

Netflix-EDA

March 23, 2025

1 Netflix-Exploratory Data Analysis (EDA) and Visualization Using Python

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

```
[2]: df=pd.read_csv(r"C:\Users\myse1\Downloads\netflix_titles_2021 - 1
↳netflix_titles_2021.csv")
```

```
[3]: df.head()
```

```
[3]:  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...

```

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

```

[4]: 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

```

```
[5]: 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

```

```
[6]: df.describe(include="all")
```

```
[6]:
```

	show_id	type	title	director	\
count	8807	8807	8807	6173	
unique	8807	2	8807	4528	
top	s1	Movie	Dick Johnson Is Dead	Rajiv Chilaka	
freq	1	6131	1	19	
mean	NaN	NaN	NaN	NaN	
std	NaN	NaN	NaN	NaN	
min	NaN	NaN	NaN	NaN	
25%	NaN	NaN	NaN	NaN	
50%	NaN	NaN	NaN	NaN	
75%	NaN	NaN	NaN	NaN	
max	NaN	NaN	NaN	NaN	

	cast	country	date_added	release_year	\
count	7982	7976	8797	8807.000000	
unique	7692	748	1714	NaN	
top	David Attenborough	United States	January 1, 2020	NaN	
freq	19	2818	110	NaN	
mean	NaN	NaN	NaN	2014.180198	
std	NaN	NaN	NaN	8.819312	
min	NaN	NaN	NaN	1925.000000	
25%	NaN	NaN	NaN	2013.000000	
50%	NaN	NaN	NaN	2017.000000	
75%	NaN	NaN	NaN	2019.000000	
max	NaN	NaN	NaN	2021.000000	

	rating	duration	listed_in	\
count	8803	8804	8807	
unique	17	220	514	
top	TV-MA	1 Season	Dramas, International Movies	
freq	3207	1793	362	
mean	NaN	NaN	NaN	
std	NaN	NaN	NaN	
min	NaN	NaN	NaN	
25%	NaN	NaN	NaN	
50%	NaN	NaN	NaN	
75%	NaN	NaN	NaN	
max	NaN	NaN	NaN	

	description
count	8807
unique	8775
top	Paranormal activity at a lush, abandoned prope...
freq	4
mean	NaN

std	NaN
min	NaN
25%	NaN
50%	NaN
75%	NaN
max	NaN

1.1 Data Cleaning and Handling Outliers

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

```
[8]: # Ensure 'duration' is string and fill missing values
df['duration'] = df['duration'].astype(str).fillna('Unknown')

# Convert seasons to equivalent minutes (assuming 1 season = 400 min)
def convert_duration(x):
    if 'Season' in x:
        try:
            num_seasons = int(x.split()[0])
            return f"{num_seasons * 400} min"
        except (ValueError, IndexError):
            return 'Unknown'
    return x

df['duration'] = df['duration'].apply(convert_duration)

# Extract numeric values using raw string to avoid warnings
df['duration_numeric'] = df['duration'].str.extract(r'(\d+)').astype(float)
```

```
[9]: # Fill missing values:
df['director'] = df['director'].fillna('Unknown')
df['cast'] = df['cast'].fillna('Unknown')
df['country'] = df['country'].fillna('Unknown')
df['rating'] = df['rating'].fillna('Not Rated')
df['duration_numeric'] = df['duration_numeric'].fillna(0)
```

```
[10]: # Check for duplicates
duplicates = df.duplicated().sum()
duplicates
```

```
[10]: 0
```

```
[11]: #After fiiling Missing Values
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8807 entries, 0 to 8806
```

Data columns (total 13 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	8807 non-null	object
4	cast	8807 non-null	object
5	country	8807 non-null	object
6	date_added	8797 non-null	datetime64[ns]
7	release_year	8807 non-null	int64
8	rating	8807 non-null	object
9	duration	8807 non-null	object
10	listed_in	8807 non-null	object
11	description	8807 non-null	object
12	duration_numeric	8807 non-null	float64

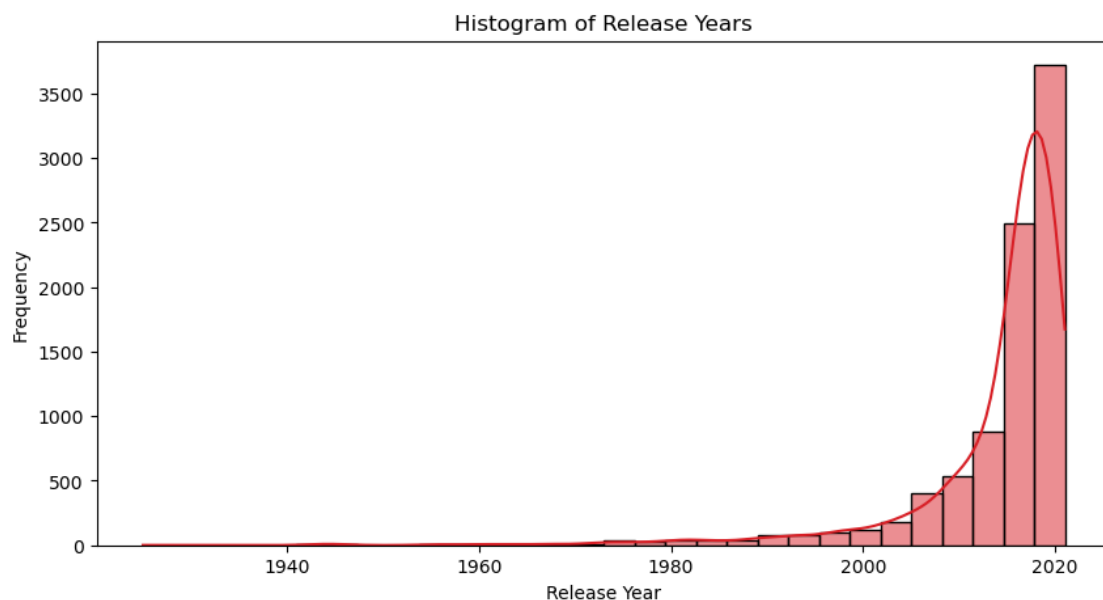
dtypes: datetime64[ns](1), float64(1), int64(1), object(10)

memory usage: 894.6+ KB

2 Visualizing and Finding insights

2.1 Plot histogram for release year

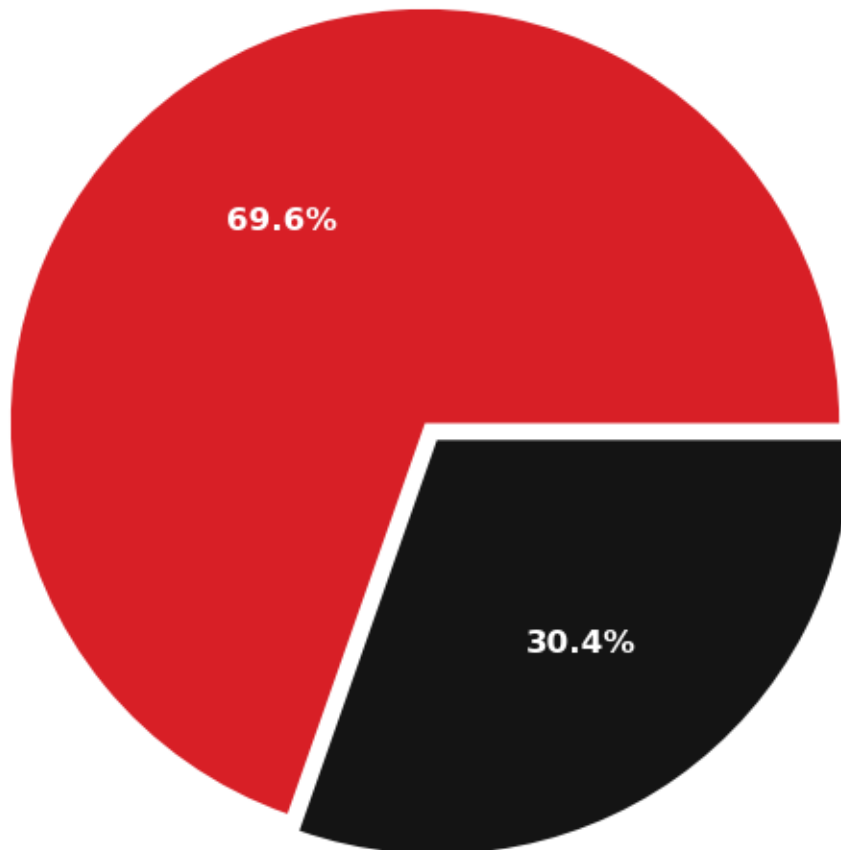
```
[12]: plt.figure(figsize=(10, 5))
sns.histplot(df['release_year'], bins=30, kde=True, color='#D81F26')
plt.title('Histogram of Release Years')
plt.xlabel('Release Year')
plt.ylabel('Frequency')
plt.show()
```



