

Business Case: Netflix - Data Exploration and Visualisation

```
#importing different libraries
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
```

```
#Loading of dataset
```

```
df =
pd.read_csv("https://d2beiqkhq929f0.cloudfront.net/public_assets/asset
s/000/000/940/original/netflix.csv")
```

```
df
```

	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	
...	
8802	s8803	Movie	Zodiac	David Fincher	
8803	s8804	TV Show	Zombie Dumb	NaN	
8804	s8805	Movie	Zombieland	Ruben Fleischer	
8805	s8806	Movie	Zoom	Peter Hewitt	
8806	s8807	Movie	Zubaan	Mozez Singh	

	cast	country
\		
0	NaN	United States
1	Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban...	South Africa
2	Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabi...	NaN
3	NaN	NaN
4	Mayur More, Jitendra Kumar, Ranjan Raj, Alam K...	India
...
8802	Mark Ruffalo, Jake Gyllenhaal, Robert Downey J...	United States
8803	NaN	NaN
8804	Jesse Eisenberg, Woody Harrelson, Emma Stone, ...	United States

8805 Tim Allen, Courteney Cox, Chevy Chase, Kate Ma... United States

8806 Vicky Kaushal, Sarah-Jane Dias, Raaghav Chanan... 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	
...	
8802	November 20, 2019	2007	R	158 min	
8803	July 1, 2019	2018	TV-Y7	2 Seasons	
8804	November 1, 2019	2009	R	88 min	
8805	January 11, 2020	2006	PG	88 min	
8806	March 2, 2019	2015	TV-14	111 min	

	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 ...	
...	...	
8802	Cult Movies, Dramas, Thrillers	
8803	Kids' TV, Korean TV Shows, TV Comedies	
8804	Comedies, Horror Movies	
8805	Children & Family Movies, Comedies	
8806	Dramas, International Movies, Music & Musicals	

	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...
...	...
8802	A political cartoonist, a crime reporter and a...
8803	While living alone in a spooky town, a young g...
8804	Looking to survive in a world taken over by zo...
8805	Dragged from civilian life, a former superhero...
8806	A scrappy but poor boy worms his way into a ty...

[8807 rows x 12 columns]

df.duplicated().sum()

0

#Insights

1. No duplicates in the data set
2. Columns 'Director','Cast' contains missing value or missing data. We need to change missing values to "No Data" so that our data stays accurate and our analysis remains fair
3. Duration of TV shows are shown in seasons and movies in minutes.

Columns information:

Show_id: Unique ID for every Movie / Tv Show

Type: Identifier - A Movie or TV Show

Title: Title of the Movie / Tv Show

Director: Director of the Movie

Cast: Actors involved in the movie/show

Country: Country where the movie/show was produced

Date_added: Date it was added on Netflix

Release_year: Actual Release year of the movie/show

Rating: TV Rating of the movie/show

Duration: Total Duration - in minutes or number of seasons

Listed_in italicized text: Genre

Description: The summary description

```
df.shape
```

```
(8807, 12)
```

This Dataset having 8807 rows and 12 columns.

#Handling missing values

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8807 entries, 0 to 8806
Data columns (total 12 columns):
#   Column          Non-Null Count  Dtype
---  ---
0   show_id         8807 non-null   object
1   type            8807 non-null   object
2   title           8807 non-null   object
```

3	director	6173	non-null	object
4	cast	7982	non-null	object
5	country	7976	non-null	object
6	date_added	8797	non-null	object
7	release_year	8807	non-null	int64
8	rating	8803	non-null	object
9	duration	8804	non-null	object
10	listed_in	8807	non-null	object
11	description	8807	non-null	object

dtypes: int64(1), object(11)
memory usage: 825.8+ KB

It can be seen that the columns 'director', 'cast', 'country' have numerous missing values in which column 'director' have the most number of missing values.

```
# Replacing missing values in the 'director' column with 'No Data'
df['director'].replace(np.nan, 'No Data', inplace=True)

# Replacing missing values in the 'cast' column with 'No Data'
df['cast'].replace(np.nan, 'No Data', inplace=True)
```

For the 'director' and 'cast' columns, replacing missing values with 'No Data' to maintain data integrity and avoid any bias in the analysis.

```
# Filling missing values in the 'country' column with the mode value
df['country'] = df['country'].fillna(df['country'].mode()[0])
```

Using the mode (the value that appears most often) to fill in any missing values in the 'country' column. This helps maintain consistency in the data and reduces the amount of data that is not usable.

```
# Finding the mode rating for movies and TV shows
movie_rating = df.loc[df['type'] == 'Movie', 'rating'].mode()[0]
tv_rating = df.loc[df['type'] == 'TV Show', 'rating'].mode()[0]

# Filling missing rating values based on the type of content
df['rating'] = df.apply(lambda x: movie_rating if x['type'] == 'Movie'
and pd.isna(x['rating'])
else tv_rating if x['type'] == 'TV Show' and
pd.isna(x['rating'])
else x['rating'], axis=1)
```

For the 'rating' column, fill in missing values based on the 'type' of the show. We assign the mode of 'rating' for movies and TV shows separately.

```
# Finding the mode duration for movies and TV shows
movie_duration_mode = df.loc[df['type'] == 'Movie', 'duration'].mode()[0]
```

```

tv_duration_mode = df.loc[df['type'] == 'TV Show', 'duration'].mode()[0]

# Filling missing duration values based on the type of content
df['duration'] = df.apply(lambda x: movie_duration_mode if x['type'] == 'Movie'
                           and pd.isna(x['duration'])
                           else tv_duration_mode if x['type'] == 'TV Show'
                           and pd.isna(x['duration'])
                           else x['duration'], axis=1)

```

For the 'duration' column, fill in missing values based on the 'type' of the show. We assign the mode of 'duration' for movies and TV shows separately.

```

# Dropping rows with missing values
df.dropna(inplace=True)

```

dropping any remaining rows with missing values to ensure a clean dataset for analysis.

```

# Converting the 'date_added' column to datetime format
df["date_added"] = pd.to_datetime(df['date_added'])

```

converting the 'date_added' column to datetime format using pd.to_datetime() to enable further analysis based on date-related attributes.

```

# Extracting month, month name, and year from the 'date_added' column
df['month_added'] = df['date_added'].dt.month
df['month_name_added'] = df['date_added'].dt.month_name()
df['year_added'] = df['date_added'].dt.year

```

#Additional Data Transformations Extracting additional attributes from the 'date_added' column to enhance our analysis capabilities. We remove the month and year values to analyze trends based on these temporal aspects.

```

# Splitting and expanding the 'cast' column
df_cast = df['cast'].str.split(',', expand=True).stack()
df_cast = df_cast.reset_index(level=1, drop=True).to_frame('cast')
df_cast['show_id'] = df['show_id']

# Splitting and expanding the 'country' column
df_country = df['country'].str.split(',', expand=True).stack()
df_country = df_country.reset_index(level=1, drop=True).to_frame('country')
df_country['show_id'] = df['show_id']

# Splitting and expanding the 'listed_in' column
df_listed_in = df['listed_in'].str.split(',', expand=True).stack()

```

```

df_listed_in = df_listed_in.reset_index(level=1,
drop=True).to_frame('listed_in')
df_listed_in['show_id'] = df['show_id']

# Splitting and expanding the 'director' column
df_director = df['director'].str.split(',', expand=True).stack()
df_director = df_director.reset_index(level=1,
drop=True).to_frame('director')
df_director['show_id'] = df['show_id']

```

#Data Transformation: Cast, Country, Listed In, and Director For a better analysis of categorical attributes, we create separate dataframes for them. This makes it easier to explore and analyze them at a more relaxed pace.

#Q1) How has the number of movies released per year changed over the last 20-30 years?

