THE NUMBERS.COM

BOT DATA EXTRACTION, 1000 MOVIES ANALYSIS



BY:

DIVYAM BAJAJ

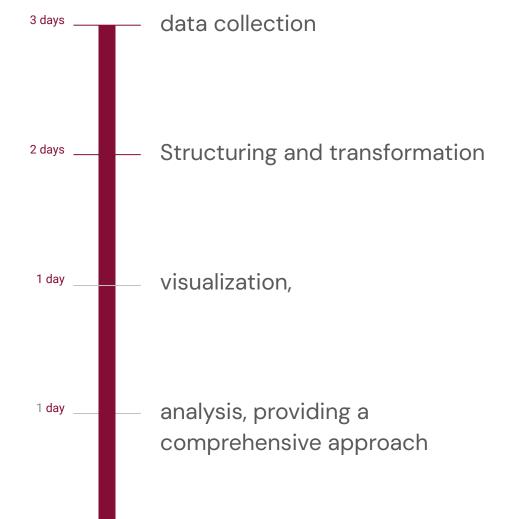


Objective of the Project

To analyze a dataset of over 1,000 movies based on IMDb ratings, worldwide box office performance, and profitability, examining various factors such as genre, profit margin, sources, production methods, production companies, and directors.



Workflow Overview



Data Collection Process

A Selenium bot was developed to extract data from The-Numbers.com, effectively navigating complex HTML structures across multiple pages to collect data

on at least 1,000 movies.

```
from urllib.parse import urljoin
from selenium import webdriver
from selenium.webdriver.chrome.service import Service as ChromeService
from webdriver manager.chrome import ChromeDriverManager
from selenium.webdriver.common.by import By
from selenium.webdriver.support.ui import WebDriverWait
from selenium.webdriver.support import expected conditions as EC
from selenium.webdriver.common.action chains import ActionChains
from selenium.common.exceptions import NoSuchElementException
import re
from time import sleep
chrome options = webdriver.ChromeOptions()
  chrome options.add argument("--start-maximized")
chrome options.add argument(argument="headless")
  chrome options.add argument("--ignore-certificate-errors")
main url = "https://www.the-numbers.com/movie/budgets/all"
driver = webdriver.Chrome(service=ChromeService(executable path=ChromeDriverManager()
```

Retrieving IMDb Ratings

IMDb ratings were retrieved via API calls using a search method to match titles, which were then integrated into the main dataset for analysis.

```
import requests
imdb = []
for x in df["Movie"]:
    try:
        base url = "http://www.omdbapi.com"
        parameters = {"t": x, "apikey": "56da1f94"}
        response = requests.get(url=base_url, params = parameters)
        detail = response.json()
        imdb.append(object/detail["imdbRating"])
    except:
        imdb.append(object/"")
print(imdb)
```

Data Structuring Techniques

Data was organized into dataframes using Pandas

an	d NumF	y arra	ys, pre	paring	the da	ataset	for							
th	orough	analysi	is and	visualiz	zation.									
Release_da	e Movie	Budget	mestic_gra	ldwide_gr	source	director	iction_con	uction_co	uction_m	genre	language	st_perform	imdb	profit
Dec 16, 201	Star Wars	5.33E+08	9.37E+08	2.06E+09	Original Sc	J.J. Abram	Lucasfilm,	United Sta	Animation	Adventure	English	United Kin	7.167	1.52E+09

Dec 9, 2022	Avatar: Th	4.6E+08	6.84E+08	2.32E+09	Original S	c James Car	r Lightstorn	United Sta Animation Action	English	China	7.5	1.86E+09
Jun 28, 2023	Indiana Jo	4.02E+08	1.74E+08	3.84E+08	Original S	cJames Ma	Lucasfilm,	United Sta Live Actior Advent	ure English	United Kin	6.5	-1.8E+07
Apr 23, 2019	Avengers:	4E+08	8.58E+08	2.75E+09	Based on	Joe Russo	Marvel St	United Sta Animation Action	English	China	8.4	2.35E+09
May 20, 2011	Pirates of	3.79E+08	2.41E+08	1.05E+09	Based on	Rob Mars	l Walt Disn	United Sta Live Actior Advent	ure English		6.6	6.67E+08
Apr 22, 2015	Avengers:	3.65E+08	4.59E+08	1.4E+09	Based on	Joss Whe	c Marvel Stu	United Sta Animation Action	English	China	7.3	1.03E+09
May 17, 2023	Fast X	3.4E+08	1.46E+08	7.15E+08	Original S	Louis Lete	Universal	United Sta Live Action Action	English	China	5.7	3.75E+08

