

Data Attributes

Attribute	Description
id	id of film
imdb_id	Engine displacement - the size of an engine in liters
popularity	Each model builds their popularity value slightly differently. For movies: Number of votes for the day, number of views for the day, number of users who marked it as a "favourite" for the day, number of users who added it to their "watchlist" for the day, release date, number of total votes, previous days score
budget	The money spend on production process
revenue	Earned money from the film
original_title	
cast	
homepage	
director	
tagline	
keywords	
overview	
runtime	
genres	
production_companies	
vote_count	Count of Ratings
vote_average	Average Ratings
release_year	
budget_adj	
revenue_adj	

Questions related to this data

- Which movies have the highest revenue or profit?
- Which genres have the highest profit?
- Movie with Highest And Lowest Budget?
- Which movie get the highest or lowest votes (Ratings).
- Is there any relationship between the popularity and the budget?
- Which genres are most popular year by year?
- Which directors directed the most popular movie in the last years?

Import Packages

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
```

Gathering Data

```
In [2]: df = pd.read_csv('data/tmdb-movies.csv')
```

Assessing Data

- number of samples in each dataset : 10866
- number of columns in each dataset : 21

```
In [3]: df.shape
```

```
Out[3]: (10866, 21)
```

- features with missing values : ('imdb_id', 'cast', 'homepage', 'director', 'tagline', 'keywords', 'overview', 'genres', 'production_companies')

```
In [4]: df.isnull().any()
```

```
Out[4]: id                False
imdb_id                True
popularity             False
budget                False
revenue               False
original_title         False
cast                  True
homepage              True
director              True
tagline               True
keywords              True
overview              True
runtime               False
genres                True
production_companies  True
release_date          False
vote_count            False
vote_average          False
release_year          False
budget_adj            False
revenue_adj           False
dtype: bool
```

```
In [5]: df.isnull().sum()
```

```
Out[5]: id                0
```

```

imdb_id          10
popularity        0
budget            0
revenue           0
original_title    0
cast             76
homepage         7930
director          44
tagline          2824
keywords         1493
overview          4
runtime           0
genres           23
production_companies 1030
release_date      0
vote_count        0
vote_average      0
release_year      0
budget_adj        0
revenue_adj       0
dtype: int64

```

- duplicate rows : 1

```
In [6]: df.duplicated().any()
```

```
Out[6]: True
```

```
In [7]: df.duplicated().sum()
```

```
Out[7]: 1
```

- number of non-null unique values for features :

```
In [8]: df.nunique()
```

```

Out[8]: id          10865
imdb_id          10855
popularity       10814
budget           557
revenue          4702
original_title   10571
cast            10719
homepage         2896
director         5067
tagline          7997
keywords         8804
overview         10847
runtime          247
genres           2039
production_companies 7445
release_date     5909
vote_count       1289
vote_average      72
release_year      56
budget_adj       2614
revenue_adj      4840
dtype: int64

```

- data types of columns:

In [9]: `df.dtypes`

```
Out[9]: id                int64
imdb_id              object
popularity          float64
budget              int64
revenue             int64
original_title      object
cast                object
homepage            object
director            object
tagline             object
keywords            object
overview            object
runtime             int64
genres              object
production_companies object
release_date        object
vote_count          int64
vote_average        float64
release_year        int64
budget_adj          float64
revenue_adj         float64
dtype: object
```

In [10]: `df.info()`

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10866 entries, 0 to 10865
Data columns (total 21 columns):
#   Column                Non-Null Count  Dtype
---  -
0   id                    10866 non-null  int64
1   imdb_id              10856 non-null  object
2   popularity            10866 non-null  float64
3   budget               10866 non-null  int64
4   revenue              10866 non-null  int64
5   original_title        10866 non-null  object
6   cast                 10790 non-null  object
7   homepage              2936 non-null   object
8   director              10822 non-null  object
9   tagline               8042 non-null   object
10  keywords              9373 non-null   object
11  overview              10862 non-null  object
12  runtime               10866 non-null  int64
13  genres                10843 non-null  object
14  production_companies  9836 non-null   object
15  release_date          10866 non-null  object
16  vote_count            10866 non-null  int64
17  vote_average          10866 non-null  float64
18  release_year          10866 non-null  int64
19  budget_adj            10866 non-null  float64
20  revenue_adj           10866 non-null  float64
dtypes: float64(4), int64(6), object(11)
memory usage: 1.7+ MB
```

