

# netflix-case-study

November 8, 2023

## NETFLIX CASE STUDY

### 0.0.1 Problem Statement

“Netflix wants to improve its content recommendation system and make data-driven decisions to enhance user experience and engagement.” ##### Basic Metrics 1. **Content Distribution:** - Movies vs. TV Shows: Analyze the distribution of content between movies and TV shows. Calculate the percentage of movies and TV shows in the dataset. 2. **Country of Production:** - Explore the countries where the movies and TV shows were produced. Identify the top countries with the most content in the Netflix library. 3. **Release Year Analysis:** - Analyze the distribution of content by release year. Identify the most common release years and trends in the dataset. 4. **TV Ratings:** - Examine the distribution of TV ratings for the content. Calculate the frequency of each TV rating category (e.g., G, PG, TV-MA) in the dataset. 5. **Content Duration:** - For movies, calculate the distribution of movie durations in minutes (e.g., 90 minutes, 120 minutes). For TV shows, calculate the distribution of the number of seasons. 6. **Genre Analysis:** - Analyze the distribution of content across different genres listed in the “Listed\_in” column. Identify the most popular genres on Netflix. 7. **Content Added Over Time:** - Examine how content has been added to Netflix over time. Create a time series analysis of content additions by year and month. 8. **Director and Cast Analysis:** - Identify the most prolific directors and actors/actresses in the Netflix library. Analyze their contributions to the platform. 9. **Content Descriptions:** - Perform text analysis on the descriptions to identify common keywords or themes. This could help with content tagging and recommendation.

### 0.0.2 Importing the necessary libraries

```
[61]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings("ignore")
```

### 0.0.3 Data Reading,

```
[62]: data=pd.read_csv("/content/netflix.csv")
```

```
[63]: # Shape of the Data:
print("Shape of the data : ", data.shape)
```

Shape of the data : (8807, 12)

```
[64]: # Data types
data_types = data.dtypes
print("Data types :")
print(data_types)
```

```
Data types :
show_id      object
type         object
title        object
director     object
cast         object
country      object
date_added   object
release_year  int64
rating       object
duration     object
listed_in    object
description  object
dtype: object
```

```
[65]: # Conversion of Categorical Attributes
categorical_attributes = ['type', 'country', 'rating', 'listed_in']
for category in categorical_attributes:
    data[category] = data[category].astype('category')
```

```
[66]: # Missing values detection
missing_values = data.isnull().sum()
print("Missing Value Detection:")
print(missing_values)
```

```
Missing Value Detection:
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
```

```
[67]: # Statistical Summary
# 1. Numerical summary : Summary of all numerical columns
numerical_summary = data.describe()

numerical_summary

categorical_summary = data[catergorical_attributes].describe()

print('categorical_summary: ')
categorical_summary
```

categorical\_summary:

```
[67]:
```

	type	country	rating	listed_in
count	8807	7976	8803	8807
unique	2	748	17	514
top	Movie	United States	TV-MA	Dramas, International Movies
freq	6131	2818	3207	362

```
[68]: # Value counts for every categorical attribute
type_count = data['type'].value_counts()
print(f"Type Count: {type_count}")
```

```
Type Count: Movie      6131
TV Show      2676
Name: type, dtype: int64
```

```
[69]: country_count = data['country'].value_counts().head(5)
print(f"Country count : {country_count}")
```

```
Country count : United States      2818
India              972
United Kingdom     419
Japan              245
South Korea        199
Name: country, dtype: int64
```

```
[70]: rating_count = data['rating'].value_counts()
print(f"Rating count {rating_count}")
```

```
Rating count 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
G             41
TV-Y7-FV      6
UR            3
NC-17         3
74 min        1
84 min        1
66 min        1
Name: rating, dtype: int64

```

```
[71]: listed_in_count=data["listed_in"].value_counts().head(5)
print(f"listed_in_count: {listed_in_count}")
```

```

listed_in_count: Dramas, International Movies          362
Documentaries                                         359
Stand-Up Comedy                                       334
Comedies, Dramas, International Movies               274
Dramas, Independent Movies, International Movies     252
Name: listed_in, dtype: int64

```

```
[72]: # unique values for Categorical attributes
type_unique=data["type"].nunique()
print(f"Unique type values : {type_unique}")
```

```
Unique type values : 2
```

```
[73]: rating_unique=data["rating"].nunique()
print(f'Unique rating values : {rating_unique}')
```

```
Unique rating values : 17
```

```
[74]: listed_in_unique=data["listed_in"].nunique()
print(f"Unique Listed_in values : {listed_in_unique}")
```

```
Unique Listed_in values : 514
```

**value\_counts()** is used to calculate the frequency of each unique value in the 'Type' column, which indicates whether a title is a movie or TV show.

