

# IMDB Movie Analysis

## *Objective:*

As a data analyst intern at IMDB, you have been tasked with exploring and analyzing the IMDB Movies dataset. Your goal is to answer specific business questions, gain insights into movie trends, and deliver actionable recommendations. Using Python and libraries such as Pandas, NumPy, Seaborn, and Matplotlib, perform analysis to help IMDB better understand genre popularity, rating trends, and factors influencing movie success.

**Dataset:** <https://drive.google.com/file/d/1lruT50ZWD4PtvDbIn4VnepZvSoeO9BrA/view?usp=sharing>

## *Tasks:*

## ***1. Project Setup and Data Loading***

**Task:** Load the dataset and perform initial setup.

### *Questions:*

***What libraries are required for this project, and why are they useful in data analysis?***

**1.Pandas:** Used for data manipulation and analysis with DataFrames and Series.

**2.NumPy:** Enables fast numerical computations with multi-dimensional arrays.

**3.Seaborn:** Simplifies statistical data visualization with aesthetically pleasing plots.

**4.Matplotlib:** Provides customizable charts and graphs for data visualization.

***Load the dataset. What is the shape of the dataset? What does each row and column represent?***

- **Each row** represents a movie.
  
- **Each column** represents a specific attribute of the movie:
  - a. **names** – Movie title.
  - b. **date\_x** – Release date.
  - c. **score** – IMDb rating score.
  - d. **genre** – Genres the movie belongs to.

- e. **overview** – Brief plot summary.
- f. **crew** – List of key cast and crew members.
- g. **orig\_title** – Original title of the movie.
- h. **status** – Release status (e.g., "Released").
- i. **orig\_lang** – Original language of the movie.
- j. **budget\_x** – Production budget (in dollars).
- k. **revenue** – Box office revenue (in dollars).
- l. **country** – Country of release.

*#Using Python and libraries*

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np
```

*#Load the dataset*

```
df = pd.read_csv("imdb_movies.csv")
df
```

	names	date_x	score \
0	Creed III	03/02/2023	73.0
1	Avatar: The Way of Water	12/15/2022	78.0
2	The Super Mario Bros. Movie	04/05/2023	76.0
3	Mummies	01/05/2023	70.0
4	Supercell	03/17/2023	61.0
...	...	...	...
10173	20th Century Women	12/28/2016	73.0
10174	Delta Force 2: The Colombian Connection	08/24/1990	54.0
10175	The Russia House	12/21/1990	61.0
10176	Darkman II: The Return of Durant	07/11/1995	55.0
10177	The Swan Princess: A Royal Wedding	07/20/2020	70.0

	genre \
0	Drama, Action
1	Science Fiction, Adventure, Action
2	Animation, Adventure, Family, Fantasy, Comedy
3	Animation, Comedy, Family, Adventure, Fantasy
4	Action
...	...
10173	Drama
10174	Action
10175	Drama, Thriller, Romance

10176 Action, Adventure, Science Fiction, Thriller, ...  
 10177 Animation, Family, Fantasy

overview \

0 After dominating the boxing world, Adonis Cree...  
 1 Set more than a decade after the events of the...  
 2 While working underground to fix a water main,...  
 3 Through a series of unfortunate events, three ...  
 4 Good-hearted teenager William always lived in ...

...  
 10173 In 1979 Santa Barbara, California, Dorothea Fi...  
 10174 When DEA agents are taken captive by a ruthles...  
 10175 Barley Scott Blair, a Lisbon-based editor of R...  
 10176 Darkman and Durant return and they hate each o...  
 10177 Princess Odette and Prince Derek are going to ...

crew \

0 Michael B. Jordan, Adonis Creed, Tessa Thompso...  
 1 Sam Worthington, Jake Sully, Zoe Saldaña, Neyt...  
 2 Chris Pratt, Mario (voice), Anya Taylor-Joy, P...  
 3 Óscar Barberán, Thut (voice), Ana Esther Albor...  
 4 Skeet Ulrich, Roy Cameron, Anne Heche, Dr Quin...

...  
 10173 Annette Bening, Dorothea Fields, Lucas Jade Zu...  
 10174 Chuck Norris, Col. Scott McCoy, Billy Drago, R...  
 10175 Sean Connery, Bartholomew 'Barley' Scott Blair...  
 10176 Larry Drake, Robert G. Durant, Arnold Vosloo, ...  
 10177 Nina Herzog, Princess Odette (voice), Yuri Low...

orig\_title status \

0 Creed III Released  
 1 Avatar: The Way of Water Released  
 2 The Super Mario Bros. Movie Released  
 3 Momias Released  
 4 Supercell Released

...  
 10173 20th Century Women Released  
 10174 Delta Force 2: The Colombian Connection Released  
 10175 The Russia House Released  
 10176 Darkman II: The Return of Durant Released  
 10177 The Swan Princess: A Royal Wedding Released

orig\_lang budget\_x revenue country

0 English 75000000.0 2.716167e+08 AU  
 1 English 460000000.0 2.316795e+09 AU  
 2 English 100000000.0 7.244590e+08 AU  
 3 Spanish, Castilian 12300000.0 3.420000e+07 AU  
 4 English 77000000.0 3.409420e+08 US

...  
 10173 English 7000000.0 9.353729e+06 US

10174	English	9145817.8	6.698361e+06	US
10175	English	21800000.0	2.299799e+07	US
10176	English	116000000.0	4.756613e+08	US
10177	English	92400000.0	5.394018e+08	GB

[10178 rows x 12 columns]

*#What is the shape of the dataset*

```
print("dataset shape:",df.shape)
df.info()
```

```
dataset shape: (10178, 12)
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10178 entries, 0 to 10177
Data columns (total 12 columns):
#   Column          Non-Null Count  Dtype
---  -
0   names           10178 non-null  object
1   date_x          10178 non-null  object
2   score           10178 non-null  float64
3   genre           10093 non-null  object
4   overview        10178 non-null  object
5   crew            10122 non-null  object
6   orig_title      10178 non-null  object
7   status          10178 non-null  object
8   orig_lang       10178 non-null  object
9   budget_x        10178 non-null  float64
10  revenue         10178 non-null  float64
11  country         10178 non-null  object
dtypes: float64(3), object(9)
memory usage: 954.3+ KB
```

*#Convert data type of date\_x into datetime*

```
df["date_x"] = pd.to_datetime(df["date_x"])
```

*# Checking the datatyp*

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10178 entries, 0 to 10177
Data columns (total 12 columns):
#   Column          Non-Null Count  Dtype
---  -
0   names           10178 non-null  object
1   date_x          10178 non-null  datetime64[ns]
2   score           10178 non-null  float64
3   genre           10093 non-null  object
4   overview        10178 non-null  object
5   crew            10122 non-null  object
6   orig_title      10178 non-null  object
```

```
7    status      10178 non-null object
8    orig_lang   10178 non-null object
9    budget_x     10178 non-null float64
10   revenue     10178 non-null float64
11   country     10178 non-null object
dtypes: datetime64[ns](1), float64(3), object(8)
memory usage: 954.3+ KB
```