In [11]: `df.original_title`

```
Out[11]: 0                Jurassic World
1          Mad Max: Fury Road
```

```

2                Insurgent
3    Star Wars: The Force Awakens
4                Furious 7
...
10861    The Endless Summer
10862                Grand Prix
10863    Beregis Avtomobilya
10864    What's Up, Tiger Lily?
10865    Manos: The Hands of Fate
Name: original_title, Length: 10866, dtype: object

```

Relase years : 1960-2015

```
In [12]: df.release_year.unique()
```

```
Out[12]: array([2015, 2014, 1977, 2009, 2010, 1999, 2001, 2008, 2011, 2002, 1994,
        2012, 2003, 1997, 2013, 1985, 2005, 2006, 2004, 1972, 1980, 2007,
        1979, 1984, 1983, 1995, 1992, 1981, 1996, 2000, 1982, 1998, 1989,
        1991, 1988, 1987, 1968, 1974, 1975, 1962, 1964, 1971, 1990, 1961,
        1960, 1976, 1993, 1967, 1963, 1986, 1973, 1970, 1965, 1969, 1978,
        1966])
```

profit is calculated by revenue - budget.

```
In [13]: df['profit'] = df['revenue'] - df['budget']
```

```
In [14]: df['profit']
```

```
Out[14]: 0      1363528810
1      228436354
2      185238201
3      1868178225
4      1316249360
...
10861      0
10862      0
10863      0
10864      0
10865     -19000
Name: profit, Length: 10866, dtype: int64
```

Checking for the zeros in budget and revenue to prevent inappropriate results.

```
In [15]: df[df['budget']==0].shape[0]
```

```
Out[15]: 5696
```

```
In [16]: df[df['revenue']==0].shape[0]
```

```
Out[16]: 6016
```

```
In [17]: df[df['budget']!=0] #I will use not 0s.
```

```
Out[17]:
```

	id	imdb_id	popularity	budget	revenue	original_title
--	----	---------	------------	--------	---------	----------------

	id	imdb_id	popularity	budget	revenue	original_title	
0	135397	tt0369610	32.985763	150000000	1513528810	Jurassic World	Chris Pratt E Dallas Howard J Khar
1	76341	tt1392190	28.419936	150000000	378436354	Mad Max: Fury Road	Tom Hardy Chi Theron Hugh Ke Byrne
2	262500	tt2908446	13.112507	110000000	295238201	Insurgent	Sha Woodley James Winslet An
3	140607	tt2488496	11.173104	200000000	2068178225	Star Wars: The Force Awakens	Harrison Ford Hamill C Fisher Adar
4	168259	tt2820852	9.335014	190000000	1506249360	Furious 7	Vin Diesel Walker J Statham Miche
...
10835	5923	tt0060934	0.299911	12000000	20000000	The Sand Pebbles	McQueen Ric Attenborough Ric
10841	42701	tt0062262	0.264925	75000	0	The Shooting	Will Hutchins Perkins Nicholson
10848	2161	tt0060397	0.207257	5115000	12000000	Fantastic Voyage	Stephen Boyd R Welch Edr O'Brien Do
10855	13343	tt0059221	0.141026	700000	0	The Ghost & Mr. Chicken	Don Knotts Staley Redmond Sa
10865	22293	tt0060666	0.035919	19000	0	Manos: The Hands of Fate	Harold P. Warren Neyman Reynolds D

5170 rows x 22 columns

In [18]:

```
df[df['revenue']!=0] #I will use not 0s.
```

Out[18]:

	id	imdb_id	popularity	budget	revenue	original_title	
0	135397	tt0369610	32.985763	150000000	1513528810	Jurassic World	Chris Pratt E Dallas Howard J Khar
1	76341	tt1392190	28.419936	150000000	378436354	Mad Max: Fury Road	Tom Hardy Chi Theron Hugh Ke Byrne