**unique()** is used to get an array of unique values in the 'Rating' column, providing a list of different content ratings available.

**nunique()** is used to count the number of unique values in the 'Country' column, indicating how many different countries are represented in the dataset.

```
[75]: data_copy = data.copy()
```

## 0.0.4 Pre-processing the data

```
[76]: # Cast
constraint=data_copy['cast'].apply(lambda x: str(x).split(',')).tolist()
data_new=pd.DataFrame(data=constraint,index=data["title"])
data_new=data_new.stack()
data_new=pd.DataFrame(data_new)
data_new.reset_index(inplace=True)
data_new=data_new[["title",0]]
data_new

data_new.columns=['title','cast']
data_new
data_cast=data_copy.merge(data_new,on='title',how='inner')
data_cast.sample(10)
```

```
[76]:
```

	show_id	type	title \
21362	s2692	TV Show	Brews Brothers
49520	s6694	TV Show	Encerrados
14279	s1771	Movie	Wheels of Fortune
19325	s2421	Movie	Choked: Paisa Bolta Hai
13237	s1634	Movie	The 2nd
17667	s2196	Movie	It's Her Day
41678	s5562	Movie	The Most Hated Woman in America
50640	s6851	Movie	Ghost Whispers
55677	s7545	Movie	My Schoolmate, the Barbarian
57119	s7772	TV Show	Power Rangers RPM

	director \
21362	NaN
49520	NaN
14279	Shaun Paul Piccinino
19325	Anurag Kashyap
13237	Brian Skiba
17667	Aniedi Anwah
41678	Tommy O'Haver
50640	Kazuchika Kise
55677	Siu-hung Chung, Wong Jing
57119	NaN

	cast_x	country \
21362	Alan Aisenberg, Mike Castle, Carmen Flood, Mar...	United States
49520	Luis Machín, Maria Merlino, Fabián Vena, Ana C...	NaN
14279	Matt Jones, Noureen DeWulf, John Ducey, Matty ...	United States
19325	Saiyami Kher, Roshan Mathew, Amruta Subhash, U...	India
13237	Ryan Phillippe, Casper Van Dien, Jack Griffo, ...	United States
17667	Bovi Ugboma, Amanda Ebeye, Shaffy Bello, Adunn...	Nigeria

41678	Melissa Leo, Josh Lucas, Vincent Kartheiser, J...	United States
50640	Maaya Sakamoto, Ikkyu Juku, Kenichirou Matsuda...	Japan
55677	Nicholas Tse, Stephen Fung, Joey Yung, Samuel ...	Hong Kong
57119	Eka Darville, Ari Boyland, Rose McIver, Milo C...	United States

	date_added	release_year	rating	duration	\
21362	April 10, 2020	2020	TV-MA	1 Season	
49520	June 18, 2018	2018	TV-MA	1 Season	
14279	November 1, 2020	2020	R	107 min	
19325	June 5, 2020	2020	TV-14	114 min	
13237	November 30, 2020	2020	TV-MA	93 min	
17667	July 29, 2020	2016	TV-MA	112 min	
41678	March 24, 2017	2017	TV-MA	93 min	
50640	February 24, 2019	2013	TV-14	56 min	
55677	August 16, 2018	2001	TV-14	90 min	
57119	January 1, 2016	2009	TV-Y7	1 Season	

	listed_in	\
21362	TV Comedies	
49520	International TV Shows, Spanish-Language TV Sh...	
14279	Comedies, Sports Movies	
19325	Dramas, International Movies	
13237	Action & Adventure	
17667	Comedies, International Movies, Romantic Movies	
41678	Dramas	
50640	Action & Adventure, Anime Features, Internatio...	
55677	Action & Adventure, Comedies, International Mo...	
57119	Kids' TV	