## 2. *Data Overview and Basic Exploration*

**Task:** Explore the structure and composition of the dataset.

**Questions:**

1. *Use .info() to understand the data types and missing values. What potential issues can you spot?*
2. *Describe the main characteristics of each column using .describe(). What can you infer from the mean, median, and distribution of numerical columns?*

```
# 1. Check the shape of the dataset
dataset_shape = df.shape

# 2. Check data types of each column
data_types = df.dtypes

# 3. Check for missing values in each column
missing_values = df.isnull().sum()

# 4. Get summary statistics for numerical columns
summary_statistics = df.describe()

# 5. Get the number of unique values in categorical columns
unique_values = df.nunique()

# Compile results
database_info = {
    "Shape (Rows, Columns)": dataset_shape,
    "Data Types": data_types,
    "Missing Values": missing_values,
    "Summary Statistics": summary_statistics,
    "Unique Values Count": unique_values
}

database_info

{'Shape (Rows, Columns)': (10178, 12),
 'Data Types': names          object
date_x          datetime64[ns]
```

```

score                float64
genre                object
overview             object
crew                 object
orig_title           object
status               object
orig_lang            object
budget_x             float64
revenue              float64
country              object
dtype: object,
'Missing Values': names      0
date_x                0
score                 0
genre                 85
overview              0
crew                  56
orig_title            0
status                0
orig_lang             0
budget_x              0
revenue               0
country               0
dtype: int64,
'Summary Statistics':
score      budget_x      revenue      date_x
count      10178      10178.000000      1.017800e+04
1.017800e+04
mean      2008-06-15 06:16:37.445470720      63.497052      6.488238e+07
2.531401e+08
min      1903-05-15 00:00:00      0.000000      1.000000e+00
0.000000e+00
25%      2001-12-25 06:00:00      59.000000      1.500000e+07
2.858898e+07
50%      2013-05-09 00:00:00      65.000000      5.000000e+07
1.529349e+08
75%      2019-10-17 00:00:00      71.000000      1.050000e+08
4.178021e+08
max      2023-12-31 00:00:00      100.000000      4.600000e+08
2.923706e+09
std      NaN      13.537012      5.707565e+07
2.777880e+08,
'Unique Values Count': names      9660
date_x      5688
score      79
genre      2303
overview    9905
crew        9927
orig_title  9736

```

```
status      3
orig_lang   54
budget_x    2316
revenue     8227
country     60
dtype: int64}
```

### 3. Data Cleaning

**Task:** Address missing values, data types, and outliers.

**Questions:**

*Which columns contain missing values? How would you handle them?*

*Are there any columns where data types need conversion (e.g., date, ratings)? Explain your decision.*

```
#Missing Values: "genre has 85 missing value"
#Filling Genre with "Not available"
df['genre'] = df['genre'].fillna('not_available')
```

```
#Missing Values:"crew has 56 missing values"
#Filling crew with "Not available"
df['crew'] = df['crew'].fillna('not_available')
```

```
#Checking the missing values again
df.isnull().sum()
```

```
names      0
date_x     0
score      0
genre      0
overview   0
crew       0
orig_title 0
status     0
orig_lang  0
budget_x   0
revenue    0
country    0
dtype: int64
```

## 4. Univariate Analysis: Explore each column individually.

**Task:** Perform univariate analysis on numerical and categorical variables.

**Questions:**

**What is the distribution of movie by years? Plot a histogram and describe its shape.**

**What are the most common genres in the dataset? Use a bar chart to show their distribution.**

```
#add new column
df["years"] = df['date_x'].dt.strftime("%Y")
df
```

	names	date_x	score	\
0	Creed III	2023-03-02	73.0	
1	Avatar: The Way of Water	2022-12-15	78.0	
2	The Super Mario Bros. Movie	2023-04-05	76.0	
3	Mummies	2023-01-05	70.0	
4	Supercell	2023-03-17	61.0	
...	...	...	...	
10173	20th Century Women	2016-12-28	73.0	
10174	Delta Force 2: The Colombian Connection	1990-08-24	54.0	
10175	The Russia House	1990-12-21	61.0	
10176	Darkman II: The Return of Durant	1995-07-11	55.0	
10177	The Swan Princess: A Royal Wedding	2020-07-20	70.0	

	genre	\
0	Drama, Action	
1	Science Fiction, Adventure, Action	
2	Animation, Adventure, Family, Fantasy, Comedy	
3	Animation, Comedy, Family, Adventure, Fantasy	
4	Action	
...	...	
10173	Drama	
10174	Action	
10175	Drama, Thriller, Romance	
10176	Action, Adventure, Science Fiction, Thriller, ...	
10177	Animation, Family, Fantasy	

	overview	\
0	After dominating the boxing world, Adonis Cree...	
1	Set more than a decade after the events of the...	
2	While working underground to fix a water main,...	
3	Through a series of unfortunate events, three ...	
4	Good-hearted teenager William always lived in ...	
...	...	
10173	In 1979 Santa Barbara, California, Dorothea Fi...	



10174 When DEA agents are taken captive by a ruthles...  
 10175 Barley Scott Blair, a Lisbon-based editor of R...  
 10176 Darkman and Durant return and they hate each o...  
 10177 Princess Odette and Prince Derek are going to ...

	crew \
0	Michael B. Jordan, Adonis Creed, Tessa Thompso...
1	Sam Worthington, Jake Sully, Zoe Saldña, Neyt...
2	Chris Pratt, Mario (voice), Anya Taylor-Joy, P...
3	Óscar Barberán, Thut (voice), Ana Esther Albor...
4	Skeet Ulrich, Roy Cameron, Anne Heche, Dr Quin...
...	...
10173	Annette Bening, Dorothea Fields, Lucas Jade Zu...
10174	Chuck Norris, Col. Scott McCoy, Billy Drago, R...
10175	Sean Connery, Bartholomew 'Barley' Scott Blair...
10176	Larry Drake, Robert G. Durant, Arnold Vosloo, ...
10177	Nina Herzog, Princess Odette (voice), Yuri Low...

	orig_title	status \
0	Creed III	Released
1	Avatar: The Way of Water	Released
2	The Super Mario Bros. Movie	Released
3	Momias	Released
4	Supercell	Released
...	...	...
10173	20th Century Women	Released
10174	Delta Force 2: The Colombian Connection	Released
10175	The Russia House	Released
10176	Darkman II: The Return of Durant	Released
10177	The Swan Princess: A Royal Wedding	Released

	orig_lang	budget_x	revenue	country	decade
years					
0	English	75000000.0	2.716167e+08	AU	2020
2023					
1	English	460000000.0	2.316795e+09	AU	2020
2022					
2	English	100000000.0	7.244590e+08	AU	2020
2023					
3	Spanish, Castilian	12300000.0	3.420000e+07	AU	2020
2023					
4	English	77000000.0	3.409420e+08	US	2020
2023					
...	...	...	...	...	...
...					
10173	English	7000000.0	9.353729e+06	US	2010
2016					
10174	English	9145817.8	6.698361e+06	US	1990
1990					
10175	English	21800000.0	2.299799e+07	US	1990

1990					
10176	English	116000000.0	4.756613e+08	US	1990
1995					
10177	English	92400000.0	5.394018e+08	GB	2020

[10178 rows x 14 columns]

*#What is the distribution of movie by years? Plot a histogram and describe its shape*

*#Plotting the distribution of movie by years*

```
df = df.sort_values(by = "years")
plt.figure(figsize=(12, 6))
sns.histplot(df['years'].dropna(), bins=30, kde=True, color="green")
plt.xlabel("Year")
plt.xticks(rotation = 80, fontsize = 6)
plt.ylabel("Number of Movies")
plt.title("Distribution of Movies by Years")
plt.show()
```