Apr 22, 2015	Avengers:	3.65E+08	4.59E+08	1.4E+09	Based on	Joss Wned	. Marvel St	United Sta	Animation	Action	English	China	7.3	1.03E+0
May 17, 2023	Fast X	3.4E+08	1.46E+08	7.15E+08	Original So	Louis Lete	Universal	United Sta	Live Action	Action	English	China	5.7	3.75E+08
May 23, 2018	Solo: A Sta	3.3E+08	2.14E+08	3.93E+08	Spin-Off	Ron Howa	Lucasfilm	United Sta	Animation	Adventu	re English	United Kin	6.9	6275134
Apr 25, 2018	Avengers:	3E+08	6.79E+08	2.05E+09	Based on	Joe Russo	Marvel St	United Sta	Animation	Action	English	China	8.4	1.75E+09

3E+08 3.09E+08 9.61E+08 Based on Gore Verb Walt Disne United Sta Live Action Adventure English May 24, 2007 Pirates of Nov 13, 2017 Justice Lea 3E+08 2.29E+08 6.56E+08 Based on (Zack Snyd) DC Films, I United Sta Live Action Action English

7.1 6.61E+08 China

6.1 3.56E+08 Jul 11, 2023 Mission: It 2.9E+08 1.73E+08 5.67E+08 Based on Christoph Paramoun United Sta Live Action Action China English 7.7 2.77E+08

Dec 14, 2016 Rogue On 2.8E+08 5.34E+08 1.06E+09 Spin-Off Felicity Jor Lucasfilm United Sta Animation Adventure English United Kin 7.8 7.75E+08 Dec 18, 2019 Star Wars 2,75E+08 5,15E+08 1,07E+09 Original ScJ.J. Abram Lucasfilm, United Sta Animation Adventure English United Kin 6.118 7.95E+08

Missing Data Transformation

Missing values were addressed using Random Forest prediction, while regex was employed to convert special characters (e.g., "\$") to numerical data, ensuring data

```
integrity.
             def random_forest_impute(database, target_column):
                 df train = database.loc[database[target column].notna()]
                 df missing = database.loc[database[target_column].isna()]
                 predictors = ['Budget']
                 X train = df train[predictors]
                 y train = df train[target column]
                 rf = RandomForestRegressor(n estimators=100, random state=42)
                 rf.fit(X=X_train, y=y_train)
                 # Predict missing values using the trained model
                 X missing = df missing[predictors]
                 predicted values = rf.predict(X=X missing)
                 database.loc[database[target_column].isna(), target_column] = predicted_values
                 return database
```

Data Visualization Tools

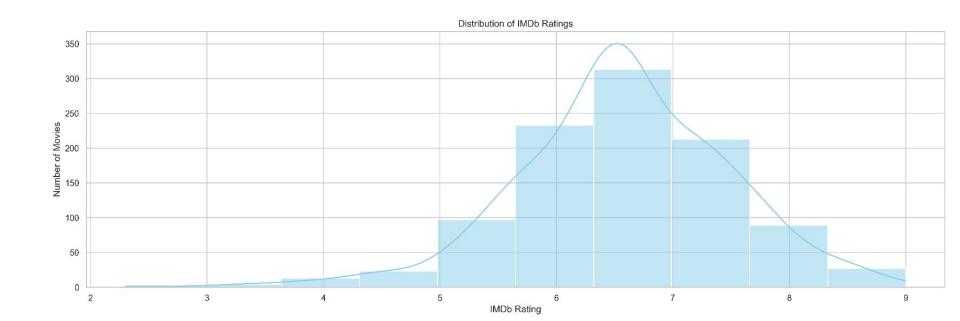
Using Matplotlib, Seaborn, and Power BI, various visuals were created, including:

- Histogram: Distribution of IMDb ratings.
- Correlation Matrix: Analyzing IMDb and profit by genre.
- Bar Graphs: Worldwide gross by genre and performance by production method.
- Line Chart: Top-performing countries by earnings.
- Map: Geographic distribution of top-performing countries, genres, and production companies.
- Pie & Donut Charts: Relationship between genres and sources.
- Box Plot: IMDb ratings by genre and source.
- Bubble and Table Charts: Movie distribution by genre, source, and rating.