2.2 Visualizing content type using a pie chart

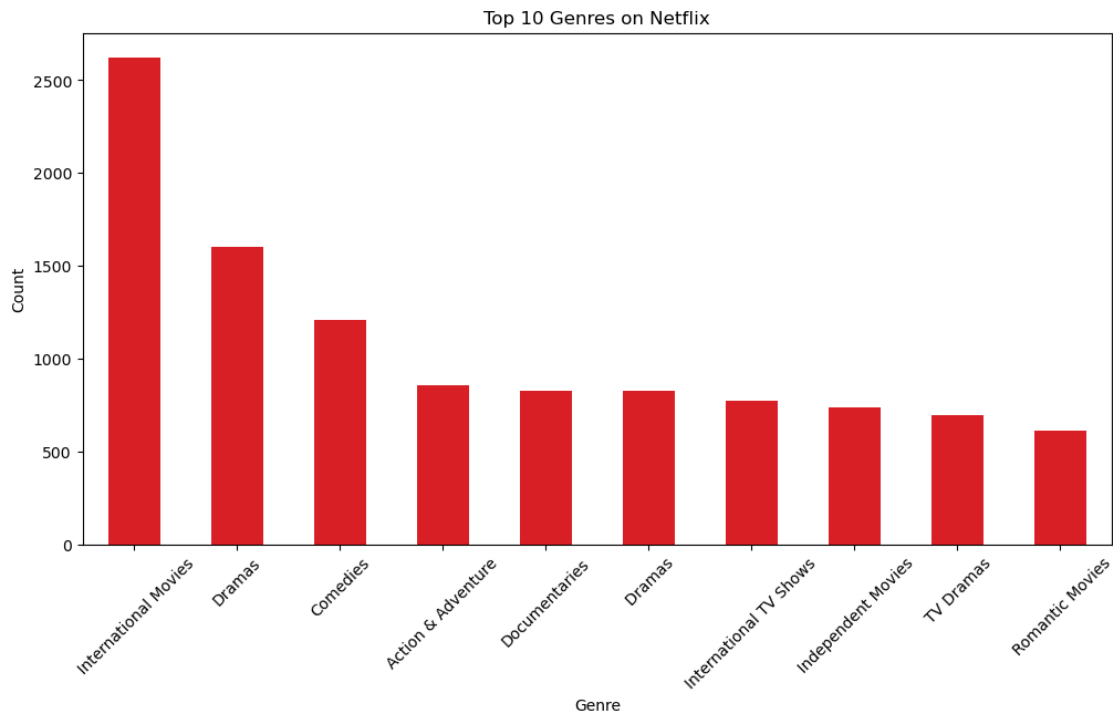
```
[13]: plt.figure(figsize=(7, 7))
df['type'].value_counts().plot.pie(autopct='%1.1f%%',
    ↪ colors=['#D81F26', '#141414'], textprops={'color': 'white', 'size':
    ↪ 12, 'fontweight': 'bold'}, explode=[0, 0.05])
plt.title('Distribution of Content Type (Movies vs TV Shows)')
plt.ylabel('')
plt.show()
```

Distribution of Content Type (Movies vs TV Shows)



2.3 Visualizing Top genres using a bar chart

```
[14]: plt.figure(figsize=(12, 6))
df['listed_in'].str.split(',').explode().value_counts().head(10).
    plot(kind='bar', color='#D81F26')
plt.title('Top 10 Genres on Netflix')
plt.xlabel('Genre')
plt.ylabel('Count')
plt.xticks(rotation=45)
plt.show()
```



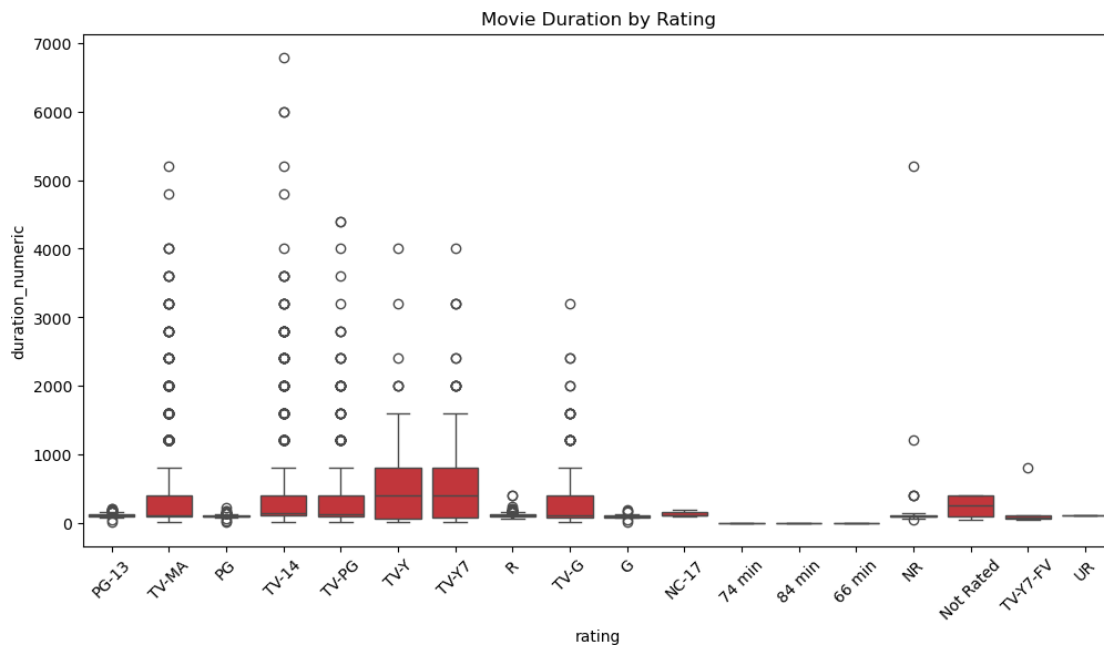
3 Here are the insights from the visualizations:

3.1 Content Type:

Movies account for approximately 70% of the content on Netflix, while TV shows make up the remaining 30%. ## Genres: The most common genres include Dramas, Comedies, and Documentaries, reflecting Netflix's strong focus on diverse storytelling.

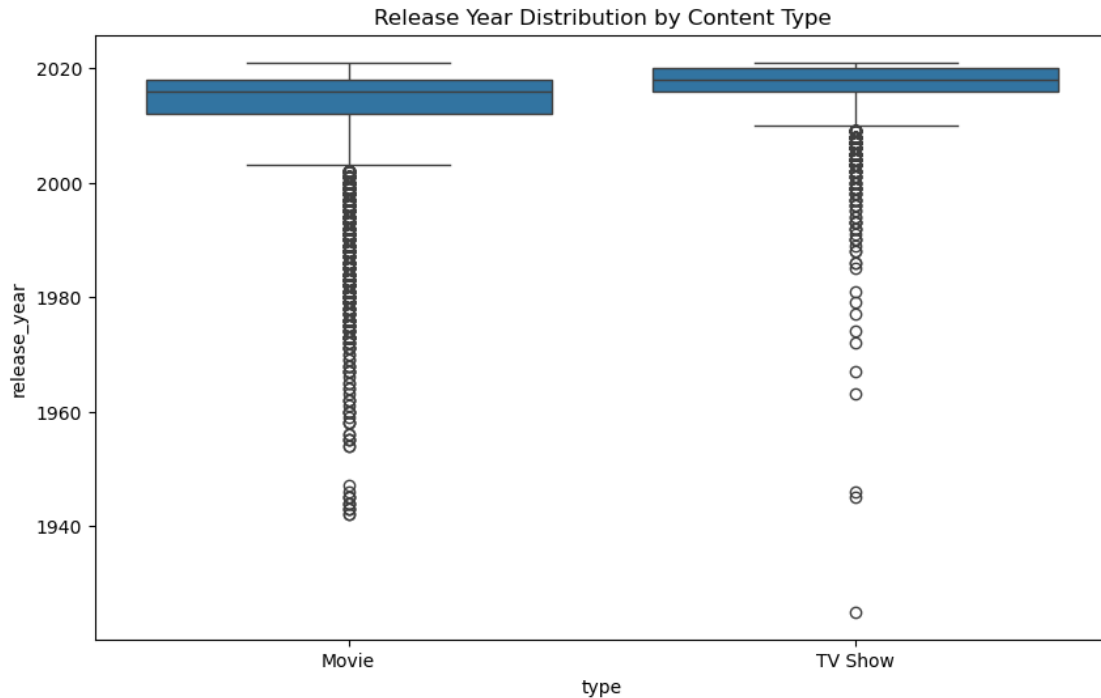
3.2 Boxplot for Movie Duration by Rating

```
[15]: plt.figure(figsize=(12, 6))
sns.boxplot(x='rating', y='duration_numeric', data=df, color='#D81F26')
plt.title('Movie Duration by Rating')
plt.xticks(rotation=45)
plt.show()
```



3.3 Boxplot for release years by content type

```
[16]: plt.figure(figsize=(10, 6))
sns.boxplot(x='type', y='release_year', data=df)
plt.title('Release Year Distribution by Content Type')
plt.show()
```

