

IMDB Movie Analysis



by Kiran Kumar



Project Description

As a data analyst we have to analyze what factors influence the success of a movie on IMDB Score or Ratings. The impact of this problem is significant for movie producers, directors, and investors who want to understand what makes a movie successful to make informed decisions in their future projects.



Approach

In the given dataset there were so many duplicates, null cells and movie names were having special characters, there was no proper alignment. So first we cleaned the data for getting right result. The following steps were followed:

- There were 45 duplicates which were removed
- The cells with null values were removed
- Cleaned movie title names which were having special characters
- Removed unwanted columns which were unnecessary before it was 27, after cleaning we have 10 columns
- Before cleaning the rows were 5044 and after cleaning the rows were 3724

So after cleaning the data we moved forward for our analysis

Before Cleaning

ıl A	В	С	D	E	F C	G H	T.	J	K	L	М	N O	P	Q R	S T	U	V	W	X	Υ	Z	AA	AB	AC
color	director_r nu	um_criticdura	tion d	lirector_f	actor_3_faactor	r_2_n actor_	1_fagross	genres	actor_1_n	movie_tit	num_vote c	ast_total actor_3_r	facenumb pl	ot_keyw movie_i	m num_user languag	e country	content	r budget	title_year	actor_2_fa	imdb_scor	aspect_rate	movie_faceb	ook_like
Color	James Car	723	178	0	855 Joel I	David 1	000 7.61E+	08 Action A	c CCH Poun	AvatarÂ	886204	4834 Wes Stud	i 0 av	atar fut http://v	w 3054 English	USA	PG-13	2.37E+08	2009	936	7.9	1.78	33000	
Color	Gore Verb	302	169	563	1000 Orlar	ndo B 40	000 3.09E+	08 Action A	c Johnny D	Pirates of	471220	48350 Jack Dave	0 gc	ddess rhttp://v	w 1238 English	USA	PG-13	3E+08	2007	5000	7.1	2.35	0	
Color	Sam Meno	602	148	0	161 Rory	Kinne 11	000 2E+	08 Action A	c Christoph	SpectreÂ	275868	11700 Stephani	1 bo	mb esr http://v	w 994 English	UK	PG-13	2.45E+08	2015	393	6.8	2.35	85000	
Color	Christoph	813	164	22000	23000 Chris	stian I 27	000 4.48E+	08 Action T	h Tom Hard	The Dark F	1144337	106759 Joseph G	0 de	ception http://v	w 2701 English	USA	PG-13	2.5E+08	2012	23000	8.5	2.35	164000	
	Doug Walke	r		131	Rob \	Walke	131	Documer	n Doug Wal	Star Wars:	8	143	0	http://v	/W					12	7.1		0	
Color	Andrew S	462	132	475	530 Sama	antha	640 730586	79 Action A	c Daryl Sab	John Carte	212204	1873 Polly Wal	l 1 al	en ame http://v	/w 738 English	USA	PG-13	2.64E+08	2012	632	6.6	2.35	24000	
Color	Sam Raim	392	156	0	4000 Jame	es Fra 24	000 3.37E+	08 Action A	c J.K. Simm	Spider-Ma	383056	46055 Kirsten D	ι 0 sa	ndman http://v	/w 1902 English	USA	PG-13	2.58E+08	2007	11000	6.2	2.35	0	
Color	Nathan Gr	324	100	15	284 Donr	na Mu	799 2.01E+	08 Adventu	Read Garr	€ TangledÂ	294810	2036 M.C. Gair	1 17	th centi http://v	w 387 English	USA	PG	2.6E+08	2010	553	7.8	1.85	29000	
Color	Joss Whee	635	141	0	19000 Robe	ert Do 26	000 4.59E+	08 Action A	c Chris Hen	Avengers:	462669	92000 Scarlett J	d 4 ar	tificial ir http://v	vv 1117 English	USA	PG-13	2.5E+08	2015	21000	7.5	2.35	118000	