	id	imdb_id	popularity	budget	revenue	original_title	
2	262500	tt2908446	13.112507	110000000	295238201	Insurgent	Shia Woodley James Winslet An
3	140607	tt2488496	11.173104	200000000	2068178225	Star Wars: The Force Awakens	Harrison Ford Hamill C Fisher Adar
4	168259	tt2820852	9.335014	190000000	1506249360	Furious 7	Vin Diesel Walker J Statham Miche
...	
10822	396	tt0061184	0.670274	7500000	33736689	Who's Afraid of Virginia Woolf?	Elizabeth Taylor Ric Burton Ge Segal
10828	5780	tt0061107	0.402730	3000000	13000000	Torn Curtain	Paul Newman Andrew Kedrova Hans
10829	6644	tt0061619	0.395668	4653000	6000000	El Dorado	John Wayne R Mitchum J Caan Charle
10835	5923	tt0060934	0.299911	12000000	20000000	The Sand Pebbles	Paul McQueen Ric Attenborough Ric
10848	2161	tt0060397	0.207257	5115000	12000000	Fantastic Voyage	Stephen Boyd R Welch Edr O'Brien Do

4850 rows x 22 columns

Cleaning Column Labels

I will drop the columns I do not need for this analysis which are: (imdb_id , homepage , tagline , keywords , overview , runtime , production_companies , release_date , budget_adj , revenue_adj)

```
In [19]: df.drop(['imdb_id','homepage','tagline','keywords','overview','runtime','pro
```

```
In [20]: df.head()
```

Out[20]:	id	popularity	budget	revenue	original_title	cast	director
0	135397	32.985763	150000000	1513528810	Jurassic World	Chris Pratt Bryce Dallas Howard Irrfan Khan Vi...	Colin Trevorrow

	id	popularity	budget	revenue	original_title	cast	director	
1	76341	28.419936	150000000	378436354	Mad Max: Fury Road	Tom Hardy Charlize Theron Hugh Keays-Byrne Nic...	George Miller	Acti
2	262500	13.112507	110000000	295238201	Insurgent	Shailene Woodley Theo James Kate Winslet Ansel...	Robert Schwentke	
3	140607	11.173104	200000000	2068178225	Star Wars: The Force Awakens	Harrison Ford Mark Hamill Carrie Fisher Adam D...	J.J. Abrams	Acti
4	168259	9.335014	190000000	1506249360	Furious 7	Vin Diesel Paul Walker Jason Statham Michelle ...	James Wan	

Data Cleaning

Replace Zero Values with Null Values for Budget & Revenue

```
In [21]: df['budget'] = df['budget'].replace(0, np.NaN)
df['revenue'] = df['revenue'].replace(0, np.NaN)
```

```
In [22]: df[df['budget']==0].shape[0] #Checking
```

Out[22]: 0

```
In [23]: df[df['revenue']==0].shape[0] #Checking
```

Out[23]: 0

Drop Nulls

```
In [24]: df = df.dropna(subset=['cast', 'director', 'genres', 'budget', 'revenue'])
```

```
In [25]: df.isnull().sum().any()
```

Out[25]: False

Drop Duplicates

```
df = df.drop_duplicates()
```

```
In [26]: df.duplicated().sum().any()
```


Out[26]: True

EDA with Visuals

Q : Is there any relationship between the popularity and the budget?

```
In [27]: correlation = df["popularity"].corr(df["budget"])

correlation
```

Out[27]: 0.4465702124386731

A: The correlation between popularity and budget is 0.4465. It is not close to 1 enough to be in a strong relation. I assume that there is no significant relation between them.

Which movies have the highest profit of all the time?

```
In [28]: sns.set_theme(style="whitegrid")
```

```
In [35]: df.columns
```

```
Out[35]: Index(['id', 'popularity', 'budget', 'revenue', 'original_title', 'cast',
              'director', 'genres', 'vote_count', 'vote_average', 'release_year',
              'profit'],
              dtype='object')
```