	description	cast_y
21362	Two rival brothers must work together to keep ...	Marques Ray
49520	This fictional series tells the stories of wha...	Ana Chalentano
14279	To claim a big inheritance, a down-on-his-luck...	Christina Moore
19325	A bank employee weighed down by her jobless hu...	Amruta Subhash
13237	A Secret Service agent is drawn into a terrori...	Samaire Armstrong
17667	After a man promises his fiancé a dream weddin...	Femi Durojaiye
41678	This drama follows the controversial life of o...	Vincent Kartheiser
50640	Freed of her responsibilities for the 501 Orga...	Kazuya Nakai
55677	A brainy private school student is transferred...	Samuel Pang
57119	The Power Rangers' new member, Dillon, isn't s...	Rose McIver

```
[77]: data_new.columns=['title','cast']
data_new
data_cast=data_copy.merge(data_new,on='title',how='inner')
data_cast
```

```

[77]:      show_id      type      title      director \
0          s1      Movie  Dick Johnson Is Dead  Kirsten Johnson
1          s2  TV Show      Blood & Water      NaN
2          s2  TV Show      Blood & Water      NaN
3          s2  TV Show      Blood & Water      NaN
4          s2  TV Show      Blood & Water      NaN
...
64946    s8807      Movie      Zubaan      Mozez Singh
64947    s8807      Movie      Zubaan      Mozez Singh
64948    s8807      Movie      Zubaan      Mozez Singh
64949    s8807      Movie      Zubaan      Mozez Singh
64950    s8807      Movie      Zubaan      Mozez Singh

                                cast_x      country \
0                                NaN  United States
1    Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban...  South Africa
2    Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban...  South Africa
3    Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban...  South Africa
4    Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban...  South Africa
...
64946  Vicky Kaushal, Sarah-Jane Dias, Raaghav Chanan...  India
64947  Vicky Kaushal, Sarah-Jane Dias, Raaghav Chanan...  India
64948  Vicky Kaushal, Sarah-Jane Dias, Raaghav Chanan...  India
64949  Vicky Kaushal, Sarah-Jane Dias, Raaghav Chanan...  India
64950  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  2 Seasons
3  September 24, 2021      2021  TV-MA  2 Seasons
4  September 24, 2021      2021  TV-MA  2 Seasons
...
64946      March 2, 2019      2015  TV-14    111 min
64947      March 2, 2019      2015  TV-14    111 min
64948      March 2, 2019      2015  TV-14    111 min
64949      March 2, 2019      2015  TV-14    111 min
64950      March 2, 2019      2015  TV-14    111 min

                                listed_in \
0                                Documentaries
1  International TV Shows, TV Dramas, TV Mysteries
2  International TV Shows, TV Dramas, TV Mysteries
3  International TV Shows, TV Dramas, TV Mysteries
4  International TV Shows, TV Dramas, TV Mysteries
...
64946  Dramas, International Movies, Music & Musicals

```

```

64947 Dramas, International Movies, Music & Musicals
64948 Dramas, International Movies, Music & Musicals
64949 Dramas, International Movies, Music & Musicals
64950 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      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...
...
64946 A scrappy but poor boy worms his way into a ty...
64947 A scrappy but poor boy worms his way into a ty...
64948 A scrappy but poor boy worms his way into a ty...
64949 A scrappy but poor boy worms his way into a ty...
64950 A scrappy but poor boy worms his way into a ty...

                                cast_y
0                                nan
1                                Ama Qamata
2                                Khosi Ngema
3                                Gail Mabalane
4                                Thabang Molaba
...
64946 Manish Chaudhary
64947 Meghna Malik
64948 Malkeet Rauni
64949 Anita Shabdish
64950 Chittaranjan Tripathy

[64951 rows x 13 columns]

```

```
[ ]:
```

```

[79]: # final pre-processed data
pre_Data = pd.read_csv('/content/Final_data.csv')

```

The data is pre-processed by splitting the ‘Actor,’ ‘Director,’ and ‘Country’ columns to create lists of values for each entry.

