

netflix-bussiness-case

May 27, 2024

```
[88]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
```

```
[89]: from google.colab import drive
drive.mount('/content/drive')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

```
[90]: df=pd.read_csv('/content/netflix-case study.csv')
```

```
[91]: df.head(5)
```

```
[91]: 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...      NaN
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...

```

```
[92]: df.shape
```

```
[92]: (8807, 12)
```

```
[93]: df.isnull().sum()
```

```

[93]: show_id      0
      type        0
      title       0
      director    2634
      cast        825
      country     831
      date_added   10
      release_year  0
      rating       4
      duration     3
      listed_in    0
      description  0
      dtype: int64

```

Splitting the directors by title

```
[94]: df_director_r=pd.DataFrame(df['director'].apply(lambda x: str(x).split(',')).
      ↪tolist(),index=df['title'])
```

```

[95]: df_director=df_director_r.stack().reset_index()
      df_director.drop('level_1',axis=1,inplace=True)
      df_director.rename(columns = {0:'director'},inplace= True)
      df_director.head(5)

```

```

[95]:           title      director
0  Dick Johnson Is Dead  Kirsten Johnson
1      Blood & Water           nan
2      Ganglands      Julien Leclercq
3  Jailbirds New Orleans           nan

```

Splitting cast by title

```
[96]: df_cast_r=pd.DataFrame(df['cast'].apply(lambda x: str(x).split(',')).
      ↪tolist(),index=df['title'])
```

```
[97]: df_cast=df_cast_r.stack().reset_index()
df_cast.drop('level_1',axis=1,inplace=True)
df_cast.rename(columns = {0:'cast'},inplace= True)
df_cast.head(5)
```

```
[97]:
```

	title	cast
0	Dick Johnson Is Dead	nan
1	Blood & Water	Ama Qamata
2	Blood & Water	Khosi Ngema
3	Blood & Water	Gail Mabalane
4	Blood & Water	Thabang Molaba

Splitting country by title

```
[98]: df_country_r=pd.DataFrame(df['country'].apply(lambda x: str(x).split(',')).
      ↪tolist(),index=df['title'])
df_country=df_country_r.stack().reset_index()
df_country.drop('level_1',axis=1,inplace= True)
df_country.rename(columns={0:'country'},inplace=True)
df_country.head(5)
```

```
[98]:
```

	title	country
0	Dick Johnson Is Dead	United States
1	Blood & Water	South Africa
2	Ganglands	nan
3	Jailbirds New Orleans	nan
4	Kota Factory	India

Splitting listed_in by title

```
[99]: df_listed_in_r=pd.DataFrame(df['listed_in'].apply(lambda x: str(x).split(',')).
      ↪tolist(),index=df['title'])
df_listed_in=df_listed_in_r.stack().reset_index()
df_listed_in.drop('level_1',axis=1,inplace=True)
df_listed_in.rename(columns={0:'listed_in'},inplace=True)
df_listed_in.head(5)
```

```
[99]:
```

	title	listed_in
0	Dick Johnson Is Dead	Documentaries
1	Blood & Water	International TV Shows
2	Blood & Water	TV Dramas

3	Blood & Water	TV Mysteries
4	Ganglands	Crime TV Shows

```
[100]: df.columns
```

```
[100]: Index(['show_id', 'type', 'title', 'director', 'cast', 'country', 'date_added',
           'release_year', 'rating', 'duration', 'listed_in', 'description'],
          dtype='object')
```

Merging the columns

```
[101]: df_new1=df_director.merge(df_cast,how='inner',on = 'title')
df_new1
df_new2=df_listed_in.merge(df_country,how='inner',on='title')
df_new2
df_new=df_new1.merge(df_new2,how= 'inner',on='title')
df_new.head(5)
```

```
[101]:
```

	title	director	cast \
0	Dick Johnson Is Dead	Kirsten Johnson	nan
1	Blood & Water	nan	Ama Qamata
2	Blood & Water	nan	Ama Qamata
3	Blood & Water	nan	Ama Qamata
4	Blood & Water	nan	Khosi Ngema

	listed_in	country
0	Documentaries	United States
1	International TV Shows	South Africa
2	TV Dramas	South Africa
3	TV Mysteries	South Africa
4	International TV Shows	South Africa

```
[102]: df.columns
```

```
[102]: Index(['show_id', 'type', 'title', 'director', 'cast', 'country', 'date_added',
           'release_year', 'rating', 'duration', 'listed_in', 'description'],
          dtype='object')
```

Creating the df_final DataFrame

```
[103]: df_final=df_new.merge(df[['show_id', 'type', 'title', 'date_added',
           'release_year', 'rating', 'duration',
           ↪ 'description']],how='inner',on='title')
df_final.sample(5)
```

```
[103]:
```

	title	director	cast \
133435	Results	Andrew Bujalski	Brooklyn Decker
74241	MFKZ	Shojiro Nishimi	Kenn Michael

88840	Neon Genesis Evangelion	nan	Takehito Koyasu
41563	Alone/Together	Antoinette Jadaone	Adrian Alandy
77979	Shot Caller	Ric Roman Waugh	Omari Hardwick

	listed_in	country	show_id	type	\
133435	Independent Movies	United States	s5890	Movie	
74241	Comedies	Japan	s3102	Movie	
88840	International TV Shows	Japan	s3725	TV Show	
41563	International Movies	Philippines	s1733	Movie	
77979	Dramas	United States	s3236	Movie	

	date_added	release_year	rating	duration	\
133435	October 22, 2015	2015	R	105 min	
74241	December 25, 2019	2017	R	95 min	
88840	June 21, 2019	1995	TV-MA	1 Season	
41563	November 5, 2020	2019	TV-14	103 min	
77979	November 24, 2019	2017	R	121 min	

	description
133435	After an ugly divorce, a wealthy but unhappy m...
74241	Mysterious goons chase teenage slacker Angelin...
88840	Fifteen years after the Second Impact, Shinji ...
41563	Eight years after their breakup, college sweet...
77979	Trying to go straight, a once-successful busin...

Replacing 'Nan' values with Unknown