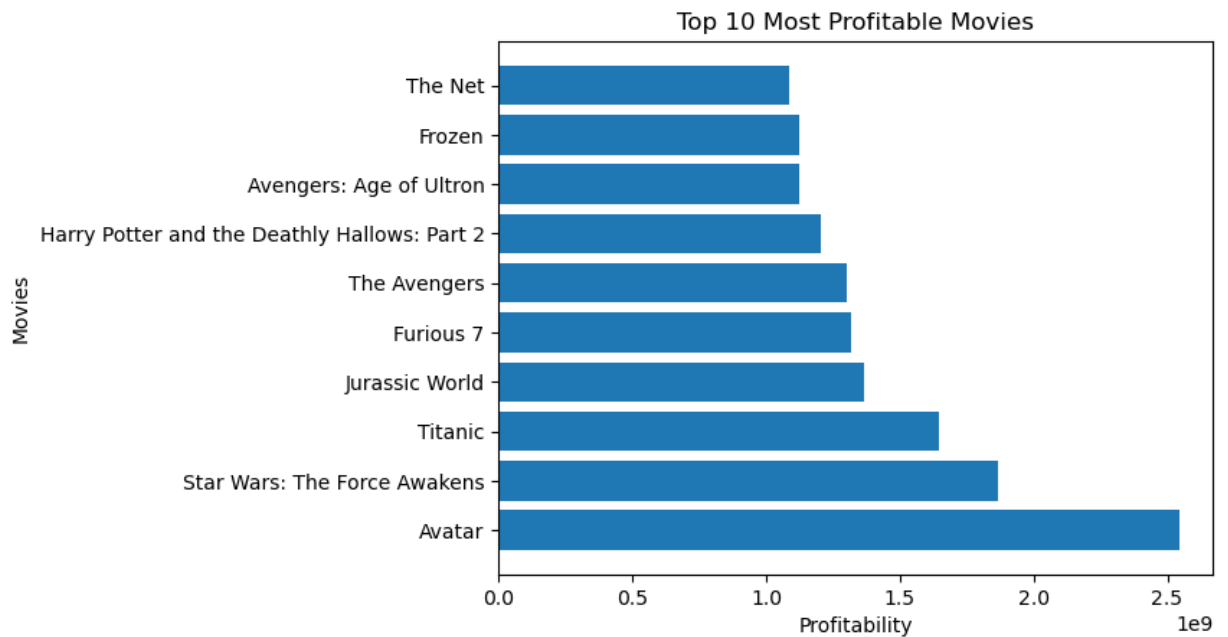
```
In [34]: type(df['profit'])
```

Out[34]: pandas.core.series.Series

```
In [64]: df = df.sort_values(by=['profit'], ascending=False)
```

```
In [65]: x = list(df['original_title'].head(10))
y = list(df['profit'].head(10))
```

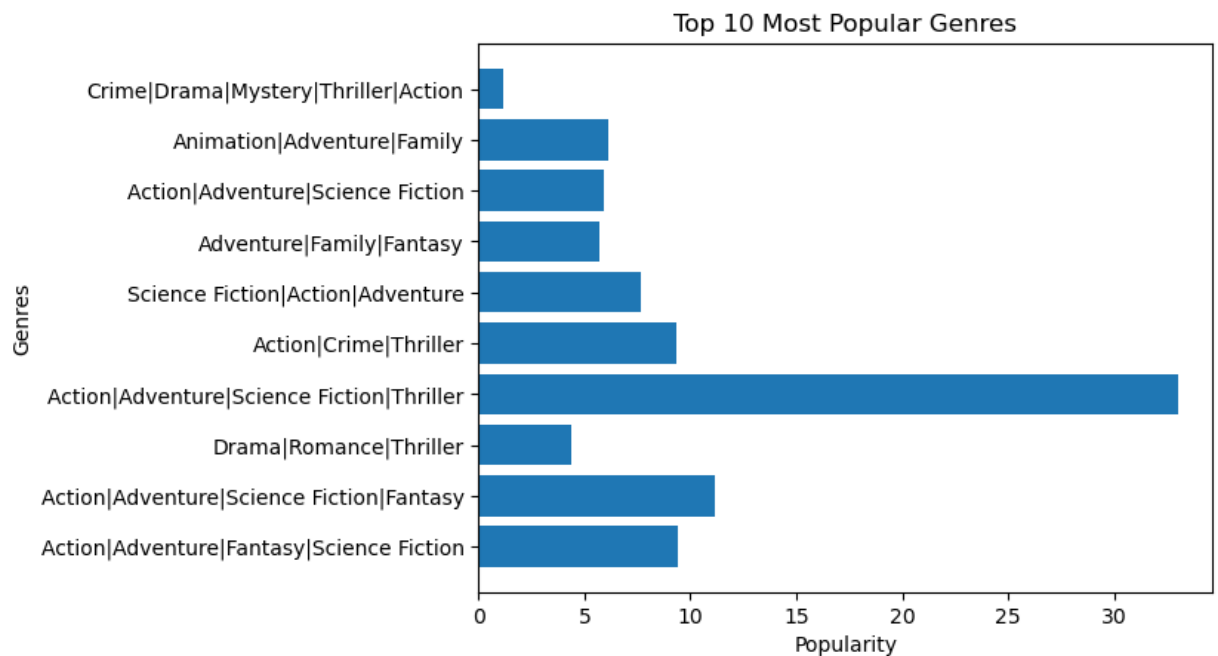
```
In [68]: plt.barh(x,y)
plt.title('Top 10 Most Profitable Movies')
plt.ylabel('Movies')
plt.xlabel('Profitability')
plt.show()
```