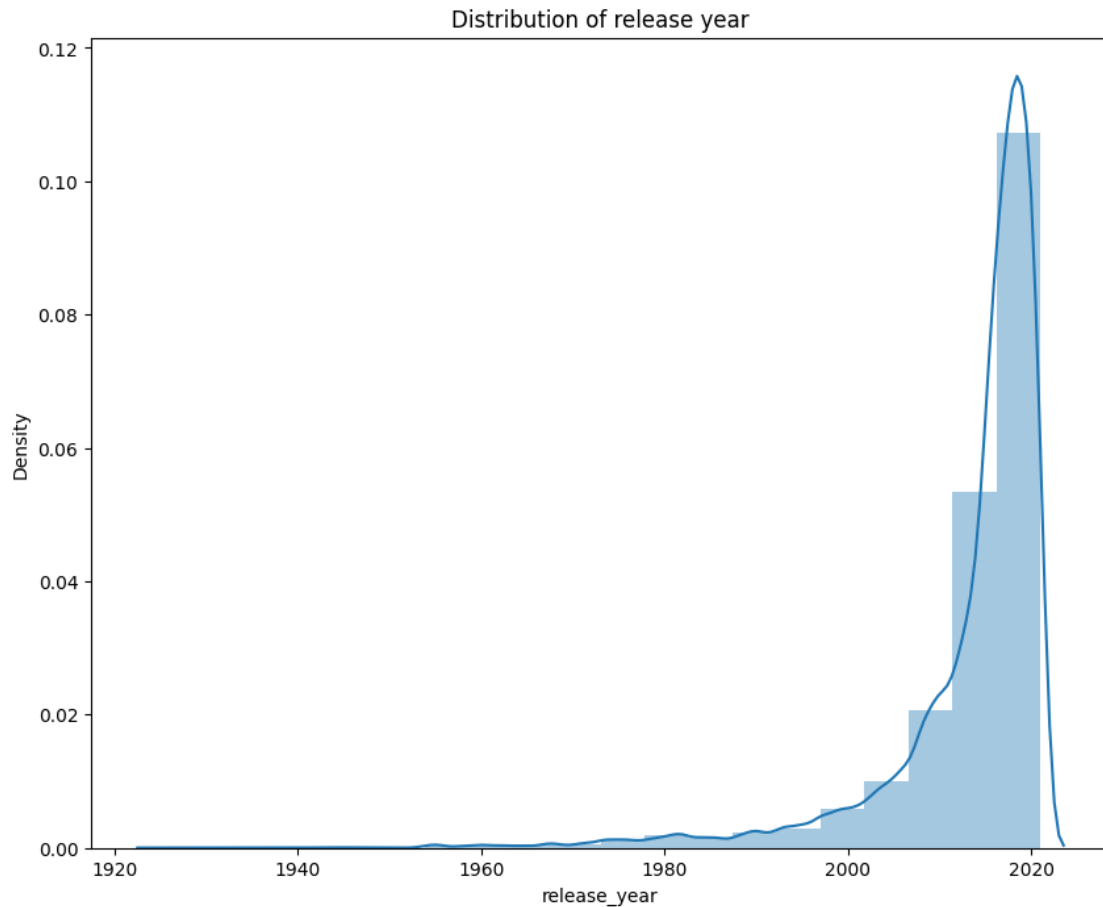
**Univariate analysis** is performed for the continuous variable ‘Release\_year’ using a distribution plot (distplot). Boxplots are created for the categorical variable ‘Rating’ to visualize its distribution.

**Bivariate analysis** is demonstrated by creating a count plot to visualize the relationship between ‘Type’ and ‘Rating.’

Note that the ‘Type’ variable is categorical, and the correlation analysis (heatmap and pairplot) is not applicable in this context, as these plots are typically used for continuous variables.

