

Netflix Data Analysis Project

Project Analyst

SWAPNISH PANDEY

(Master Of Computer Applications)

Project Summary & Key Insights

- This project presents a comprehensive analysis of Netflix's content catalog using Python and data visualization techniques. The dataset includes thousands of titles spanning movies and TV shows, with metadata such as genre, duration, country of origin, release year, and date added to the platform. The goal of this analysis was to uncover patterns in content strategy, viewer engagement, and platform evolution over time.
- The study begins by examining the distribution of content types. It was observed that movies dominate the platform, but TV shows have shown consistent growth, especially in international markets. The average duration of movies was found to be approximately 90 minutes, aligning with standard feature-length expectations. Interestingly, a trend analysis revealed that movie durations have slightly declined in recent years, possibly reflecting changing viewer attention spans and the rise of mobile-first consumption.
- For TV shows, the most common number of seasons is one, indicating a high volume of limited series or pilot content. This suggests that Netflix frequently experiments with new formats and concepts, using viewer feedback to determine renewals. Genre distribution over the years showed that Drama and Comedy remain dominant, while Documentaries and International content have grown significantly, reflecting Netflix's global expansion and diversified audience base.
- The analysis also explored content launch strategy. It was found that July and December are peak months for content additions, aligning with summer breaks and holiday seasons. This insight is valuable for planning promotional campaigns and major releases. Additionally, the genre-country relationship revealed that the United States leads in Drama and Comedy, while India and South Korea are prominent in Romance and Action genres. This supports Netflix's strategy of tailoring content to regional preferences and leveraging local production hubs.
- Overall, this project demonstrates how data-driven insights can inform content acquisition, production planning, and viewer engagement strategies. By understanding trends in duration, genre, and geography, Netflix can optimize its catalog to meet evolving audience demands. The analysis not only highlights current strengths but also uncovers opportunities for future growth, especially in emerging markets and underrepresented genres.

```

!pip install pandas

Defaulting to user installation because normal site-packages is not
writeable
Requirement already satisfied: pandas in c:\users\hp\appdata\roaming\
python\python313\site-packages (2.3.3)
Requirement already satisfied: numpy>=1.26.0 in c:\users\hp\appdata\
roaming\python\python313\site-packages (from pandas) (2.3.2)
Requirement already satisfied: python-dateutil>=2.8.2 in c:\users\hp\
appdata\roaming\python\python313\site-packages (from pandas)
(2.9.0.post0)
Requirement already satisfied: pytz>=2020.1 in c:\users\hp\appdata\
roaming\python\python313\site-packages (from pandas) (2025.2)
Requirement already satisfied: tzdata>=2022.7 in c:\users\hp\appdata\
roaming\python\python313\site-packages (from pandas) (2025.2)
Requirement already satisfied: six>=1.5 in c:\users\hp\appdata\
roaming\python\python313\site-packages (from python-dateutil>=2.8.2-
>pandas) (1.17.0)

import pandas as pd
print(pd.__version__)

```

2.3.3

```

df = pd.read_csv(r"C:\Users\HP\Downloads\netflix.csv")
df.head()

      show_id    type          title        director \
0         s1  Movie  Dick Johnson Is Dead  Kirsten Johnson
1         s2  TV Show        Blood & Water           NaN
2         s3  TV Show          Ganglands  Julien Leclercq
3         s4  TV Show   Jailbirds New Orleans           NaN
4         s5  TV Show          Kota Factory           NaN

                           cast        country \
0                         NaN  United States
1  Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban...  South Africa
2  Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabi...
3                         NaN           NaN
4  Mayur More, Jitendra Kumar, Ranjan Raj, Alam K...  India

      date_added  release_year rating duration \
0  September 25, 2021       2020  PG-13     90 min
1  September 24, 2021       2021  TV-MA    2 Seasons
2  September 24, 2021       2021  TV-MA    1 Season
3  September 24, 2021       2021  TV-MA    1 Season
4  September 24, 2021       2021  TV-MA    2 Seasons

                           listed_in \
0                  Documentaries
1  International TV Shows, TV Dramas, TV Mysteries

```

```

2 Crime TV Shows, International TV Shows, TV Act...
3 Docuseries, Reality TV
4 International TV Shows, Romantic TV Shows, TV ...

                           description
0 As her father nears the end of his life, filmm...
1 After crossing paths at a party, a Cape Town t...
2 To protect his family from a powerful drug lor...
3 Feuds, flirtations and toilet talk go down amo...
4 In a city of coaching centers known to train I...

df['date_added'] = pd.to_datetime(df['date_added'], errors='coerce')
df['month_added'] = df['date_added'].dt.month
df['year_added'] = df['date_added'].dt.year

df['duration_type'] = df['duration'].apply(lambda x: 'Season' if
'Season' in str(x) else 'Minutes')
df['duration_value'] = df['duration'].str.extract(r'(\d+)').astype(float)

df = df.copy()
df['director'] = df['director'].fillna('Unknown')
df['cast'] = df['cast'].fillna('Unknown')
df['country'] = df['country'].fillna('Unknown')

## 1:- Content Strategy

## Q1 :- What is the ratio of movies vs TV shows on Netflix?

type_counts = df['type'].value_counts()
type_percent = df['type'].value_counts(normalize=True) * 100

print("Content Type Counts:\n", type_counts)
print("\nContent Type Percentages:\n", type_percent)

type_counts.plot(kind='pie', autopct='%.1f%%', startangle=90,
colors=['#66b3ff', '#ff9999'], figsize=(6,6))
plt.title('Movies vs TV Shows Ratio')
plt.ylabel('')
plt.show()

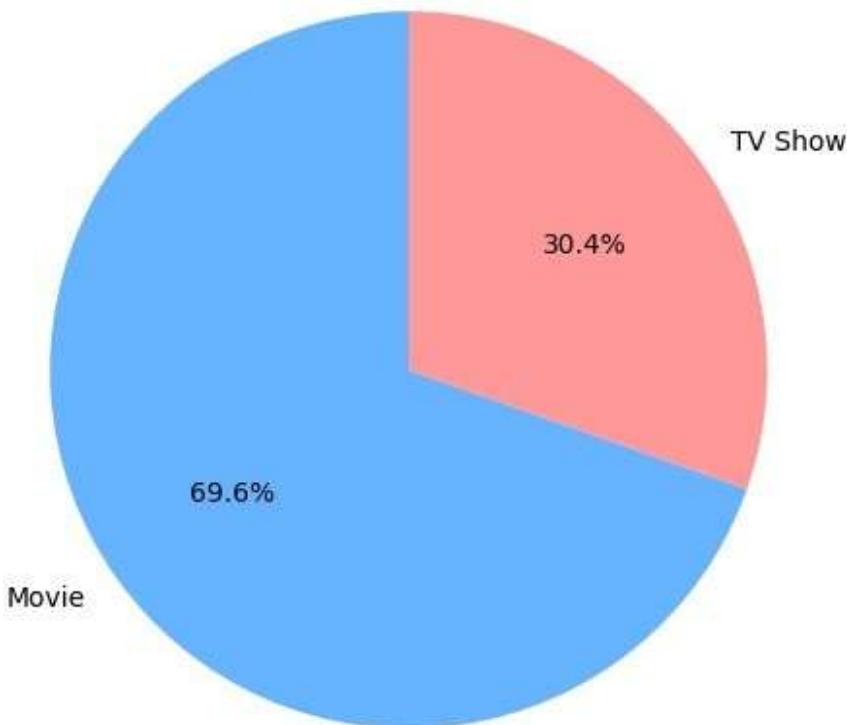
Content Type Counts:
type
Movie      6131
TV Show    2676
Name: count, dtype: int64

Content Type Percentages:
type
Movie     69.615079

