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

## **Data Understanding**

- 1. Loading csv file in juypter notebook
- 2. Dataframe shape
- 3. Head and Tail
- 4. dtypes
- 5. Describe

```
In [4]: data = pd.read_csv(r"C:\Users\tapas\OneDrive\Desktop\imdb_top_1000.csv")
    data
```

Poster\_Link Series\_Title Released\_Year Certificate Runtime

	POSter_Link	Series_Title	Keleased_Year	Certificate	Kuntime	'
0	https://m.media- amazon.com/images/M/MV5BMDFkYT	The Shawshank Redemption	1994	А	142 min	]
1	https://m.media- amazon.com/images/M/MV5BM2MyNj	The Godfather	1972	А	175 min	( [.
2	https://m.media- amazon.com/images/M/MV5BMTMxNT	The Dark Knight	2008	UA	152 min	<i>,</i> (
3	https://m.media- amazon.com/images/M/MV5BMWMwMG	The Godfather: Part II	1974	А	202 min	[
4	https://m.media- amazon.com/images/M/MV5BMWU4N2	12 Angry Men	1957	U	96 min	1
•••						
995	https://m.media- amazon.com/images/M/MV5BNGEwMT	Breakfast at Tiffany's	1961	А	115 min	Co E Ror
996	https://m.media- amazon.com/images/M/MV5BODk3Yj	Giant	1956	G	201 min	C W
997	https://m.media- amazon.com/images/M/MV5BM2U3Yz	From Here to Eternity	1953	Passed	118 min	E Ron
998	https://m.media- amazon.com/images/M/MV5BZTBmMj	Lifeboat	1944	NaN	97 min	С

Poster\_Link Series\_Title Released\_Year Certificate Runtime

	999	https://m.medi amazon.com/images/M/MV5BMTY5OD		193	5 Na	N 86 mi	n M <u>y</u>
	1000	rows × 16 columns					
In [5]:	data	a.shape					
Out[5]:	(100	90, 16)					
In [6]:	data	a.head()					
Out[6]:		Poster_Link	Series_Title	Released_Year	Certificate	Runtime	Genre
	0	https://m.media- amazon.com/images/M/MV5BMDFkYT	The Shawshank Redemption	1994	А	142 min	Drama
	1	https://m.media- amazon.com/images/M/MV5BM2MyNj	The Godfather	1972	А	175 min	Crime Drama
	2	https://m.media- amazon.com/images/M/MV5BMTMxNT	The Dark Knight	2008	UA	152 min	Action Crime Drama
	<b>3</b>	https://m.media- mazon.com/images/M/MV5BMWMwMG	The Godfather: Part II	1974	А	202 min	Crime Drama
	4	https://m.media- amazon.com/images/M/MV5BMWU4N2	12 Angry Men	1957	U	96 min	Crime Drama
◀							•
In [7]:	data	a.tail()					

0 1 [ = ]	B ( 11 1	c ·	D 1 1 1 1 1 1	c	B 41	
Out[7]:	Poster_Link	Series_litle	Released_Year	Certificate	Runtime	Ge

995	https://m.media- amazon.com/images/M/MV5BNGEwMT	Breakfast at Tiffany's	1961	А	115 min	Come Dra Roma
996	https://m.media- amazon.com/images/M/MV5BODk3Yj	Giant	1956	G	201 min	Dra West
997	https://m.media- amazon.com/images/M/MV5BM2U3Yz	From Here to Eternity	1953	Passed	118 min	Dra Romai
998	https://m.media- amazon.com/images/M/MV5BZTBmMj	Lifeboat	1944	NaN	97 min	Dra
999	https://m.media- amazon.com/images/M/MV5BMTY5OD	The 39 Steps	1935	NaN	86 min	Cri Myst Thr



In [8]: data.dtypes

object Poster\_Link Out[8]: Series\_Title object Released\_Year object Certificate object Runtime object Genre object IMDB\_Rating float64 Overview object Meta\_score float64 Director object Star1 object Star2 object Star3 object Star4 object No\_of\_Votes int64

dtype: object

object

In [9]: data.describe()