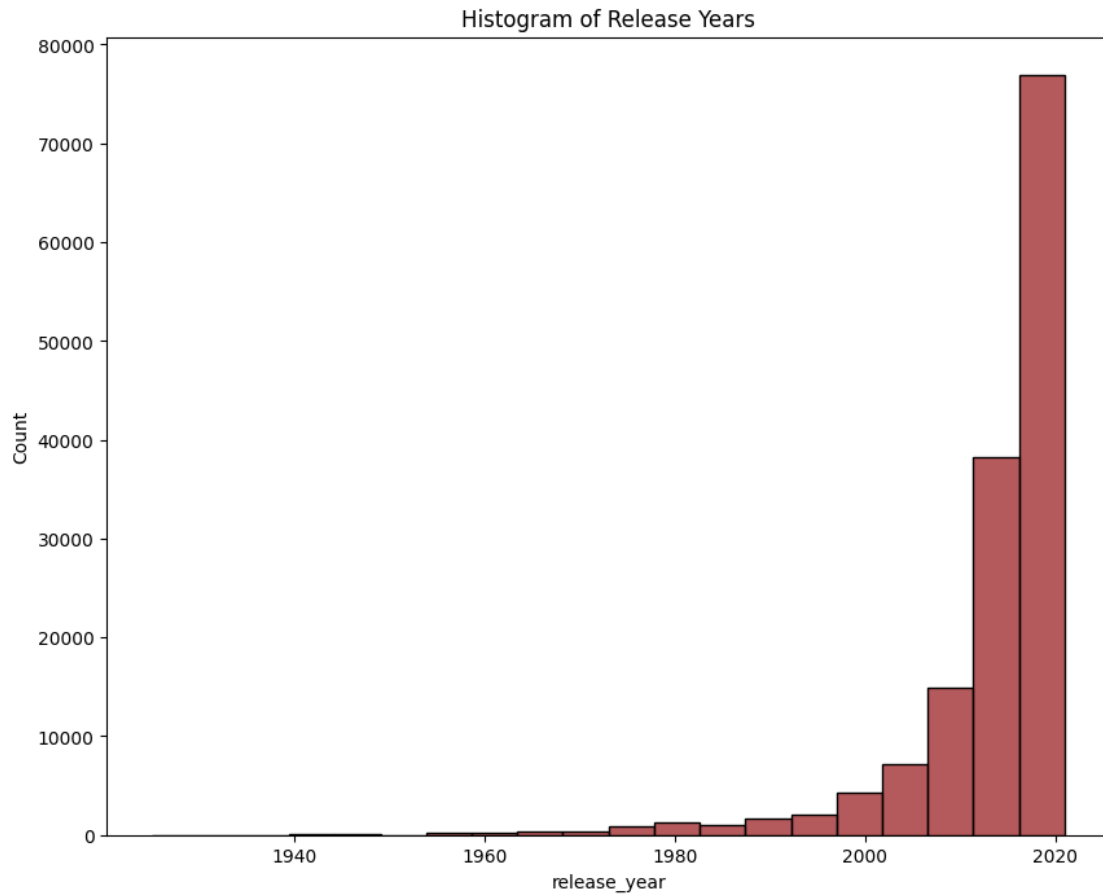


```
[80]: # univariate analysis for continuous variables
# continuous variables : 'release_year'
plt.figure(figsize=(10, 8))
sns.distplot(pre_Data['release_year'], bins = 20, kde=True)
plt.title('Distribution of release year')
plt.show()
```

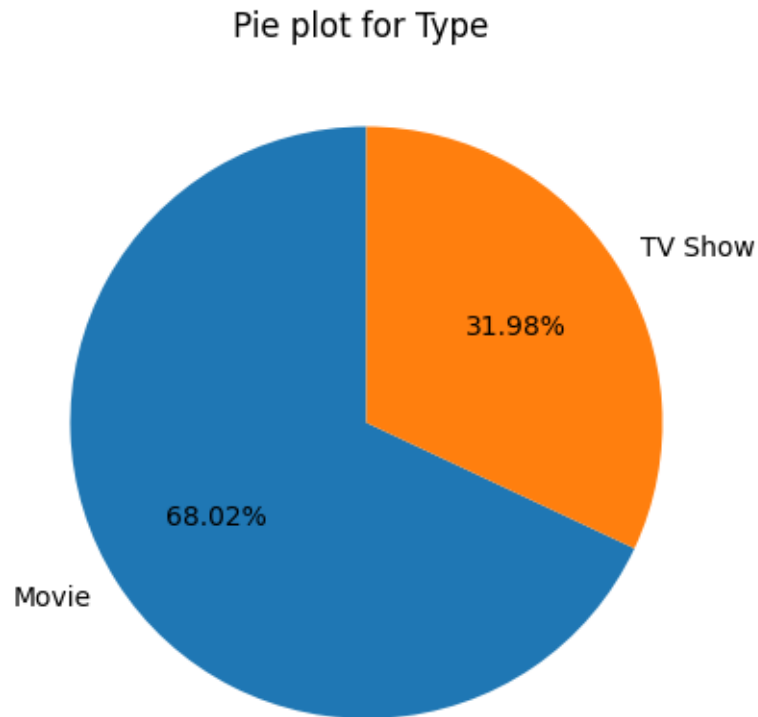


```
[81]: pre_Data['date_added'] = pd.to_datetime(pre_Data['date_added'])
```

```
[82]: # histogram for continuous variables
release_years = pre_Data['release_year']
plt.figure(figsize=(10, 8))
sns.histplot(release_years, bins = 20, color="#9B2226")
plt.title('Histogram of Release Years')
plt.show()
```