```
[104]: df_final['director'].replace(['nan'], ['Unknown Director'], inplace = True)
df_final['cast'].replace(['nan'], ['Unknown Cast'], inplace = True)
df_final['date_added'].replace(['nan'], ['Unknown Date'], inplace = True)
df_final['country'].replace(['nan'], ['Unknown Country'], inplace=True)

df_final.head(5)
```

```
[104]:
```

	title	director	cast	\
0	Dick Johnson Is Dead	Kirsten Johnson	Unknown Cast	
1	Blood & Water	Unknown Director	Ama Qamata	
2	Blood & Water	Unknown Director	Ama Qamata	
3	Blood & Water	Unknown Director	Ama Qamata	
4	Blood & Water	Unknown Director	Khosi Ngema	

	listed_in	country	show_id	type	date_added	\
0	Documentaries	United States	s1	Movie	September 25, 2021	
1	International TV Shows	South Africa	s2	TV Show	September 24, 2021	
2	TV Dramas	South Africa	s2	TV Show	September 24, 2021	
3	TV Mysteries	South Africa	s2	TV Show	September 24, 2021	
4	International TV Shows	South Africa	s2	TV Show	September 24, 2021	

	release_year	rating	duration	\
0	2020	PG-13	90 min	
1	2021	TV-MA	2 Seasons	
2	2021	TV-MA	2 Seasons	
3	2021	TV-MA	2 Seasons	
4	2021	TV-MA	2 Seasons	

	description
0	As her father nears the end of his life, filmm...
1	After crossing paths at a party, a Cape Town t...
2	After crossing paths at a party, a Cape Town t...
3	After crossing paths at a party, a Cape Town t...
4	After crossing paths at a party, a Cape Town t...

Checking for null values

```
[105]: df_final[df_final['duration'].isnull()]
```

```
[105]:
```

	title	director	cast	\
126582	Louis C.K. 2017	Louis C.K.	Louis C.K.	
131648	Louis C.K.: Hilarious	Louis C.K.	Louis C.K.	
131782	Louis C.K.: Live at the Comedy Store	Louis C.K.	Louis C.K.	

	listed_in	country	show_id	type	date_added	\
126582	Movies	United States	s5542	Movie	April 4, 2017	
131648	Movies	United States	s5795	Movie	September 16, 2016	
131782	Movies	United States	s5814	Movie	August 15, 2016	

	release_year	rating	duration	\
126582	2017	74 min	NaN	
131648	2010	84 min	NaN	
131782	2015	66 min	NaN	

	description
126582	Louis C.K. muses on religion, eternal love, gi...
131648	Emmy-winning comedy writer Louis C.K. brings h...
131782	The comic puts his trademark hilarious/thought...

```
[106]: df_final[df_final['duration'].isnull()]
```

```
[106]:
```

	title	director	cast	\
126582	Louis C.K. 2017	Louis C.K.	Louis C.K.	
131648	Louis C.K.: Hilarious	Louis C.K.	Louis C.K.	
131782	Louis C.K.: Live at the Comedy Store	Louis C.K.	Louis C.K.	

	listed_in	country	show_id	type	date_added	\
--	-----------	---------	---------	------	------------	---

126582	Movies	United States	s5542	Movie	April 4, 2017
131648	Movies	United States	s5795	Movie	September 16, 2016
131782	Movies	United States	s5814	Movie	August 15, 2016

	release_year	rating	duration	\	
126582	2017	74 min	NaN		
131648	2010	84 min	NaN		
131782	2015	66 min	NaN		

	description
126582	Louis C.K. muses on religion, eternal love, gi...
131648	Emmy-winning comedy writer Louis C.K. brings h...
131782	The comic puts his trademark hilarious/thought...

```
[107]: import pandas as pd

# Assuming df_final is already defined and loaded with the relevant data

# Fill missing 'duration' values with 'rating' values
df_final['duration'] = df_final['duration'].fillna(df_final['rating'])

# Update specific rows in 'rating' column
df_final.loc[126582, 'rating'] = "unknown rating"
df_final.loc[131648, 'rating'] = "unknown rating"
df_final.loc[131782, 'rating'] = "unknown rating"
```

```
[108]: for ele in df_final[df_final['rating'].isnull()]:
        df_final['rating'] = df_final['rating'].fillna('unknown rating')
```

```
[109]: for ele in df_final[df_final['date_added'].isnull()]:
        df_final['date_added'] = df_final['date_added'].fillna('unknown date')
```

Confirming no further nulls in the df_final

```
[110]: df_final.isnull().sum()
```

```
[110]: title          0
       director      0
       cast          0
       listed_in     0
       country       0
       show_id       0
       type          0
       date_added    0
       release_year  0
       rating        0
       duration      0
```

```
description      0
dtype: int64
```

Final data frame after cleaning as 'df_final'

```
[111]: df_final.head()
```

```
[111]:
```

	title	director	cast	\
0	Dick Johnson Is Dead	Kirsten Johnson	Unknown Cast	
1	Blood & Water	Unknown Director	Ama Qamata	
2	Blood & Water	Unknown Director	Ama Qamata	
3	Blood & Water	Unknown Director	Ama Qamata	
4	Blood & Water	Unknown Director	Khosi Ngema	

	listed_in	country	show_id	type	date_added	\
0	Documentaries	United States	s1	Movie	September 25, 2021	
1	International TV Shows	South Africa	s2	TV Show	September 24, 2021	
2	TV Dramas	South Africa	s2	TV Show	September 24, 2021	
3	TV Mysteries	South Africa	s2	TV Show	September 24, 2021	
4	International TV Shows	South Africa	s2	TV Show	September 24, 2021	

	release_year	rating	duration	\
0	2020	PG-13	90 min	
1	2021	TV-MA	2 Seasons	
2	2021	TV-MA	2 Seasons	
3	2021	TV-MA	2 Seasons	
4	2021	TV-MA	2 Seasons	

	description
0	As her father nears the end of his life, filmm...
1	After crossing paths at a party, a Cape Town t...
2	After crossing paths at a party, a Cape Town t...
3	After crossing paths at a party, a Cape Town t...
4	After crossing paths at a party, a Cape Town t...

1. What type of content is available in different countries?