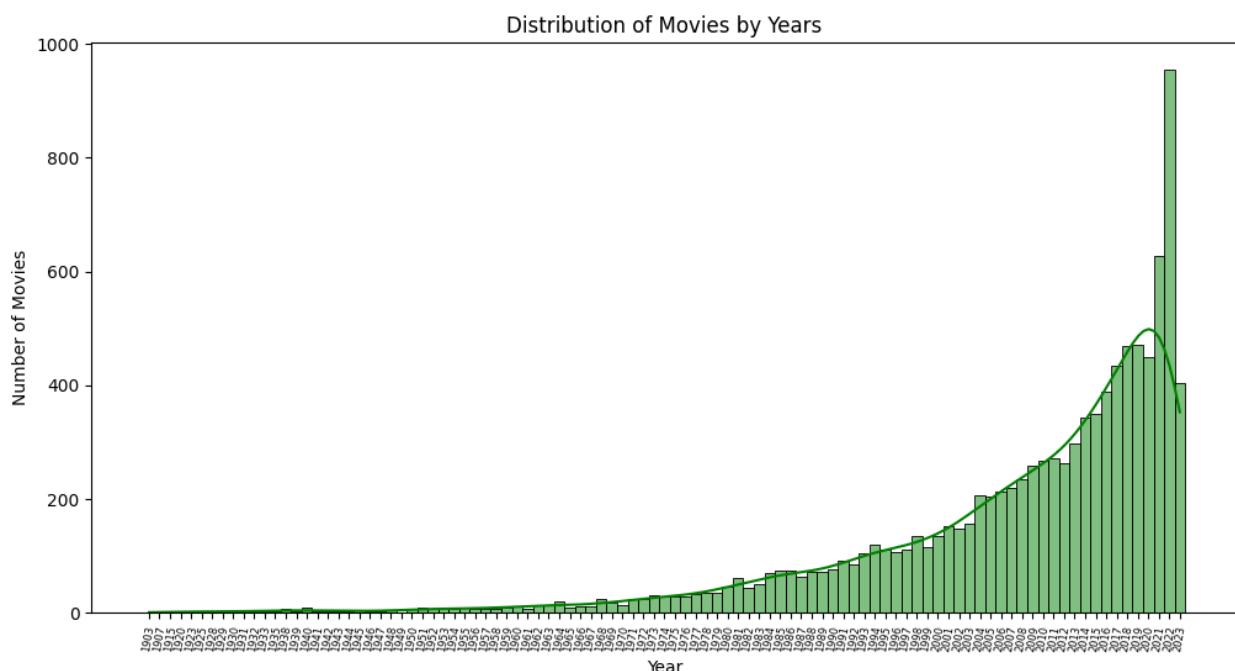
*#Plot a histogram and describe its shape*

*print("""The histogram shows the distribution of movie releases over the years.*

*The shape of the distribution appears to be right-skewed positively skewed,*

*indicating that more movies have been released in recent years compared to earlier decades.*

*The number of releases starts off lower in the early years, gradually increasing, and then surging significantly in the modern era.""")*



The histogram shows the distribution of movie releases over the years.

The shape of the distribution appears to be right-skewed positively skewed,

indicating that more movies have been released in recent years compared to earlier decades.

The number of releases starts off lower in the early years, gradually increasing, and then surging significantly in the modern era.

*#What are the most common genres in the dataset? Use a bar chart to show their distribution.*

*# Group by genre and count the date\_x of movies*

```
gb = df.groupby("genre").agg({"date_x": "count"})
```

```
gb = gb.sort_values(by = "date_x", ascending = False)
```

```
gb = gb.head(20)
```

*# the most common genres in the dataset*

```
plt.figure(figsize = (12,4))
```

```
sns.barplot(x = gb.index, y = gb["date_x"], data = gb ,hue =
```

```
gb.index,palette = "viridis")
```

```
plt.xlabel("Genre")
```

```
plt.ylabel("count of date_x")
```

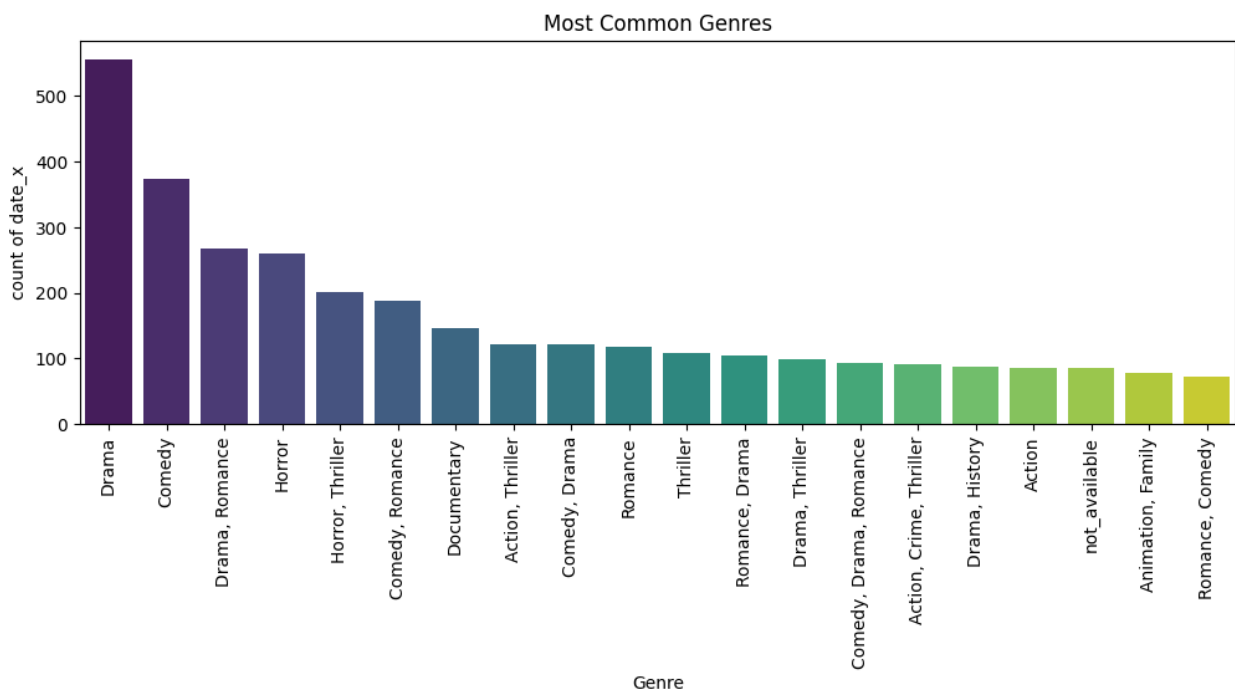
```
plt.title("Most Common Genres")
```

```
plt.xticks(rotation = 90)
```

```
plt.show()
```

*# most common genres*

```
print("The most common genre in the dataset is : Drama")
```



The most common genre in the dataset is : Drama

## 5. Bivariate Analysis: Explore relationships between two variables.

**Task:** Use scatter plots, box plots, and correlation analysis.

**Questions:**

*Is there a relationship between a movie's years and its score? Plot a scatter plot and describe any observed trend.*

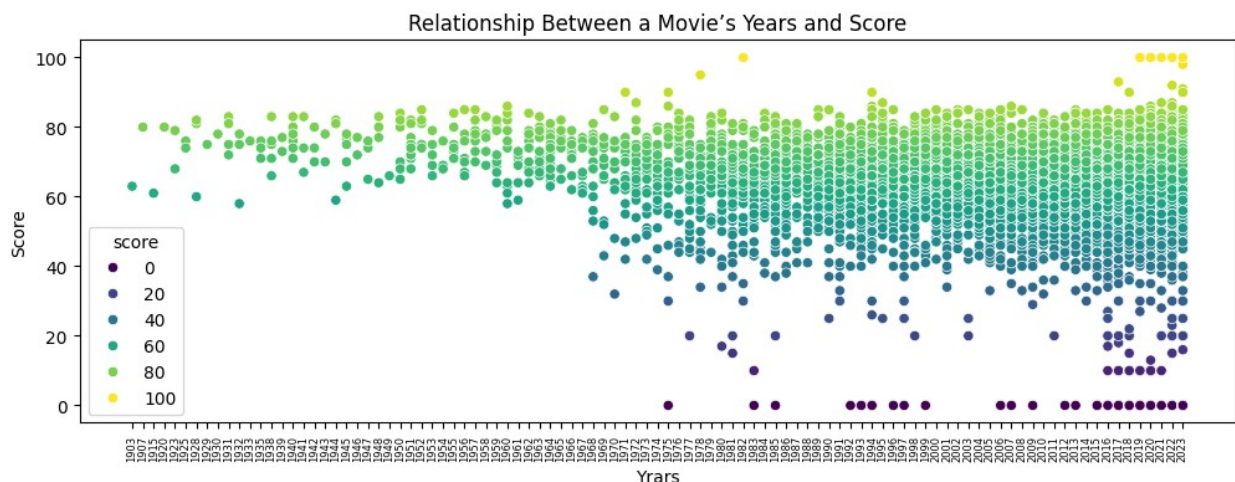
*How do ratings vary by genre? Use a boxplot to visualize the differences in ratings across genres.*