Gross

ut[9]:		IMDB_Rating	Meta_score	No_of_Votes
	count	1000.000000	843.000000	1.000000e+03
	mean	7.949300	77.971530	2.736929e+05
	std	0.275491	12.376099	3.273727e+05
	min	7.600000	28.000000	2.508800e+04
	25%	7.700000	70.000000	5.552625e+04
	50%	7.900000	79.000000	1.385485e+05
	75%	8.100000	87.000000	3.741612e+05
	max	9.300000	100.000000	2.343110e+06

In [10]:	<pre>data.describe(include='object')</pre>									
Out[10]:		Poster_Link	Series_Title	Released_Year	Certificate	Runtime	Gei			
	count	1000	1000	1000	899	1000	1(			
	unique	1000	999	100	16	140	Ź			
	top	https://m.media- amazon.com/images/M/MV5BMDFkYT	Drishyam	2014	U	100 min	Dra			
	freq	1	2	32	234	23				
4							<b>•</b>			

# **Data Preparation**

- 1. Checking for null values
- 2. Converting dtypes
- 3. Dropping irrelevant columns and rows
- 4. Renaming columns

In [11]: data.isnull().sum()

```
Poster_Link
                            0
Out[11]:
         Series_Title
                            0
         Released_Year
                            0
         Certificate
                          101
         Runtime
                            0
         Genre
                            0
         IMDB_Rating
                            0
         Overview
                            0
         Meta_score
                          157
         Director
                            0
         Star1
                            0
         Star2
                            0
         Star3
                            0
         Star4
                            0
         No_of_Votes
                            0
                          169
         Gross
         dtype: int64
In [12]: print(data['Gross'].head(1))
         data['Gross'] = data['Gross'].str.replace(',',')
         print(data['Gross'].head(1))
         data['Gross'] = data['Gross'].astype('float64')
         data['Gross'] = data['Gross'].replace('Not Rated', 0)
              28,341,469
         Name: Gross, dtype: object
              28341469
         Name: Gross, dtype: object
         data['Gross'] = data['Gross'].replace(np.nan, 0)
In [14]:
In [15]:
         data['Gross'] = data['Gross'].astype('int64')
In [16]:
         data['Gross'].dtype
         dtype('int64')
Out[16]:
         data.drop(['Poster_Link', 'Overview'], axis = 1)
In [17]:
```

Out[17]:		Series_Title	Released_Year	Certificate	Runtime	Genre	IMDB_Rating	Meta_score	Dire
	0	The Shawshank Redemption	1994	А	142 min	Drama	9.3	80.0	F Dara
	1	The Godfather	1972	А	175 min	Crime, Drama	9.2	100.0	Fr Cop
	2	The Dark Knight	2008	UA	152 min	Action, Crime, Drama	9.0	84.0	Christo N
	3	The Godfather: Part II	1974	А	202 min	Crime, Drama	9.0	90.0	Fr Cop
	4	12 Angry Men	1957	U	96 min	Crime, Drama	9.0	96.0	Si Lı
	•••								
	995	Breakfast at Tiffany's	1961	А	115 min	Comedy, Drama, Romance	7.6	76.0	l Edv
	996	Giant	1956	G	201 min	Drama, Western	7.6	84.0	Ge Ste
	997	From Here to Eternity	1953	Passed	118 min	Drama, Romance, War	7.6	85.0	Zinner
	998	Lifeboat	1944	NaN	97 min	Drama, War	7.6	78.0	A Hitch
	999	The 39 Steps	1935	NaN	86 min	Crime, Mystery, Thriller	7.6	93.0	A Hitch

1000 rows × 14 columns



## **Performing EDA**

In [19]: data.corr()

C:\Users\tapas\AppData\Local\Temp\ipykernel\_15340\2627137660.py:1: FutureWarning:
The default value of numeric\_only in DataFrame.corr is deprecated. In a future ver
sion, it will default to False. Select only valid columns or specify the value of
numeric\_only to silence this warning.
 data.corr()

Out[19]:		IMDB_Rating	Meta_score	No_of_Votes	Gross
	IMDB_Rating	1.000000	0.268531	0.494979	0.082381
	Meta_score	0.268531	1.000000	-0.018507	-0.053659
	No_of_Votes	0.494979	-0.018507	1.000000	0.602128
	Gross	0.082381	-0.053659	0.602128	1.000000