Color	David Yate	375	153	282	10000 Danie	iel Rac 25	000 3.02E+	08 Adventu	r Alan Rick	Harry Pott	321795	58753 Rupert G	i 3 bl	ood bo http://v	w 973 English	UK	PG	2.5E+08	2009	11000	7.5	2.35	10000	
Color	Zack Snyd	673	183	0	2000 Laure	en Co 15	000 3.3E+	08 Action A	c Henry Cav	Batman v	371639	24450 Alan D. P	u 0 ba	sed on http://v	w 3018 English	USA	PG-13	2.5E+08	2016	4000	6.9	2.35	197000	
Color	Bryan Sing	434	169	0	903 Marle	on Br 18	000 2E+	08 Action A	c Kevin Spa	Superman	240396	29991 Frank Lan	€ 0 cr	ystal ep http://v	/w 2367 English	USA	PG-13	2.09E+08	2006	10000	6.1	2.35	0	
Color	Marc Forst	403	106	395	393 Math	hieu A		08 Action A			330784	2023 Rory Kinn	1 ac	tion her http://v	/w 1243 English	UK	PG-13	2E+08	2008	412	6.7	2.35	0	
Color	Gore Verb	313	151	563	1000 Orlan	ndo B 40	000 4.23E+	08 Action A	c Johnny D	Pirates of	522040	48486 Jack Dave	2 bo	x office http://v	/w 1832 English	USA	PG-13	2.25E+08	2006	5000	7.3	2.35	5000	
Color	Gore Verb	450	150	563	1000 Ruth	Wils 40	000 892899	10 Action A	c Johnny D	The Lone I	181792	45757 Tom Wilk	i 1 ho	rse out http://v	/w 711 English	USA	PG-13	2.15E+08	2013	2000	6.5	2.35	48000	
Color	Zack Snyd	733	143	0	748 Chris	stoph 15	000 2.91E+	08 Action A	c Henry Cav	Man of Ste	548573	20495 Harry Len	r 0 ba	sed on http://v	vv 2536 English	USA	PG-13	2.25E+08	2013	3000	7.2	2.35	118000	
Color	Andrew A	258	150	80	201 Pierf	france 22	000 1.42E+	08 Action A	c Peter Din	The Chron	149922	22697 DamiÃin	4 br	other brhttp://v	w 438 English	USA	PG	2.25E+08	2008	216	6.6	2.35	0	
Color	Joss Whee	703	173	0	19000 Robe	ert Do 26	000 6.23E+	08 Action A	c Chris Hen	The Aveng	995415	87697 Scarlett J	a 3 al	en inva http://v	w 1722 English	USA	PG-13	2.2E+08	2012	21000	8.1	1.85	123000	
Color	Rob Marsh	448	136	252	1000 Sam	Clafli 40	000 2.41E+	08 Action A	c Johnny D	Pirates of	370704	54083 Stephen	6 4 bl	ackbear http://v	w 484 English	USA	PG-13	2.5E+08	2011	11000	6.7	2.35	58000	
Color	Barry Soni	451	106	188	718 Mich	nael St 10	000 1.79E+	08 Action A	c Will Smit	Men in Bla	268154	12572 Nicole Sc	h 1 al	en crin http://v	w 341 English	USA	PG-13	2.25E+08	2012	816	6.8	1.85	40000	
Color	Peter Jack	422	164	0	773 Adan	m Bro 5	000 2.55E+	08 Adventu	re Aidan Tur	The Hobbi	354228	9152 James Ne	e 0 ar	my elf http://v	w 802 English	New Zea	ala PG-13	2.5E+08	2014	972	7.5	2.35	65000	
Color	Marc Web	599	153	464	963 Andr	rew G 15	000 2.62E+	08 Action A	c Emma Sto	The Amaz	451803	28489 Chris Zylk	a O liz	ard out http://v	/w 1225 English	USA	PG-13	2.3E+08	2012	10000	7	2.35	56000	
Color	Ridley Scc	343	156	0	738 Willi	iam H	891 1.05E+	08 Action A	c Mark Add	Robin Hoc	211765	3244 Scott Grir	r 0 11	90s arc http://v	/w 546 English	USA	PG-13	2E+08	2010	882	6.7	2.35	17000	
Color	Peter Jack	509	186	0	773 Adan	m Bro 5	000 2.58E+	08 Adventu	re Aidan Tur	The Hobbi	483540	9152 James Ne	6 dv	varf elf http://v	vv 951 English	USA	PG-13	2.25E+08	2013	972	7.9	2.35	83000	