```
[112]: #Non-graphical analysis
type_uni=df_final['type'].nunique()
country_uni=df_final['country'].nunique()
print(f'The no.of unique values of type of content is :{type_uni} ')
print(f'The no.of unique country content :{country_uni} ')
```

```
The no.of unique values of type of content is :2
The no.of unique country content :198
```

```
[113]: country_wise=df_final.groupby(['country'])['type'].value_counts()
country_wise_df=country_wise.reset_index(name='counts')
```



```

country_wise_df

# Remove the 'counts' column
country_wise_df.drop(columns=['counts'], inplace=True, errors='ignore')

# Merge rows for countries with both "Movie" and "TV Show"
country_wise_df = country_wise_df.groupby('country')['type'].apply(lambda x: '␣'
↪& '.join(map(str, x))).reset_index()

# Function to map types to numeric categories
def map_category(types):
    if 'Movie' in types and 'TV Show' in types:
        return 2
    elif 'Movie' in types:
        return 0
    elif 'TV Show' in types:
        return 1
    else:
        return None

# Apply the mapping function to create the 'category' column
country_wise_df['category'] = country_wise_df['type'].apply(map_category)

country_wise_df.drop(columns=['type'], inplace=True)
country_wise_df.rename(columns={'category': 'type'}, inplace=True)
country_wise_df

```

```

[113]:
      country  type
0           2
1  Afghanistan  0
2    Albania    0
3    Algeria    0
4    Angola     0
..         ...  ...
193   Uruguay    2
194  Venezuela    0
195   Vietnam    0
196 West Germany  0
197   Zimbabwe    0

[198 rows x 2 columns]

```

```

[114]: import geopandas as gpd

world = gpd.read_file(gpd.datasets.get_path('naturalearth_lowres'))

```

```

# Merge world map with the data
world = world.merge(country_wise_df, how='left', left_on='name',
                    right_on='country')

# Plotting
fig, ax = plt.subplots(1, 1, figsize=(10, 6))

# Plot countries with type 0 in one color
world[world['type'] == 0].plot(ax=ax, color='green', legend=True)

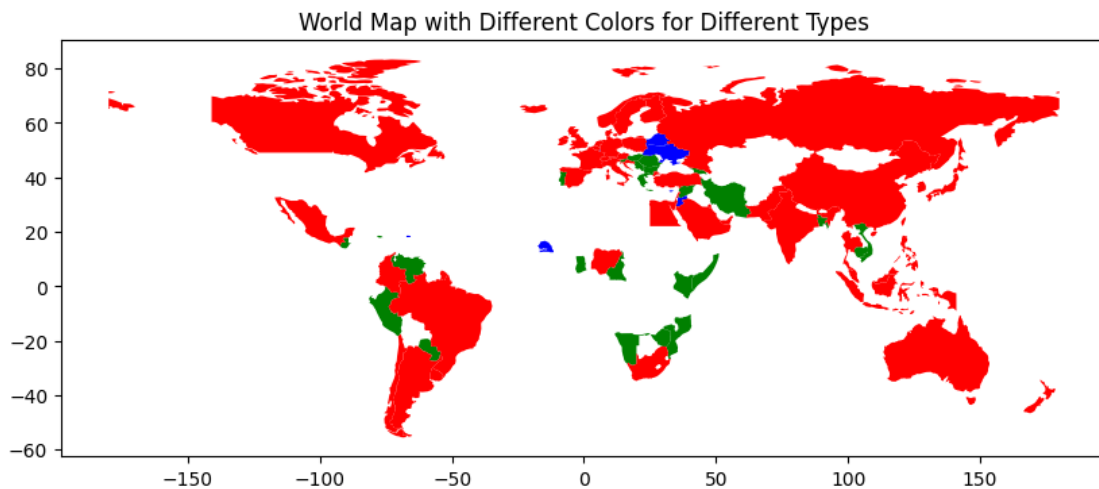
# Plot countries with type 1 in another color
world[world['type'] == 1].plot(ax=ax, color='blue', legend=True)

# Plot countries with type 2 in another color
world[world['type'] == 2].plot(ax=ax, color='red', legend=True)

ax.set_title('World Map with Different Colors for Different Types')
plt.show()

```

<ipython-input-114-a2bd7936e57b>:4: FutureWarning: The geopandas.dataset module is deprecated and will be removed in GeoPandas 1.0. You can get the original 'naturalearth_lowres' data from <https://www.naturalearthdata.com/downloads/110m-cultural-vectors/>.
world = gpd.read_file(gpd.datasets.get_path('naturalearth_lowres'))



2. How has the number of movies released per year changed over the last 20-30 years?

```

[115]: # Extract relevant columns and remove duplicates
movies_yr = df_final[['title', 'type', 'release_year']].copy()
movies_yr.drop_duplicates(subset=['title'], inplace=True)

```

```

movies_yr.reset_index(drop=True, inplace=True)

# Group by type and release year, then count the occurrences
movies_yr = movies_yr.groupby(['type', 'release_year']).size().
    ↪reset_index(name='count')

# Filter the DataFrame to include only movies
movies_yr_df = movies_yr[movies_yr['type'] == 'Movie'].copy()

# Sort the DataFrame by release year
movies_yr_df.sort_values(by='release_year', inplace=True)

# Define the time period for the last 30 years
end_year = 2021
start_year = end_year - 30

# Filter the DataFrame for the specified time period
movies_yr_df = movies_yr_df[
    (movies_yr_df['release_year'] >= start_year) &
    (movies_yr_df['release_year'] <= end_year)
]