```

```
TV Show      30.384921
Name: proportion, dtype: float64
```

Movies vs TV Shows Ratio



```
##Q2:- Which genres are most popular on Netflix globally?
```

```
genre_series =
df['listed_in'].dropna().str.split(',').explode().str.strip()
top_genres = genre_series.value_counts().head(10)

print("Top Genres:\n", top_genres)

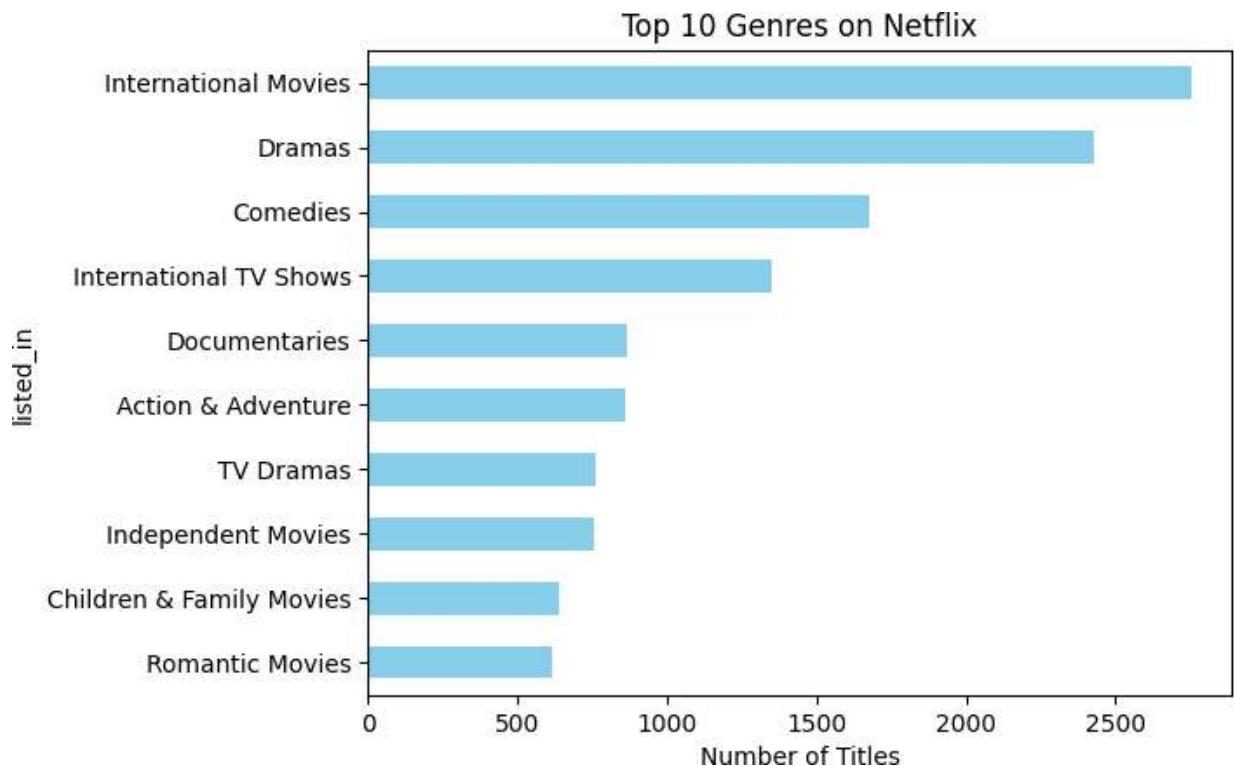
top_genres.plot(kind='barh', color='skyblue')
plt.title('Top 10 Genres on Netflix')
plt.xlabel('Number of Titles')
plt.gca().invert_yaxis()
plt.show()

Top Genres:
listed_in
```

```

International Movies      2752
Dramas                  2427
Comedies                1674
International TV Shows   1351
Documentaries            869
Action & Adventure       859
TV Dramas                763
Independent Movies        756
Children & Family Movies   641
Romantic Movies           616
Name: count, dtype: int64

```



```

##Q3:- Which years saw the highest release of content on Netflix ?

release_years = df['release_year'] =
pd.to_datetime(df['release_year'], errors='coerce').dt.year
year_counts = df['release_year'].value_counts().sort_index()

year_counts.plot(kind='bar', figsize=(12,6), color='coral')
plt.title('Content Releases by Year')
plt.xlabel('Release Year')
plt.ylabel('Number of Titles')
plt.show()