Color	Chris Weit	251	113	129	1000 Eva G	Green 16	000 700835	19 Adventu	r Christoph	The Golde	149019	24106 Kristin Sc	c 2 ch	ildren ehttp://v	w 666 English	USA	PG-13	1.8E+08	2007	6000	6.1	2.35	0	
Color	Peter Jack	446	201	0	84 Thon	mas Kı 6	000 2.18E+	08 Action A	c Naomi W	King Kong	316018	7123 Evan Park	0 ar	imal na http://v	vv 2618 English	New Zea	ala PG-13	2.07E+08	2005	919	7.2	2.35	0	
Color	James Car	315	194	0	794 Kate	Wins 29	000 6.59E+	08 Drama R	c Leonardo	TitanicÂ	793059	45223 Gloria Stu	0 ar	tist love http://v	w 2528 English	USA	PG-13	2E+08	1997	14000	7.7	2.35	26000	
Color	Anthony F	516	147	94	11000 Scarl	lett Jc 21	000 4.07E+	08 Action A	c Robert Do	Captain A	272670	64798 Chris Eva	n 0 ba	sed on (http://v	/w 1022 English	USA	PG-13	2.5E+08	2016	19000	8.2	2.35	72000	
Color	Peter Berg	377	131	532	627 Alexa	andei 14	000 651731	60 Action A	c Liam Nee	Battleship	202382	26679 Tadanobi	u 0 bo	x office http://v	/w 751 English	USA	PG-13	2.09E+08	2012	10000	5.9	2.35	44000	
Color	Colin Trev	644	124	365	1000 Judy	Gree 3	000 6.52E+	08 Action A	c Bryce Dal	Jurassic W	418214	8458 Omar Sy	0 di	nosaur http://v	/w 1290 English	USA	PG-13	1.5E+08	2015	2000	7	2	150000	
Color	Sam Meno	750	143	0	393 Hele	n McC	883 3.04E+	08 Action A	c Albert Fir	SkyfallÂ	522030	2039 Rory Kinn	0 br	awl chi http://v	/w 1498 English	UK	PG-13	2E+08	2012	563	7.8	2.35	80000	
Color	Sam Raim	300	135	0	4000 Jame	es Fra 24	000 3.73E+	08 Action A	c J.K. Simm	Spider-Ma	411164	43388 Kirsten D	1 de	ath dochttp://v	/w 1303 English	USA	PG-13	2E+08	2004	11000	7.3	2.35	0	
Color	Shane Bla	608	195	1000	3000 Jon F	Favrea 21	000 4.09E+	08 Action A	c Robert Do	Iron Man	557489	30426 Don Chea	a 3 ar	mor ex http://v	/w 1187 English	USA	PG-13	2E+08	2013	4000	7.2	2.35	95000	
Color	Tim Burto	451	108	13000	11000 Alan	Rickr 40	000 3.34E+	08 Adventu	re Johnny De	Alice in W	306320	79957 Anne Hat	l 0 al	ce in wchttp://v	vv 736 English	USA	PG	2E+08	2010	25000	6.5	1.85	24000	
Color	Brett Ratn	334	104	420	560 Kelse	ey Gra 20	000 2.34E+	08 Action A	c Hugh Jack	X-Men: Th	383427	21714 Daniel Cu	0 ba	ttle muhttp://v	w 1912 English	Canada	PG-13	2.1E+08	2006	808	6.8	2.35	0	
Color	Dan Scanle	376	104	37	760 Tyler	r Labii 12	000 2.68E+	08 Adventu	Steve Bus	Monsters	235025	14863 Sean Hay	€ 0 ch	eating http://v			G	2E+08	2013	779	7.3	1.85	44000	
Color	Michael B	366	150	0	464 Kevir	n Dun	894 4.02E+	08 Action A	c Glenn Mo	Transform	323207	3218 Ramon Re	0 au	tobot c http://v	/w 1439 English	USA	PG-13	2E+08	2009	581	6	2.35	0	
Color	Michael B	378	165	0	808 Soph	nia My	974 2.45E+	08 Action A	c Bingbing	Transform	242420	3988 Kelsey Gr	a 2 bl	ockbust http://v			PG-13	2.1E+08	2014	956	5.7	2.35	56000	
Color	Com Daim	525	120	0	11000 Mila	Kunic 44	000 2 2554	08 Advontu	e Tim Halm	Oztho Cre	175400	72441 James Fr	4 61	ouel ma bttp://u	n E11 Fnalich	LICA	DC.	2.155100	2012	15000	6.4	2.25	60000	