Correlation Matrix: Analyzing IMDb and profit by genre

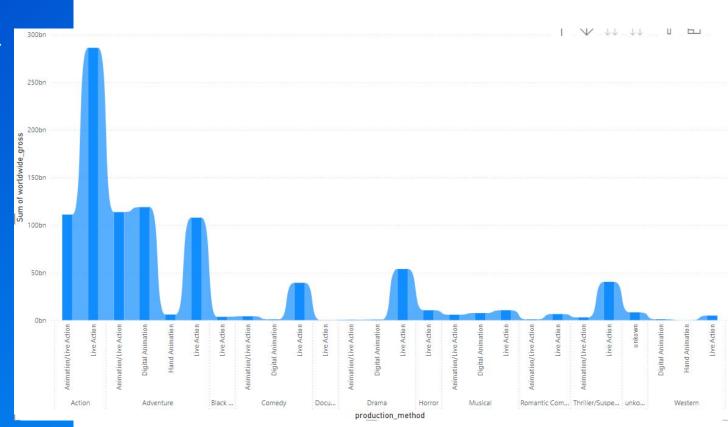
			C	Correlatio	n Matrix	of IMDb	Ratings	, Profit, a	and Geni	е				- 1.00
imdb	1.00	0.39	0.08	0.01	-0.14	0.04	0.16	-0.07	0.03	-0.03	0.03	0.00		1.00
profit	0.39	1.00	0.16	-0.04	-0.11	-0.02	-0.09	-0.04	0.06	-0.04	-0.09	-0.06		- 0.75
genre_Adventure	0.08	0.16	1.00	-0.05	-0.20	-0.02	-0.21	-0.09	-0.10	-0.08	-0.18	-0.07		
enre_Black Comedy	0.01	-0.04	-0.05	1.00	-0.02	-0.00	-0.02	-0.01	-0.01	-0.01	-0.02	-0.01		- 0.50
genre_Comedy	-0.14	-0.11	-0.20	-0.02	1.00	-0.01	-0.08	-0.03	-0.04	-0.03	-0.07	-0.03		- 0.25
genre_Documentary	0.04	-0.02	-0.02	-0.00	-0.01	1.00	-0.01	-0.00	-0.00	-0.00	-0.01	-0.00		
genre_Drama	0.16	-0.09	-0.21	-0.02	-0.08	-0.01	1.00	-0.04	-0.04	-0.03	-0.07	-0.03		- 0.00
genre_Horror	-0.07	-0.04	-0.09	-0.01	-0.03	-0.00	-0.04	1.00	-0.02	-0.01	-0.03	-0.01		- -0.25
genre_Musical	0.03	0.06	-0.10	-0.01	-0.04	-0.00	-0.04	-0.02	1.00	-0.02	-0.04	-0.01		0.50
_Romantic Comedy	-0.03	-0.04	-0.08	-0.01	-0.03	-0.00	-0.03	-0.01	-0.02	1.00	-0.03	-0.01		- -0.50
e_Thriller/Suspense	0.03	-0.09	-0.18	-0.02	-0.07	-0.01	-0.07	-0.03	-0.04	-0.03	1.00	-0.03		- -0.75
genre_Western	0.00	-0.06	-0.07	-0.01	-0.03	-0.00	-0.03	-0.01	-0.01	-0.01	-0.03	1.00		4.05
	qpmi	profit	dventure	Comedy	Comedy	umentary	Drama	e_Horror	Musical	Comedy	esuedsn	Western		1.00

Histogram: Distribution of IMDb ratings.

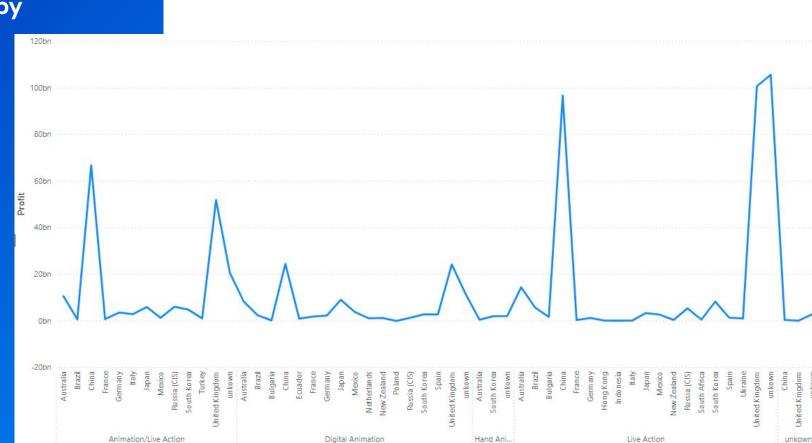


Bar Graphs:
Worldwide gross by genre and performance by production

method.