*Is there a correlation between the number of votes a budget and revenue? Create a scatter plot and calculate the correlation coefficient. What can you conclude?*

*# Is there a relationship between a movie's years and its score? Plot a scatter plot and describe any observed trend.*

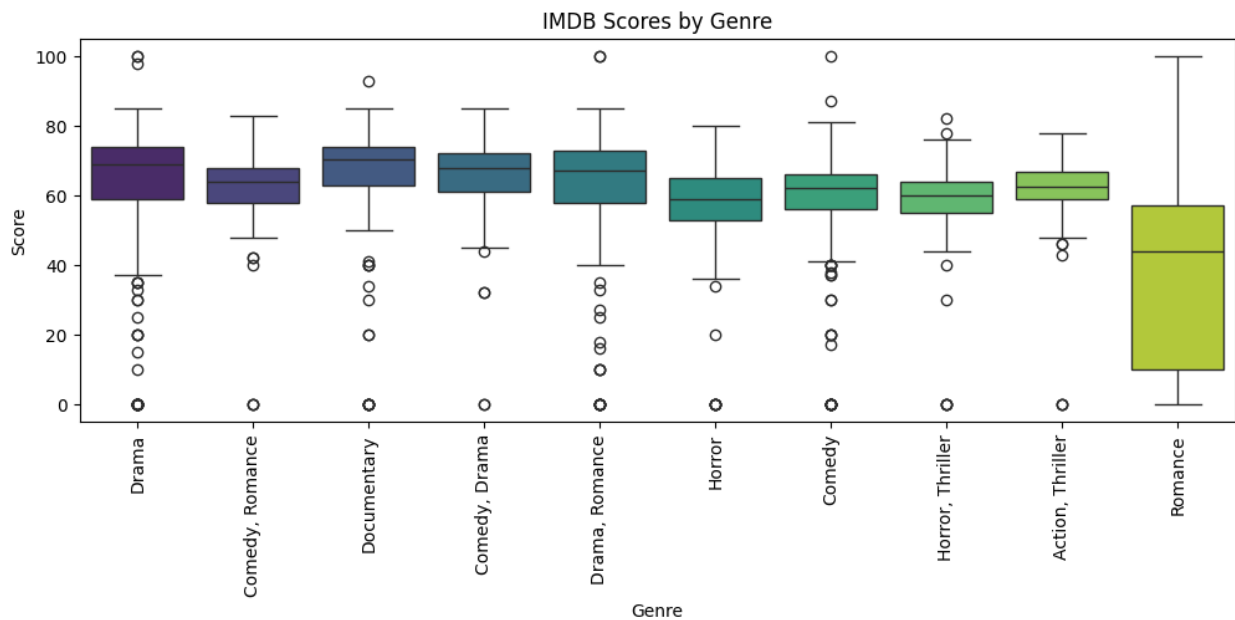
```
df = df.sort_values(by = "years")
plt.figure(figsize=(12,4))
sns.scatterplot(x = "years", y = "score", data = df, hue = "score",
palette = "viridis" )
plt.xticks(rotation = 90, fontsize = 6)
plt.title("Relationship Between a Movie's Years and Score")
plt.xlabel("Yrars")
plt.ylabel("Score")
plt.show
```

```
<function matplotlib.pyplot.show(close=None, block=None)>
```



*#how do ratings vary by genre? Use a boxplot to visualize the differences in ratings across genres.*

```
top_genre = df["genre"].value_counts().head(10).index
plt.figure(figsize=(12, 4))
sns.boxplot(data=df[df["genre"].isin(top_genre)], x="genre",
y="score", hue="genre", palette="viridis")
plt.title("IMDB Scores by Genre")
plt.xlabel("Genre")
plt.ylabel("Score")
plt.xticks(rotation=90)
plt.show()
```



*# Is there a correlation between the number of votes a budget and revenue? Create a scatter plot and calculate the correlation coefficient. What can you conclude?*

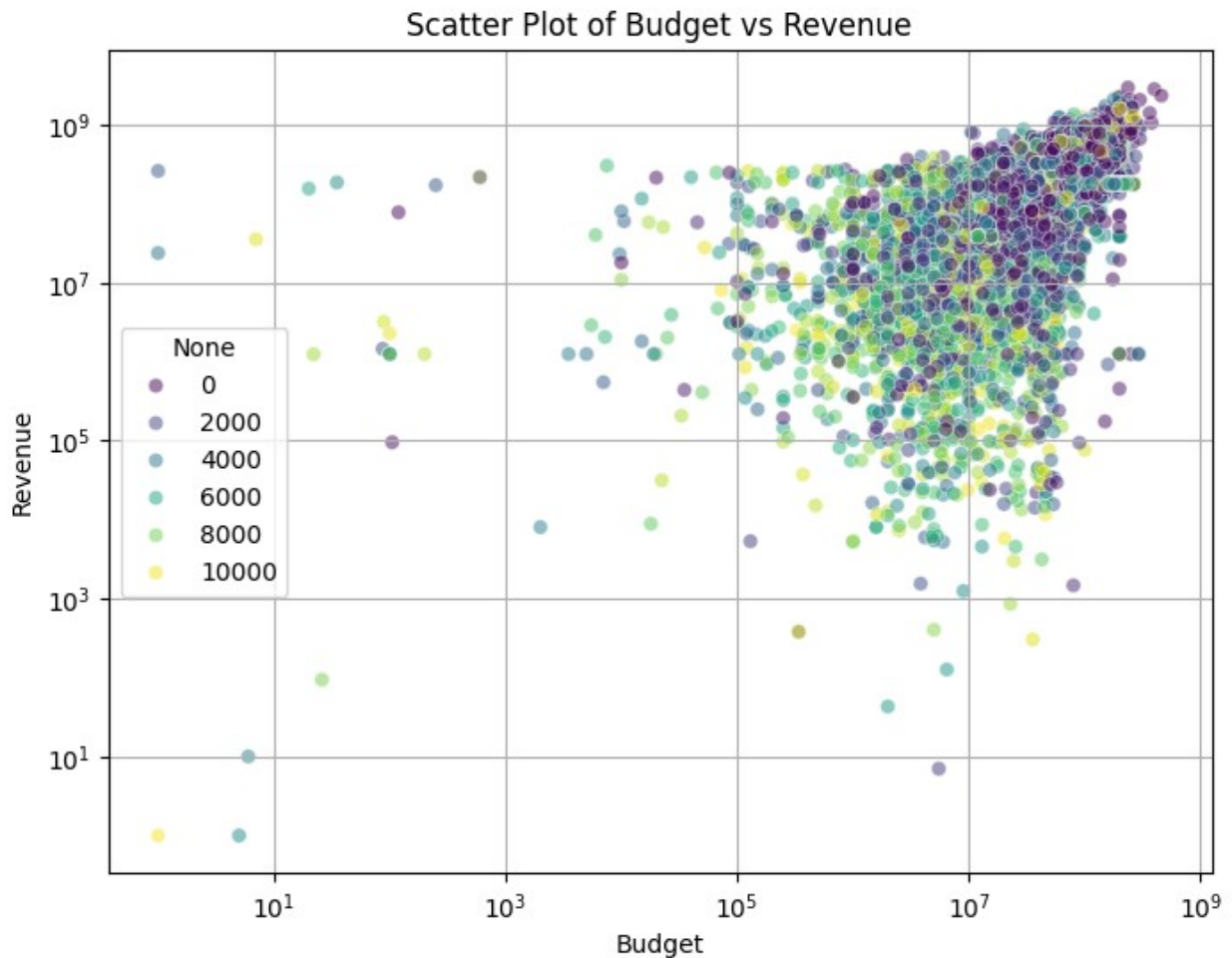
*# Scatter plot*

```
plt.figure(figsize=(8,6))
sns.scatterplot(x=df['budget_x'], y=df['revenue'], alpha=0.5, hue =
df.index, palette = "viridis")
plt.xlabel('Budget')
plt.ylabel('Revenue')
plt.title('Scatter Plot of Budget vs Revenue')
plt.xscale('log') # Log scale for better visualization
plt.yscale('log')
plt.grid(True)
plt.show()
```