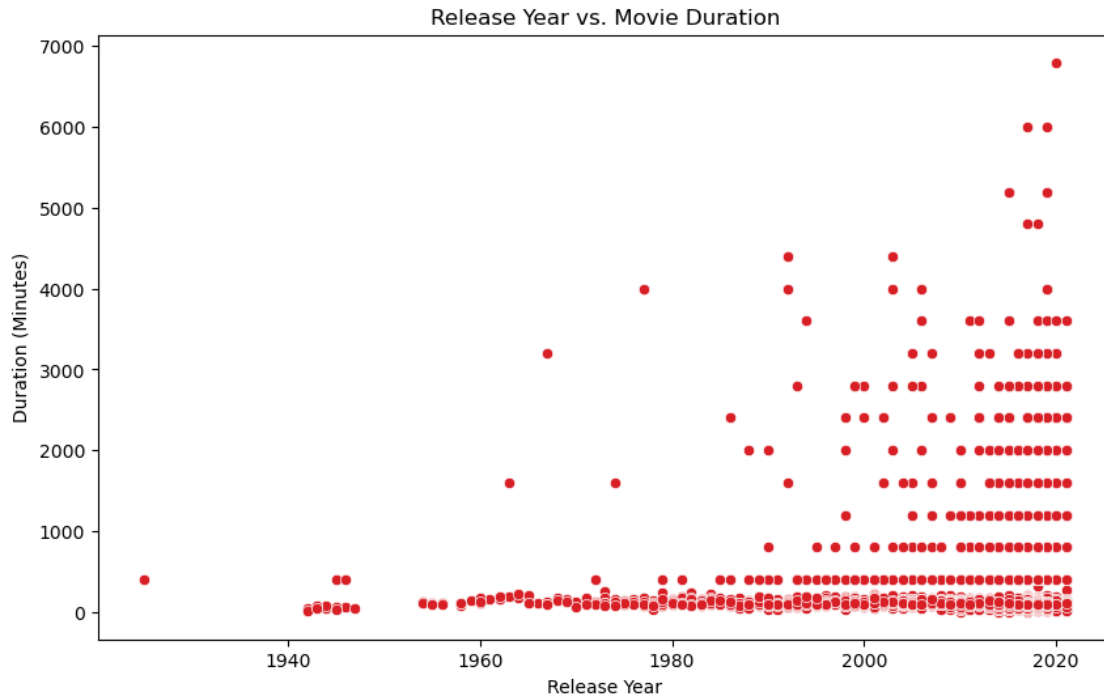
4 The boxplots provide these insights:

4.1 Movie Duration by Rating:

TV-MA and TV-14 movies tend to have longer durations compared to other ratings like PG or R. Movies with a G rating have shorter durations, typically aligned with children's content. ##
Release Year by Content Type: Both movies and TV shows have seen a significant rise in releases after 2015. However, TV shows display a slightly wider range, with older releases still available on the platform.

4.2 Scatter plot to explore relationship between release year and movie duration

```
[17]: plt.figure(figsize=(10, 6))
sns.scatterplot(x='release_year', y='duration_numeric', data=df,
               color='#D81F26', alpha=1)
plt.title('Release Year vs. Movie Duration')
plt.xlabel('Release Year')
plt.ylabel('Duration (Minutes)')
plt.show()
```



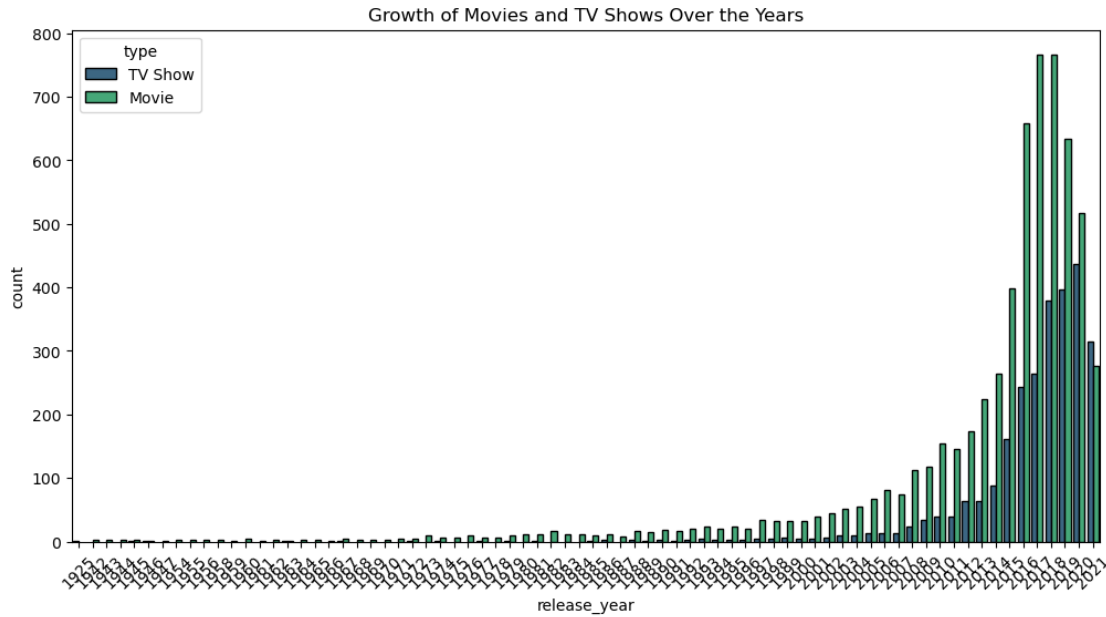
5 The analysis of relationships reveals:

5.1 Release Year vs. Movie Duration:

Most Movies/Series duration under 1000 min. Movies generally remain within the 60 to 120-minute range, irrespective of release year and most movies released in range between 2000 to 2020.

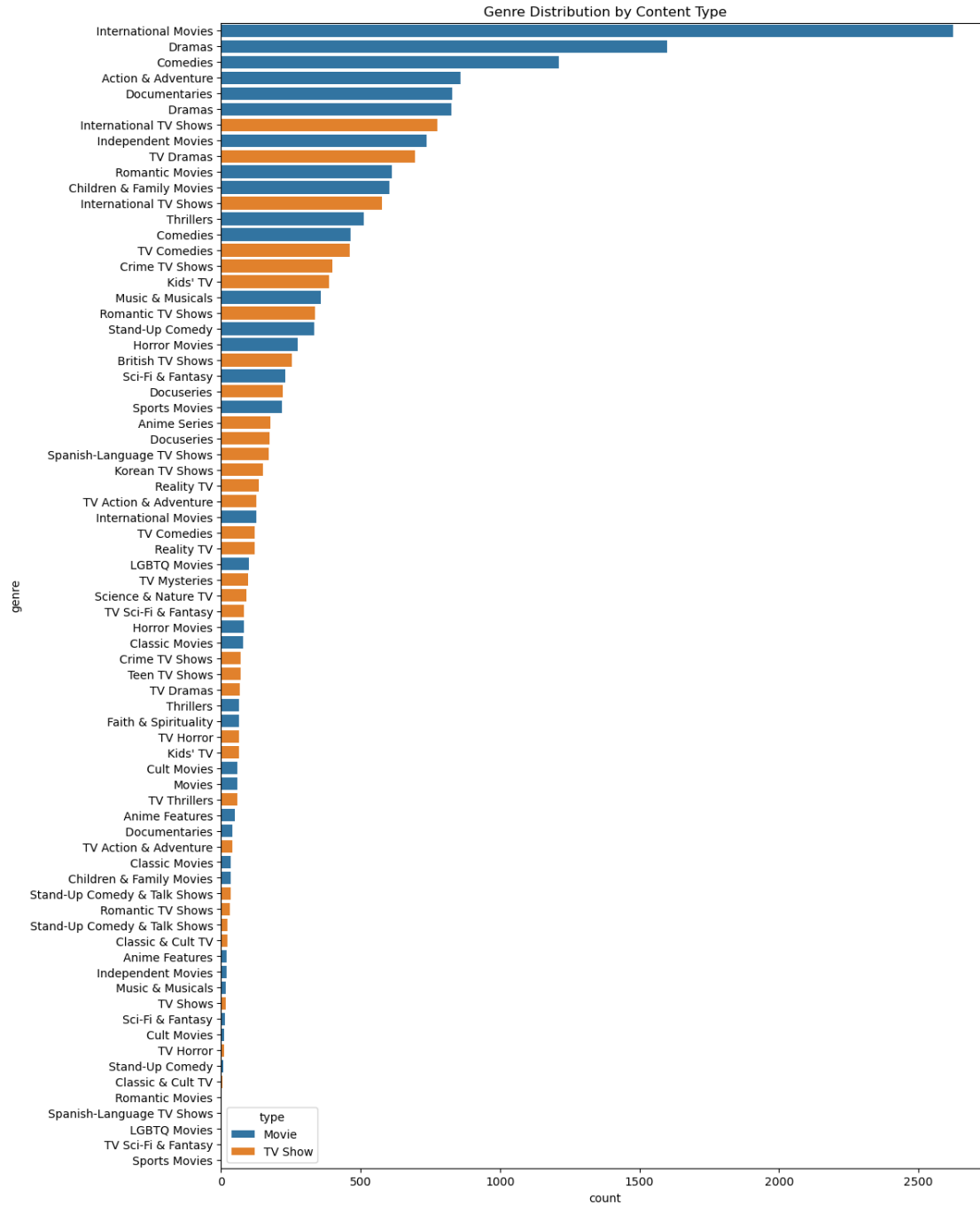
5.2 Analyzing the Growth of Movies and TV Shows Over the Years

```
[18]: plt.figure(figsize=(12, 6))
sns.countplot(x='release_year', hue='type', data=df, palette='viridis',
             edgecolor='black')
plt.title('Growth of Movies and TV Shows Over the Years')
plt.xticks(rotation=45)
plt.show()
```