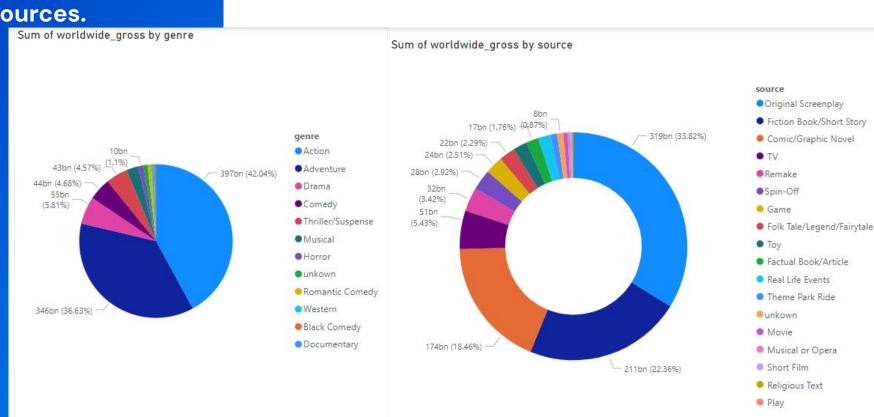
Line Chart:
Top-performing
countries by
earnings.



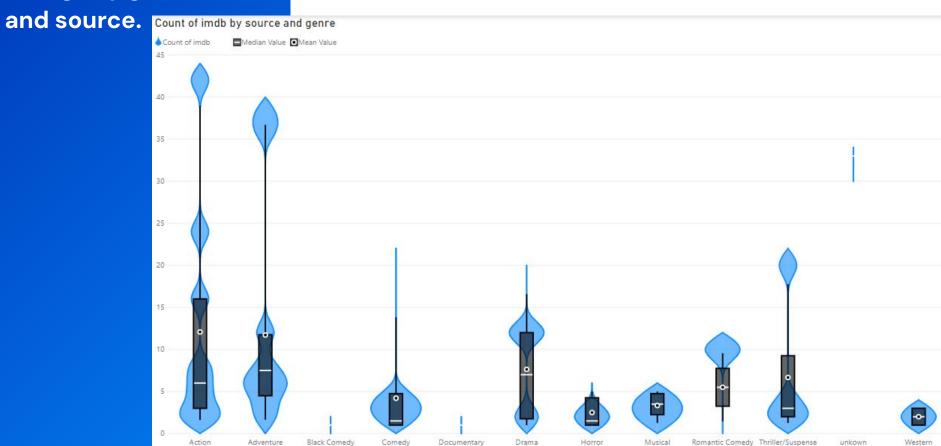
Map: Geographic distribution of top-performing countries, genres, and production companies.



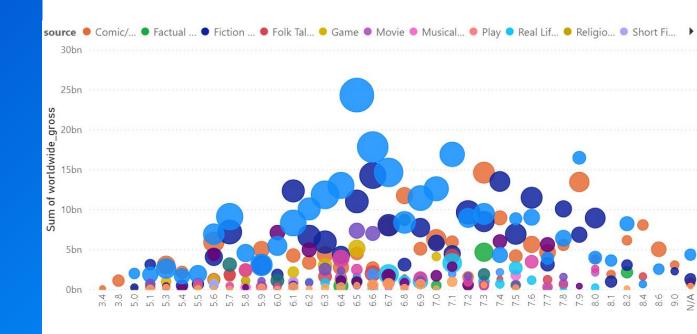
Pie & Donut Charts: Relationship between genres and sources.



Box Plot: IMDb ratings by genre



Bubble Charts: Movie distribution by genre, source, and rating.



imdb

Table for Genre and Imdb ratings with values for

Budge	et.													
genre		8.1	8.2	8.3	8.4	8.5	8.6	8.7	8.8	8.9	9.0	N/A	unkown	Total
Action	00	531000000	1260000000		1530000000	792000000	700000000	195000000			555000000	120000000	5014000000	129915600000
Adventure	00	1345000000	314000000	750000000		1516600000	600000000	330000000	94000000	186000000	188000000	1545000000	5365400000	108725000000
Black Comedy			300000000										700000000	1916203077

genre		8.1	8.2	8.3	8.4	8.5	8.6	8.7	8.8	8.9	9.0	N/A	unkown	Total
Action	00	531000000	1260000000		1530000000	792000000	700000000	195000000			555000000	120000000	5014000000	1299156000
Adventure	00	1345000000	314000000	750000000		1516600000	600000000	330000000	94000000	186000000	188000000	1545000000	5365400000	1087250000
Black Comedy			300000000										700000000	19162030
												576000000	122700000	467245000

Action	00	531000000	1260000000		1530000000	792000000	700000000	195000000			555000000	120000000	5014000000	1299156000
Adventure	00	1345000000	314000000	750000000		1516600000	600000000	330000000	94000000	186000000	188000000	1545000000	5365400000	1087250000
Black Comedy			300000000										700000000	19162030
Comedy												576000000	1237000000	167315000
D														4600000