*# Compute correlation coefficient*

```
correlation = df['budget_x'].corr(df['revenue'])
```

```
print(f'Correlation coefficient between budget and revenue:  
{correlation:.2f}')
```



Correlation coefficient between budget and revenue: 0.67

## 6.Genre-Specific Analysis

**Task:** Delve deeper into the genre of movies.

**Questions:**

*Which genre has the highest average rating? Calculate the average rating for each genre and plot the results.*

*How does the popularity of genres vary over time? Plot the number of movies released per genre each year.*

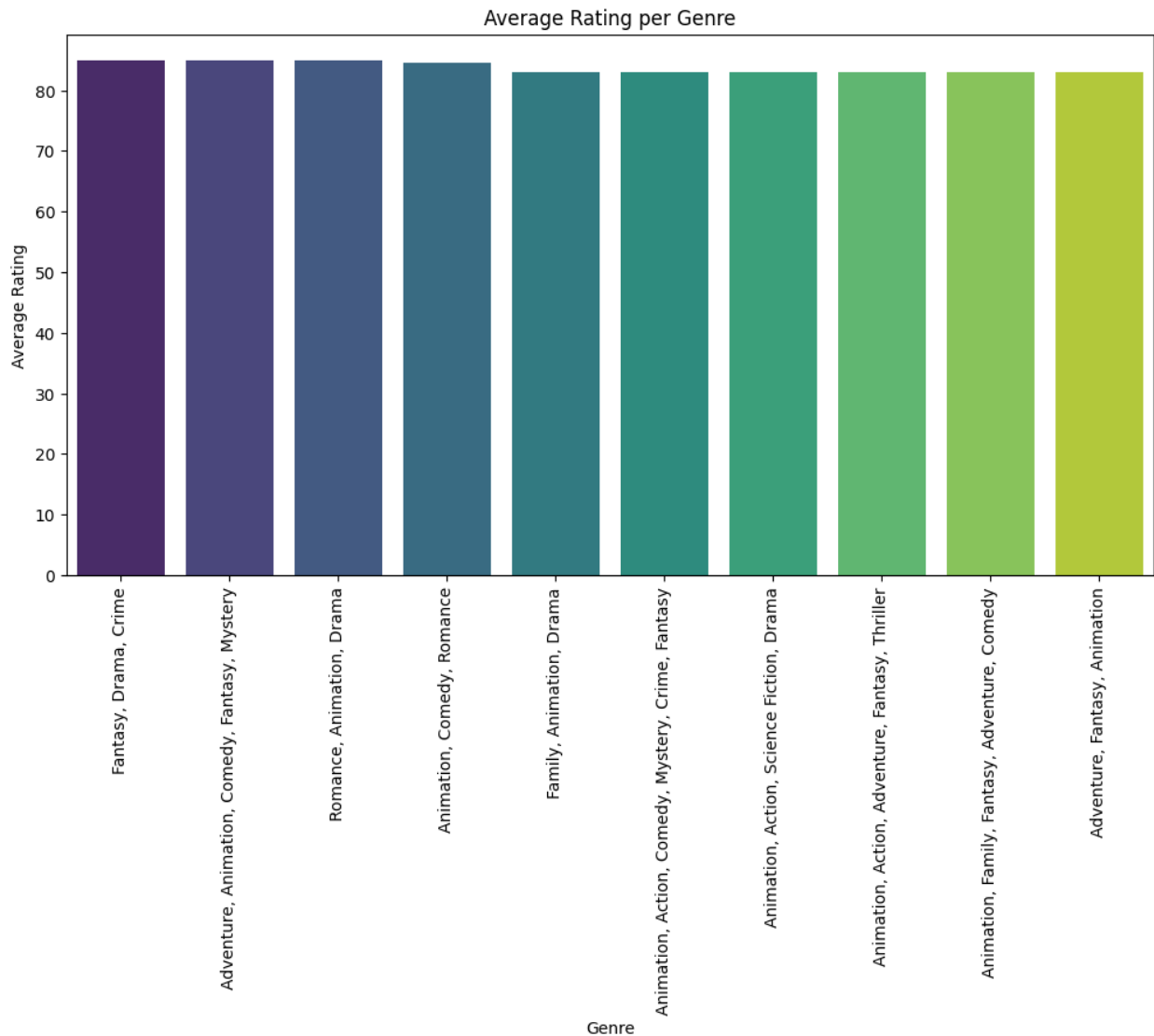
```

# Calculate average rating per genre
genre_avg_rating = df.groupby('genre')
['score'].mean().sort_values(ascending=False)
genre_avg_rating = genre_avg_rating.head(10)
print(genre_avg_rating)

genre
Fantasy, Drama, Crime      85.000000
Adventure, Animation, Comedy, Fantasy, Mystery  85.000000
Romance, Animation, Drama  85.000000
Animation, Comedy, Romance 84.666667
Family, Animation, Drama   83.000000
Animation, Action, Comedy, Mystery, Crime, Fantasy 83.000000
Animation, Action, Science Fiction, Drama  83.000000
Animation, Action, Adventure, Fantasy, Thriller  83.000000
Animation, Family, Fantasy, Adventure, Comedy  83.000000
Adventure, Fantasy, Animation 83.000000
Name: score, dtype: float64

# Plot average rating per genre
plt.figure(figsize=(12,6))
sns.barplot(x=genre_avg_rating.index, y=genre_avg_rating.values,
hue=genre_avg_rating.index, palette='viridis', legend=False)
plt.xlabel('Genre')
plt.ylabel('Average Rating')
plt.title('Average Rating per Genre')
plt.xticks(rotation=90)
plt.show()

```



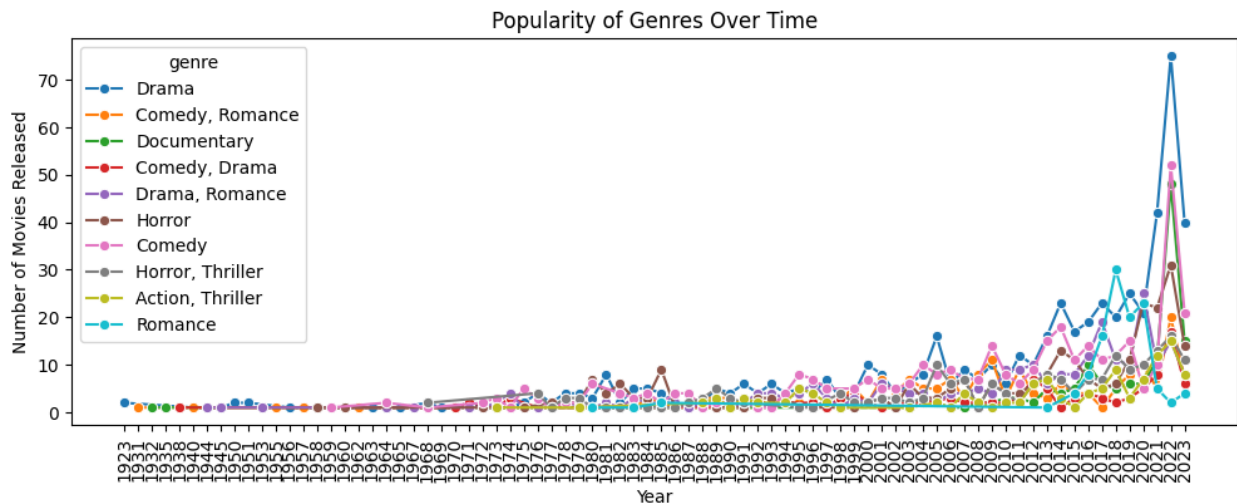
```
#How does the popularity of genres vary over time? Plot the number of
# movies released per genre each year.
# Count number of movies per genre each year
genre_yearly_count = df.groupby(['years',
'genre']).size().reset_index(name='movie_count')
top_genres = genre_yearly_count.groupby('genre')
['movie_count'].sum().nlargest(10).index

# Filter the movies data to include only these top genres
filtered_movies =
genre_yearly_count[genre_yearly_count['genre'].isin(top_genres)]

# Plot genre popularity over time
plt.figure(figsize=(12,4))
sns.lineplot(data=filtered_movies, x='years', y='movie_count',
hue='genre',marker='o')
```



```
plt.xlabel('Year')
plt.xticks(rotation=90)
plt.ylabel('Number of Movies Released')
plt.title('Popularity of Genres Over Time')
plt.show()
```