5.3 Distribution of Genres (Movies vs TV Shows)

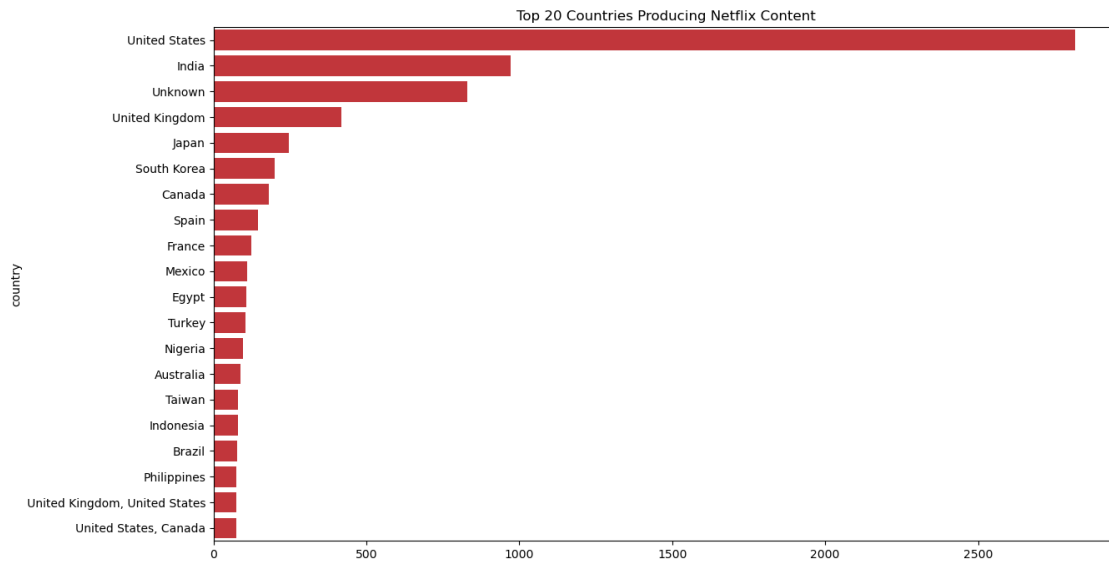
```
[19]: df_exploded = df.assign(genre=df['listed_in'].str.split(',')).explode('genre')
plt.figure(figsize=(12, 18))
sns.countplot(y='genre', hue='type', data=df_exploded,
              order=df_exploded['genre'].value_counts().index)
plt.title('Genre Distribution by Content Type')
plt.show()
```



5.4 Distribution of Content by Country

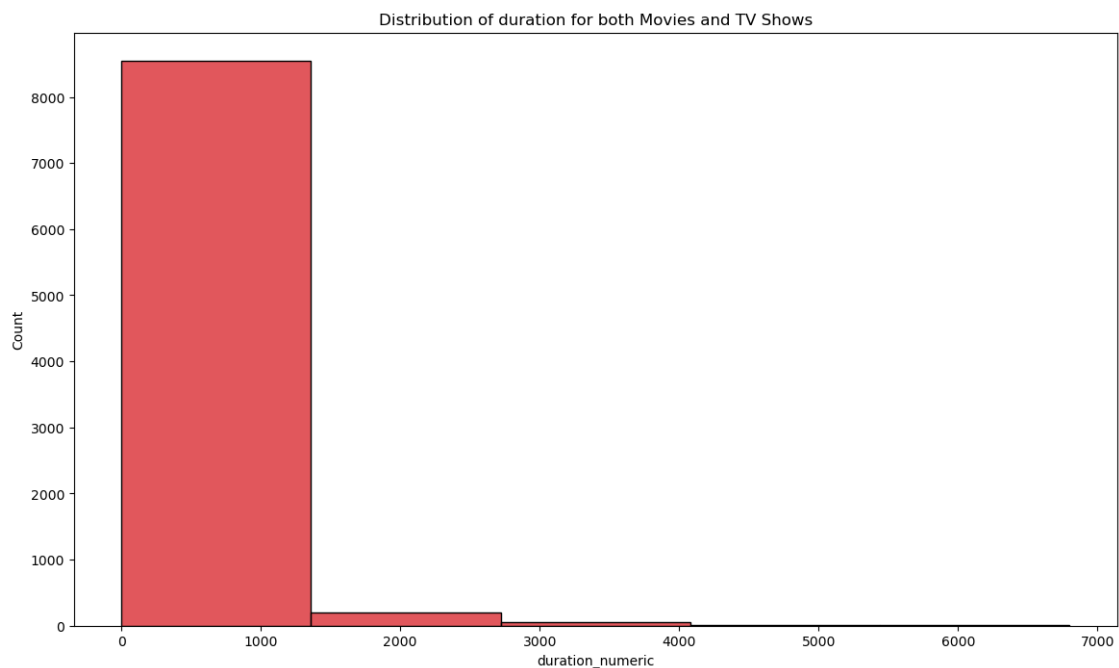
```
[20]: plt.figure(figsize=(14, 8))
df_country = df['country'].value_counts().head(20)
sns.barplot(x=df_country.values, y=df_country.index,color='#D81F26')
plt.title('Top 20 Countries Producing Netflix Content')
```

```
plt.show()
```



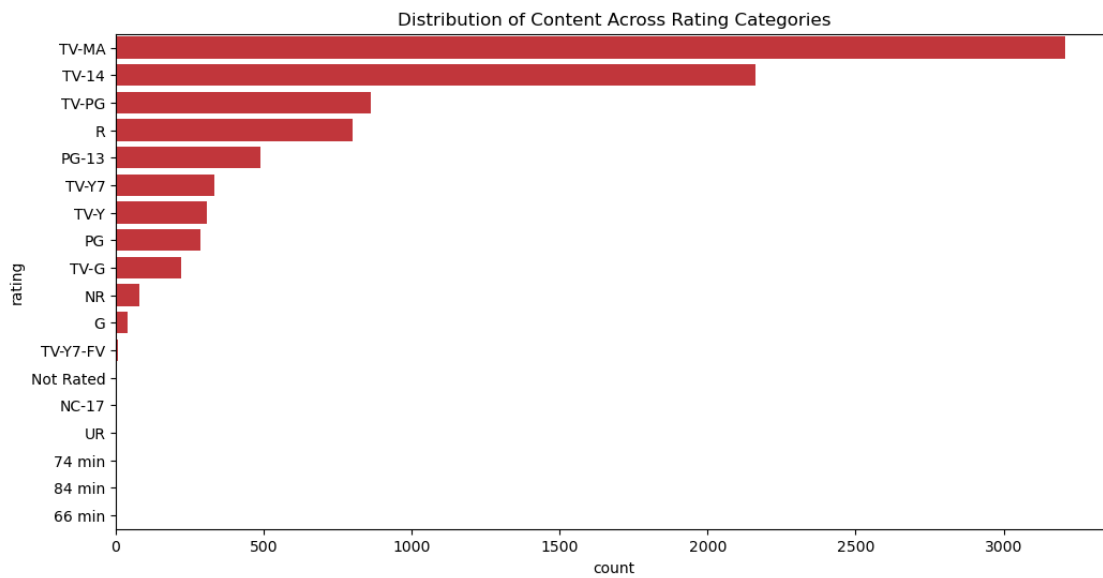
5.5 Distribution of duration for both Movies and TV Shows

```
[21]: plt.figure(figsize=(14, 8))
sns.histplot(df['duration_numeric'], bins=5, color='#D81F26')
plt.title("Distribution of duration for both Movies and TV Shows")
plt.show()
```