After Cleaning

	A	В	C	D	E	F	G	Н		J
	director_name	duration		genres	movie_title	language	country	budget		imdb_score
	2 James Cameron		760505847		Avatar	English	USA	237000000		7.9
	Gore Verbinski		309404152		Pirates of the Caribbean: At World's End	English	USA	300000000		7.1
ı	4 Sam Mendes	148	200074175	Action	Spectre	English	UK	245000000		6.8
	Christopher Nolan	164	448130642		The Dark Knight Rises	English	USA	250000000		8.5
	6 Andrew Stanton	132	73058679	Action	John Carter	English	USA	263700000		6.6
	7 Sam Raimi	156	336530303	Action	Spider-Man 3	English	USA	258000000		6.2
	Nathan Greno	100	200807262	Adventure	Tangled	English	USA	260000000	2010	7.8
	Joss Whedon	141	458991599	Action	Avengers: Age of Ultron	English	USA	250000000	2015	7.5
	0 David Yates	153	301956980	Adventure	Harry Potter and the Half-Blood Prince	English	UK	250000000	2009	7.5
	1 Zack Snyder	183	330249062	Action	Batman v Superman: Dawn of Justice	English	USA	250000000	2016	6.9
	2 Bryan Singer	169	200069408	Action	Superman Returns	English	USA	209000000	2006	6.1
	3 Marc Forster	106	168368427	Action	Quantum of Solace	English	UK	200000000	2008	6.7
/	4 Gore Verbinski	151	423032628	Action	Pirates of the Caribbean: Dead Man's Chest	English	USA	225000000	2006	7.3
	5 Gore Verbinski	150	89289910	Action	The Lone Ranger	English	USA	215000000	2013	6.5
	6 Zack Snyder	143	291021565	Action	Man of Steel	English	USA	225000000	2013	7.2
	7 Andrew Adamson	150	141614023	Action	The Chronicles of Narnia: Prince Caspian	English	USA	225000000	2008	6.6
	8 Joss Whedon	173	623279547	Action	The Avengers	English	USA	220000000	2012	8.1
	9 Rob Marshall	136	241063875	Action	Pirates of the Caribbean: On Stranger Tides	English	USA	250000000	2011	6.7
	0 Barry Sonnenfeld	106	179020854	Action	Men in Black 3	English	USA	225000000	2012	6.8
	1 Peter Jackson	164	255108370	Adventure	The Hobbit: The Battle of the Five Armies	English	New Zealand	250000000	2014	7.5
	2 Marc Webb	153	262030663	Action	The Amazing Spider-Man	English	USA	230000000	2012	7
	3 Ridley Scott	156	105219735	Action	Robin Hood	English	USA	200000000	2010	6.7
	4 Peter Jackson	186	258355354	Adventure	The Hobbit: The Desolation of Smaug	English	USA	225000000	2013	7.9