## 7. Year and Trend Analysis

**Task:** Analyze trends over time.

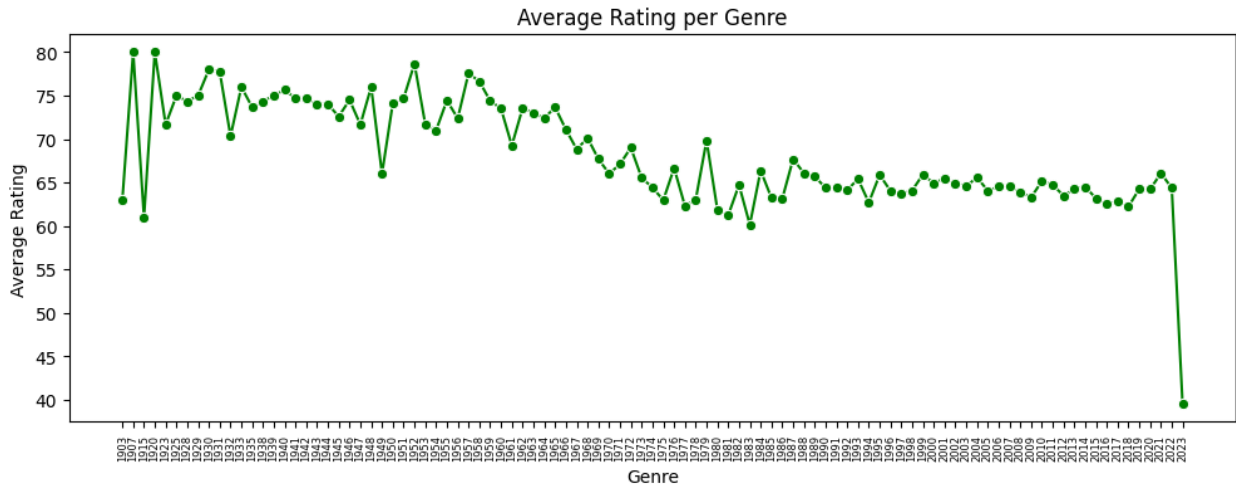
**Questions:**

**How has the average movie rating changed over the years? Plot the average rating for each year.**

**Which years had the highest and lowest number of movie releases? Plot the number of movies released each year.**

```
#How has the average movie rating changed over the years? Plot the
average rating for each year.
# Calculate average rating per genre
genre_avg_rating = df.groupby('years')['score'].mean().reset_index()

# Plot average rating per genre
plt.figure(figsize=(12,4))
sns.lineplot(data=genre_avg_rating, x='years', y='score', marker='o',
color='green')
plt.xlabel('Genre')
plt.ylabel('Average Rating')
plt.title('Average Rating per Genre')
plt.xticks(rotation=90, fontsize=6)
plt.show()
```



*#Which years had the highest and lowest number of movie releases? Plot the number of movies released each year.*

*# Count number of movies released per year*

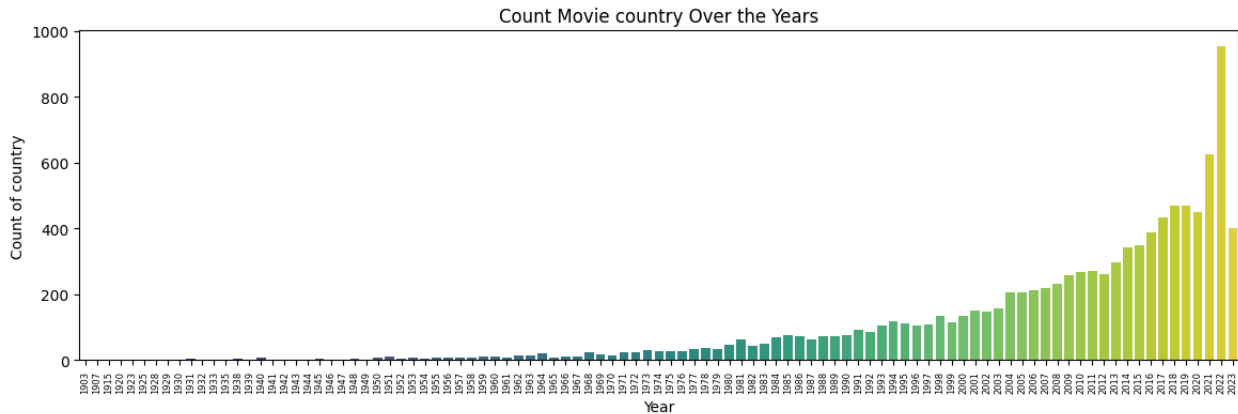
```
gb = df.groupby("years").agg({"country": "count"})
```

*# Plot the number of movies released each year using a bar plot*

```
plt.figure(figsize=(14,4))
sns.barplot(x=gb.index, y=gb['country'], hue=gb.index,
palette='viridis')
plt.ylabel('Count of country')
plt.xlabel('Year')
plt.title('Count Movie country Over the Years')
plt.xticks(rotation=90, fontsize=6)
plt.show()
```

*# Find highest and lowest release years*

```
highest_release_year = gb.idxmax()
lowest_release_year = gb.idxmin()
print(f'Year with highest number of releases: {highest_release_year}
({gb.max()} movies)')
print(f'Year with lowest number of releases: {lowest_release_year}
({gb.min()} movies)')
```



```
Year with highest number of releases: country    2022
dtype: object (country    954
dtype: int64 movies)
Year with lowest number of releases: country    1903
dtype: object (country    1
dtype: int64 movies)
```

## 8. Multivariate Analysis: Analyze multiple variables together.

**Task:** Combine insights from multiple columns to explore complex relationships.

**Questions:**

*Which genres are most popular in each decade? Create a bar plot showing the most frequent genres by decade.*

*Plot a heatmap or pairplot to examine relationships between budget, revenue, scores.*