```

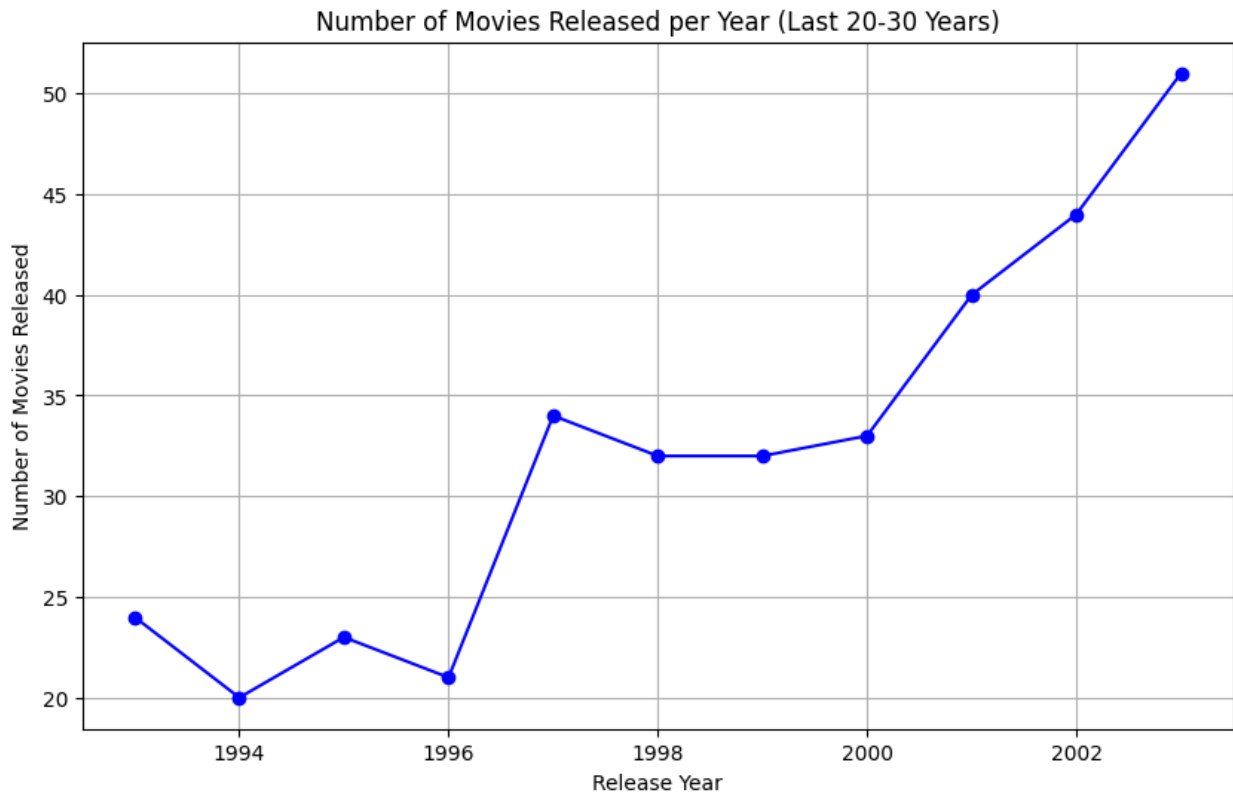
# Filter out the rows containing TV shows
movies_data = df[df['type'] == 'Movie']

# Group the data by release year and count the number of movies in each year
movies_per_year =
movies_data['release_year'].value_counts().sort_index()

# Filter the data for the last 20-30 years
current_year = pd.Timestamp.now().year
years_to_consider = range(current_year - 30, current_year - 20 + 1)
movies_per_year =
movies_per_year[movies_per_year.index.isin(years_to_consider)]

# Plot the number of movies released per year
plt.figure(figsize=(10, 6))
plt.plot(movies_per_year.index, movies_per_year.values, marker='o',
linestyle='-', color='b')
plt.xlabel('Release Year')
plt.ylabel('Number of Movies Released')
plt.title('Number of Movies Released per Year (Last 20-30 Years)')
plt.grid(True)
plt.show()

```



Insight

- Between the years 1993 and 1996, the annual count of movie releases remained within the range of 20 to 25.
- However, a notable spike occurred in 1997, where the number of movie releases surged substantially, ranging from 31 to 34.
- From the year 2000 onward, a consistent and prominent rise in the yearly count of movie releases can be observed.

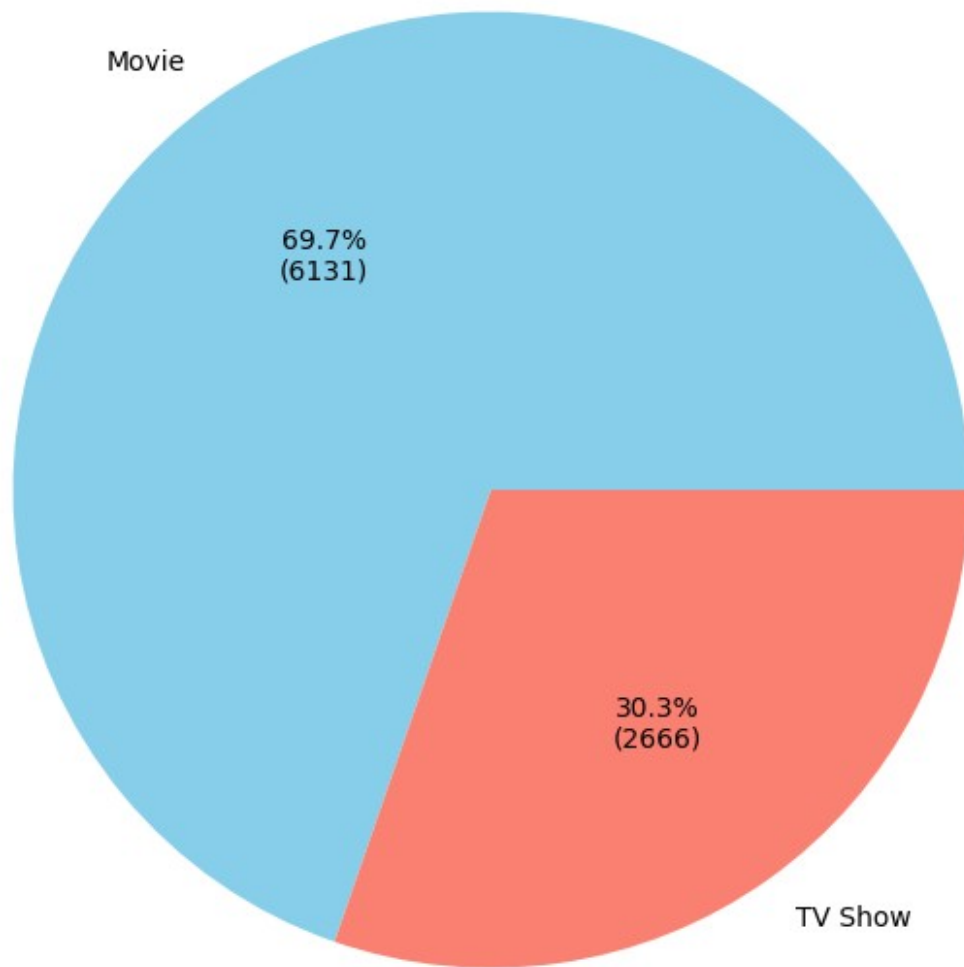
##Q2). Comparison of tv shows vs. movies.