# Display the first few rows of the filtered DataFrame
movies_yr_df.head()

```

```

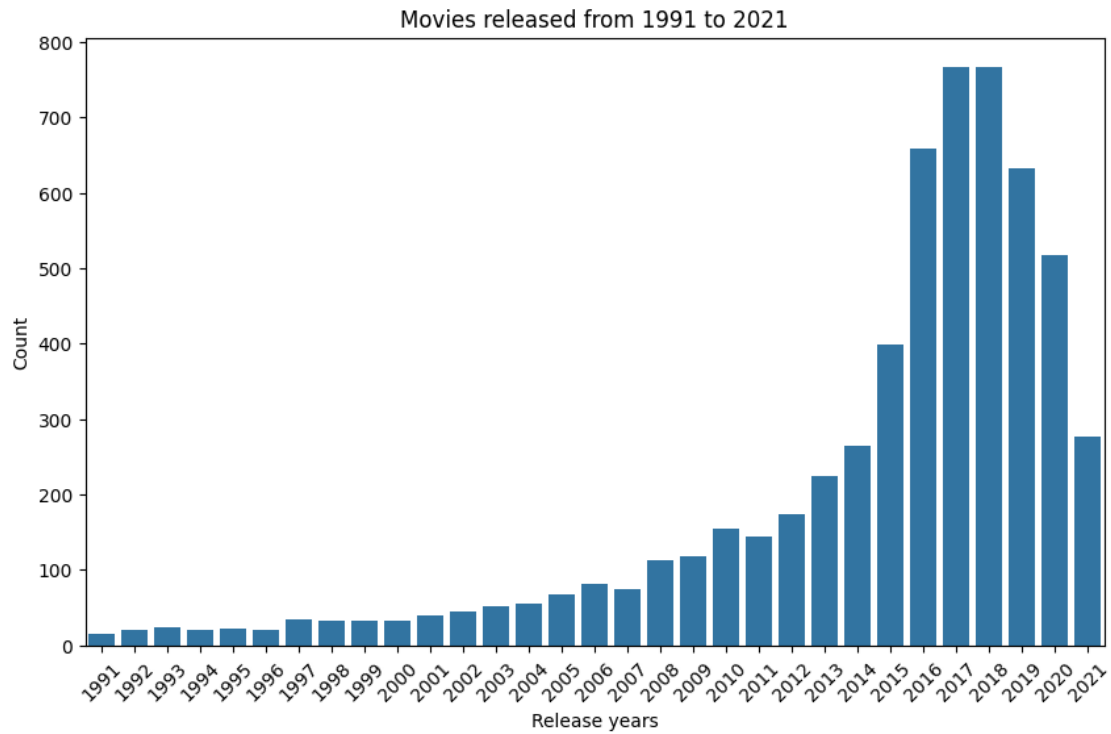
[115]:
   type  release_year  count
42  Movie           1991     16
43  Movie           1992     20
44  Movie           1993     24
45  Movie           1994     20
46  Movie           1995     23

```

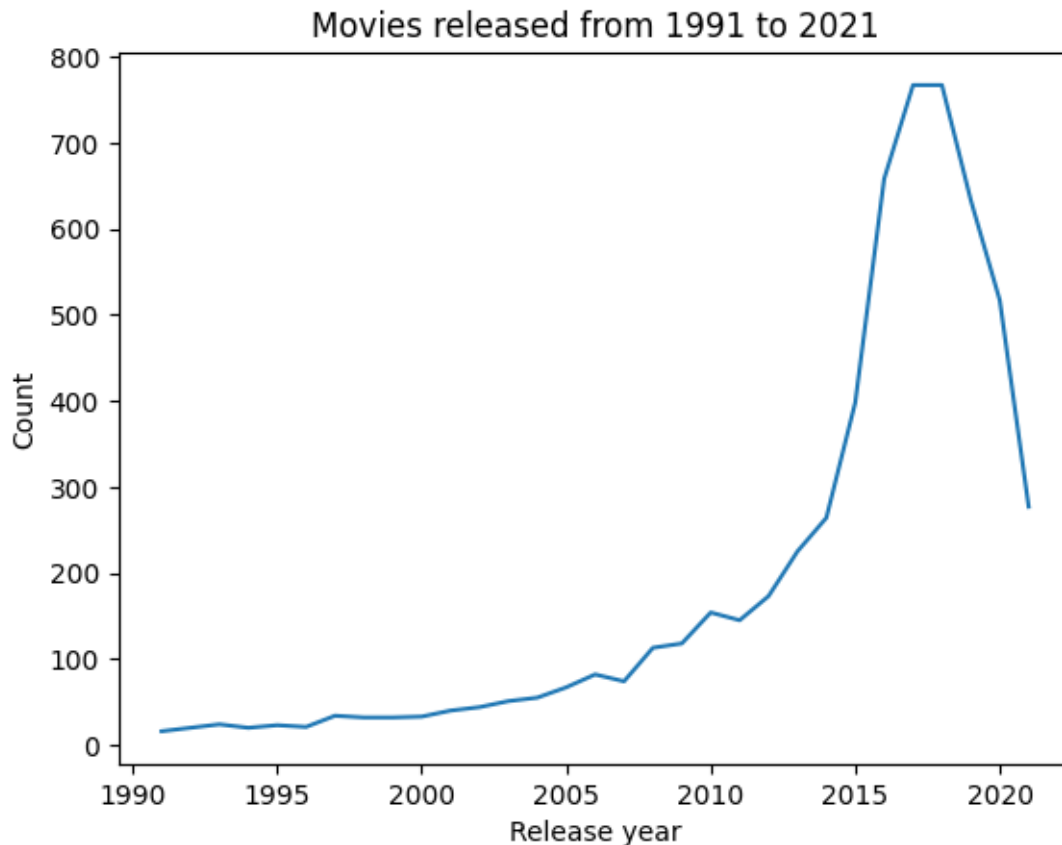
```

[116]: plt.figure(figsize=(10, 6))
sns.barplot(x='release_year', y='count', data=movies_yr_df)
plt.xlabel('Release years')
plt.ylabel('Count')
plt.title('Movies released from 1991 to 2021')
plt.xticks(rotation=45)
plt.show()

```



```
[117]: sns.lineplot(x='release_year',y='count',data=movies_yr_df)
plt.title('Movies released from 1991 to 2021')
plt.xlabel('Release year')
plt.ylabel('Count')
plt.show()
```



3.Comparison of tv shows vs. movies.

```
[118]: # Extracting only title, type and release_year columns from cleaned data
movie_shows = df_final[['title', 'type', 'release_year']].copy()

# Deleting duplicate rows
movie_shows.drop_duplicates(subset=['title'], inplace=True)

# Resetting the index after deleting duplicate rows
movie_shows = movie_shows.reset_index(drop=True)

# Grouping by type (Movies and TV shows) and value counting the number of
↳ movies and TV shows per year
movie_shows = movie_shows.groupby('type')['release_year'].value_counts()

# Resetting index again
movie_shows = movie_shows.reset_index(name='count')

# Splitting the dataframe by movie and TV show
movie_shows_df = movie_shows[movie_shows['type'] == 'Movie'].copy()
```

```

tv_shows_df = movie_shows[movie_shows['type'] == 'TV Show'].copy()