#### **Top voted Movies**

```
In [21]: top_voted = data.sort_values(['No_of_Votes'], ascending = False)

In [22]: fig,axs=plt.subplots(figsize=(20,5))
sns.barplot(x = top_voted['Movies_Title'][:10], y = top_voted['No_of_Votes'][:10])
plt.title("Top voted Movies", weight = "bold")
plt.xticks(rotation=90)
plt.show()

Top voted Movies

Top voted Movies

Top voted Movies

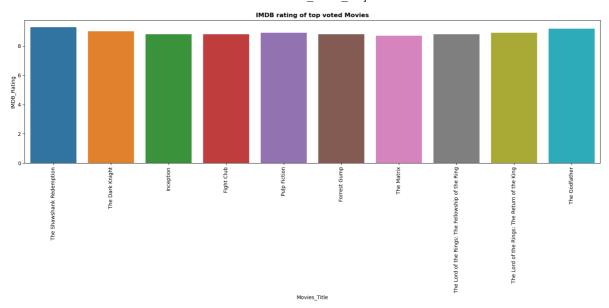
No.vec. Title

No.vec. Title
```

#### IMDB rating of top voted movies

IMDb's scores are based on users' ratings

```
In [23]: fig,axs=plt.subplots(figsize = (20,5))
    sns.barplot(x = top_voted['Movies_Title'][:10], y = top_voted['IMDB_Rating'][:10])
    plt.title("IMDB rating of top voted Movies", weight = "bold")
    plt.xticks(rotation=90)
    plt.show()
```

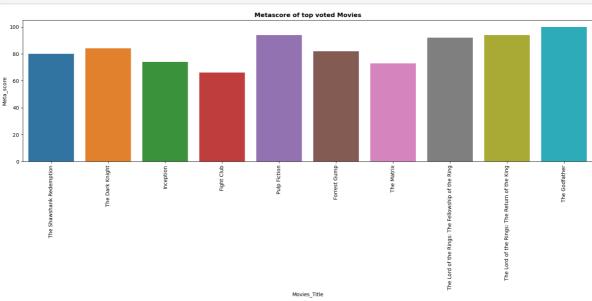


```
#top 10 most rated movies
In [24]:
          data.sort_values(by = 'No_of_Votes', ascending = False).head(10)['Movies_Title']
                                         The Shawshank Redemption
Out[24]:
         2
                                                   The Dark Knight
         8
                                                         Inception
         9
                                                        Fight Club
         6
                                                      Pulp Fiction
         11
                                                      Forrest Gump
         14
                                                        The Matrix
         10
               The Lord of the Rings: The Fellowship of the Ring
         5
                    The Lord of the Rings: The Return of the King
         1
                                                     The Godfather
         Name: Movies_Title, dtype: object
```

#### Metascore of top rated movies

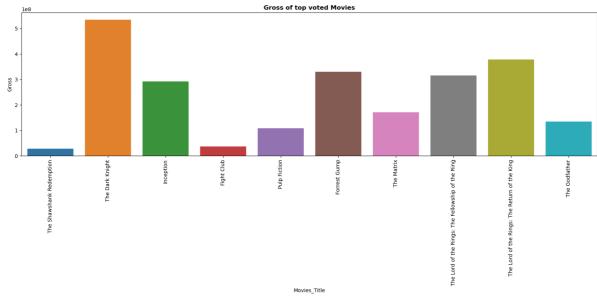
Metacritic's main scores are based on reviewers' ratings

```
fig,axs=plt.subplots(figsize = (20,5))
sns.barplot(x = top_voted['Movies_Title'][:10], y = top_voted['Meta_score'][:10])
plt.title("Metascore of top voted Movies", weight = "bold")
plt.xticks(rotation=90)
plt.show()
```



#### Gross of top voted Movies

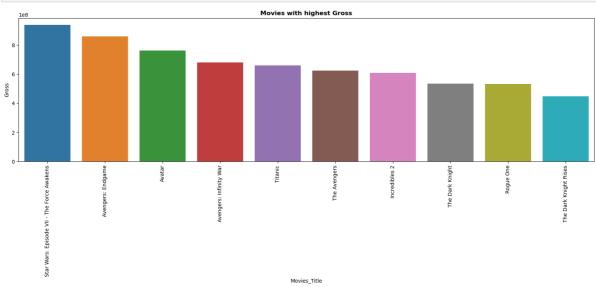
```
In [26]: fig,axs=plt.subplots(figsize = (20,5))
    sns.barplot(x = top_voted['Movies_Title'][:10], y = top_voted['Gross'][:10])
    plt.title("Gross of top voted Movies", weight = "bold")
    plt.xticks(rotation=90)
    plt.show()
```



