |
| Kung Fu Hustle | Cantonese |
| The White Ribbon | German |
| The Triplets of Belleville | French |
| Run Lola Run | German |
| The Lunchbox | Hindi |
| A Very Long Engagement | French |
| Ponyo | Japanese |
| The Great Beauty | Italian |
| The Red Violin | French |
| Good Bye Lenin! | German |
| Y Tu Mamá También | Spanish |
| Lady Vengeance | Korean |
| The Flowers of War | Mandarin |
| The Kite Runner | Dari |
| House of Flying Daggers | Mandarin |
| Headhunters | Norwegian |
| | |

| March of the Penguins | French |
|----------------------------------|------------|
| The Legend of Drunken Master | Cantonese |
| The Raid: Redemption | Indonesian |
| The Painted Veil | Mandarin |
| 2046 | Cantonese |
| The Orphanage | Spanish |
| Maria Full of Grace | Spanish |
| Red Cliff | Mandarin |
| The Baader Meinhof Complex | German |
| Mongol: The Rise of Genghis Khan | Mongolian |
| Paris, je t'aime | French |
| Shaolin Soccer | Cantonese |
| The Good, the Bad, the Weird | Korean |
| L'auberge espagnole | French |
| Soul Kitchen | German |
| Tsotsi | Zulu |
| District B13 | French |
| Timecrimes | Spanish |
| The Passion of the Christ | Aramaic |
| The Protector | Thai |
| Curse of the Golden Flower | Mandarin |
| The Host | Korean |
| Trollhunter | Norwegian |
| Jab Tak Hai Jaan | Hindi |
| El Mariachi | Spanish |
| High Tension | French |
| Coco Before Chanel | French |
| Rumble in the Bronx | Cantonese |
| Wasabi | French |
| [Rec] 2 | Spanish |
| Night Watch | Russian |
| The Interpreter | Aboriginal |
| Dead Snow | Norwegian |
| The Legend of Zorro | Spanish |
| In the Land of Blood and Honey | Bosnian |
| | |

3. Best Directors:

The top 10 directors along with their IMDB scores are:

| top10director | score |
|--------------------------|-------------|
| Charles Chaplin | 8.6 |
| Tony Kaye | 8.6 |
| Alfred Hitchcock | 8.5 |
| Damien Chazelle | 8.5 |
| Majid Majidi | 8.5 |
| Ron Fricke | 8.5 |
| Sergio Leone | 8.433333333 |
| Christopher Nolan | 8.425 |
| Asghar Farhadi | 8.4 |
| Marius A. Markevicius | 8.4 |



4. Popular genres:

The top 5 genres with their average gross values (in million \$) is given below:

| genre_1 | genre_2 | Average_gross |
|-----------|-----------|---------------|
| Family | Sci-Fi | 434.9 |
| Adventure | Sci-Fi | 228.6375 |
| Adventure | Animation | 116.9974359 |
| Adventure | Family | 116.5314286 |
| Action | Adventure | 110.3844125 |

The most viewed movies belong to family and Sci-Fi genres.

5. Critic favourite and audience favourite actors:

| actor_1_name | AVERAGE of num_critic_for_reviews | AVERAGE of num_user_for_reviews |
|-------------------|-----------------------------------|---------------------------------|
| Brad Pitt | 245 | 742.3529412 |
| Leonardo DiCaprio | 330.1904762 | 914.4761905 |
| Meryl Streep | 181.4545455 | 297.1818182 |

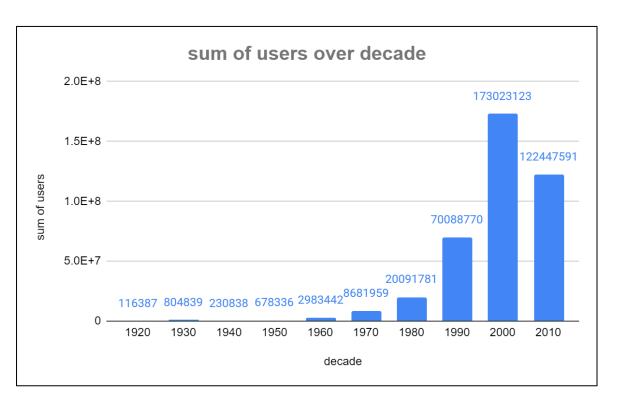
Leonardo DiCaprio is the most famous actor among the audience as well as the critics.

6. Change in voted users over decades:

| decade | df_by_decade |
|--------|--------------|
| 1920 | 116387 |
| 1930 | 804839 |
| 1940 | 230838 |
| 1950 | 678336 |
| 1960 | 2983442 |
| 1970 | 8681959 |
| 1980 | 20091781 |
| 1990 | 70088770 |
| 2000 | 173023123 |
| 2010 | 122447591 |

df_by_decade is the total number of voted users in that particular decade.

Maximum number of votes by users was obtained in the decade 2000.



| Result – | | |
|---|--|--|
| | MDB Movie Analysis helps us to analyse various properties of movies which includes their atings, number of likes, top movies, best actors and directors and so on. | |
| This project has helped me learn how to analyse and make decisions during broadcasting of a movie which will lead to the growth of the company (VOD distributor). | | |
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