# Sorting the dataframes by release year
movie_shows_df.sort_values(by='release_year', inplace=True)
tv_shows_df.sort_values(by='release_year', inplace=True)

movie_shows_df.head(), tv_shows_df.head()

```

```

[118]: (
      type  release_year  count
62  Movie           1942      2
59  Movie           1943      3
58  Movie           1944      3
56  Movie           1945      3
68  Movie           1946      1,
      type  release_year  count
107 TV Show           1925      1
106 TV Show           1945      1
108 TV Show           1946      1
110 TV Show           1963      1
111 TV Show           1967      1)

```

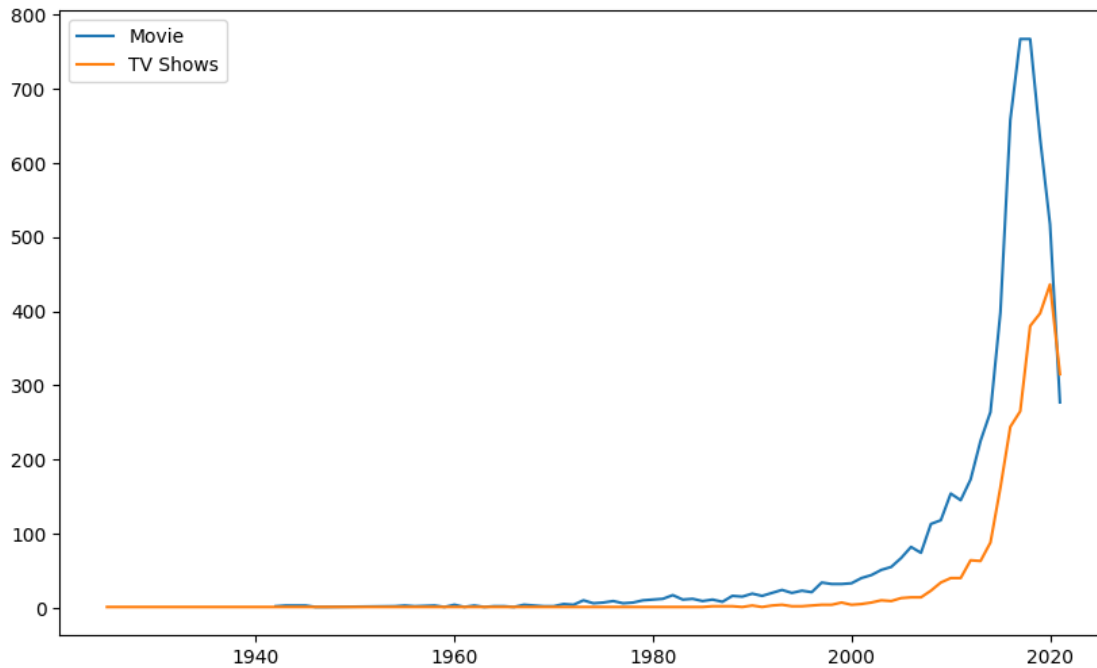
```

[119]: plt.figure(figsize=(10, 6))
# Plot the dataframes
plt.plot(movie_shows_df['release_year'], movie_shows_df['count'], label='Movie')
plt.plot(tv_shows_df['release_year'], tv_shows_df['count'], label='TV Shows')

# Add a legend
plt.legend()

# Show the plot
plt.show()

```



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

```
[120]: # Extract relevant columns and remove duplicates
best_time = df_final[['title', 'date_added', 'type']].copy()
best_time.drop_duplicates(subset=['title'], inplace=True)
best_time.reset_index(drop=True, inplace=True)

# Remove rows where type is 'Movie'
best_time = best_time[best_time['type'] != 'Movie'].copy()

# Extract the month from the 'date_added' column
best_time['month_added'] = best_time['date_added'].apply(lambda x: x.split('_')[
    ↪ 0])

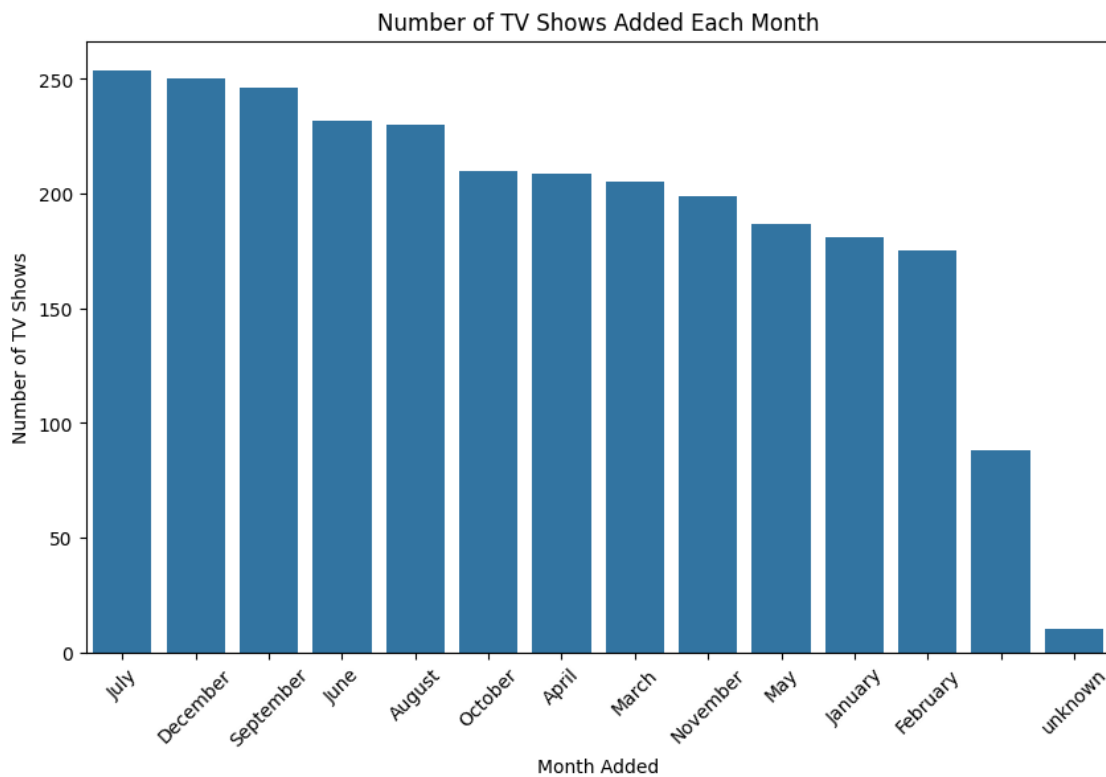
# Count the number of TV shows added each month
best = best_time['month_added'].value_counts().reset_index()
best.columns = ['month_added', 'count']

# Extract the top 5 months
top_months = best.nlargest(5, 'count')['month_added'].tolist()
print(f"The best time to launch a TV show is: \n{top_months}")
```

The best time to launch a TV show is:

```
['July', 'December', 'September', 'June', 'August']
```

```
[121]: plt.figure(figsize=(10, 6))
sns.barplot(x='month_added', y='count', data=best, order=best['month_added'])
plt.xlabel('Month Added')
plt.ylabel('Number of TV Shows')
plt.title('Number of TV Shows Added Each Month')
plt.xticks(rotation=45)
plt.show()
```



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

```
[122]: analysis_dir = df_final[['director', 'rating', 'type', 'title']].copy()
analysis_dir.drop_duplicates(subset=['title'], inplace=True)
analysis_dir.reset_index(drop=True, inplace=True)

analysis_dir = df_final.groupby(['director', 'rating', 'type']).size().
    ↪ reset_index(name='count').sort_values(by=['director', 'rating', 'type'])
movie_shows_dir_df = analysis_dir[analysis_dir['type'] == 'Movie'].copy()
movie_shows_dir_df.drop(movie_shows_dir_df[movie_shows_dir_df['rating'] == '
    ↪ unknown rating'].index, inplace=True)

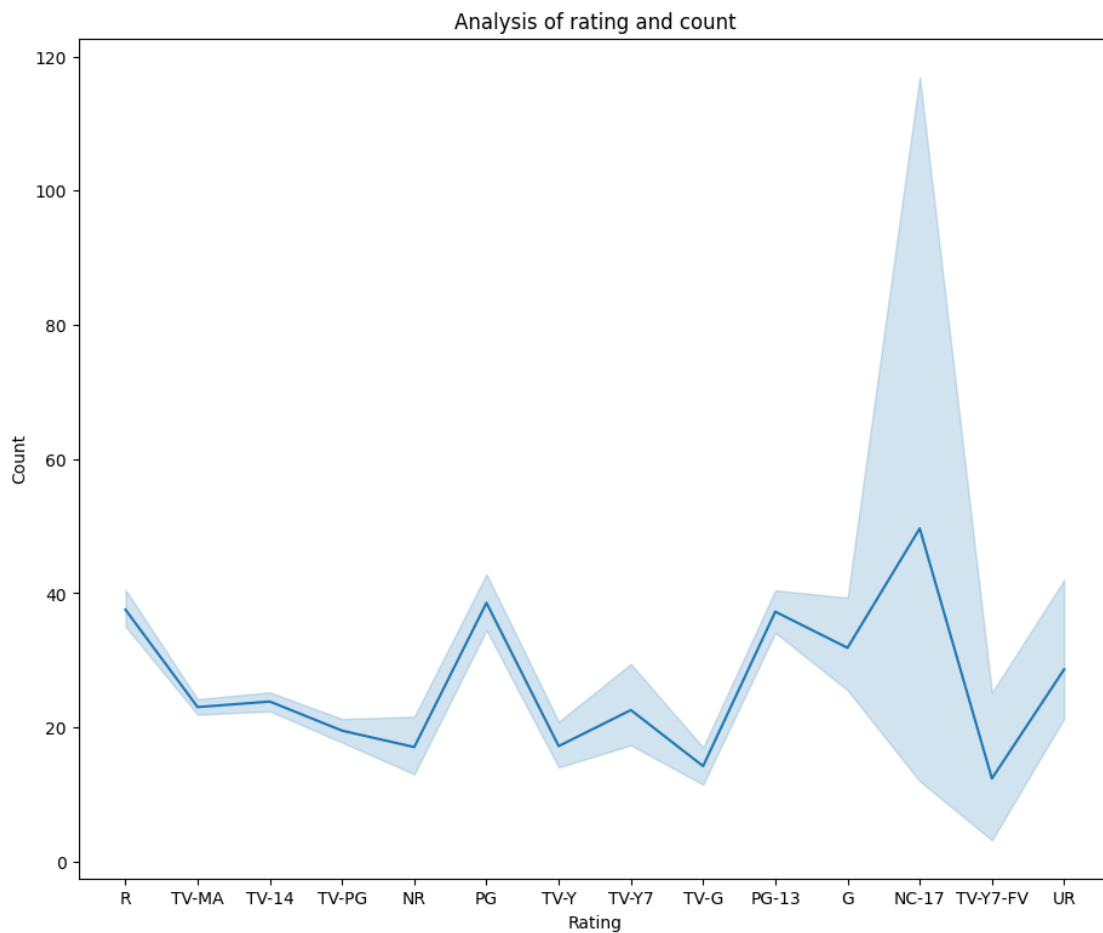
movie_shows_dir_df.head(5)
```



```
[122]:
```

	director	rating	type	count
0	Aaron Moorhead	R	Movie	6
1	Aaron Moorhead	TV-MA	Movie	18
2	Aaron Woolf	TV-14	Movie	2
3	Abbas Alibhai Burmawalla	TV-14	Movie	18
4	Abdullah Al Noor	TV-MA	Movie	30

```
[123]: plt.figure(figsize=(11, 9))
sns.lineplot(x='rating',y='count',data=movie_shows_dir_df)
plt.title('Analysis of rating and count')
plt.xlabel('Rating')
plt.ylabel('Count')
plt.show()
```



6. Does Netflix has more focus on TV Shows than movies in recent years

```
[124]: recent_years = df_final[df_final['release_year'] >= 2016]
recent_years = recent_years.drop_duplicates(subset=["title"])
```

```

type_counts = recent_years['type'].value_counts()
movie = type_counts.get('Movie', 0)
tv_show = type_counts.get('TV Show', 0)
print(f"Number of movies in recent years: {movie}")
print(f"Number of TV shows in recent years: {tv_show}")

if movie > tv_show:
    print("Netflix has more focus on movies in recent years.")
elif movie < tv_show:
    print("Netflix has more focus on TV shows in recent years.")

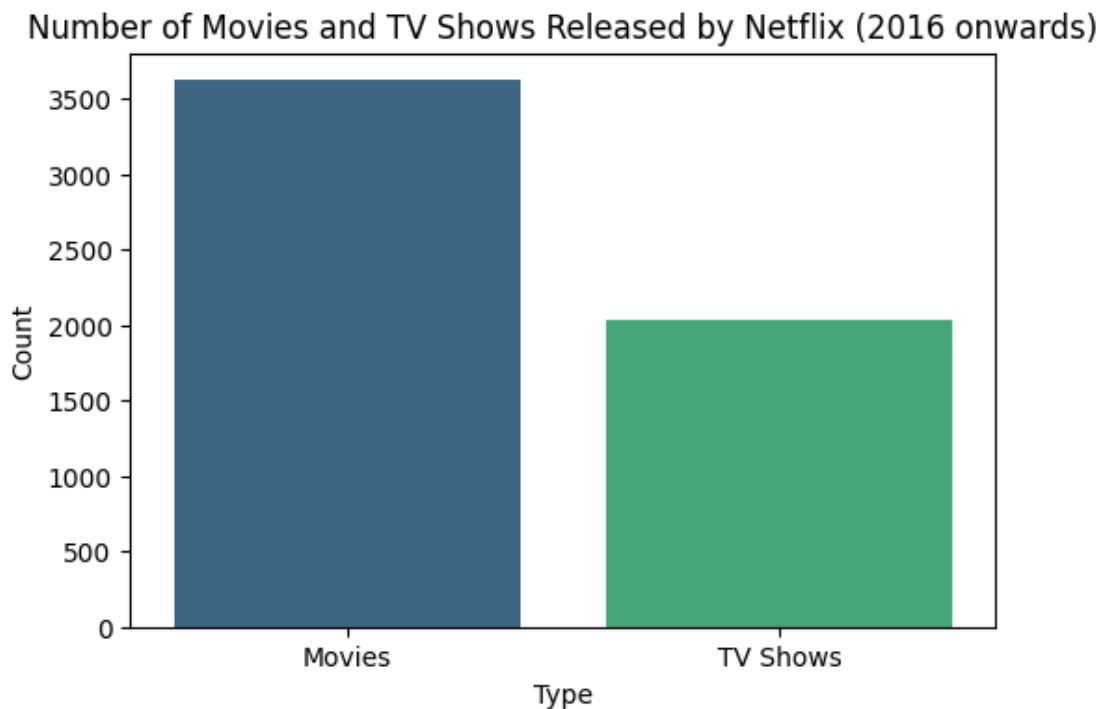
```

Number of movies in recent years: 3619
 Number of TV shows in recent years: 2037
 Netflix has more focus on movies in recent years.

```

[125]: plt.figure(figsize=(6, 4))
sns.barplot(x=['Movies', 'TV Shows'], y=[movie, tv_show], hue=['Movies', 'TV_
↳Shows'], palette='viridis', legend=False)
plt.xlabel('Type')
plt.ylabel('Count')
plt.title('Number of Movies and TV Shows Released by Netflix (2016 onwards)')
plt.show()

```



7 .Understanding what content is available in different countries

```
[126]: df_understanding=df_final.groupby(['country'])['listed_in'].value_counts()
df_understanding_1=df_understanding.reset_index(name='counts')
df_understanding_1.sample(5)
```

```
[126]:
```

	country	listed_in	counts
936	United Arab Emirates	Thrillers	8
2180	Switzerland	Faith & Spirituality	1
141	Canada	Kids' TV	58
2007	Russia	TV Action & Adventure	8
1773	Malaysia	TV Comedies	7

```
[128]: # Create a figure and subplots
plt.figure(figsize=(18, 6))

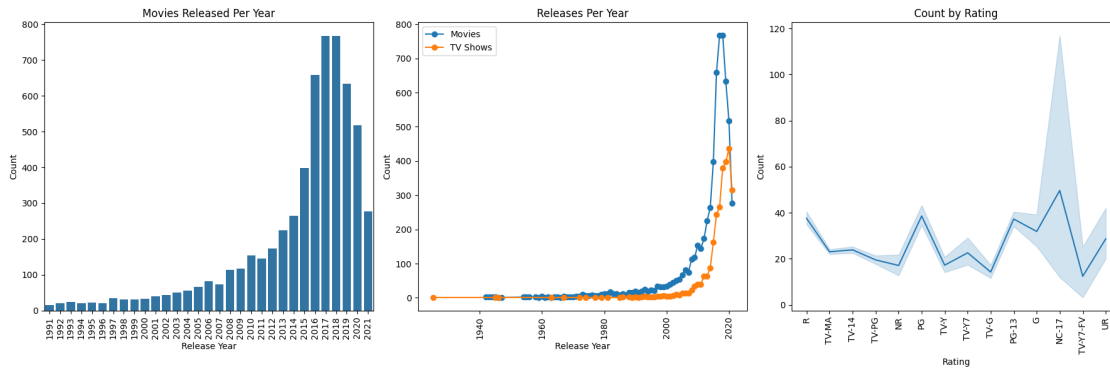
# Plot first subplot
plt.subplot(1, 3, 1)
sns.barplot(x='release_year', y='count', data=movies_yr_df)
plt.xlabel('Release Year')
plt.ylabel('Count')
plt.title('Movies Released Per Year')
plt.xticks(rotation=90)

# Plot second subplot
plt.subplot(1, 3, 2)
plt.plot(movie_shows_df['release_year'], movie_shows_df['count'], label='Movies', marker='o')
plt.plot(tv_shows_df['release_year'], tv_shows_df['count'], label='TV Shows', marker='o')
plt.xlabel('Release Year')
plt.ylabel('Count')
plt.title('Releases Per Year')
plt.legend()
plt.xticks(rotation=90)

# Plot third subplot
plt.subplot(1, 3, 3)
sns.lineplot(x='rating', y='count', data=movie_shows_dir_df)
plt.xlabel('Rating')
plt.ylabel('Count')
plt.title('Count by Rating')
plt.xticks(rotation=90)