Tech Stack Used

In this analytics process we used MS- Excel 2021 for analyzing the dataset and for presentation we used MS-Powerpoint 2021





A. Movie Genre Analysis: Analyze the distribution of movie genres and their impact on the IMDB score.

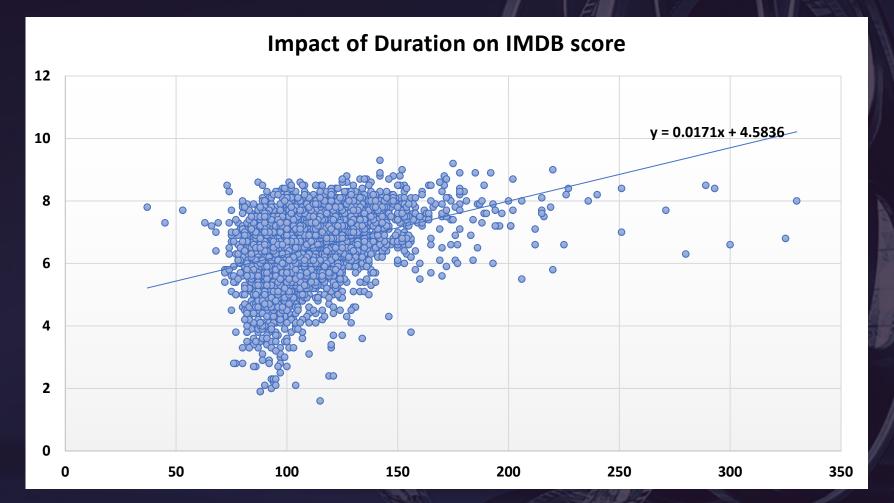
Task: Determine the most common genres of movies in the dataset.

Then, for each genre, calculate descriptive statistics (mean, median, mode, range, variance, standard deviation) of the IMDB scores.

Unique Genres	Total Genres	MEAN(Average)	MEDIAN	MODE	MAX	MIN	RANGE	VARIANCE	STD DEV.
Comedy	984	6.17	6.30	6.40	8.80	1.90	6.90	1.05	1.03
Action	951	6.29	6.30	6.10	9.00	2.10	6.90	1.06	1.03
Drama	659	6.83	6.90	6.70	8.80	2.10	6.70	0.82	0.91
Adventure	366	6.56	6.70	7.30	8.60	2.30	6.30	1.26	1.12
Crime	253	6.94	7.00	7.30	9.30	3.30	6.00	0.75	0.87
Biography	204	7.16	7.20	7.40	8.90	4.50	4.40	0.48	0.70
Horror	159	5.84	5.90	6.70	8.50	2.30	6.20	1.05	1.03
Animation	45	6.74	7.00	7.10	8.00	4.50	3.50	0.92	0.96
Fantasy	37	6.28	6.50	6.80	7.90	4.30	3.60	0.78	0.88
Documentary	26	6.80	7.45	7.50	8.50	1.60	6.90	2.83	1.68
Mystery	23	6.65	6.70	7.10	8.50	3.30	5.20	1.14	1.07
Sci-Fi	7	6.63	6.40	6.70	8.20	5.00	3.20	1.05	1.02
Family	3	6.50	5.90	6.70	7.90	5.70	2.20	0.99	0.99
Western	2	8.10	8.10	6.70	8.90	7.30	1.60	0.64	0.80
Musical	2	6.75	6.75	6.70	7.20	6.30	0.90	0.20	0.45
Romance	1	7.10	7.10	6.70	7.10	7.10	0.00	0.00	0.00
Thriller	1	4.80	4.80	6.70	4.80	4.80	0.00	0.00	0.00

B. Movie Duration Analysis: Analyze the distribution of movie durations and its impact on the IMDB score Task: Analyze the distribution of movie durations and identify the relationship between movie duration and IMDB score

Duration of Movie							
Average	110.26						
Median	106.00						
Stdev	22.68						



• Here the Trendline is going upwards means, longer length movies are getting good IMDB Score, so there is a positive relation for movie duration and IMDB score.