```

```
-----
NameError                               Traceback (most recent call
last)
Cell In[5], line 1
----> 1 release_years = df['release_year'] =
pd.to_datetime(df['release_year'], errors='coerce').dt.year
    2 year_counts = df['release year'].value_counts().sort_index()
    4 year_counts.plot(kind='bar', figsize=(12,6), color='coral')

NameError: name 'pd' is not defined

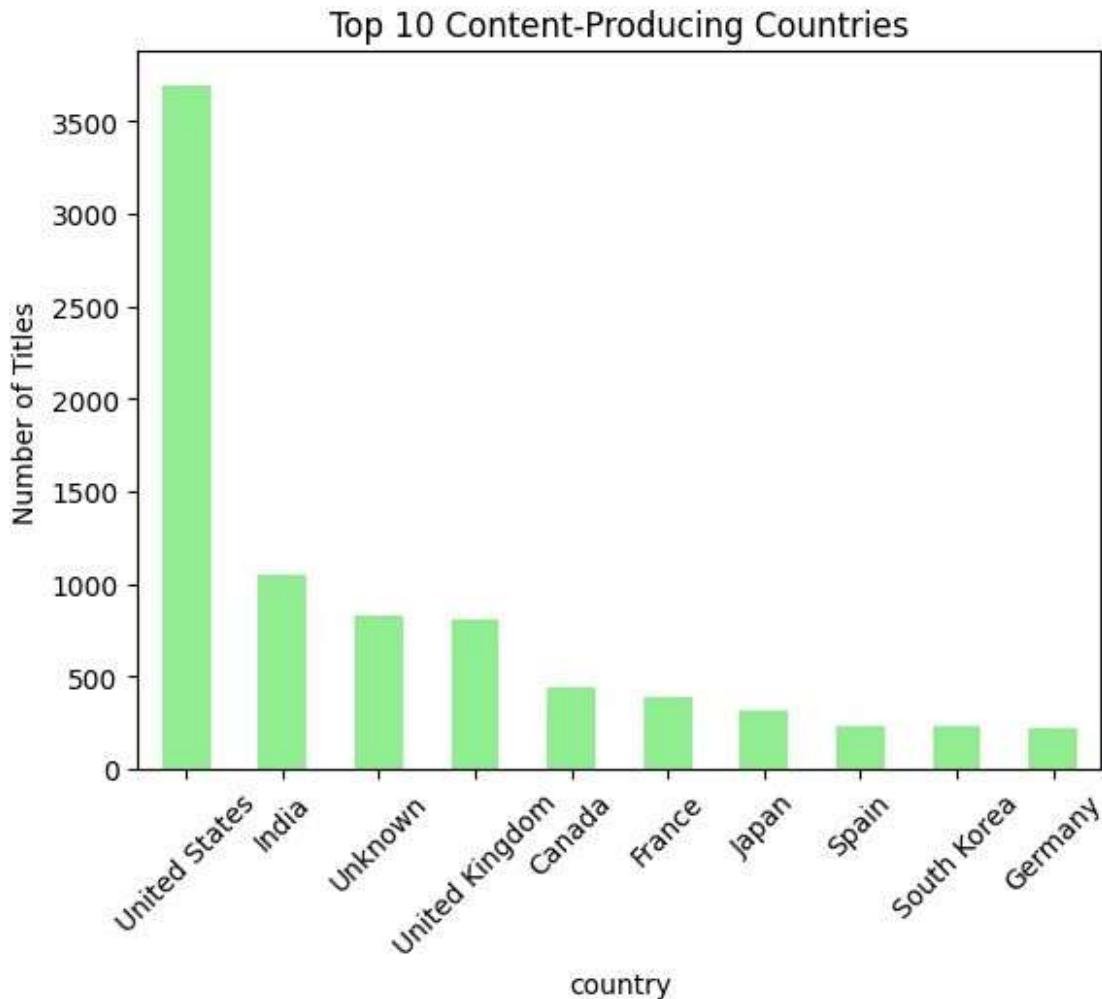
#Q4:- Which Countries Produce the most Netflix content ?

country_series =
df['country'].dropna().str.split(',').explode().str.strip()
top_countries = country_series.value_counts().head(10)

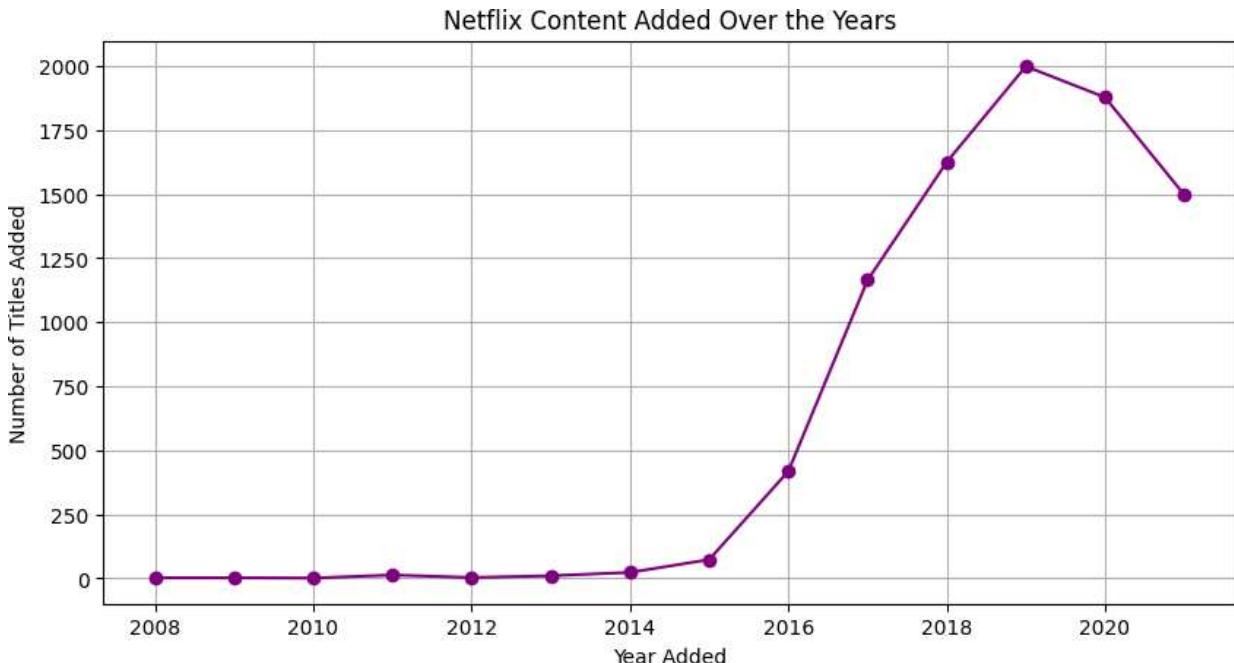
print("Top Countries:\n", top_countries)

top_countries.plot(kind='bar', color='lightgreen')
plt.title('Top 10 Content-Producing Countries')
plt.ylabel('Number of Titles')
plt.xticks(rotation=45)
plt.show()

Top Countries:
   country
United States      3690
India              1046
Unknown             831
United Kingdom     806
Canada              445
France              393
Japan                318
Spain                232
South Korea          231
Germany              226
Name: count, dtype: int64
```



```
##Q5:-How has the trend of adding new content evolved year by year?  
year_added_counts = df['year_added'].value_counts().sort_index()  
  
year_added_counts.plot(kind='line', marker='o', color='purple',  
figsize=(10,5))  
plt.title('Netflix Content Added Over the Years')  
plt.xlabel('Year Added')  
plt.ylabel('Number of Titles Added')  
plt.grid(True)  
plt.show()
```



```
##6:-6.Which ratings (e.g., TV-MA, PG, etc.) are most frequent on Netflix?
```

```
rating_counts = df['rating'].value_counts().head(10)

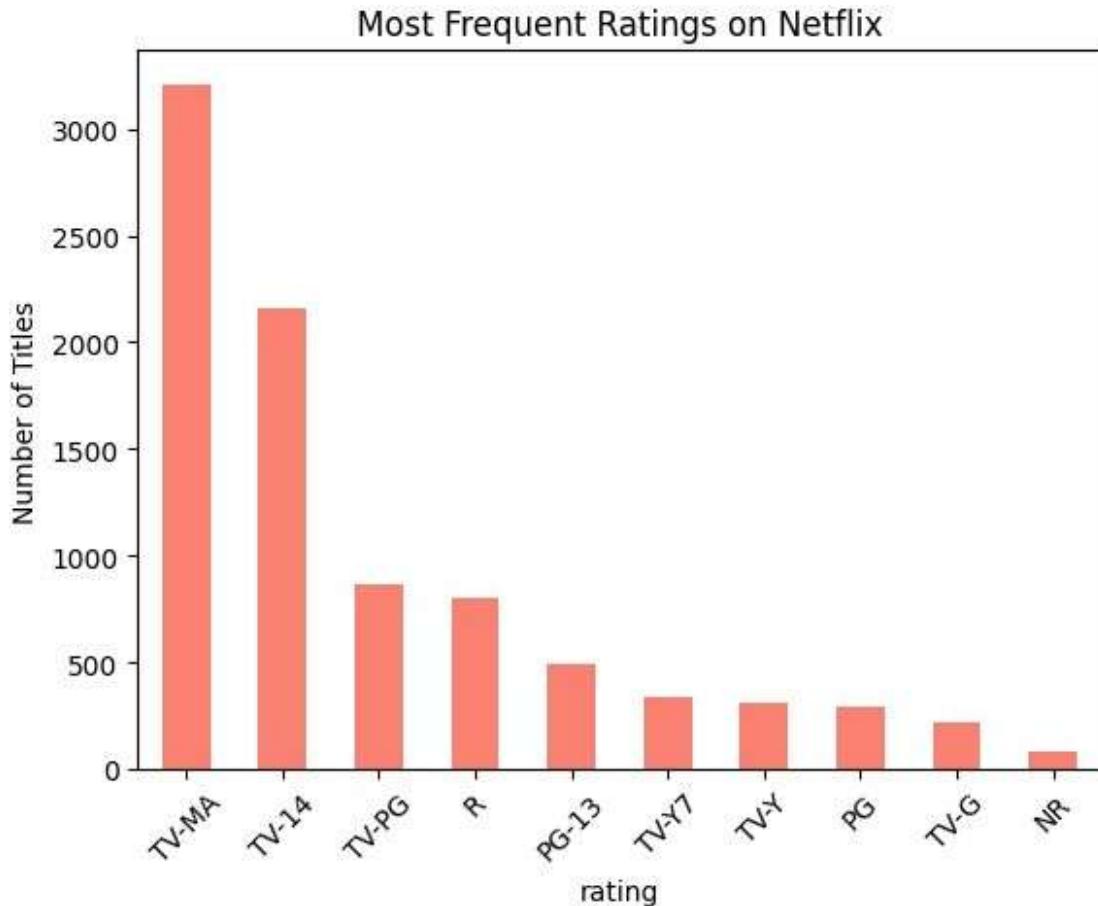
print("Top Ratings:\n", rating_counts)

rating_counts.plot(kind='bar', color='salmon')
plt.title('Most Frequent Ratings on Netflix')
plt.ylabel('Number of Titles')
plt.xticks(rotation=45)
plt.show()
```