Adventure	10	1345000000	314000000	750000000	1516600000	600000000	330000000	94000000	186000000	188000000	1545000000	5365400000	108/25000000
Black Comedy			300000000									700000000	1916203077
Comedy											576000000	1237000000	16731500000
Documentary													160000000
Drama	00		234000000	172000000	360000000	130000000		195000000				1620400000	21827200000
Horror													3953500000

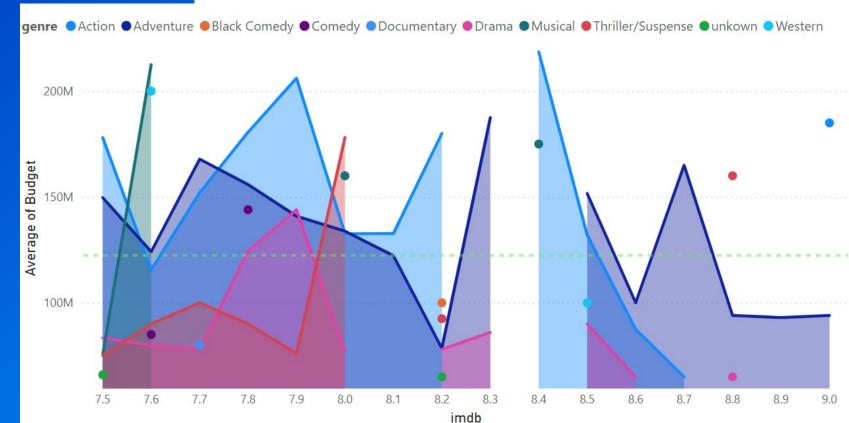
Comedy								576000000	1237000000	16731500000
Documentary										160000000
Drama	00	234000000	172000000		360000000	130000000	195000000		1620400000	21827200000
Horror	111									3953500000
Musical	00			350000000					375000000	6400000000
Romantic Comedy								120000000	65000000	2490000000
Thriller/Suspense	00	740000000					480000000		425000000	18780884985

Documentary										160000000
Drama	00	234000000	172000000		360000000	130000000	195000000		1620400000	21827200000
Horror										3953500000
Musical	00		3500	000000					375000000	6400000000
Romantic Comedy								120000000	65000000	2490000000
Thriller/Suspense	00	740000000					48000000		425000000	18780884985
unkown		65000000						110000000	468100000	5311282975
Western				3	300000000					3991000000
E-174 0.55										

Total	0	1876000000	2913000000	922000000	1880000000	2968600000	1430000000	525000000	769000000	186000000	743000000	2471000000	15269900000	320202171037
Western						300000000								3991000000
unkown			65000000									110000000	468100000	5311282975
Thriller/Suspense	00		740000000						480000000				425000000	18780884985
Romantic Comedy												120000000	65000000	2490000000
Musical	00				350000000								375000000	6400000000
Horror														3953500000
Drama	00		234000000	172000000		360000000	130000000		195000000				1620400000	21827200000
Documentary														160000000
Comedy												576000000	1237000000	16731500000

Graph for Genre and Imdb ratings with values for

Budget.



Key Findings from the Analysis

Key findings include a strong correlation between genre and IMDb ratings, regional performance insights, and the ranking of top production companies, highlighting trends in the movie industry.

- Genre Correlation: Drama correlates highly with IMDb ratings, while adventure links to higher profits.
- IMDb Ratings: Most movies fall within the 6-7 IMDb rating range.
- Regional Performance: Action movies perform best in China;
 China and the UK lead in live-action and animation.
- Top Production Companies: Walt Disney, Warner Bros, Marvel Studios, Universal Pictures, Paramount, and DreamWorks rank as global leaders.
- Genre & Source Performance: Action is the top genre, followed by adventure, with original screenplays as the leading source.
- Action Movies have the highest Average Budgets and Adventure genre movies have average budgets around \$125M.

Challenges Encountered

Challenges included developing a bot for dynamic pages, writing precise XPaths, using multiple tabs in Selenium, extracting country performance data, and handling financial data anomalies, which were all addressed during the project.

- Developing a bot to handle dynamic pages and navigate complex HTML.
- Writing precise XPaths for elements without clear indexing.
- Using multiple tabs in Selenium for data collection.
- Extracting country performance from charts with regex.
- Handling "\$0" values in financial data with regex and predictive imputation.
- Training Random Forest Model for Missing Values.
- Plotting Correlation Matrix including categorical variable.