C. Language Analysis: Situation: Examine the distribution of movies based on their language.

Task: Determine the most common languages used in movies and analyze their impact on the IMDB score using descriptive statistics.

There are total 34 languages movies in the given Dataset, among them we have taken top 10 languages movies which are very high in number. These are:

Languages	Total Movies	Mean	Median	Stdev	
English	3566	6.43	6.50	1.05	
French	34	7.36	7.30	0.51	
Spanish	23	7.08	7.20	0.84	
Mandarin	14	7.02	7.25	0.74	
Japanese	10	7.66	8.00	0.94	
German	10	7.77	7.80	0.68	
Cantonese	7	7.34	7.30	0.32	
Italian	7	7.19	7.00	1.07	
Korean	5	7.70	7.70	0.51	
Hindi	5	7.22	7.40	0.72	

- Among all the languages "English" is the most common language used in movies
- The average IMDB score of English is 6.43 and all other languages are mentioned in the table
- The Highest average IMDB score is language is "German"

D. Director Analysis: Influence of directors on movie ratings.

<u>Task</u>: Identify the top directors based on their average IMDB score and analyze their contribution to the success of movies using percentile calculations.

There are total 1659 Directors in the given dataset after, among them we have taken top 10 directors whose IMDB score is very high and movies are above 10. They are:

Director	Total Movies	Avg. IMDB
Martin Scorsese	15	7.68
Steven Spielberg	24	7.54
Richard Linklater	11	7.33
Clint Eastwood	19	7.21
Rob Reiner	11	7.02
Woody Allen	19	7.00
Oliver Stone	12	6.91
John Carpenter	12	6.83
Spike Lee	15	6.73
Steven Soderbergh	13	6.71
Barry Levinson	11	6.58
Wes Craven	11	6.11

The contribution of Directors to the success of movies is 75% (Done using percentile functions)

E. Budget Analysis: Explore the relationship between movie budgets and their financial success.

Task: Analyze the correlation between movie budgets and gross earnings, and identify the movies with the highest profit margin.

The Correlation between Movie budget and gross earnings is 0.0983(so it is mostly on the positive side), with this we can see that as budget increases it will bring profits to the movie.

The top 10 with highest profit margin are:

budget	gross	Profit	Highest Profit	Highest profit Movie
237000000	760505847	523505847	523505847	Avatar
150000000	652177271	502177271		
20000000	658672302	458672302		
11000000	460935665	449935665		
10500000	434949459	424449459		
220000000	623279547	403279547		
45000000	422783777	377783777		
115000000	474544677	359544677		
185000000	533316061	348316061		
78000000	407999255	329999255		
	150000000 200000000 11000000 10500000 220000000 45000000 115000000	227000000 760505847 150000000 652177271 200000000 658672302 11000000 460935665 10500000 434949459 220000000 623279547 45000000 474544677 185000000 533316061	287000000 760305347 SY8305847 150000000 652177271 502177271 200000000 658672302 458672302 11000000 460935665 449935665 10500000 434949459 424449459 220000000 623279547 403279547 45000000 474544677 359544677 185000000 533316061 348316061	budget gross Profit Profit 287000000 780505277 52350527 52350527 150000000 652177271 502177271 200000000 658672302 458672302 11000000 460935665 449935665 10500000 434949459 424449459 220000000 623279547 403279547 45000000 422783777 377783777 115000000 474544677 359544677 185000000 533316061 348316061





Result

The purpose of this presentation is to analyze IMDB movie ratings and identify what factors influence them. By the end of this presentation, you will have a better understanding of what makes a good movie. This project made me think as an analyst and do the work on it to get the desired solutions. Thank you for giving me this opportunity to explore myself to understand the Data Analytics.

Note: You can checkout my working file here

https://docs.google.com/spreadsheets/d/18cPMhpff Ac86F3kcAXK91hnSNoOtxMoKgs4fSe8XAcg/edit?usp =sharing

THANK 40U