Top Ratings:

rating	
TV-MA	3207
TV-14	2160
TV-PG	863
R	799
PG-13	490
TV-Y7	334
TV-Y	307
PG	287
TV-G	220
NR	80

Name: count, dtype: int64



#Q7:-Do some countries tend to produce more mature content (TV-MA) ?

```
tvma_df = df[df['rating'] == 'TV-MA']
tvma_countries =
tvma_df['country'].dropna().str.split(',').explode().str.strip().value_
counts().head(10)

print("Top TV-MA Producing Countries:\n", tvma_countries)

tvma_countries.plot(kind='bar', color='darkred')
plt.title('Countries Producing Most TV-MA Content')
plt.ylabel('Number of TV-MA Titles')
plt.xticks(rotation=45)
plt.show()
```

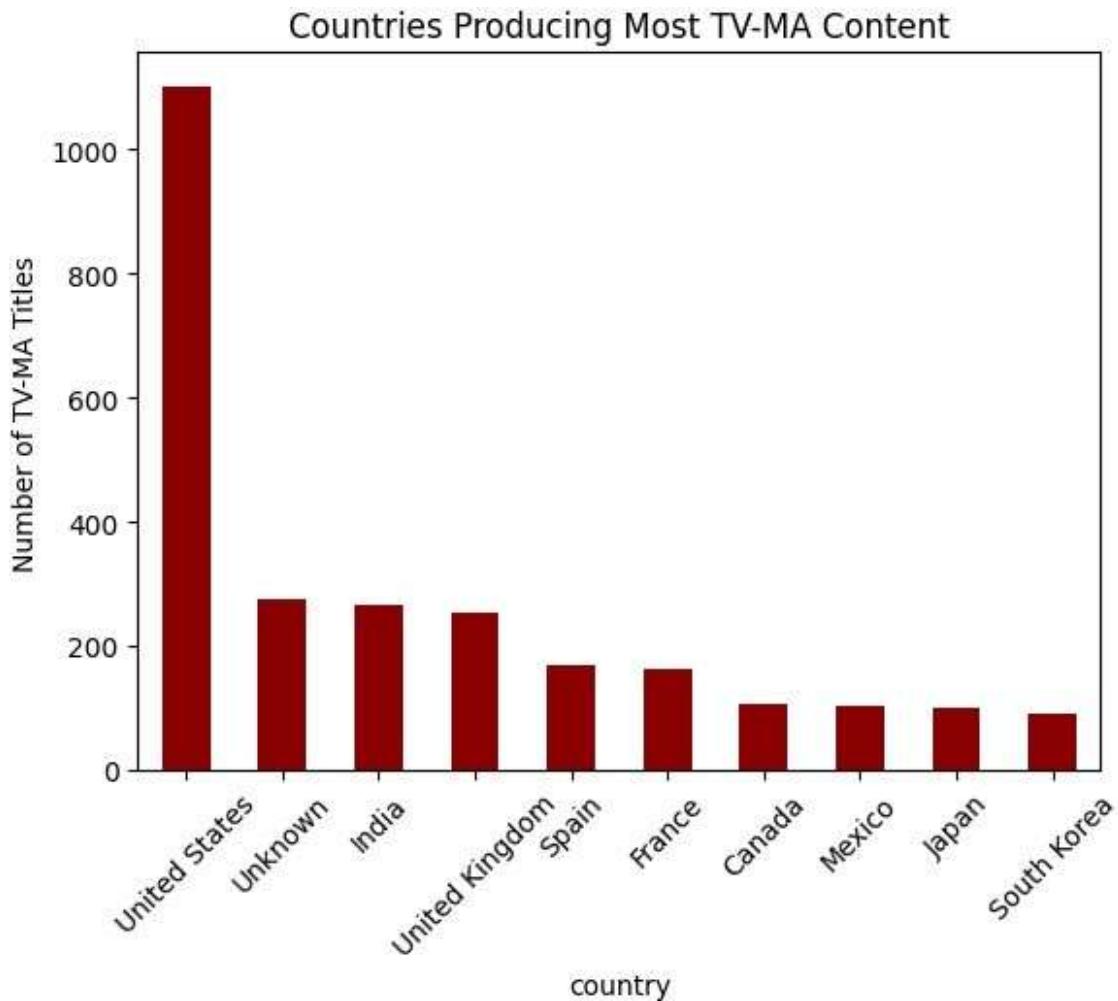
Top TV-MA Producing Countries:

country	
United States	1101
Unknown	276
India	266
United Kingdom	253
Spain	170

```

France          163
Canada          107
Mexico          102
Japan           101
South Korea     92
Name: count, dtype: int64

```



#Q8:-Which genres are more associated with TV shows .

```

genre_split = df[['type', 'listed_in']].dropna()
genre_split =
genre_split.assign(genres=genre_split['listed_in'].str.split(',')).
explode('genres')
genre_split['genres'] = genre_split['genres'].str.strip()

genre_pivot = genre_split.groupby(['type',
'genres']).size().unstack(fill_value=0)
genre_pivot = genre_pivot.loc[:, 

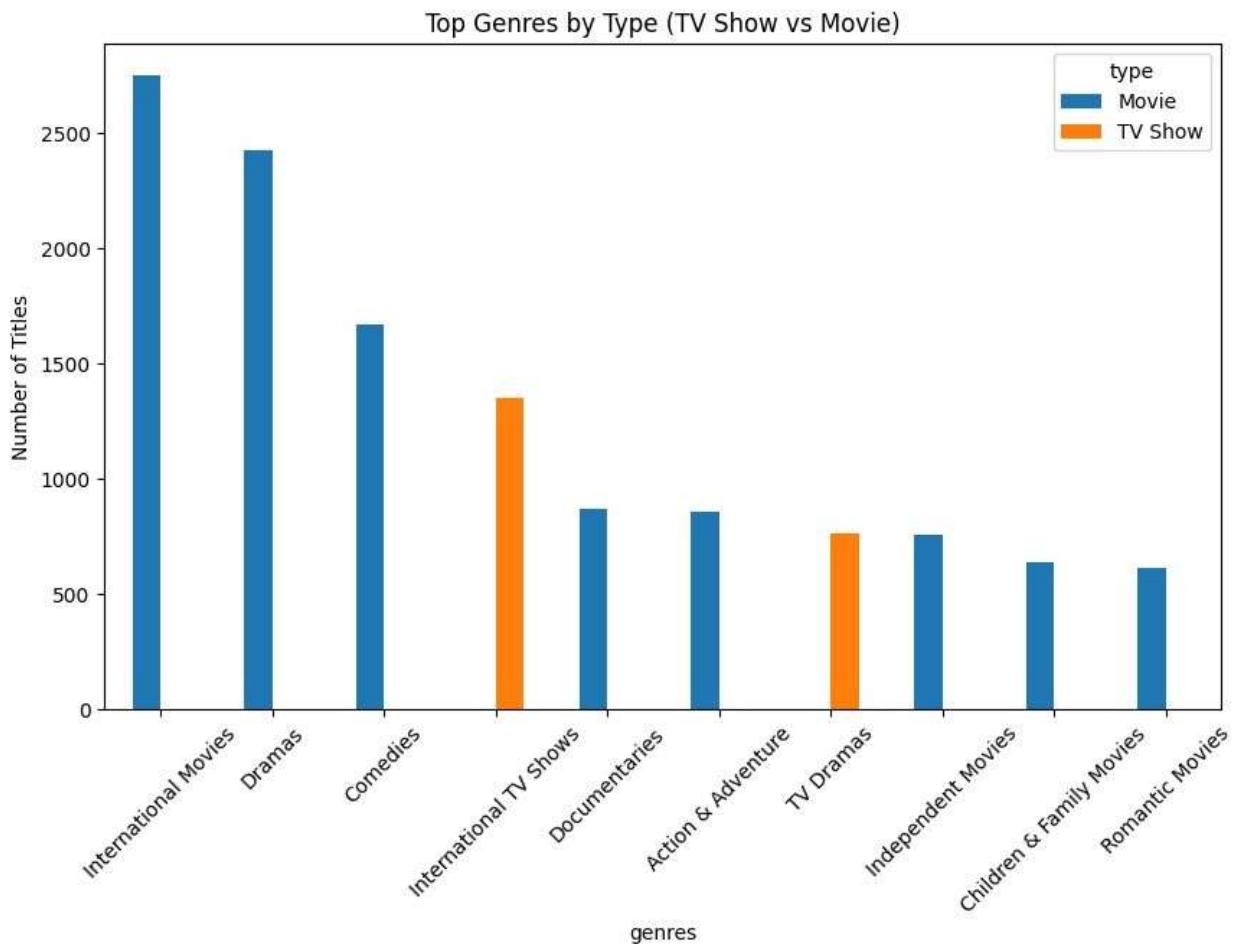
```

```

genre_pivot.sum().sort_values(ascending=False).head(10).index]

genre_pivot.T.plot(kind='bar', figsize=(10, 6))
plt.title('Top Genres by Type (TV Show vs Movie)')
plt.ylabel('Number of Titles')
plt.xticks(rotation=45)
plt.show()