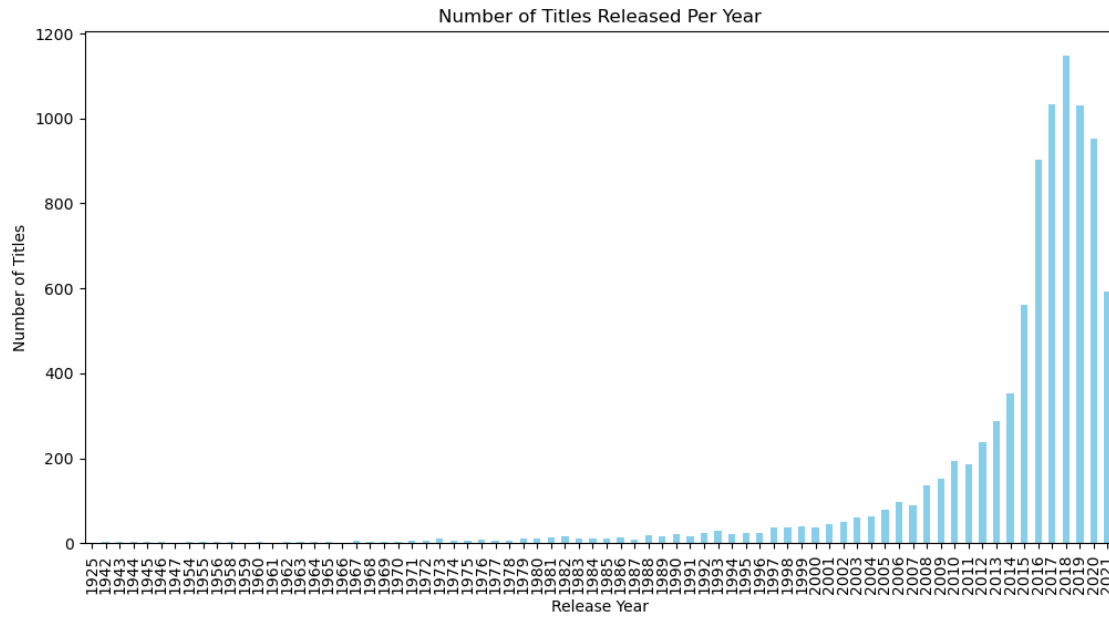
5.6 Rating Distribution

```
[22]: plt.figure(figsize=(12, 6))
sns.countplot(y='rating', data=df, order=df['rating'].value_counts().index,
             color='#D81F26')
plt.title('Distribution of Content Across Rating Categories')
plt.show()
```



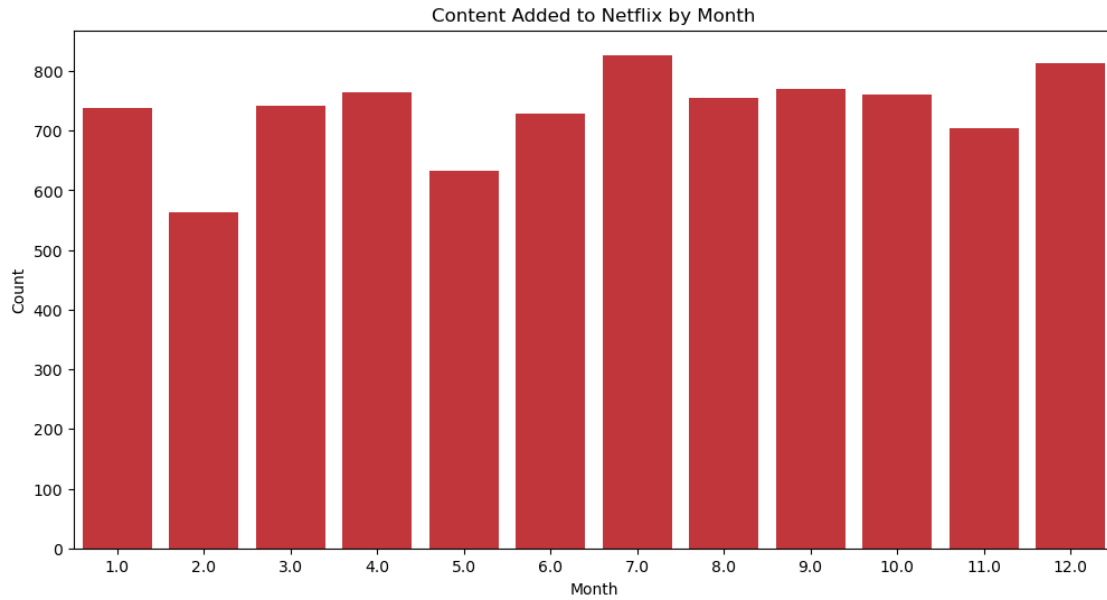
5.7 Number of Titles Released Per Year

```
[23]: plt.figure(figsize=(12, 6))
df_release = df['release_year'].value_counts().sort_index()
df_release.plot(kind='bar', color='skyblue')
plt.title('Number of Titles Released Per Year')
plt.xlabel('Release Year')
plt.ylabel('Number of Titles')
plt.show()
```



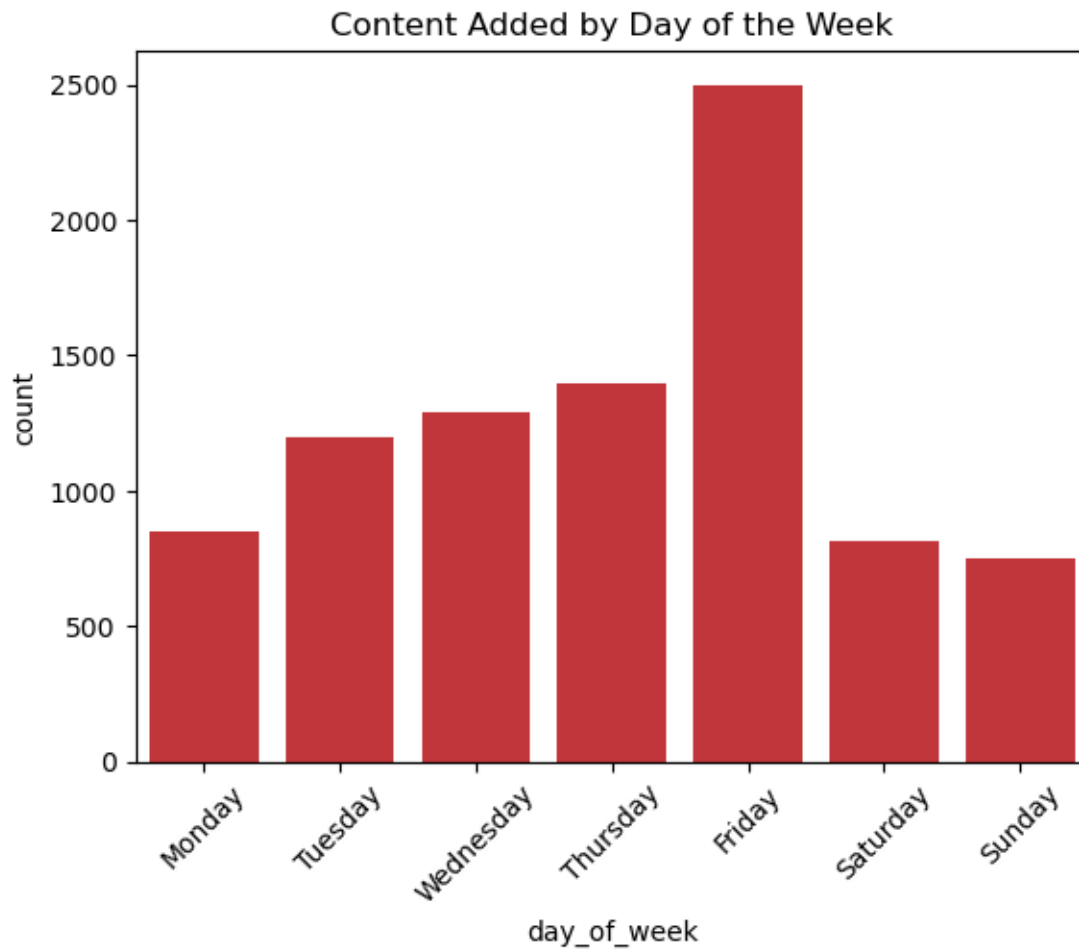
5.8 Content Added to Netflix by Month

```
[24]: df['date_added'] = pd.to_datetime(df['date_added'])
df['month_added'] = df['date_added'].dt.month
plt.figure(figsize=(12, 6))
sns.countplot(x='month_added', data=df, color='#D81F26')
plt.title('Content Added to Netflix by Month')
plt.xlabel('Month')
plt.ylabel('Count')
plt.show()
```



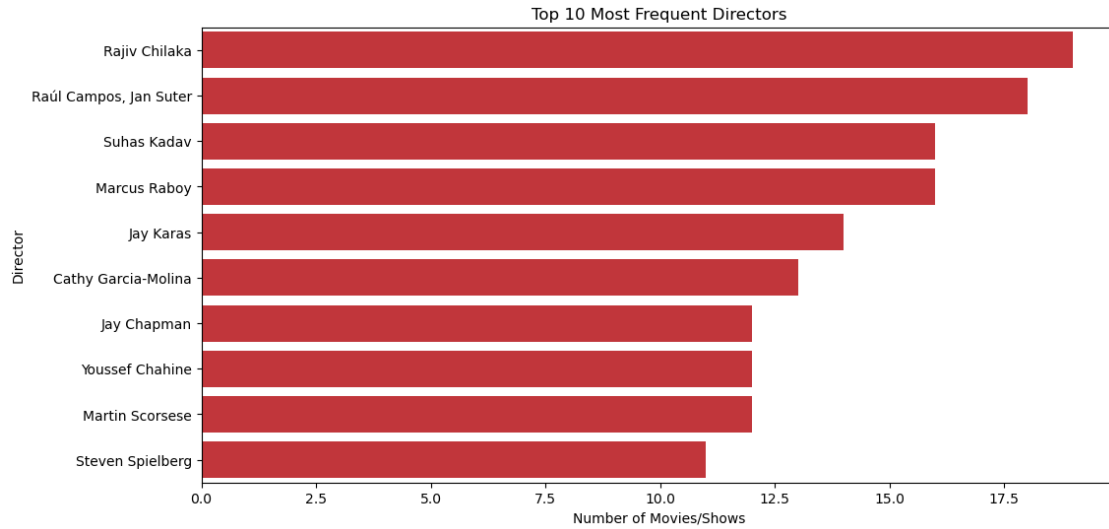
5.9 Day of the Week Analysis

```
[25]: df['day_of_week'] = df['date_added'].dt.day_name()
sns.countplot(x='day_of_week', data=df, order=['Monday', 'Tuesday', 'Wednesday', 'Thursday', 'Friday', 'Saturday', 'Sunday'], color='#D81F26')
plt.title('Content Added by Day of the Week')
plt.xticks(rotation=45)
plt.show()
```