```
# Compare TV shows vs. movies count
show_counts = df['type'].value_counts()

# Calculate the count of TV shows and movies
show_counts = df['type'].value_counts()

# Create a pie chart
plt.figure(figsize=(8, 8))
plt.pie(show_counts, labels=show_counts.index, autopct=lambda p:
'{:.1f}%\n({:.0f})'.format(p, p * sum(show_counts) / 100),
colors=['skyblue', 'salmon'])
plt.title('Comparison of TV Shows vs. Movies on Netflix')
plt.show()
```

Comparison of TV Shows vs. Movies on Netflix



```
# Compare TV shows vs. movies by release year
import matplotlib.pyplot as plt
from matplotlib.patches import Patch

palette = {'TV Show': 'blue', 'Movie': 'orange'}
```



```

plt.figure(figsize=(12, 6))

# Create an "ax" object to work with
ax = sns.histplot(data=df, x='release_year', hue='type',
element='poly', bins=30, palette=palette)

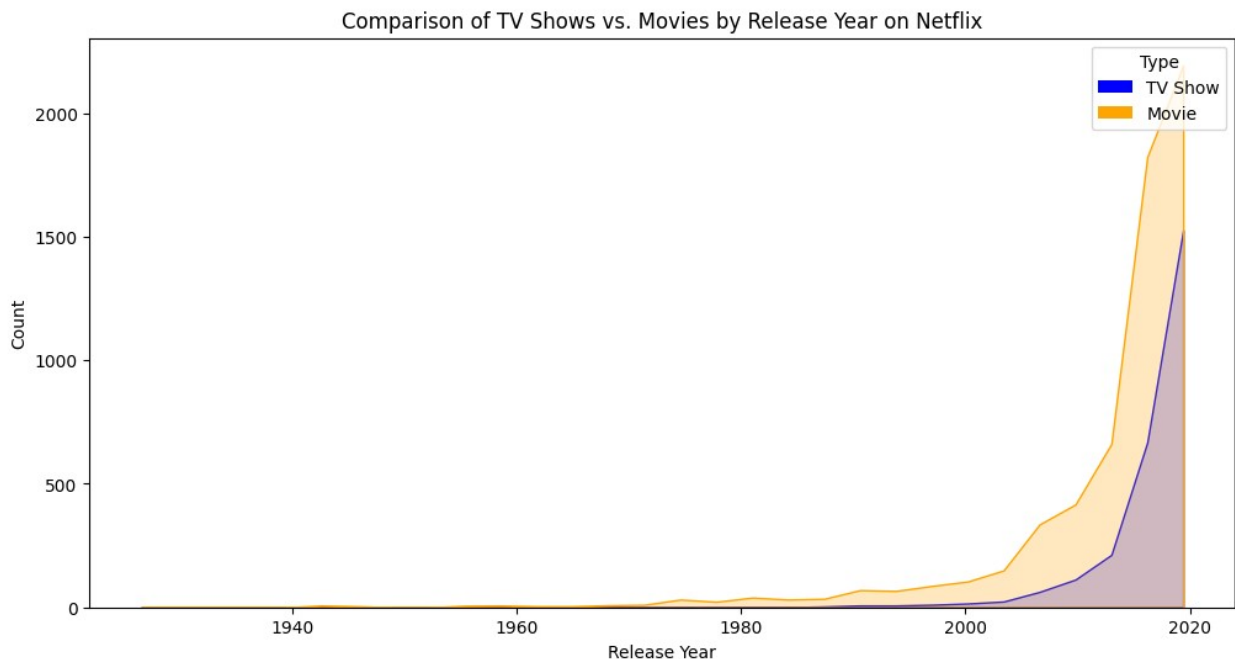
# Set title and labels
plt.title('Comparison of TV Shows vs. Movies by Release Year on Netflix')
plt.xlabel('Release Year')
plt.ylabel('Count')

# Adjust x-axis and y-axis label font size
plt.xticks(fontsize=10)
plt.yticks(fontsize=10)

# Manually add a legend
legend_elements = [Patch(facecolor=palette[label], label=label) for
label in palette]
ax.legend(handles=legend_elements, title='Type', fontsize=10)

# Show the plot
plt.show()

```



```

# Compare TV shows vs. movies by country
top_countries = df['country'].value_counts().head(10)

```

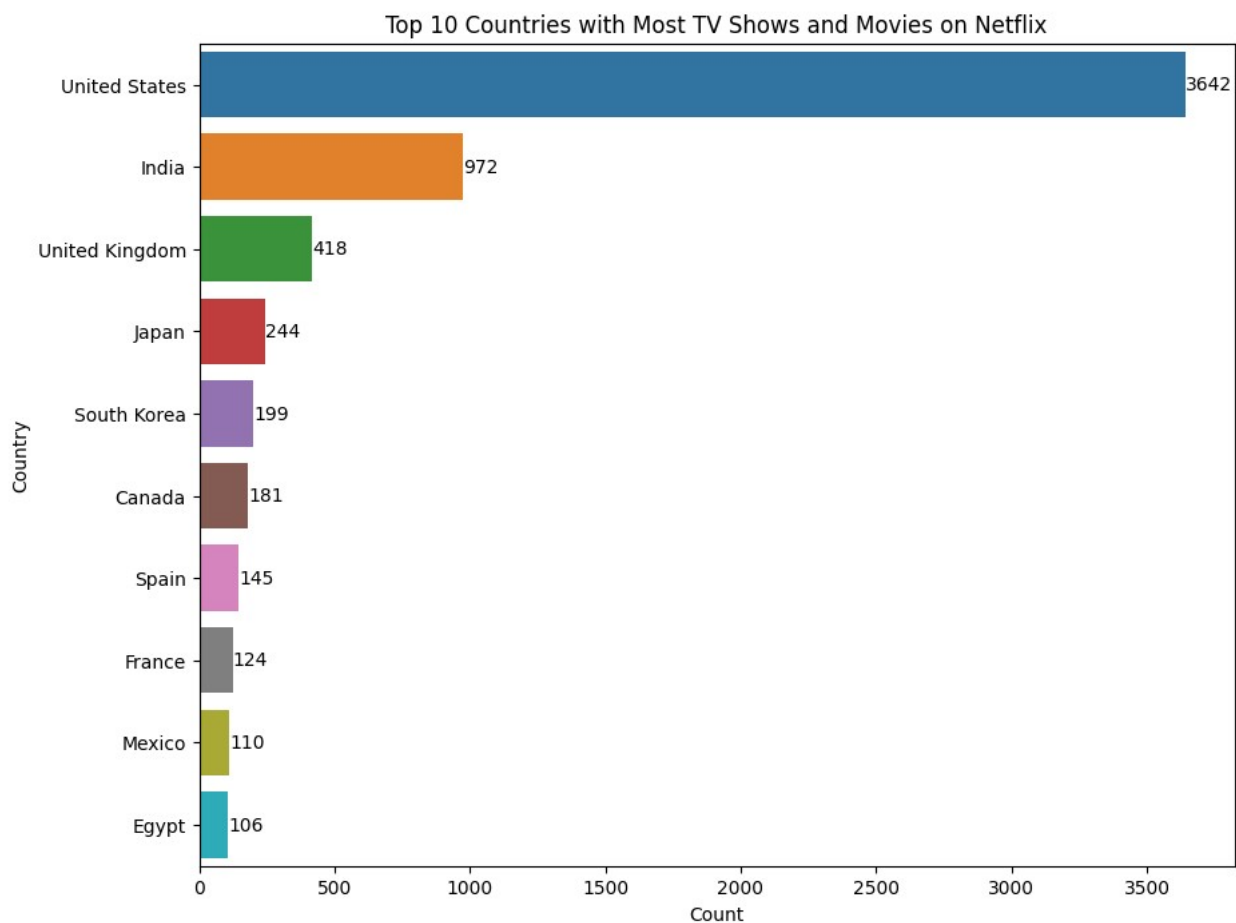
```

plt.figure(figsize=(10, 8))
ax = sns.barplot(x=top_countries.values, y=top_countries.index)
plt.title('Top 10 Countries with Most TV Shows and Movies on Netflix')
plt.xlabel('Count')
plt.ylabel('Country')