```



#9:-Which genres dominate the U.S. vs other countries?

```

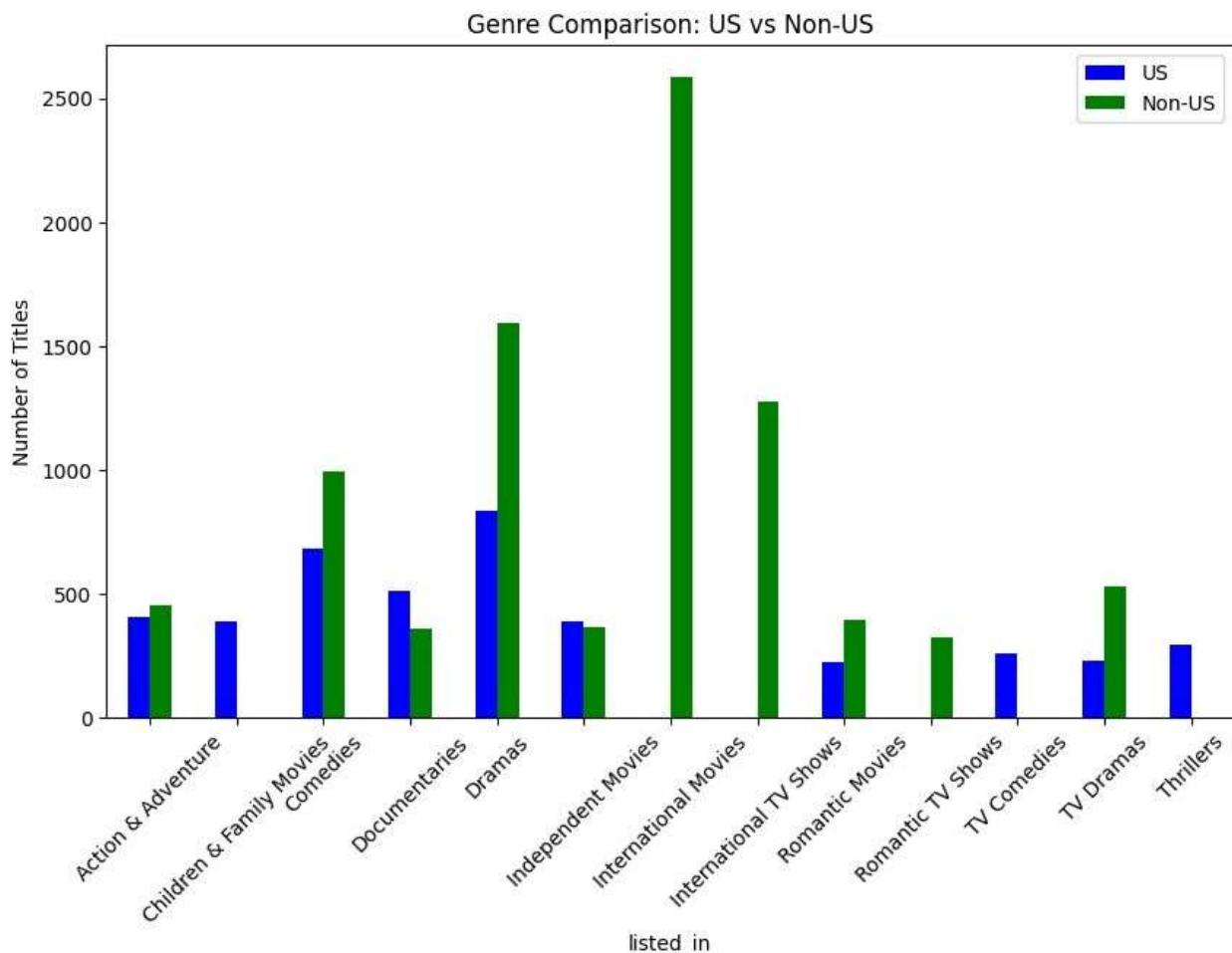
us_df = df[df['country'].str.contains('United States', na=False)]
non_us_df = df[~df['country'].str.contains('United States', na=False)]

us_genres =
us_df['listed_in'].dropna().str.split(',').explode().str.strip().value
_counts().head(10)
non_us_genres =
non_us_df['listed_in'].dropna().str.split(',').explode().str.strip().v
alue_counts().head(10)