#### Top movies by Gross

```
In [27]: highest_earning = data.sort_values(['Gross'], ascending = False)

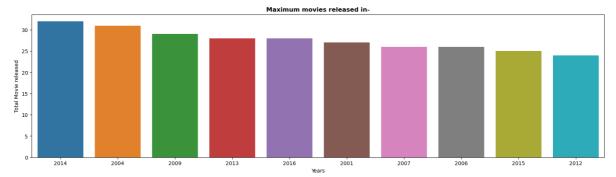
In [28]: fig,axs=plt.subplots(figsize = (20,5))
    sns.barplot(x = highest_earning['Movies_Title'][:10], y = highest_earning['Gross'][
    plt.title("Movies with highest Gross", weight = "bold")
    plt.xticks(rotation=90)
    plt.show()
```



### Maximum movies released in year

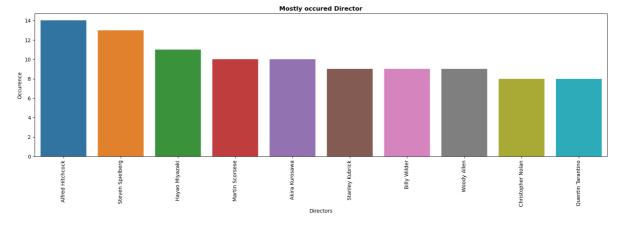
```
In [29]: fig,axs = plt.subplots(figsize = (20,5))
sns.barplot(x = data['Released_Year'].value_counts()[:10].index, y = data['Released_Year']
```

```
plt.title("Maximum movies released in-", weight = "bold")
plt.xlabel("Years")
plt.ylabel("Total Movie released")
plt.show()
```



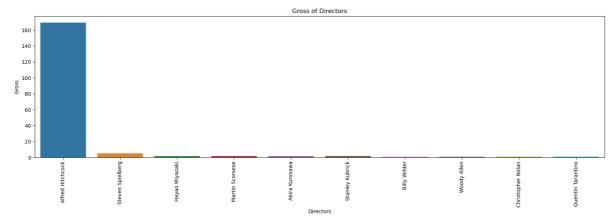
#### Mostly occured Director

```
In [30]: fig,axs=plt.subplots(figsize = (20,5))
    sns.barplot(x = data['Director'].value_counts()[:10].index, y = data['Director'].va
    plt.title("Mostly occured Director", weight = "bold")
    plt.xlabel("Directors")
    plt.ylabel("Occurence")
    plt.xticks(rotation = 90)
    plt.show()
```



#### **Directors with highest Gross**

```
In [31]: fig,axs=plt.subplots(figsize = (20,5))
    sns.barplot(x = data['Director'].value_counts()[:10].index, y = data['Gross'].value
    plt.title("Gross of Directors")
    plt.xlabel("Directors")
    plt.xticks(rotation = 90)
    plt.show()
```



#### Stars with most occurences in Movies

```
In [32]: stars=['Star1','Star2','Star3','Star4']
fig,axs=plt.subplots(4,1,figsize=(20,7))
ax=0
for x in stars:
    axs[ax].bar(data[x].value_counts()[:10].index,data[x].value_counts()[:10])
    axs[ax].set_title(x)
    axs[ax].set_title(x)
    axs[ax].set_tylabel("Appearances", weight = "bold")
    ax+=1
    plt.tight_layout()

Star1

Star2

**This is a stars in the stars in the star in
```

### **IMDB** rating distribution

```
In [33]: fig,axs=plt.subplots(figsize=(20,5))
sns.histplot(data['IMDB_Rating'],bins=30, kde = True)
plt.title("Distribution of Ratings", weight = "bold")
plt.show()
Distribution of Ratings
```

**Top 10 Genres** 

```
from collections import Counter
genre=[]
for x in data['Genre']:
    for y in x.split(','):
        genre.append(y.strip().lower())

count=Counter(genre)
    count=count.most_common()[:10]
    x,y=map(list,zip(*count))

fig,axs=plt.subplots(figsize=(20,5))
    g=sns.barplot(y=y,x=x)
    g.set_ylabel("Genres", weight = "bold")
    g.set_title("Top Ten Genres", weight = "bold")
    plt.show()
```