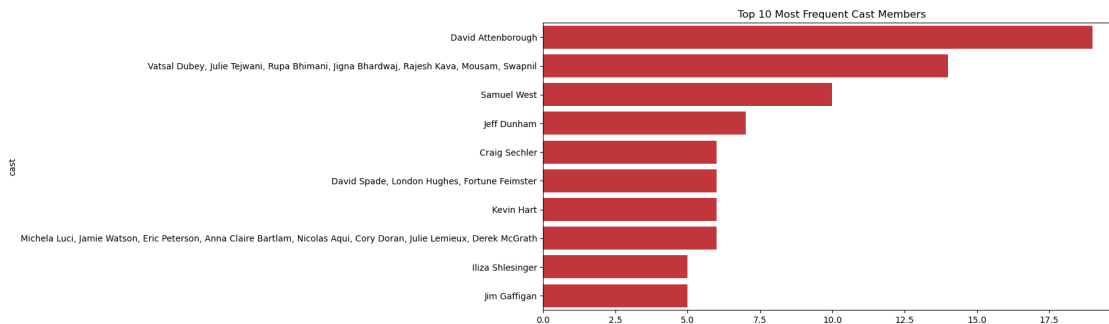
5.10 Most Frequent Directors

```
[26]: directors = df['director'].value_counts().iloc[1:11]
plt.figure(figsize=(12, 6))
sns.barplot(x=directors.values, y=directors.index,color='#D81F26')
plt.title('Top 10 Most Frequent Directors')
plt.xlabel('Number of Movies/Shows')
plt.ylabel('Director')
plt.show()
```



5.11 Frequent Cast Members

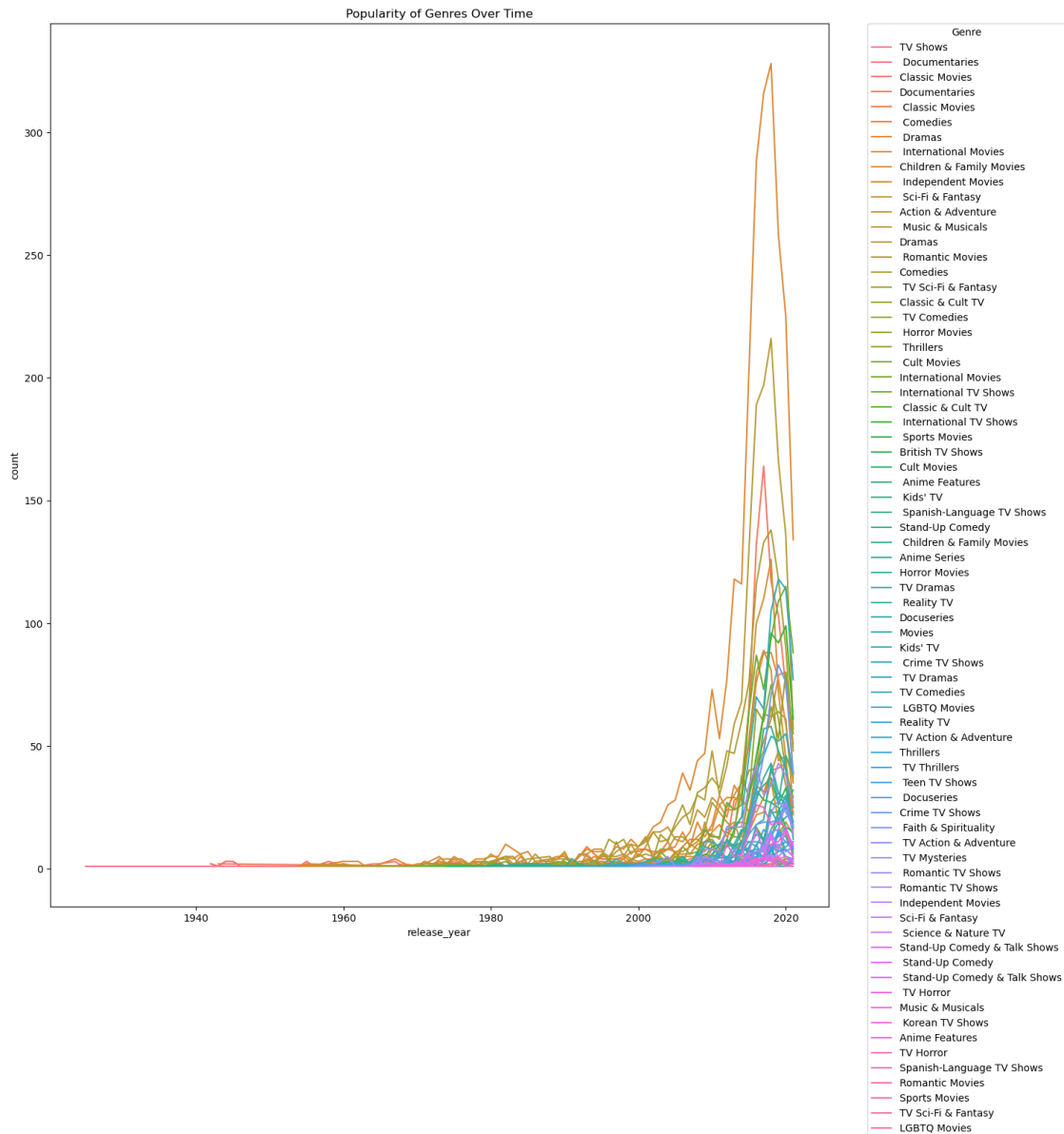
```
[27]: cast= df['cast'].value_counts().iloc[1:11]
plt.figure(figsize=(12, 6))
sns.barplot(x=cast.values, y=cast.index,color='#D81F26')
plt.title('Top 10 Most Frequent Cast Members')
plt.show()
```



5.12 Popularity of Genres Over Time

```
[28]: plt.figure(figsize=(14, 16))
genre_trend = df_exploded.groupby(['release_year', 'genre']).size().
    ↪reset_index(name='count')
sns.lineplot(x='release_year', y='count', hue='genre', data=genre_trend)
plt.legend(title="Genre", bbox_to_anchor=(1.05, 1), loc='upper left',
    ↪borderaxespad=0)
```

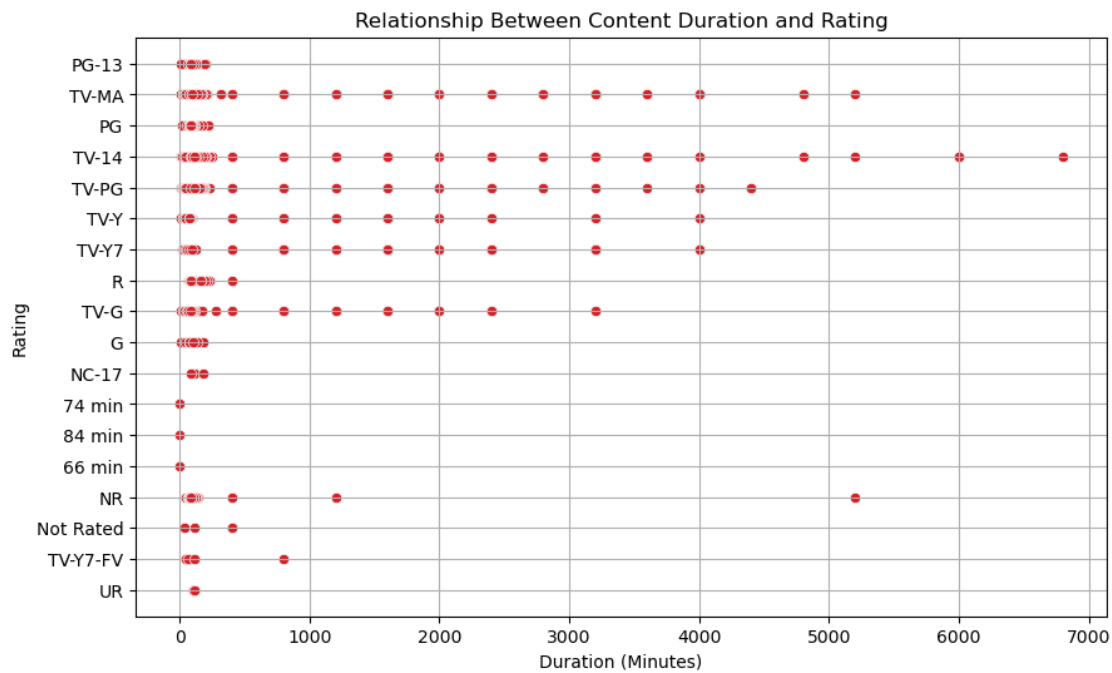
```
plt.title('Popularity of Genres Over Time')
plt.show()
```