# Add count values on the bars
for p in ax.patches:
    ax.annotate(str(int(p.get_width())), (p.get_width(), p.get_y() +
p.get_height() / 2.),
                ha='left', va='center', fontsize=10, color='black')

plt.show()

```



```

# Extract unique genres
unique_genres = df['listed_in'].str.split(', ').explode().unique()

# Separate the data for TV shows and movies
tv_shows = df[df['type'] == 'TV Show']
movies = df[df['type'] == 'Movie']

```

```

# Calculate genre counts for TV shows
tv_show_genre_counts = tv_shows['listed_in'].str.split(',').explode().value_counts()

# Calculate genre counts for movies
movie_genre_counts = movies['listed_in'].str.split(',').explode().value_counts()

# Calculate total counts for percentage calculation
total_tv_shows = len(tv_shows)
total_movies = len(movies)

# Set a threshold for minimum bin percentage to show labels
threshold_percentage = 2.0 # You can adjust this threshold value as needed

# Calculate percentages for TV show genres
tv_show_percentages = (tv_show_genre_counts / total_tv_shows) * 100

# Calculate percentages for movie genres
movie_percentages = (movie_genre_counts / total_movies) * 100

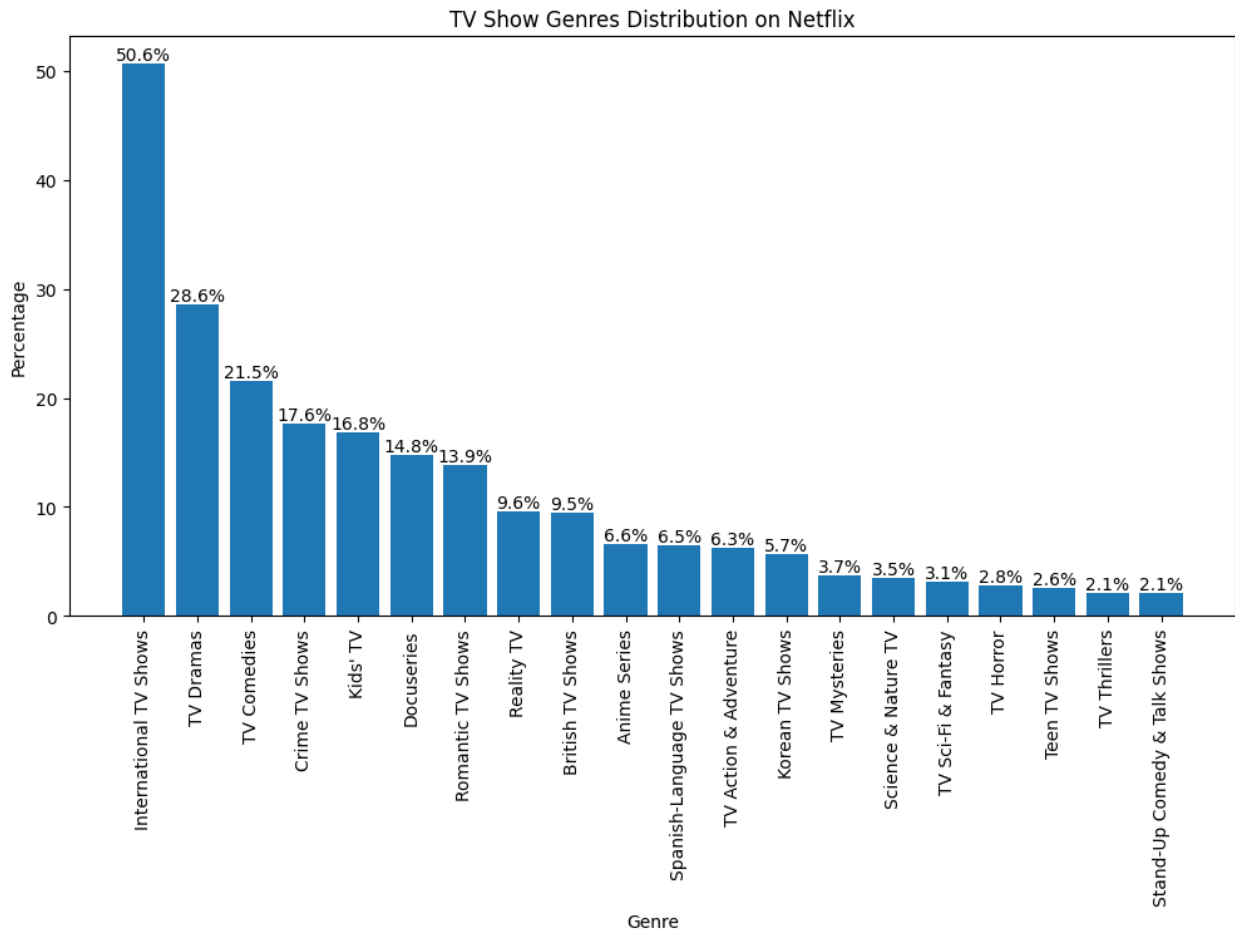
# Filter small bins and their labels based on threshold percentage
tv_show_filtered = tv_show_percentages[tv_show_percentages >= threshold_percentage]
movie_filtered = movie_percentages[movie_percentages >= threshold_percentage]

# Plot histogram for TV show genres
plt.figure(figsize=(12, 6))
plt.bar(tv_show_filtered.index, tv_show_filtered.values)
plt.title('TV Show Genres Distribution on Netflix')
plt.xlabel('Genre')
plt.ylabel('Percentage')

# Annotate plot with labels for filtered bins
for genre, percentage in tv_show_filtered.items():
    plt.annotate(f'{percentage:.1f}%', (genre, percentage),
ha='center', va='bottom')