# Adjust layout to prevent overlap
plt.tight_layout()

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

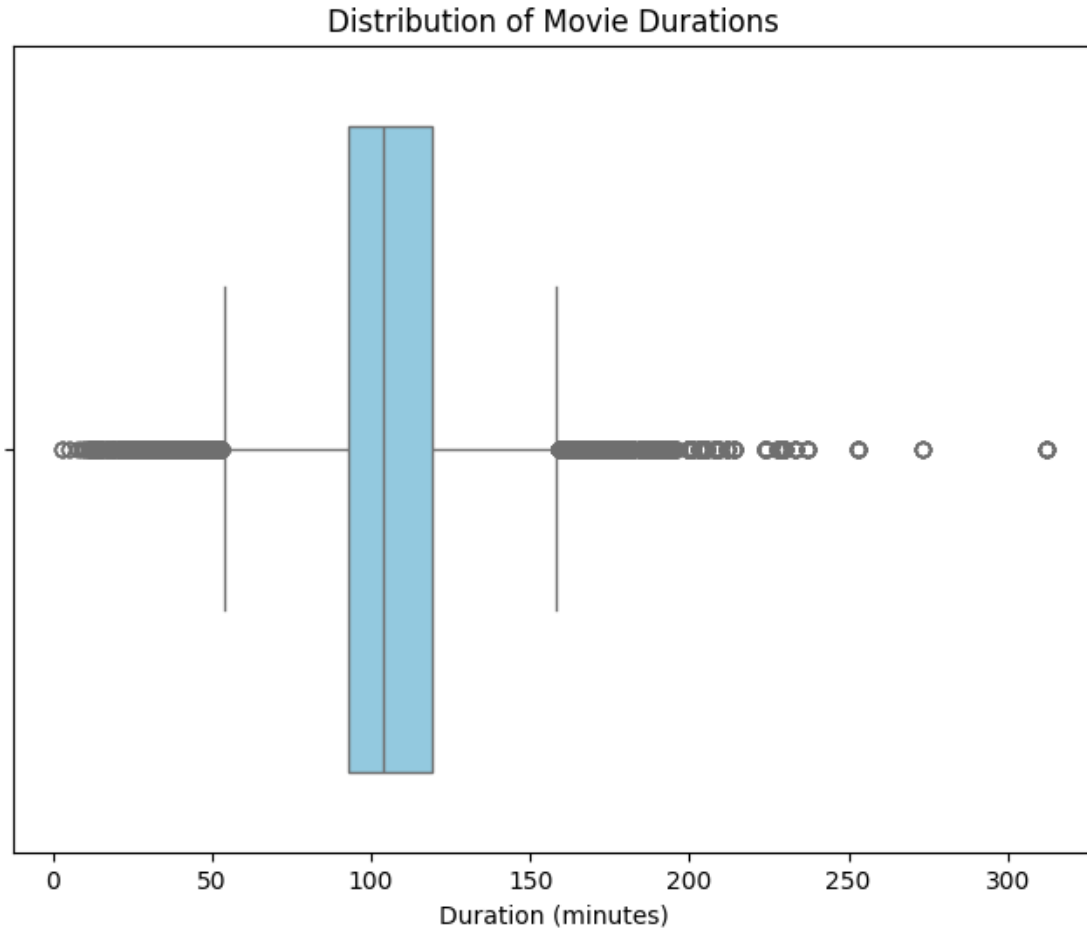


Box plot for the distribution of movie duration.

```
[129]: # Filter out TV shows
movies_data = df_final[df_final['type'] == 'Movie'].copy()

# Convert duration to numeric (remove 'min' and convert to integer)
movies_data['duration'] = movies_data['duration'].str.replace(' min', '').
    .astype(int)

# Create a boxplot
plt.figure(figsize=(8, 6))
sns.boxplot(x=movies_data['duration'], color='skyblue')
plt.title('Distribution of Movie Durations')
plt.xlabel('Duration (minutes)')
plt.show()
```



0.0.1 1. *Content Availability in Different Countries*

Insight: - Different countries have access to different types of content, including movies, TV shows, or both.

Recommendation: - *Localization Strategy:* Tailor content offerings to specific countries based on local preferences. For instance, if a country shows a higher preference for TV shows, focus on acquiring or producing more TV shows for that market. - *Content Diversity:* Ensure a mix of both movies and TV shows in each market to cater to a broader audience and avoid saturating the market with one type of content.

0.0.2 2. *Trends in Movie Releases Over the Last 20-30 Years*

Insight: - There is a noticeable trend in the number of movies released over the years, with fluctuations possibly influenced by industry changes and global events.

Recommendation: - *Historical Analysis:* Use historical data to predict future trends in movie releases and plan content acquisition accordingly. - *Seasonal Releases:* Identify peak times for movie releases and schedule major releases during these periods to maximize viewership.

0.0.3 3. Comparison of TV Shows vs. Movies

Insight: - There has been a notable difference in the release patterns of movies and TV shows over the years.

Recommendation: - *Content Balance:* Maintain a balance between movies and TV shows to cater to different audience preferences. - *Genre Analysis:* Analyze popular genres within both categories to optimize content acquisition and production.

0.0.4 4. Best Time to Launch a TV Show

Insight: - Certain months have higher counts of TV show releases, indicating possibly favorable times for launching new shows.

Recommendation: - *Optimal Release Timing:* Plan the launch of new TV shows during months with historically high viewership to capitalize on audience engagement. - *Marketing Campaigns:* Align marketing efforts with these optimal times to maximize the impact of new releases.

0.0.5 5. Analysis of Actors/Directors in Different Types of Shows/Movies

Insight: - Certain directors and actors are associated with specific types of content and ratings.

Recommendation: - *Talent Acquisition:* Focus on collaborating with directors and actors who have a proven track record of success in the type of content Netflix wants to promote. - *Star Power:* Use well-known actors and directors to attract viewership, especially in new markets or for new content types.

0.0.6 6. Focus on TV Shows vs. Movies in Recent Years

Insight: - There has been a shift in focus towards TV shows in recent years.

Recommendation: - *Strategic Focus:* If the trend shows a growing interest in TV shows, continue to invest in high-quality TV show production and acquisition. - *Content Expansion:* Consider expanding the TV show library, including different genres and international TV shows, to cater to diverse tastes.

0.0.7 7. Understanding Content Available in Different Countries

Insight: - The content available varies significantly between countries, reflecting local tastes and preferences.

Recommendation: - *Localized Content:* Increase investment in local content production and acquisitions to cater to regional preferences. - *Global Distribution:* Ensure popular content is available globally, with localized subtitles and dubbing to enhance accessibility and viewership.

0.0.8 General Recommendations:

- *Data-Driven Decisions:* Continuously use data analytics to monitor viewer preferences and trends, enabling more informed decisions about content acquisition and production.
- *User Feedback:* Regularly collect and analyze user feedback to understand their preferences and improve the content library accordingly.

- *Marketing Strategies:* Tailor marketing strategies based on regional insights and viewing habits to maximize engagement and subscription rates.
- *Content Innovation:* Explore new and innovative content formats, such as interactive shows or short-form content, to stay ahead in the competitive streaming market.

By leveraging these insights and recommendations, Netflix can optimize its content strategy to enhance user engagement, satisfaction, and overall market competitiveness.