```

```
# Combine for comparison
genre_compare = pd.DataFrame({'US': us_genres, 'Non-US': non_us_genres}).fillna(0)

genre_compare.plot(kind='bar', figsize=(10, 6), color=['blue', 'green'])
plt.title('Genre Comparison: US vs Non-US')
plt.ylabel('Number of Titles')
plt.xticks(rotation=45)
plt.show()
```



```
#10:-What genres are most popular in the last 3 years?

recent_df = df[df['year_added'] >= (df['year_added'].max() - 2)]
recent_genres =
recent_df['listed_in'].dropna().str.split(',').explode().str.strip().value_counts().head(10)

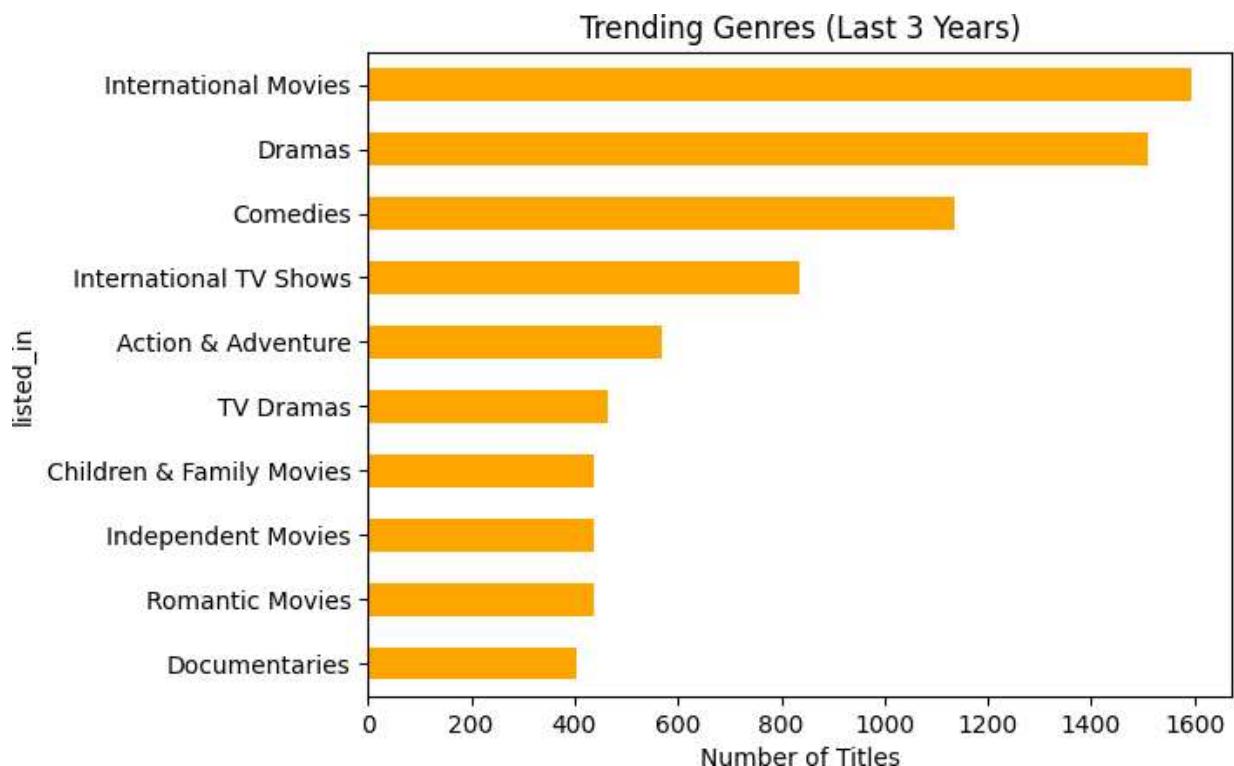
print("Top Genres (Last 3 Years):\n", recent_genres)
```

```

recent_genres.plot(kind='barh', color='orange')
plt.title('Trending Genres (Last 3 Years)')
plt.xlabel('Number of Titles')
plt.gca().invert_yaxis()
plt.show()

Top Genres (Last 3 Years):
   listed_in
International Movies      1593
Dramas                  1511
Comedies                 1135
International TV Shows    836
Action & Adventure       568
TV Dramas                 463
Children & Family Movies  439
Independent Movies        438
Romantic Movies            437
Documentaries              405
Name: count, dtype: int64

```



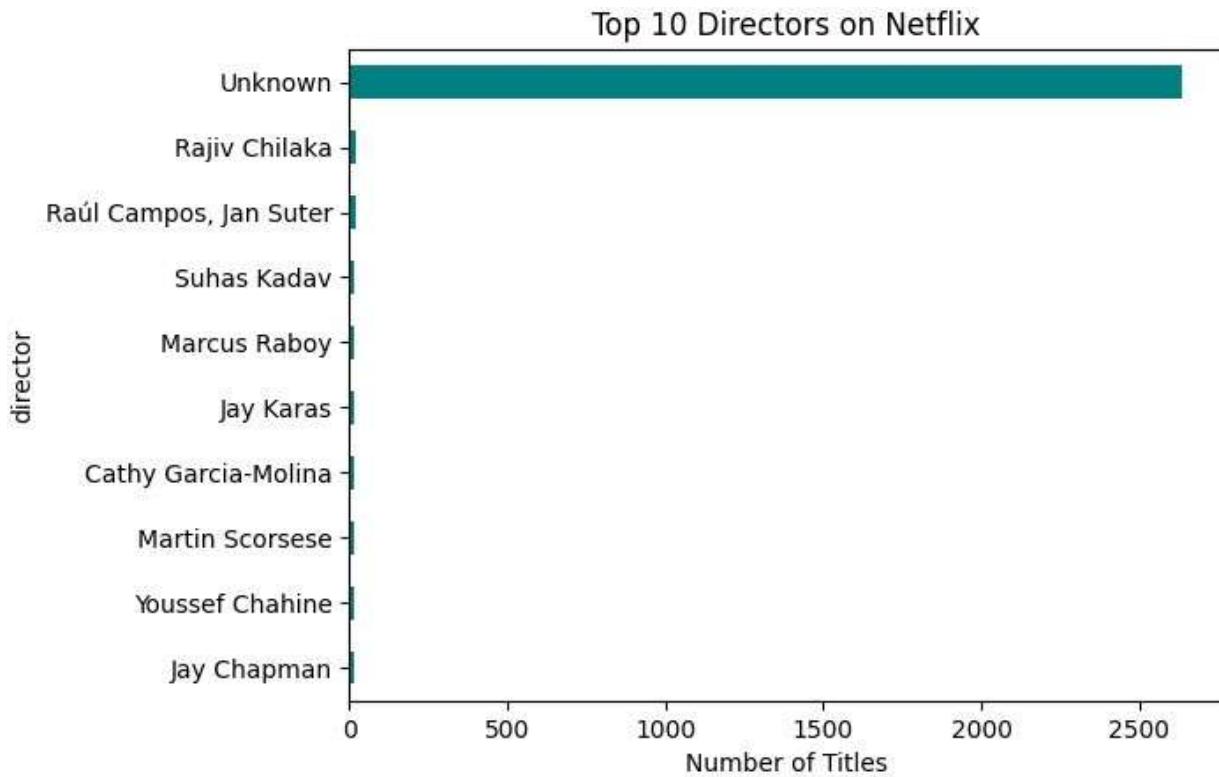
```

#11:-Who are the top 10 directors with the most Netflix content?

top_directors = df['director'].dropna().value_counts().head(10)
top_directors.plot(kind='barh', color='teal')
plt.title('Top 10 Directors on Netflix')

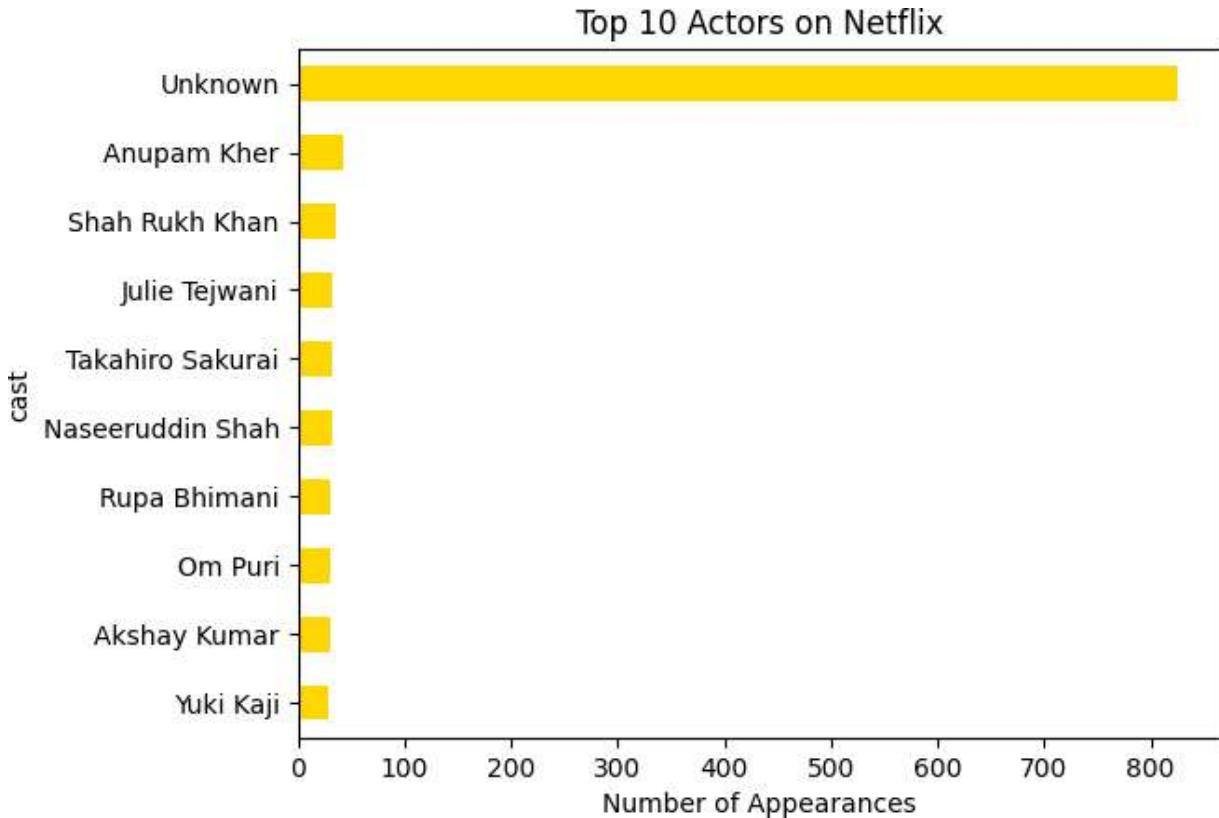
```

```
plt.xlabel('Number of Titles')
plt.gca().invert_yaxis()
plt.show()
```



#12:-Which actors appear most frequently in Netflix shows?

```
actor_series =
df['cast'].dropna().str.split(',').explode().str.strip()
top_actors = actor_series.value_counts().head(10)
top_actors.plot(kind='barh', color='gold')
plt.title('Top 10 Actors on Netflix')
plt.xlabel('Number of Appearances')
plt.gca().invert_yaxis()
plt.show()
```



```
#13:-Which director-genre pairs are most frequent?
```

```
director_genre = df[['director', 'listed_in']].dropna()
director_genre =
director_genre.assign(genres=director_genre['listed_in'].str.split(',')
)).explode('genres')
director_genre['genres'] = director_genre['genres'].str.strip()
pair_counts = director_genre.groupby(['director',
'genres']).size().sort_values(ascending=False).head(10)
print(pair_counts)

director    genres
Unknown    International TV Shows      1223
          TV Dramas                  702
          TV Comedies                539
          Kids' TV                  433
          Crime TV Shows             401
          Romantic TV Shows          341
          Docuseries                 335
          Reality TV                  249
          British TV Shows            228
          Anime Series                 165
dtype: int64
```

```
#14:-How many Titles have unknown directors or cast members?
```

```

unknown_directors = df['director'].isna().sum()
unknown_cast = df['cast'].isna().sum()
print(f"Titles with Unknown Director: {unknown_directors}")
print(f"Titles with Unknown Cast: {unknown_cast}")

Titles with Unknown Director: 0
Titles with Unknown Cast: 0

##15:-What is the average duration of Movies on Netflix?

avg_duration = df[df['type'] == 'Movie']['duration_value'].mean()
print(f"Average Movie Duration: {avg_duration:.2f} minutes")

Average Movie Duration: 99.58 minutes

##16:-What's the most common number for seasons for TV shows?

season_counts = df[df['type'] == 'TV Show']
['duration_value'].value_counts().head(1)
print("Most Common Season Count:\n", season_counts)

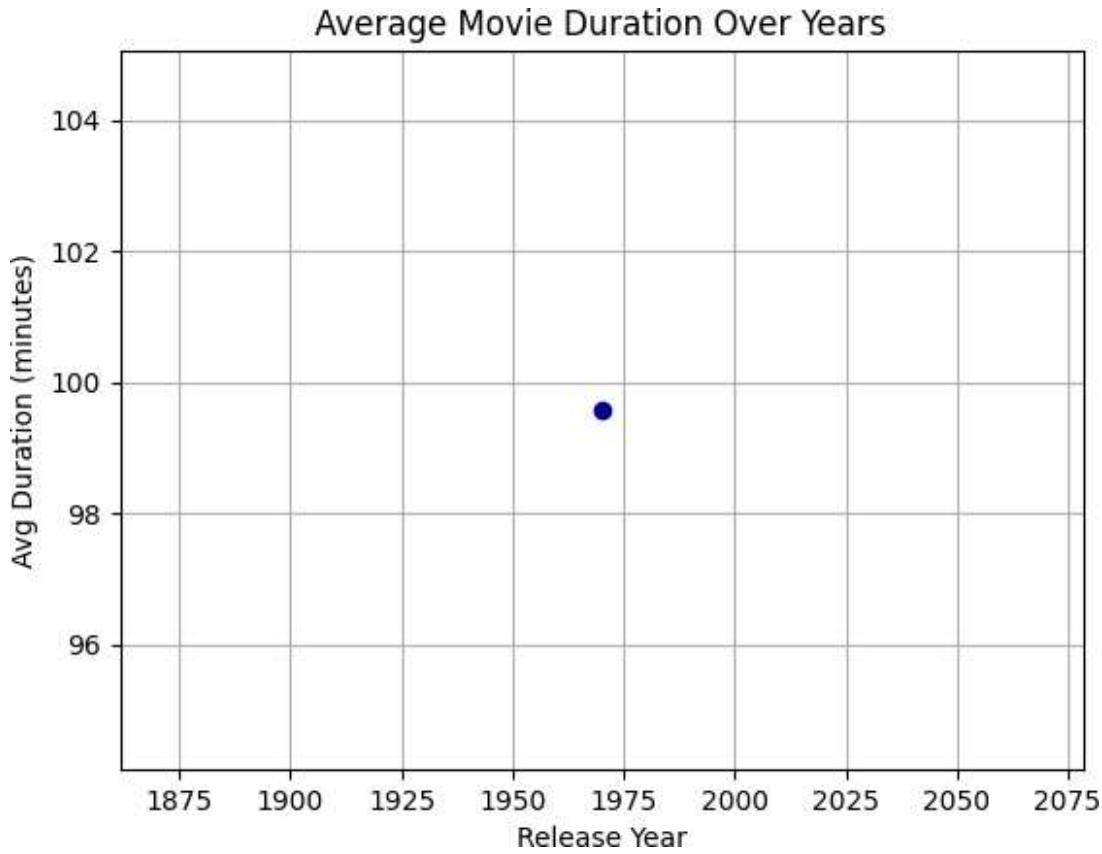
Most Common Season Count:
duration_value
1.0    1793
Name: count, dtype: int64

##17:-Is there a trend in movie durations over the years?

movie_df = df[df['type'] == 'Movie']
duration_trend = movie_df.groupby('release_year')
['duration_value'].mean()

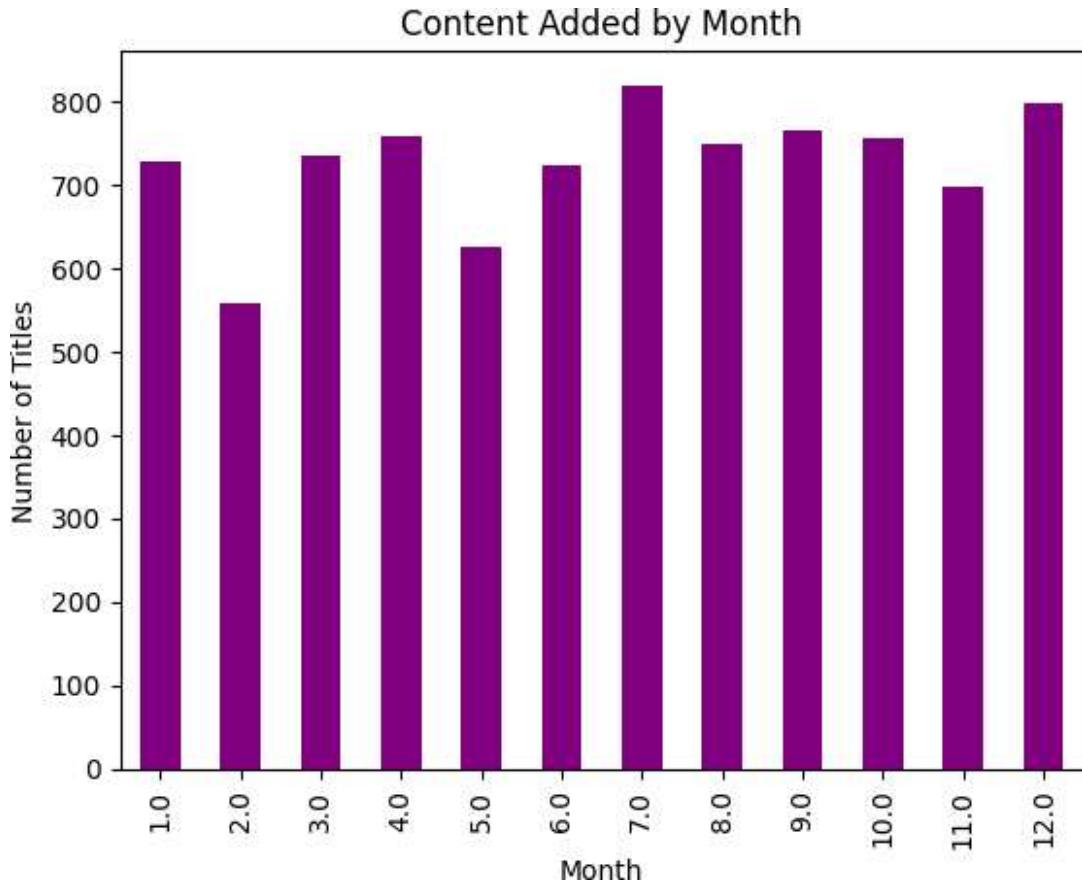
duration_trend.plot(kind='line', marker='o', color='darkblue')
plt.title('Average Movie Duration Over Years')
plt.xlabel('Release Year')
plt.ylabel('Avg Duration (minutes)')
plt.grid(True)
plt.show()