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

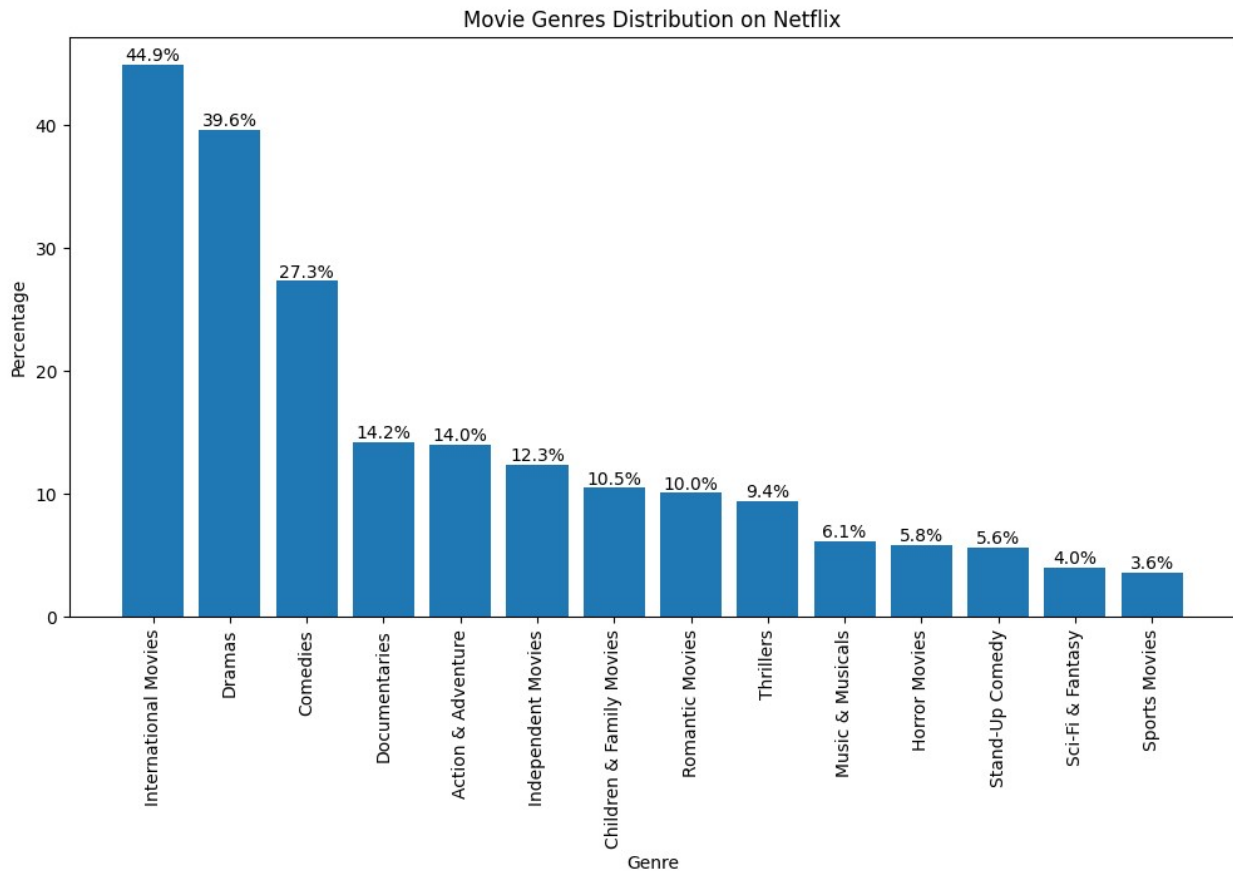
```



```
# Plot histogram for movie genres
plt.figure(figsize=(12, 6))
plt.bar(movie_filtered.index, movie_filtered.values)
plt.title('Movie Genres Distribution on Netflix')
plt.xlabel('Genre')
plt.ylabel('Percentage')

# Annotate plot with labels for filtered bins
for genre, percentage in movie_filtered.items():
    plt.annotate(f'{percentage:.1f}%', (genre, percentage),
                ha='center', va='bottom')

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



Insights

- Netflix offers a diverse collection of content. In the total of 8797 contents 69.7% (6131) are movies and 30.3%(2666) are TV shows.
- Comparing the two, the number of available Movies significantly surpasses that of TV Shows on Netflix.
- Starting from 2010, there has been a remarkable surge in the release of both Movies and TV Shows on Netflix, indicating the platform's soaring popularity.
- The United States stands out as the leading country of origin for Netflix's TV Shows and Movies. India and the United Kingdom follow closely in the second and third spots.
- When analyzing genres, a striking resemblance emerges between TV Shows and Movies. Both categories predominantly feature genres such as International, Dramas, and Comedies. Notably, 50% of TV Shows and 44% of Movies fall under the International genre, underscoring Netflix's global viewership.

Q3). What is the best time to launch a TV show?

```
# Convert Date_added to datetime format
df['date_added'] = pd.to_datetime(df['date_added'])

# Extract release month from date_added
df['release_month'] = df['date_added'].dt.month
```

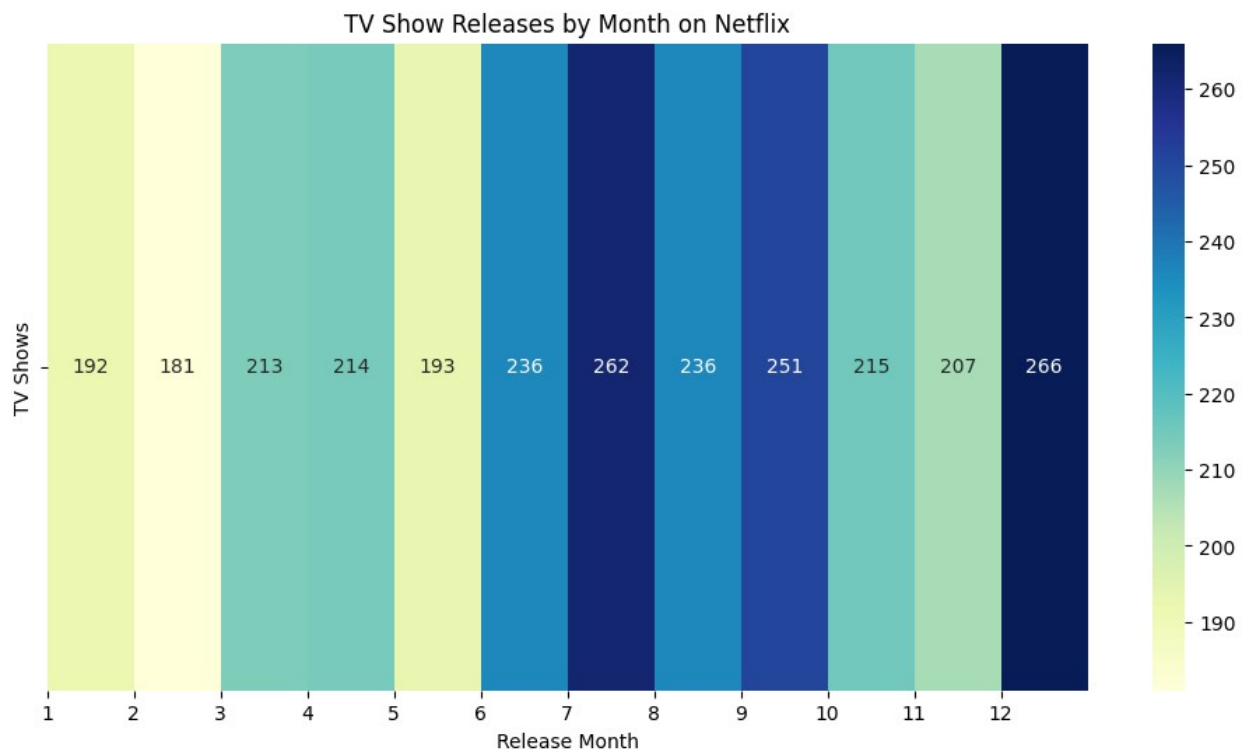
```

# Filter TV Shows
tv_shows = df[df['type'] == 'TV Show']

# Count TV Shows by release month
release_month_counts =
tv_shows['release_month'].value_counts().sort_index()

# Create a heatmap
plt.figure(figsize=(12, 6))
sns.heatmap(release_month_counts.values.reshape(1, -1), cmap='YlGnBu',
annot=True, fmt='d', yticklabels=['TV Shows'])
plt.title('TV Show Releases by Month on Netflix')
plt.xlabel('Release Month')
plt.ylabel('')
plt.xticks(range(12), labels=[str(i + 1) for i in range(12)]) #
Adjust month labels
plt.show()

```



Insight

- The optimal time to launch a TV show is during the month of December, coinciding with the global holiday season. This period allows for a larger potential audience due to annual vacations around the world.
- Additionally, it's worth noting that a significant number of TV shows are released between June and September, inclusive. This trend can be attributed to the summer

vacation period in the United States, a major source of origin for many Netflix movies and TV shows. If the primary target audience for the TV show is predominantly in the U.S., the release window of June to September could be strategically chosen to align with the American summer vacation. However, if the intended audience is more diverse or global, December emerges as an optimal month for the TV show release due to the broader appeal of holiday-related leisure time.