*Are there specific genres or release years with higher-rated movies? Group by genre and year, then analyze the average rating.*

```
#Which genres are most popular in each decade? Create a bar plot
showing the most frequent genres by decade.
import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np

# Convert 'date_x' to datetime format
df['date_x'] = pd.to_datetime(df['date_x'], errors='coerce')

# Extract decade from the year
df['decade'] = (df['date_x'].dt.year // 10) * 10
```

```

# Drop rows where decade is NaN
df = df.dropna(subset=['decade'])

# Explode the 'genre' column (split multiple genres)
df['genre'] = df['genre'].astype(str) # Ensure it's a string before
splitting
df_exploded =
df.assign(genre=df['genre'].str.split(',')).explode('genre')

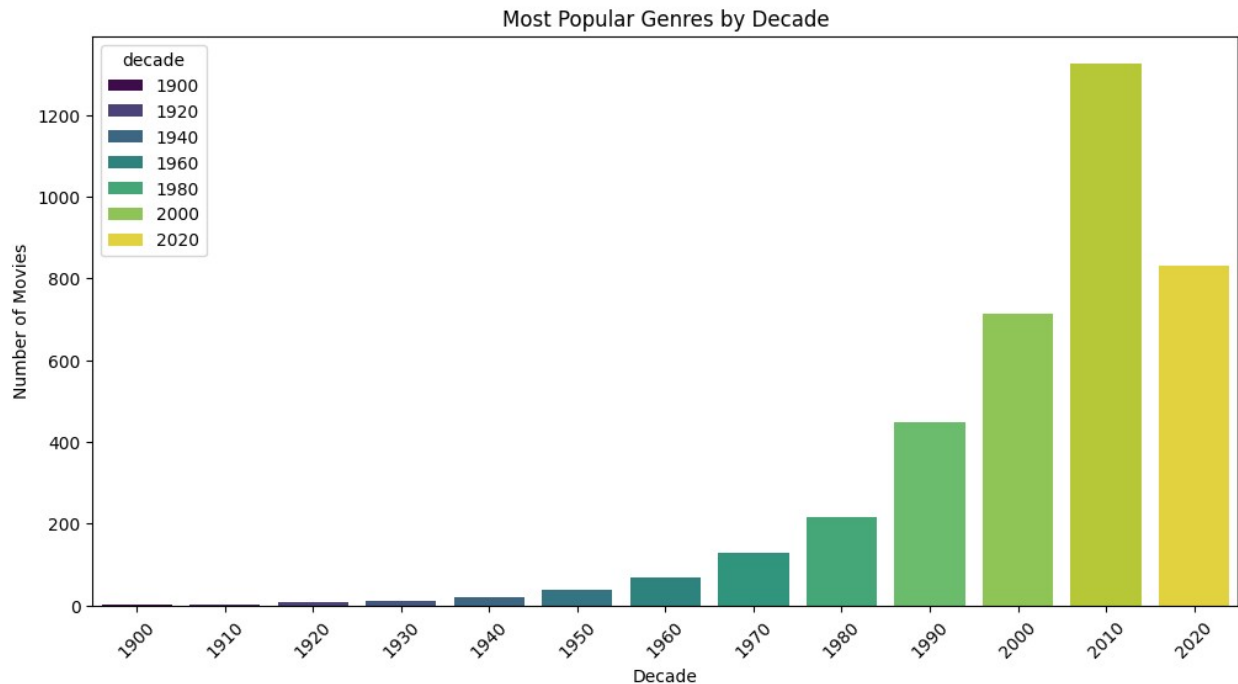
# Trim whitespace from genre names
df_exploded['genre'] = df_exploded['genre'].str.strip()

# Count genres per decade
genre_counts = df_exploded.groupby(['decade',
'genre']).size().reset_index(name='count')

# Get the most frequent genre per decade
top_genres_per_decade =
genre_counts.loc[genre_counts.groupby('decade')['count'].idxmax()]

# Plot the results
plt.figure(figsize=(12, 6))
sns.barplot(data=top_genres_per_decade, x='decade', y='count',
hue='decade', palette='viridis')
plt.xlabel("Decade")
plt.ylabel("Number of Movies")
plt.title("Most Popular Genres by Decade")
plt.xticks(rotation=45)
plt.show()

```



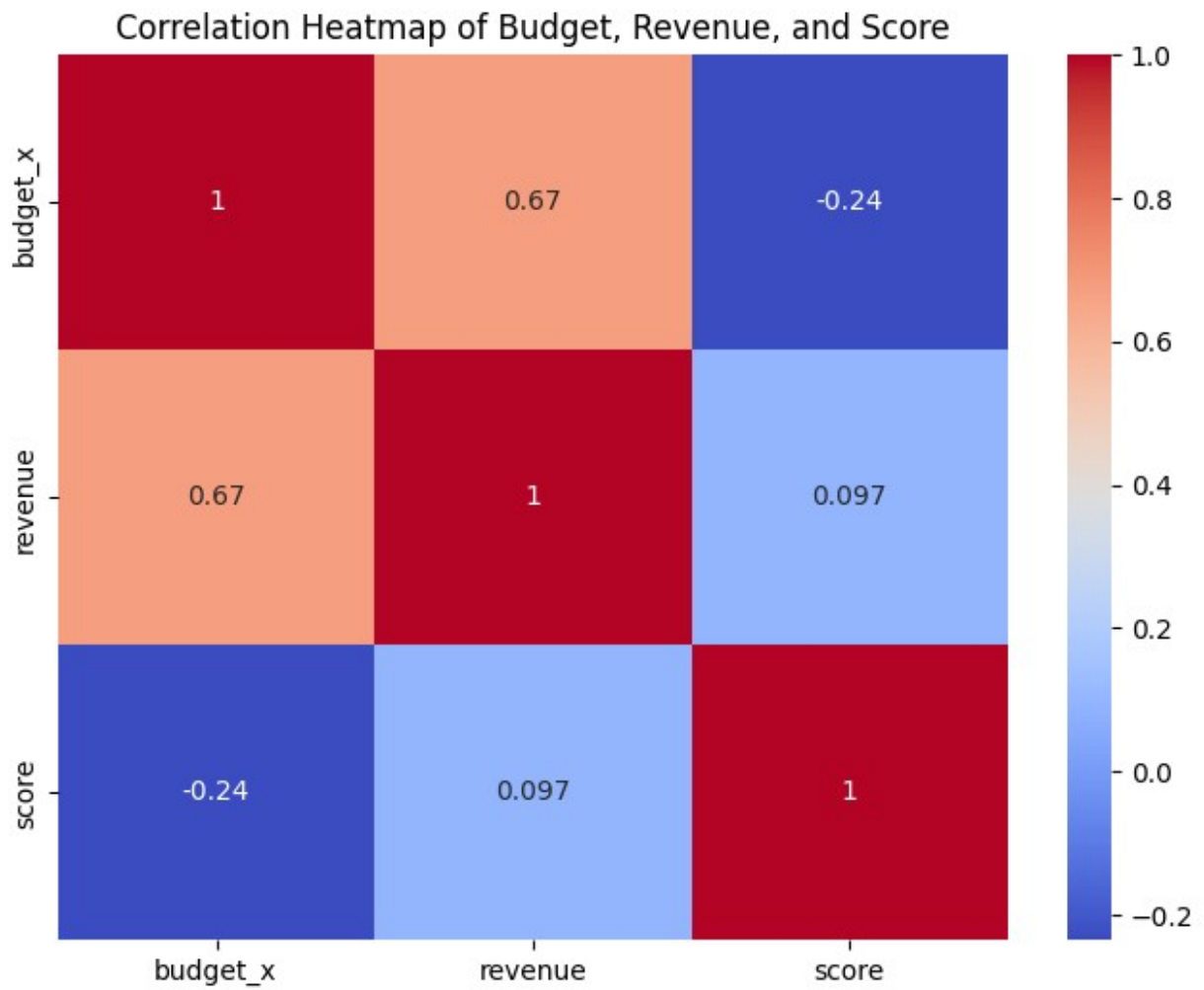
*# Plot a heatmap or pairplot to examine relationships between budget, revenue, scores.*

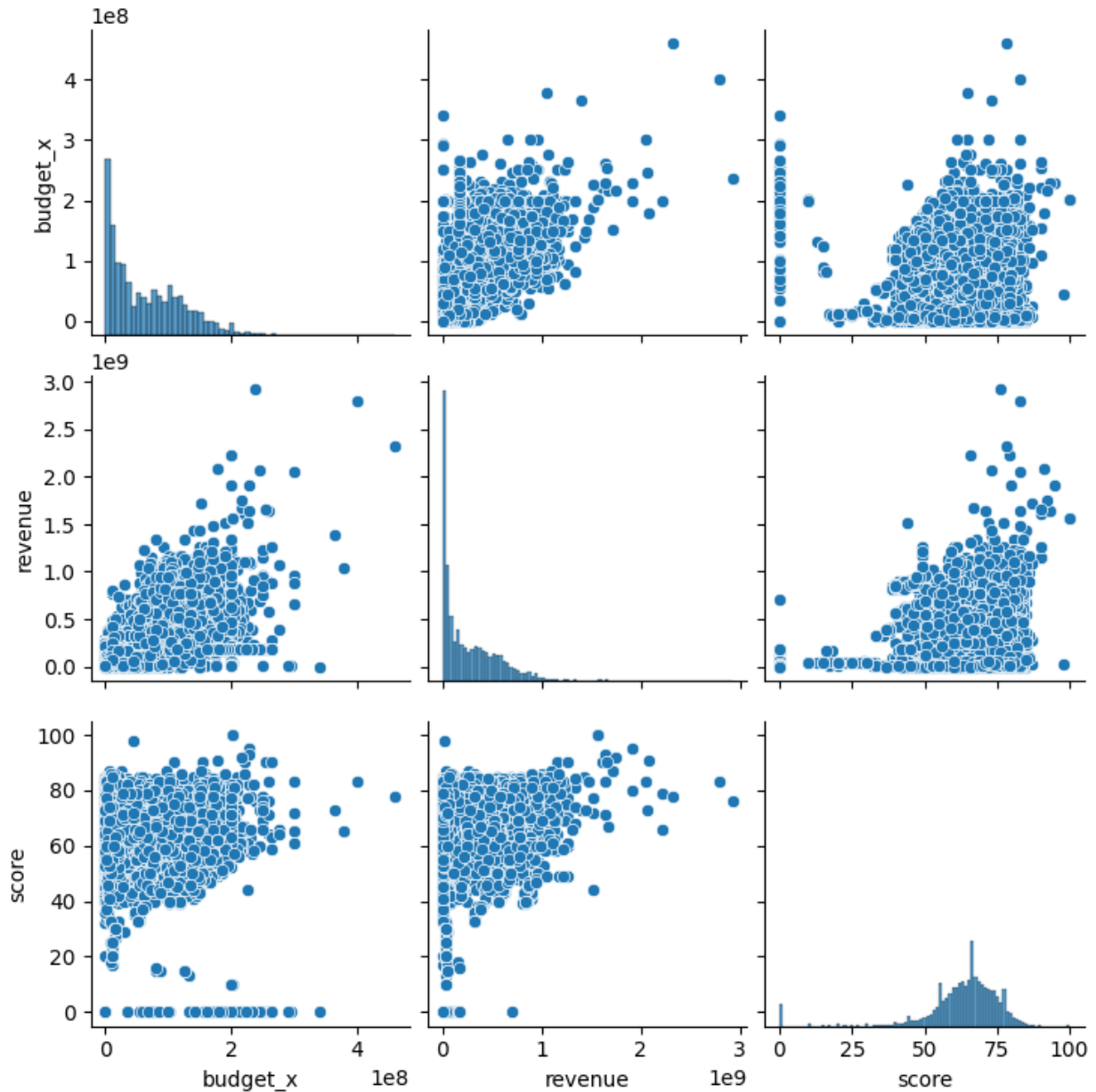
*#Heatmap of correlations*

```
plt.figure(figsize=(8,6))
sns.heatmap(df[['budget_x', 'revenue', 'score']].corr(), annot=True,
            cmap='coolwarm')
plt.title('Correlation Heatmap of Budget, Revenue, and Score')
plt.show()
```

*# Pairplot to examine relationships*

```
sns.pairplot(df[['budget_x', 'revenue', 'score']])
plt.show()
```

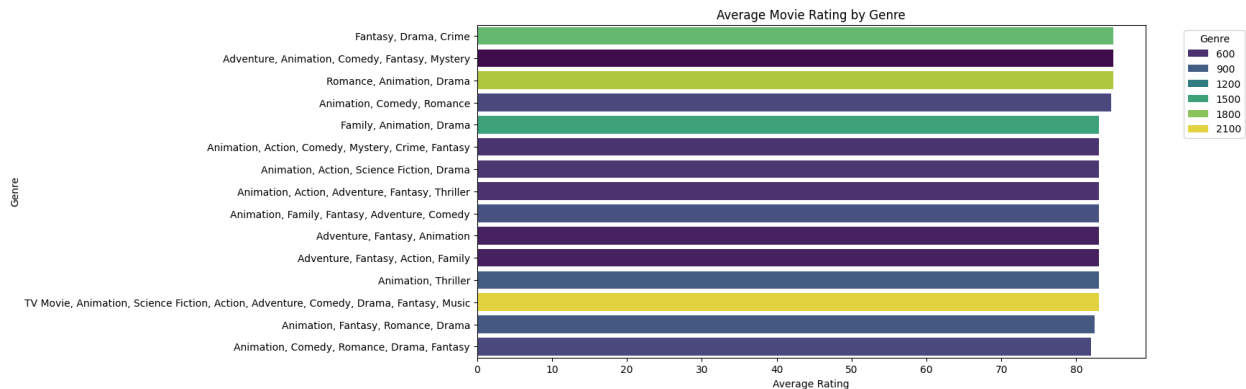




```
#Are there specific genres or release years with higher-rated movies?
#Group by genre and year, then analyze the average rating.
#Group by genre and year, then analyze the average rating
plt.figure(figsize=(12, 6))
genre_avg_rating = df.groupby('genre')['score'].mean().reset_index()
genre_avg_rating = genre_avg_rating.sort_values(by='score',
ascending=False)
genre_avg_rating = genre_avg_rating.head(15)

#Plot the results
sns.barplot(data=genre_avg_rating, x='score',
y='genre', hue=genre_avg_rating.index, palette='viridis')
```

```
plt.xlabel("Average Rating")
plt.ylabel("Genre")
plt.legend(title = 'Genre', bbox_to_anchor = (1.05, 1), loc = 'upper
left')
plt.title("Average Movie Rating by Genre")
plt.show()
```



## 9. Insights and Summary

**Task:** Summarize key findings.

### 1. Movie Releases Trend

- Over time, the number of movies released has grown **significantly**, meaning more films are being produced now than in the past.
- The highest number of movie releases happened in **2022**, showing a peak in production.
- The distribution of releases is **right-skewed**, which means earlier years had fewer releases, and the number has increased rapidly in recent decades.
- This trend suggests that **technological advancements, streaming platforms, and increased global interest** in filmmaking have contributed to more movies being made.

### 2. Popular Genres & Ratings

- Drama** is the **most common genre** across all movies, followed by **Action** and **Adventure**.
- However, even though Drama is the most frequent, movies in the **Fantasy, Animation, and Comedy** genres tend to receive **higher IMDB ratings (85 on average)**.
- This means that while Drama movies are made the most, **audiences and critics rate Fantasy and Animation movies higher** on average.



- One possible reason is that **animated and fantasy films often have better storytelling, visual appeal, and emotional impact**, leading to higher ratings.
- 

### 3. Budget vs. Revenue

- Movies with **higher budgets tend to earn more money** at the box office.
  - The **correlation is 0.67**, which is a **strong positive relationship**, meaning that **spending more on production usually results in higher revenue**.
  - However, this isn't a **perfect correlation (1.0)**, meaning some low-budget films can still perform well, and some expensive films can fail.
  - Other factors like **marketing, genre, and audience engagement** also influence how much revenue a movie makes.
- 

### 4. Trends Over Decades

- In **earlier decades, Drama was the most dominant genre**, meaning most movies focused on **serious storytelling and real-life themes**.
- In recent decades, **Action, Sci-Fi, and Fantasy genres have gained popularity**, likely due to advancements in CGI, special effects, and growing audience demand for visually engaging films.
- When looking at the **highest-rated movies**, they mostly belong to the **Fantasy, Animation, and Drama** genres.
- This suggests that **people generally enjoy imaginative and visually stunning films**, which receive better ratings compared to standard genres like Drama or Action.