```



```
#18:-In which months does Netflix add the most content?
```

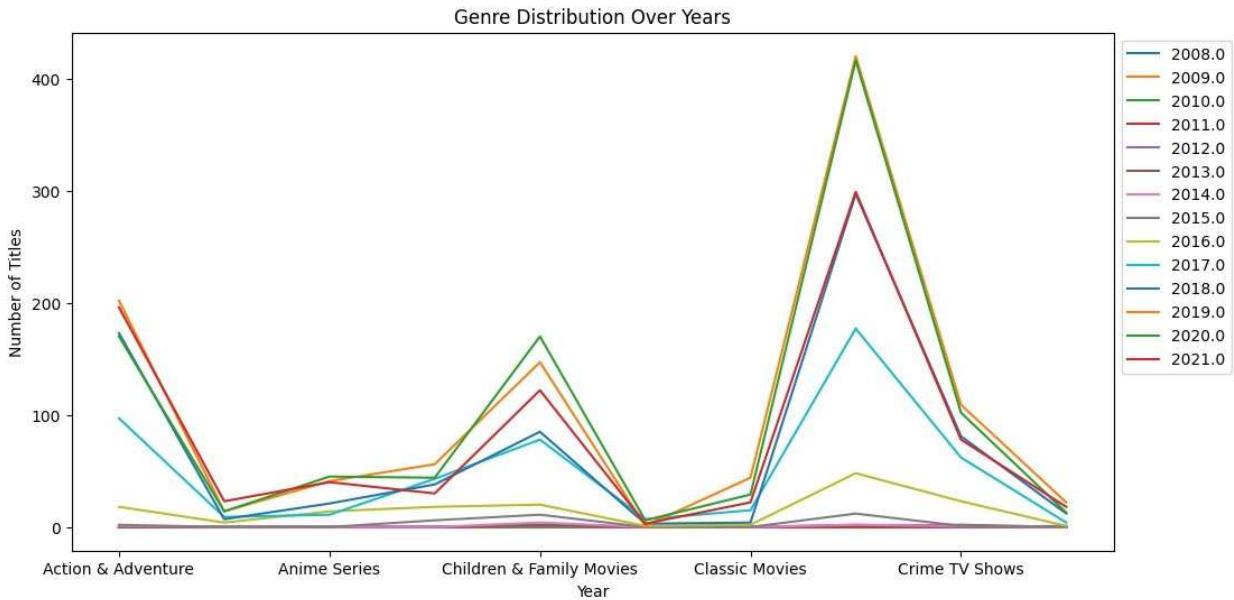
```
month_counts = df['month_added'].value_counts().sort_index()
month_counts.plot(kind='bar', color='purple')
plt.title('Content Added by Month')
plt.xlabel('Month')
plt.ylabel('Number of Titles')
plt.show()
```



```
#19:-How does the genre distribution vary across different years?
```

```
genre_year = df[['year_added', 'listed_in']].dropna()
genre_year =
genre_year.assign(genres=genre_year['listed_in'].str.split(',')).explode('genres')
genre_year['genres'] = genre_year['genres'].str.strip()

genre_trend = genre_year.groupby(['year_added',
'genres']).size().unstack(fill_value=0)
genre_trend.T.head(10).plot(figsize=(12, 6))
plt.title('Genre Distribution Over Years')
plt.ylabel('Number of Titles')
plt.xlabel('Year')
plt.legend(loc='upper left', bbox_to_anchor=(1, 1))
plt.show()
```



#20:-Which countries produce the most content in each genre?

```

country_genre = df[['country', 'listed_in']].dropna()
country_genre =
country_genre.assign(genres=country_genre['listed_in'].str.split(','))
.explode('genres')
country_genre['genres'] = country_genre['genres'].str.strip()
country_genre =
country_genre.assign(country_split=country_genre['country'].str.split(
','))
country_genre = country_genre.explode('country_split')
country_genre['country_split'] =
country_genre['country_split'].str.strip()
matrix = country_genre.groupby(['country_split',
'genres']).size().unstack(fill_value=0)
top_countries =
country_genre['country_split'].value_counts().head(5).index
matrix.loc[top_countries].T.head(10).plot(kind='bar', figsize=(12, 6))
plt.title('Top Genres by Country')
plt.ylabel('Number of Titles')
plt.xlabel('Genre')
plt.legend(title='Country', bbox_to_anchor=(1, 1))
plt.tight_layout()
plt.show()

```