Q4. Analysis of actors/directors of different types of shows/movies.

```
# Function to analyze actors/directors
def analyze_actors_directors(data, column, show_type):
    # Filter data by show type
    filtered_data = data[data['type'] == show_type]

    # Exclude rows with 'No Data' in the specified column
    filtered_data = filtered_data[filtered_data[column] != 'No Data']

    # Split the column values by comma and create a list
    split_values = filtered_data[column].str.split(',
').dropna().tolist()

    # Flatten the list of lists
    flattened_values = [item for sublist in split_values for item in
sublist]

    # Create a frequency count of actors/directors
    freq_count = pd.Series(flattened_values).value_counts()

    return freq_count

# Analysis of actors for TV Shows
tv_show_actors = analyze_actors_directors(df, 'cast', 'TV Show')
tv_show_actors = tv_show_actors.head(10)

# Analysis of actors for Movies
movie_actors = analyze_actors_directors(df, 'cast', 'Movie')
movie_actors = movie_actors.head(10)

# Analysis of directors for TV Shows
tv_show_directors = analyze_actors_directors(df, 'director', 'TV
Show')
tv_show_directors = tv_show_directors.head(10)

# Analysis of directors for Movies
movie_directors = analyze_actors_directors(df, 'director', 'Movie')
movie_directors = movie_directors.head(10)
```

```

# Create bar charts
plt.figure(figsize=(12, 8))

plt.subplot(2, 2, 1)
tv_show_actors.plot(kind='bar', color='blue')
plt.title('Actor appeared most in TV Shows')
plt.xlabel('Actor')
plt.ylabel('Count')
for i, v in enumerate(tv_show_actors):
    plt.text(i, v + 1, str(v), ha='center', va='bottom', fontsize=10)

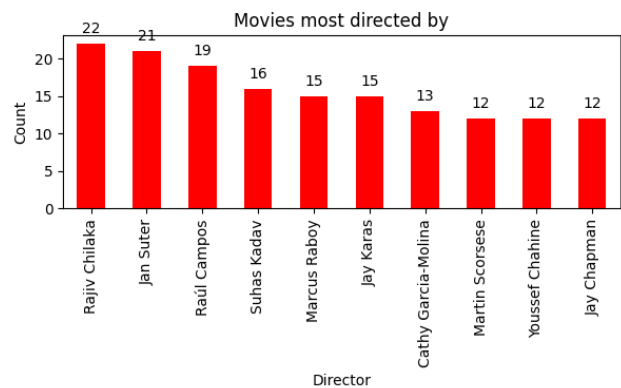
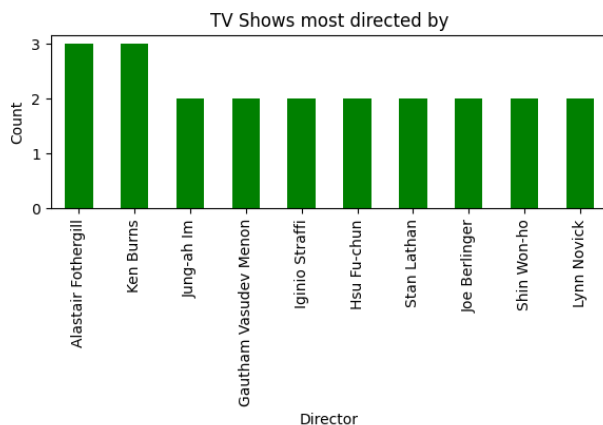
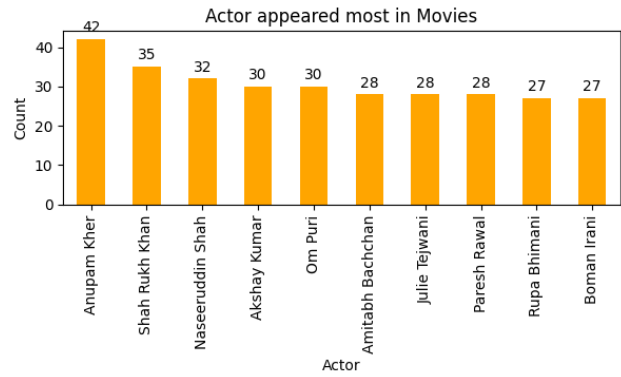
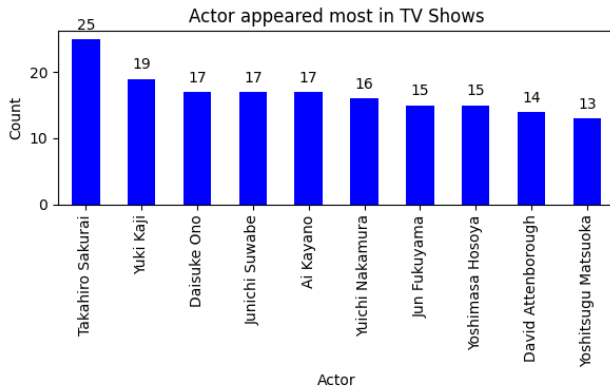
plt.subplot(2, 2, 2)
movie_actors.plot(kind='bar', color='orange')
plt.title('Actor appeared most in Movies')
plt.xlabel('Actor')
plt.ylabel('Count')
for i, v in enumerate(movie_actors):
    plt.text(i, v + 1, str(v), ha='center', va='bottom', fontsize=10)

plt.subplot(2, 2, 3)
tv_show_directors.plot(kind='bar', color='green')
plt.title('TV Shows most directed by')
plt.xlabel('Director')
plt.ylabel('Count')
#for i, v in enumerate(tv_show_directors):
#    plt.text(i, v + 1, str(v), ha='center', va='bottom', fontsize=10)

plt.subplot(2, 2, 4)
movie_directors.plot(kind='bar', color='red')
plt.title('Movies most directed by')
plt.xlabel('Director')
plt.ylabel('Count')
for i, v in enumerate(movie_directors):
    plt.text(i, v + 1, str(v), ha='center', va='bottom', fontsize=10)

plt.tight_layout()
plt.show()

```

Insight

- Among the top Actors, Takahiro Sakurai has appeared 3 TV shows, and Anupam Kher has appeared 22 film
- Among the top directors, Alastair Fothergill has directed 3 TV shows, and Rajiv Chilaka has directed 22 films