5.13 Relationship Between Content Duration and Rating

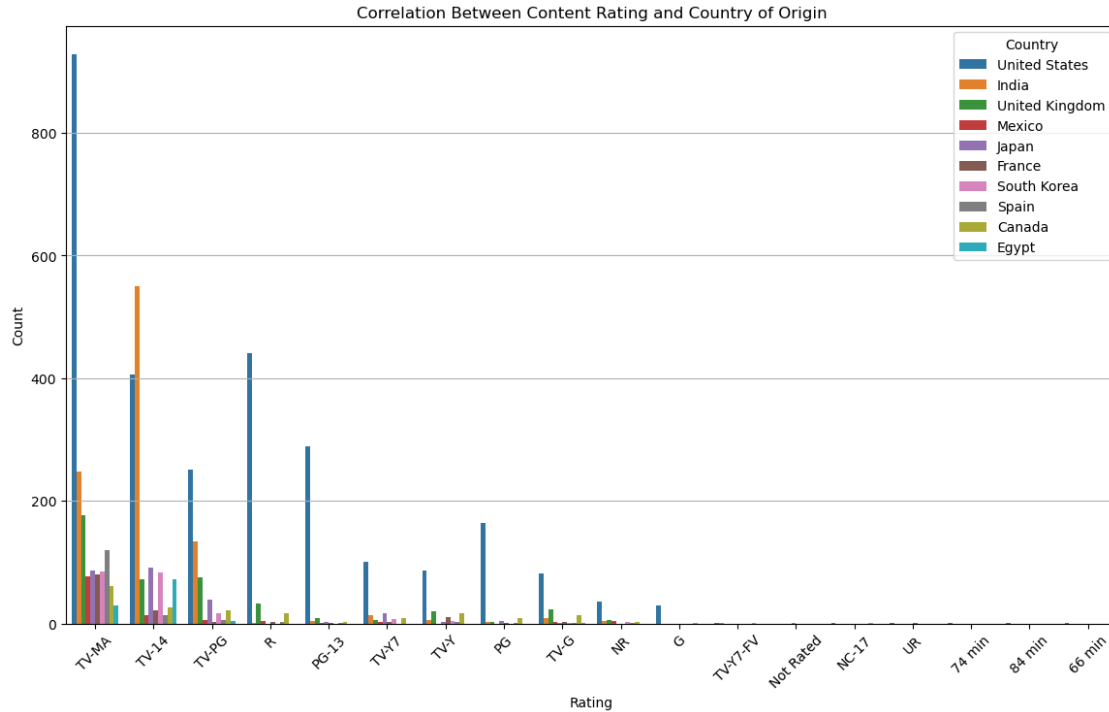
```
[29]: plt.figure(figsize=(10, 6))
sns.scatterplot(x='duration_numeric', y='rating', data=df, color='#D81F26')
plt.title('Relationship Between Content Duration and Rating')
plt.xlabel('Duration (Minutes)')
plt.ylabel('Rating')
```

```
plt.grid(True)
plt.show()
```



5.14 Plotting the relationship between countries and content ratings

```
[30]: top_countries_rating = df[df['country'] != 'Unknown']['country'].value_counts().
      ↪head(10).index
filtered_data = df[df['country'].isin(top_countries_rating)]
plt.figure(figsize=(14, 8))
sns.countplot(data=filtered_data, x='rating', hue='country', order=df['rating'].
      ↪value_counts().index)
plt.title('Correlation Between Content Rating and Country of Origin')
plt.xlabel('Rating')
plt.ylabel('Count')
plt.legend(title='Country')
plt.xticks(rotation=45)
plt.grid(axis='y')
plt.show()
```

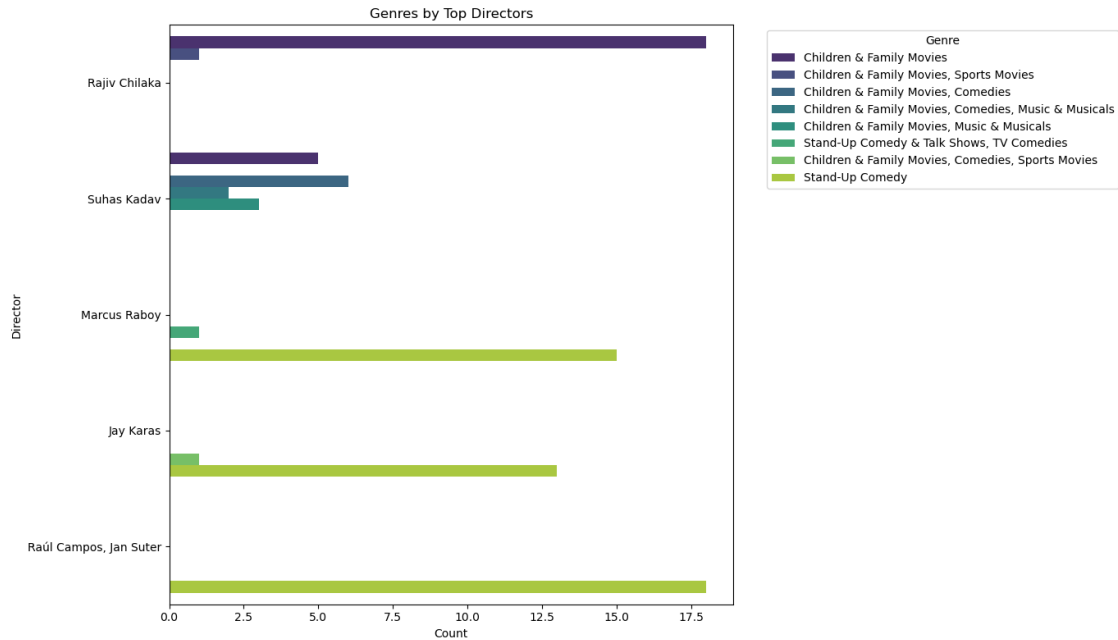


5.15 Extract top directors with count of genre

```
[31]: # Get the top 10 directors
top_directors = df[df['director'] != 'Unknown']['director'].value_counts().
    head(5).index

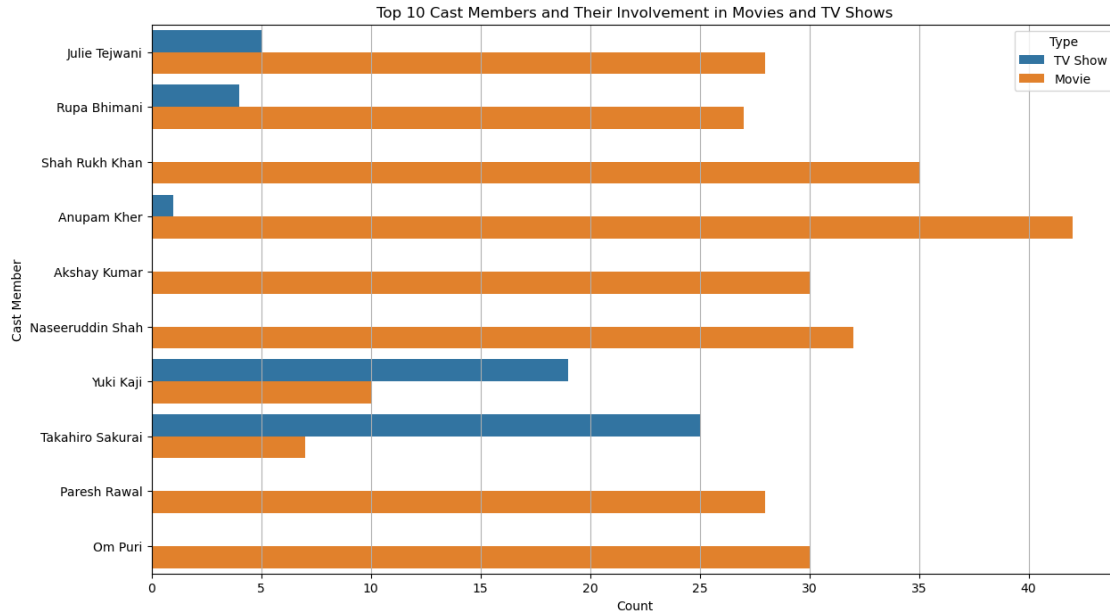
# Filter data for those directors
director_data = df[df['director'].isin(top_directors)].explode('listed_in')

plt.figure(figsize=(14, 8))
sns.countplot(data=director_data, y='director', hue='listed_in',
    palette='viridis')
plt.title('Genres by Top Directors')
plt.xlabel('Count')
plt.ylabel('Director')
plt.legend(title='Genre', bbox_to_anchor=(1.05, 1), loc='upper left')
plt.tight_layout()
plt.show()
```