Which genres have the highest profit?

In [69]:

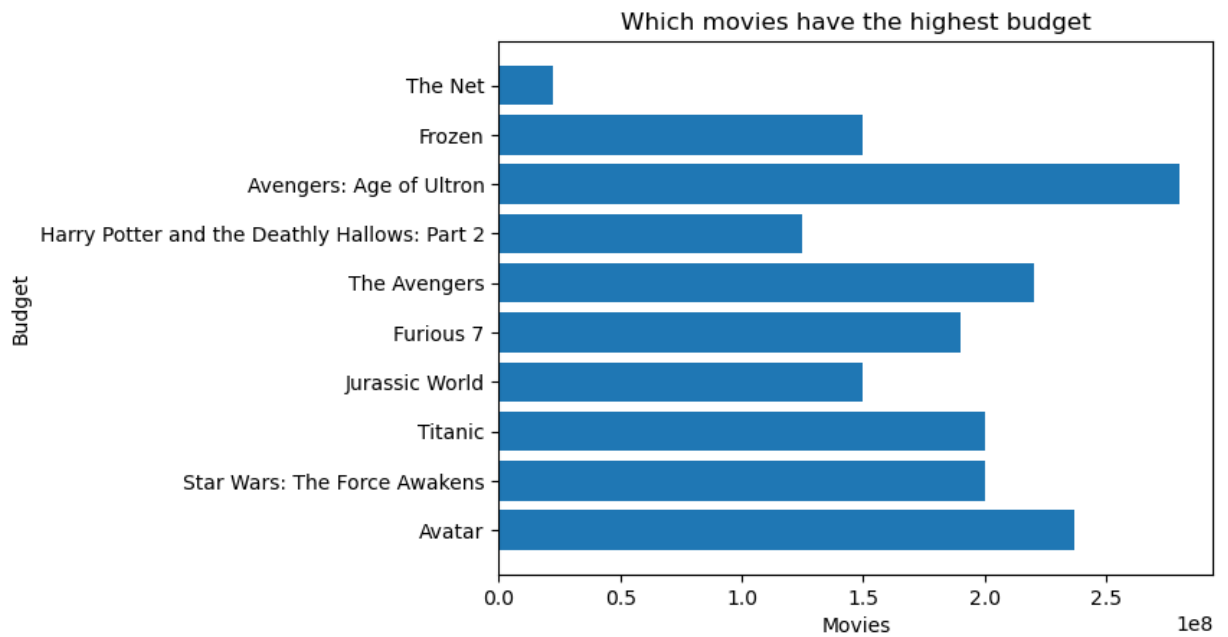
```
x = list(df['genres'].head(10))
y = list(df['popularity'].head(10))
plt.barh(x,y)
plt.title('Top 10 Most Popular Genres')
plt.ylabel('Genres')
plt.xlabel('Popularity')
plt.show()
```



Movie with Highest And Lowest Budget?

In [70]:

```
x = list(df['original_title'].head(10))
y = list(df['budget'].head(10))
plt.barh(x,y)
plt.title('Which movies have the highest budget')
plt.ylabel('Budget')
plt.xlabel('Movies')
plt.show()
```



Which movie get the highest or lowest votes (Ratings).

```
In [71]: x = list(df['original_title'].head(10))
y = list(df['vote_count'].head(10))
plt.barh(x,y)
plt.title('Which movies most loved')
plt.ylabel('Votes')
plt.xlabel('Movies')
plt.show()
```