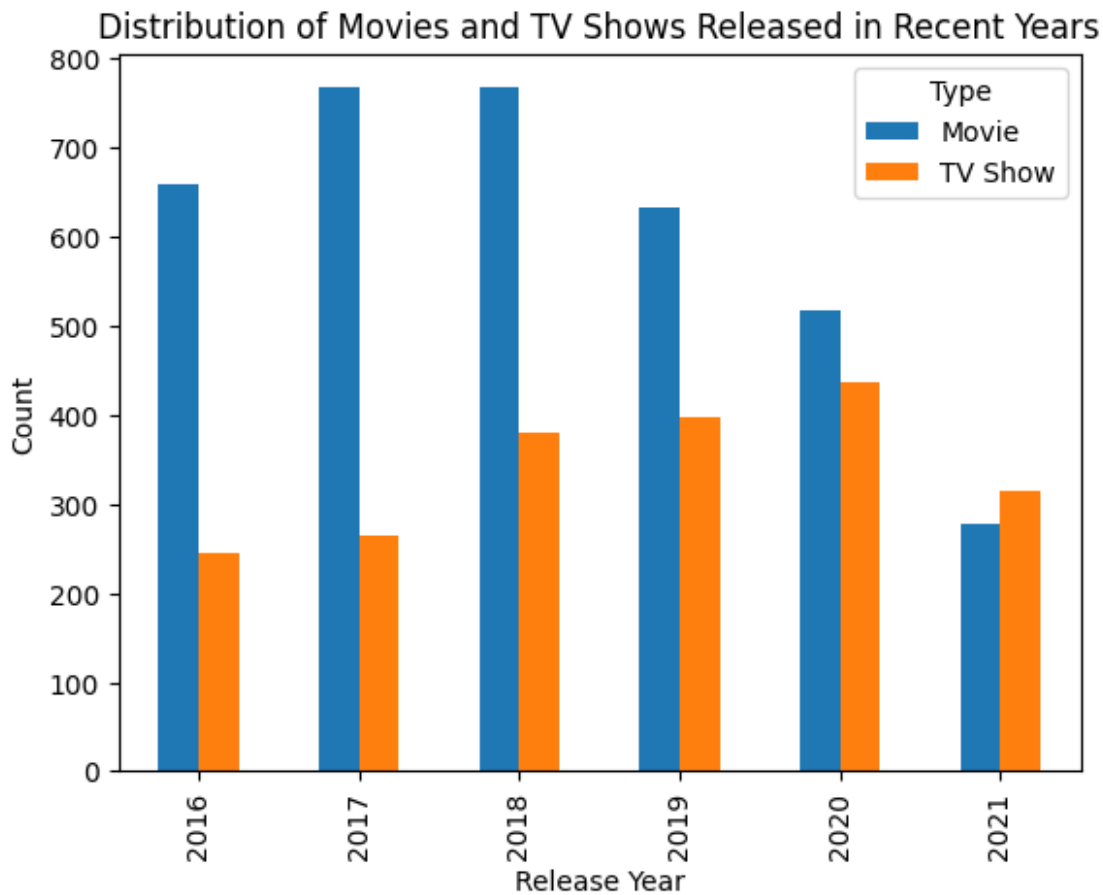
##Q5). Does Netflix has more focus on TV Shows than movies in recent years?

```
recent_years = range(2016,2024) # Change the range as needed
recent_df = df[df['release_year'].isin(recent_years)]

# Count movies and TV shows by release year and type
counts = recent_df.groupby(['release_year', 'type'])
['show_id'].count().unstack()

# Plot dual bar charts
plt.figure(figsize=(10, 6))
counts.plot(kind='bar', stacked=False)
plt.title('Distribution of Movies and TV Shows Released in Recent Years')
plt.xlabel('Release Year')
plt.ylabel('Count')
plt.legend(title='Type')
plt.show()
```

<Figure size 1000x600 with 0 Axes>



Insight

- It has been observed that in recent years, the number of TV shows being released on Netflix has been gradually increasing, while the number of movies being released has shown a gradual decrease. In the year 2021, for the first time in the history of Netflix, the total count of TV shows exceeded the total count of movies released in that year. This indicates a shift in focus for Netflix towards producing more TV shows than movies.

Q6). Understanding what content is available in different countries

```
country = df["country"].apply(lambda x: str(x).split(", ")).tolist()
#exploding the country column
df_country = pd.DataFrame(country, index = df["title"])
df_country = df_country.stack()
df_country = df_country.reset_index()
df_country.drop(columns = "level_1" , inplace = True)
df_country.columns = ["title" , "country"]
```

```

listed_in = df["listed_in"].apply(lambda x: str(x).split(",
")).tolist()
df_genre = pd.DataFrame(listed_in, index = df["title"])

df_trend_country = df.merge(df_country , on = "title")
df_trend_country.drop(columns = "country_x" , inplace = True)
df_trend_country.rename(columns = {"country_y":"country"}, inplace =
True)

temp = df_trend_country['country'].value_counts()[11].reset_index()
temp.rename(columns = {'index':'country', 'country':'count'},
inplace=True)
country_list = temp['country'].tolist()
df_top10country =
df_trend_country.loc[df_trend_country['country'].isin(country_list)]
df_top10country = df_top10country.loc[df_top10country["country"]!
="Unknown"] #dropping of rows whose value is unknown.

genre_country_df= df_trend_country.merge(df_genre , on= "title")
genre_country_df.head(5)

```

	show_id	type	title	director \
0	s1	Movie	Dick Johnson Is Dead	Kirsten Johnson
1	s2	TV Show	Blood & Water	No Data
2	s3	TV Show	Ganglands	Julien Leclercq
3	s4	TV Show	Jailbirds New Orleans	No Data
4	s5	TV Show	Kota Factory	No Data

		cast
date_added \		
0		No Data September 25, 2021
1	Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban...	September 24, 2021
2	Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabi...	September 24, 2021
3		No Data September 24, 2021
4	Mayur More, Jitendra Kumar, Ranjan Raj, Alam K...	September 24, 2021

	release_year	rating	duration \
0	2020	PG-13	90 min
1	2021	TV-MA	2 Seasons
2	2021	TV-MA	1 Season
3	2021	TV-MA	1 Season
4	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	country	\
0	As her father nears the end of his life, filmm...	United States	
1	After crossing paths at a party, a Cape Town t...	South Africa	
2	To protect his family from a powerful drug lor...	United States	
3	Feuds, flirtations and toilet talk go down amo...	United States	
4	In a city of coaching centers known to train I...	India	

	0	1
2		

0	Documentaries	None
None		

None

1	International TV Shows	TV Dramas	TV Mysteries
2	Crime TV Shows	International TV Shows	TV Action & Adventure

3	Docuseries	Reality TV
None		

4	International TV Shows	Romantic TV Shows
Comedies		

4	International TV Shows	Romantic TV Shows	TV Comedies

```

temp_genre = genre_country_df['listed_in'].value_counts()
[:10].reset_index()
temp_genre.rename(columns={'index': 'genre', 'listed_in': 'count'},
inplace=True)
genre_list = temp_genre['genre'].tolist()
df_top10_genre =
genre_country_df.loc[genre_country_df['listed_in'].isin(genre_list)]
df_top10_genre.head()

```

	show_id	type	title	\
0	s1	Movie	Dick Johnson Is Dead	
6	s7	Movie	My Little Pony: A New Generation	
7	s8	Movie	Sankofa	
8	s8	Movie	Sankofa	
9	s8	Movie	Sankofa	

	director	\
0	Kirsten Johnson	
6	Robert Cullen, José Luis Ucha	
7	Haile Gerima	
8	Haile Gerima	
9	Haile Gerima	

	cast
date_added	\

0		No Data	September 25, 2021
6	Vanessa Hudgens, Kimiko Glenn, James Marsden, ...		September 24, 2021
7	Kofi Ghanaba, Oyafunmike Ogunlano, Alexandra D...		September 24, 2021
8	Kofi Ghanaba, Oyafunmike Ogunlano, Alexandra D...		September 24, 2021
9	Kofi Ghanaba, Oyafunmike Ogunlano, Alexandra D...		September 24, 2021

	release_year	rating	duration	\
0	2020	PG-13	90 min	
6	2021	PG	91 min	
7	1993	TV-MA	125 min	
8	1993	TV-MA	125 min	
9	1993	TV-MA	125 min	

	listed_in	\
0	Documentaries	
6	Children & Family Movies	
7	Dramas, Independent Movies, International Movies	
8	Dramas, Independent Movies, International Movies	
9	Dramas, Independent Movies, International Movies	

	description	country	\
0	As her father nears the end of his life, filmm...	United States	
6	Equestria's divided. But a bright-eyed hero be...	United States	
7	On a photo shoot in Ghana, an American model s...	United States	
8	On a photo shoot in Ghana, an American model s...	Ghana	
9	On a photo shoot in Ghana, an American model s...	Burkina Faso	

	0	1	2
0	Documentaries	None	None
6	Children & Family Movies	None	None
7	Dramas	Independent Movies	International Movies
8	Dramas	Independent Movies	International Movies
9	Dramas	Independent Movies	International Movies

```
import seaborn as sns
import matplotlib.pyplot as plt
```

```
plt.figure(figsize=(12, 8))
sns.heatmap(heat_genre_final, annot=True, cmap="Blues", fmt=".0f")
```

```
plt.title("Top 10 Genre Distribution across 10 Different Countries")
plt.show()
```