5.16 Extract top cast members

```
[32]: df_exploded_cast = df.copy()
df_exploded_cast['cast'] = df_exploded_cast['cast'].str.split(',')
df_exploded_cast = df_exploded_cast.explode('cast')
top_cast = df_exploded_cast[df_exploded_cast['cast'] != 'Unknown']['cast'].
    ↪value_counts().head(10).index
cast_data = df_exploded_cast[df_exploded_cast['cast'].isin(top_cast)]
plt.figure(figsize=(14, 8))
sns.countplot(data=cast_data, y='cast', hue='type')
plt.title('Top 10 Cast Members and Their Involvement in Movies and TV Shows')
plt.xlabel('Count')
plt.ylabel('Cast Member')
plt.legend(title='Type')
plt.grid(axis='x')
plt.show()
```



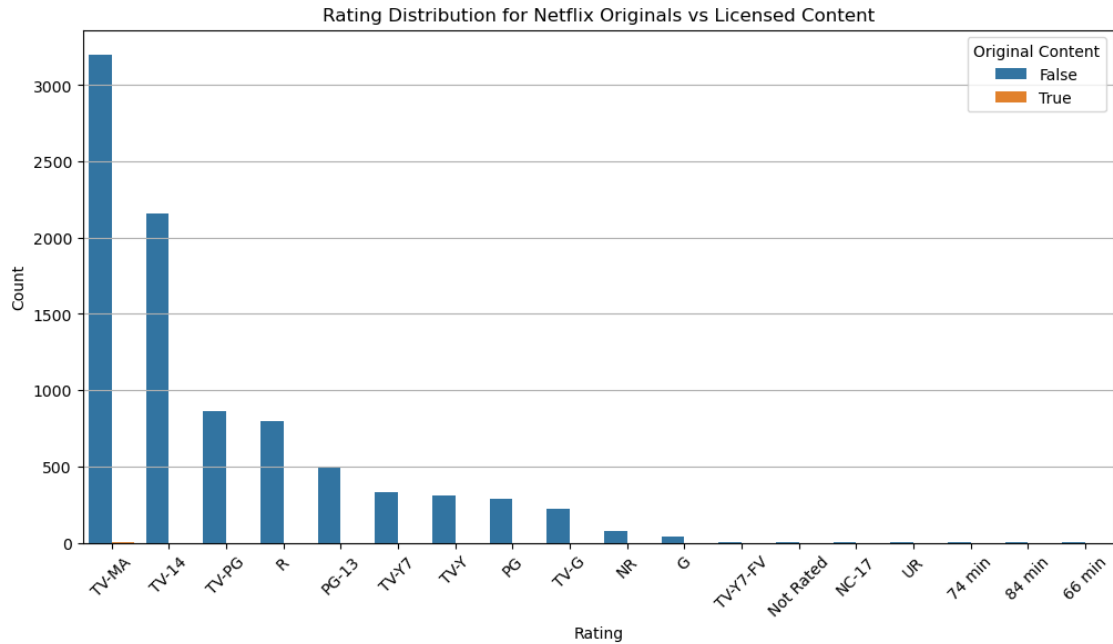
5.17 Generate word cloud for recent titles

```
[33]: from wordcloud import WordCloud
recent_titles = ' '.join(df[df['release_year'] >= 2015]['title'].dropna())
wordcloud_recent = WordCloud(width=800, height=400, background_color='black').
    generate(recent_titles)
plt.figure(figsize=(10, 6))
plt.imshow(wordcloud_recent, interpolation='bilinear')
plt.axis('off')
plt.title('Word Cloud of Recent Netflix Titles (2015 - 2021)')
plt.show()
```

5.18 Netflix Originals vs Licensed Content

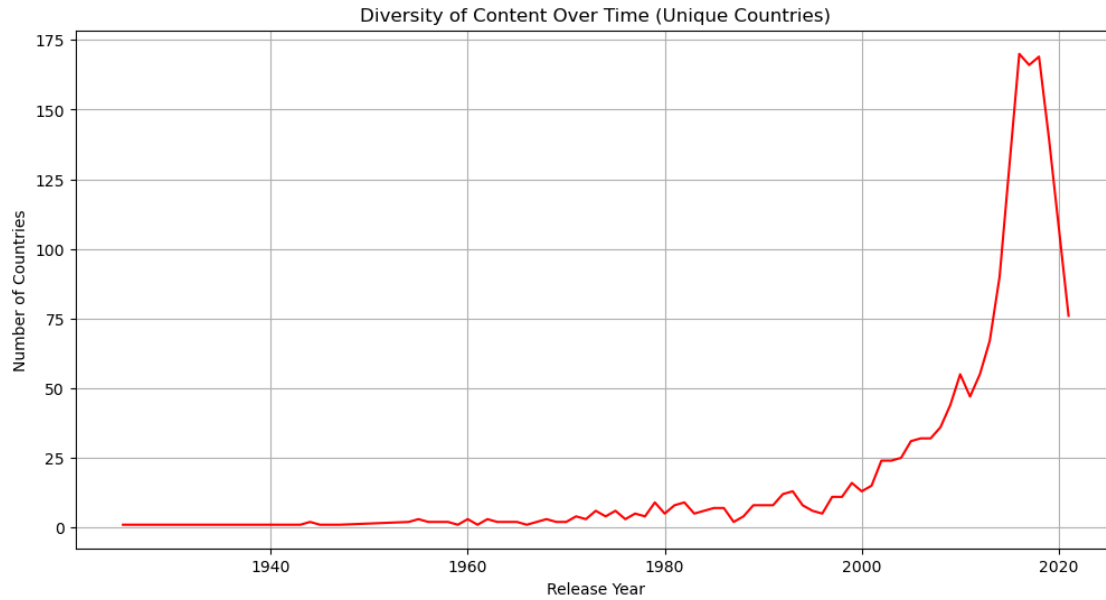
```
[34]: df['is_original'] = df['title'].str.contains('Netflix')

plt.figure(figsize=(12, 6))
sns.countplot(data=df, x='rating', hue='is_original', order=df['rating'].
    ↪value_counts().index)
plt.title('Rating Distribution for Netflix Originals vs Licensed Content')
plt.xlabel('Rating')
plt.ylabel('Count')
plt.legend(title='Original Content')
plt.xticks(rotation=45)
plt.grid(axis='y')
plt.show()
```

5.19 Diversity of Content Over Time

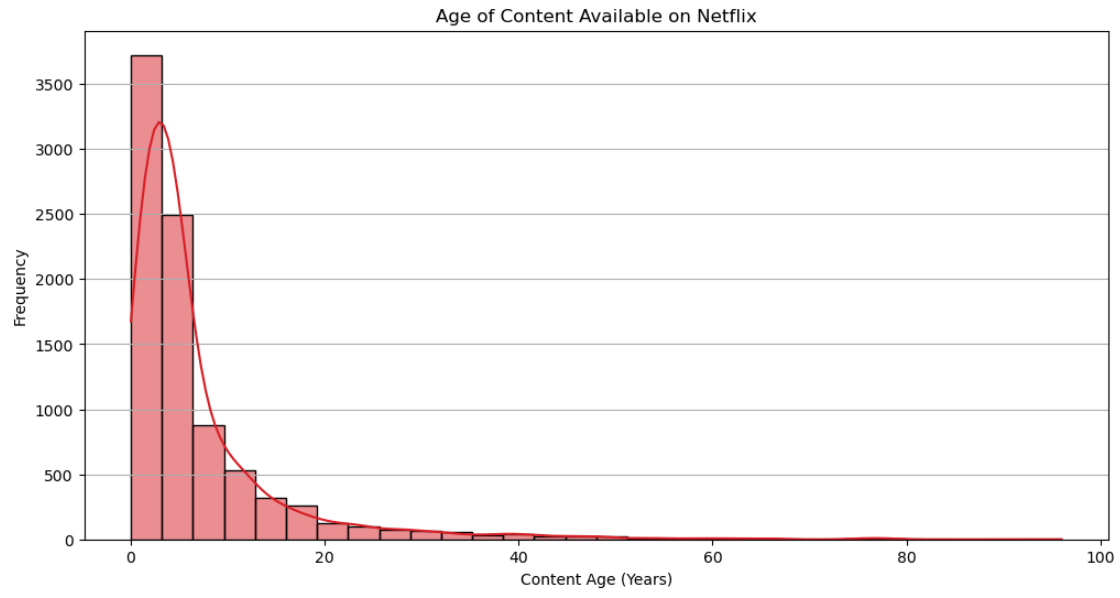
```
[35]: country_year = df.groupby('release_year')['country'].nunique()
plt.figure(figsize=(12, 6))
country_year.plot(kind='line', color='red')
plt.title('Diversity of Content Over Time (Unique Countries)')
plt.xlabel('Release Year')
plt.ylabel('Number of Countries')
plt.grid(True)
plt.show()
```



5.20 Correlation Between Age of Content and Popularity

```
[36]: df['content_age'] = 2021 - df['release_year']

plt.figure(figsize=(12, 6))
sns.histplot(df['content_age'], bins=30, kde=True, color='#D81F26')
plt.title('Age of Content Available on Netflix')
plt.xlabel('Content Age (Years)')
plt.ylabel('Frequency')
plt.grid(axis='y')
plt.show()
```



[]: