from google.colab import files
data=files.upload()

Choose Files imdb_top_1000.csv

• imdb_top_1000.csv(text/csv) - 438096 bytes, last modified: 8/1/2024 - 100% done Saving imdb_top_1000.csv to imdb_top_1000.csv

import pandas as pd
import seaborn as sns

import matplotlib.pyplot as plt

import numpy as np

df=pd.read_csv("imdb_top_1000.csv")
df.head()

	Poster_Link	Series_Title	Released_Year	Certificate	Runtime	Genre	IMDB_Rating	Overview	Meta_score
0	https://m.media- amazon.com/images/M/MV5BMDFkYT	The Shawshank Redemption	1994	А	142 min	Drama	9.3	Two imprisoned men bond over a number of years	80.0
1	https://m.media- amazon.com/images/M/MV5BM2MyNj	The Godfather	1972	А	175 min	Crime, Drama	9.2	An organized crime dynasty's aging patriarch t	100.C
2	https://m.media- amazon.com/images/M/MV5BMTMxNT	The Dark Knight	2008	UA	152 min	Action, Crime, Drama	9.0	When the menace known as the Joker wreaks havo	84.C
3	https://m.media-amazon.com/images/M/MV5BMWMwMG	The Godfather: Part II	1974	А	202 min	Crime, Drama	9.0	The early life and career of Vito Corleone in	90.0
4	https://m.media- amazon.com/images/M/MV5BMWU4N2	12 Angry Men	1957	U	96 min	Crime, Drama	9.0	A jury holdout attempts to prevent a miscarria	96.0

print(df.info())

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 16 columns):

Jaca	COTUMNIS (COCAT	TO COTUMNIS).					
#	Column	Non-Null Count	Dtype				
0	Poster_Link	1000 non-null	object				
1	Series_Title	1000 non-null	object				
2	Released_Year	1000 non-null	object				
3	Certificate	899 non-null	object				
4	Runtime	1000 non-null	object				
5	Genre	1000 non-null	object				
6	<pre>IMDB_Rating</pre>	1000 non-null	float64				
7	Overview	1000 non-null	object				
8	Meta_score	843 non-null	float64				
9	Director	1000 non-null	object				
10	Star1	1000 non-null	object				
11	Star2	1000 non-null	object				
12	Star3	1000 non-null	object				
13	Star4	1000 non-null	object				
14	No_of_Votes	1000 non-null	int64				
15	Gross	831 non-null	object				
44	(1 + < 4/2)	:-+C4/1\ - -:+/12\					

dtypes: float64(2), int64(1), object(13)

memory usage: 125.1+ KB

None

^{**} Task 1. Director's Impact on Gross Earnings: Analyze how movies directed by different directors**

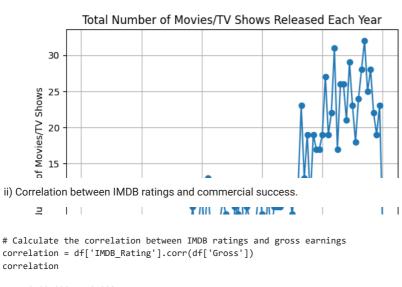
perform in terms of gross earnings. Are there any noticeable trends or patterns?bold
 text

```
df['Director'].nunique()
     548
director_gross = df.groupby('Director')['Gross'].mean().sort_values(ascending=False).head(10)
\mbox{\#} Convert the result to a DataFrame for simplicity
\label{eq:df_director_gross} = pd.DataFrame(\{'Director': director_gross.index, 'Average\_Gross': director\_gross.values\})
df_director_gross
              Director Average_Gross
                                          \blacksquare
      0 Anthony Russo
                         5.512599e+08
                                           th
      1 Gareth Edwards
                          5.321773e+08
             J.J. Abrams
                          4.743903e+08
      3
            Josh Cooley
                          4.340380e+08
            Roger Allers
      4
                          4.227838e+08
      5
              Tim Miller
                          3.630707e+08
            James Gunn
      6
                          3.614949e+08
      7 James Cameron
                          3.496473e+08
                          3.412682e+08
           Byron Howard
             David Yates
                          3.263179e+08
```

Task 2: Other Potential Analyses:

i) Genre popularity over the years

```
# Group the data by year and calculate the total number of movies/TV shows each year
total_movies_per_year = df.groupby('Release_Year').size()
total_movies_per_year
     Release_Year
     1920.0
               1
     1921.0
               1
     1922.0
               1
     1924.0
     1925.0
     2016.0
     2017.0
               22
     2018.0
               19
     2019.0
               23
     2020.0
               6
     Length: 99, dtype: int64
# Plot the total number of movies/TV shows released each year
total_movies_per_year.plot(kind='line', marker='o')
plt.title('Total Number of Movies/TV Shows Released Each Year')
plt.xlabel('Release Year')
plt.ylabel('Number of Movies/TV Shows')
plt.grid(True)
plt.show()
```



correlation = df['IMDB_Rating'].corr(df['Gross']) correlation

0.09592277110132356

iii) Impact of movie length on ratings or earnings.

```
#removing min
Duration= df['Runtime'].str.split(' ', expand=True)[0].astype(float)
Duration
            142.0
     0
            175.0
     1
     2
            152.0
     3
            202.0
            96.0
            115.0
     995
     996
            201.0
     997
            118.0
     998
             97.0
     999
             86.0
     Name: 0, Length: 1000, dtype: float64
correlation_ratings = df['Duration'].corr(df['IMDB_Rating'])
correlation_ratings
     0.24309589775419407
correlation_earnings = df['Duration'].corr(df['Gross'])
correlation_earnings
```

0.13967069065697754