Insight

- Most of the contents are available in US
- In India dramas, comedies are more popular.

```
# Extracting unique genres from the 'listed_in' column
genres = df['listed_in'].str.split(',', expand=True).stack().unique()

# Create a new DataFrame to store the genre data
genre_data = pd.DataFrame(index=genres, columns=genres, dtype=float)

# Fill the genre data DataFrame with zeros
genre_data.fillna(0, inplace=True)

# Iterate over each row in the original DataFrame and update the genre data DataFrame
for _, row in df.iterrows():
    listed_in = row['listed_in'].split(',')
    for genre1 in listed_in:
        for genre2 in listed_in:
            genre_data.at[genre1, genre2] += 1

# Create a correlation matrix using the genre data
```


Insight

- The heatmap showcases the correlation between different genres. Analyzing the heatmap reveals pronounced positive correlations between specific genres, such as TV Dramas and International TV Shows, as well as Romantic TV Shows and International TV Shows.

```
df_movies = df[df['type'] == 'Movie']
df_tv_shows = df[df['type'] == 'TV Show']
# Extract the movie lengths and TV show episodes from the 'duration'
column
movie_lengths = df_movies['duration'].str.extract('(\d+)',
expand=False).astype(int)
tv_show_episodes = df_tv_shows['duration'].str.extract('(\d+)',
expand=False).astype(int)

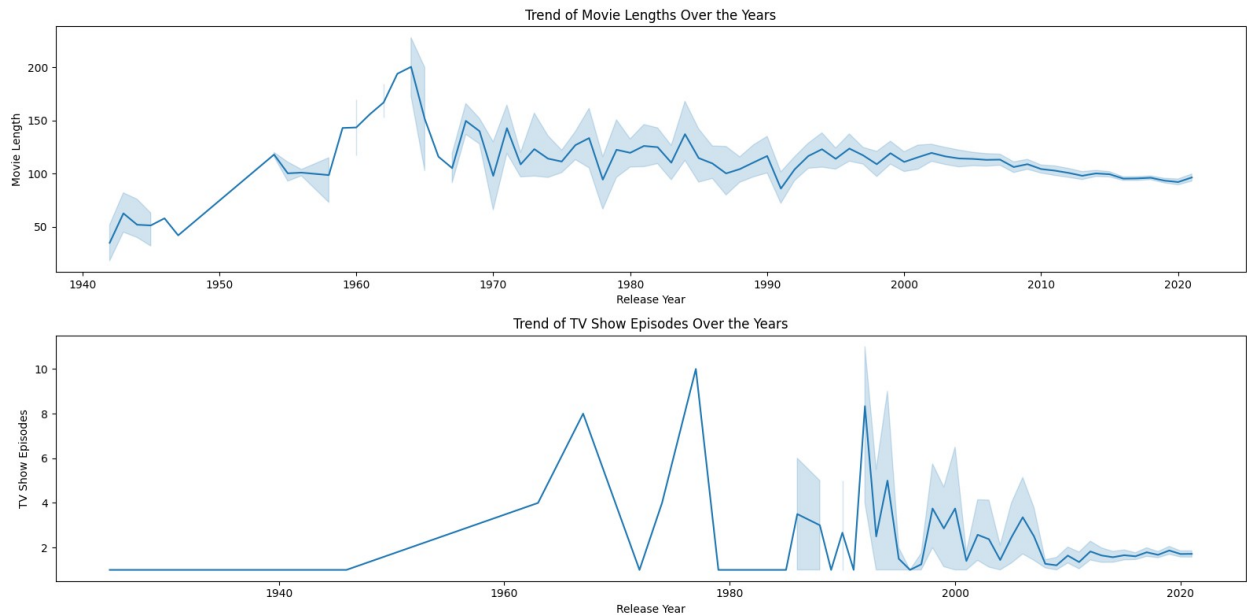
# Create line plots for movie lengths and TV show episodes
plt.figure(figsize=(16, 8))

plt.subplot(2, 1, 1)
sns.lineplot(data=df_movies, x='release_year', y=movie_lengths)
plt.xlabel('Release Year')
plt.ylabel('Movie Length')
plt.title('Trend of Movie Lengths Over the Years')

plt.subplot(2, 1, 2)
sns.lineplot(data=df_tv_shows, x='release_year', y=tv_show_episodes)
plt.xlabel('Release Year')
plt.ylabel('TV Show Episodes')
plt.title('Trend of TV Show Episodes Over the Years')

# Adjust the layout and spacing
plt.tight_layout()

# Show the plots
plt.show()
```

Insight

- In recent years Netflix is preferring TV Shows with less than Two seasons.
- Since 2000's the duration of movie is around 100 - 120 minutes.

Conclusion

- Quantity : Netflix had more movies than TV Shows in thier Library.
- Between 1993 and 1996, movie releases numbered 20 to 25 annually. In 1997, a notable surge brought releases to 31 to 34. Post-2000, a consistent and strong growth in yearly movie counts occurred.
- Netflix boasts a diverse content collection with 8797 titles, comprising 69.7% movies and 30.3% TV shows. Movies outnumber TV shows. Post-2010, a significant surge in releases reflects Netflix's rising popularity. The US leads top countries, followed by India and the UK. Genre analysis shows both TV shows and movies focus on International, Dramas, and Comedies, highlighting Netflix's global viewership.
- The ideal TV show launch time is December, aligning with the worldwide holiday season and increased vacationing audience. Notably, a significant TV show influx occurs from June to September, driven by the U.S. summer break, suitable if the target audience is mainly American. For broader or global viewership, December's broader holiday allure remains strategically optimal.
- Recent trends reveal a rise in Netflix TV show releases and a decline in movies. In 2021, a historic shift occurred as TV show launches surpassed movies for the first time, highlighting Netflix's evolving emphasis on TV content production.
- The majority of content is accessible in the US. In India, dramas and comedies hold higher popularity levels.
- Anupam Kher is one of the Top Actor and Rajiv Chilaka is the top director in Netflix Contents. This signifies Netflix's targeted approach towards the Indian audience.

Recommendation

- Based on the analysis of these points, it is recommended that Netflix continues to diversify its content library, with a strategic focus on increasing its TV show offerings. The historical trend of rising TV show releases and the milestone in 2021 where TV shows surpassed movies underline the growing demand for TV content. To cater to their global viewership, Netflix should maintain a balance between genres like International, Dramas, and Comedies, which resonate with a wide audience.
- Considering the popularity of December releases aligning with the holiday season, Netflix should strategically plan the launch of their original TV shows during this period to capture a larger audience during peak leisure time. Additionally, for the American audience, capitalizing on the surge in TV show releases from June to September, corresponding with the US summer vacation, could be advantageous.
- As India exhibits a preference for dramas and comedies, Netflix should continue curating and producing content in these genres to cater to the preferences of the Indian audience. However, the company should also be mindful of the potential for diversification in India's content preferences over time.
- Overall, maintaining a balanced content mix, strategic release timing, and adapting content strategies to specific regional preferences will help Netflix continue its growth and global dominance in the streaming industry.
- Acknowledging Anupam Kher's stature as a top actor and Rajiv Chilaka's prominence as a leading director within Netflix's content offerings, it's astute to capitalize on their influence for strategic collaborations. Incorporating acclaimed Indian artists and directors could effectively amplify Netflix's appeal to the Indian audience, fostering greater engagement and viewership. By curating exclusive content that resonates with Indian sensibilities and culture, the platform can significantly bolster its presence in the Indian market. This approach aligns with Netflix's aim to tap into the vast potential of the Indian audience and ultimately drive substantial business growth.