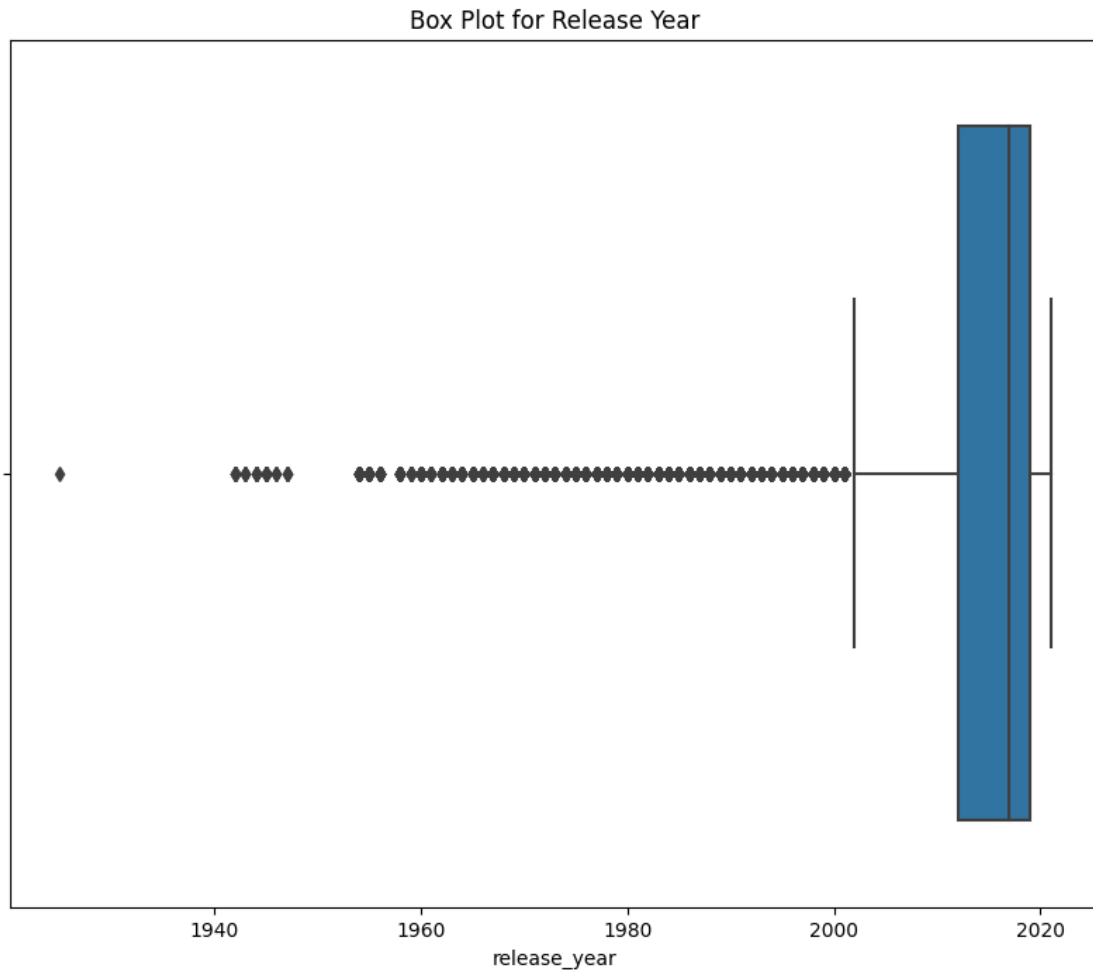
```
[124]: #PIE PLOT
plt.figure(figsize=(5, 5))
type_count=pre_Data["type"].value_counts()
plt.pie(type_count,labels=type_count.index,autopct="%.2f%",startangle=90)
plt.title("Pie plot for Type ")
plt.show()
```



In this example , we're using a pie plot to visualize the percentage of Movie and TV shows in the Netflix data

Checking for outliers often involves using descriptive statistics, data visualization, or statistical methods. **Box plots** and scatter plots can be helpful for visualizing outliers. Here's an example using a box plot to check for outliers in a numerical variable:

```
[116]: # Categorical variables
# 1.Boxplot
plt.figure(figsize=(10, 8))
sns.boxplot(x=pre_Data['release_year'])
plt.title('Box Plot for Release Year')
plt.show()
```

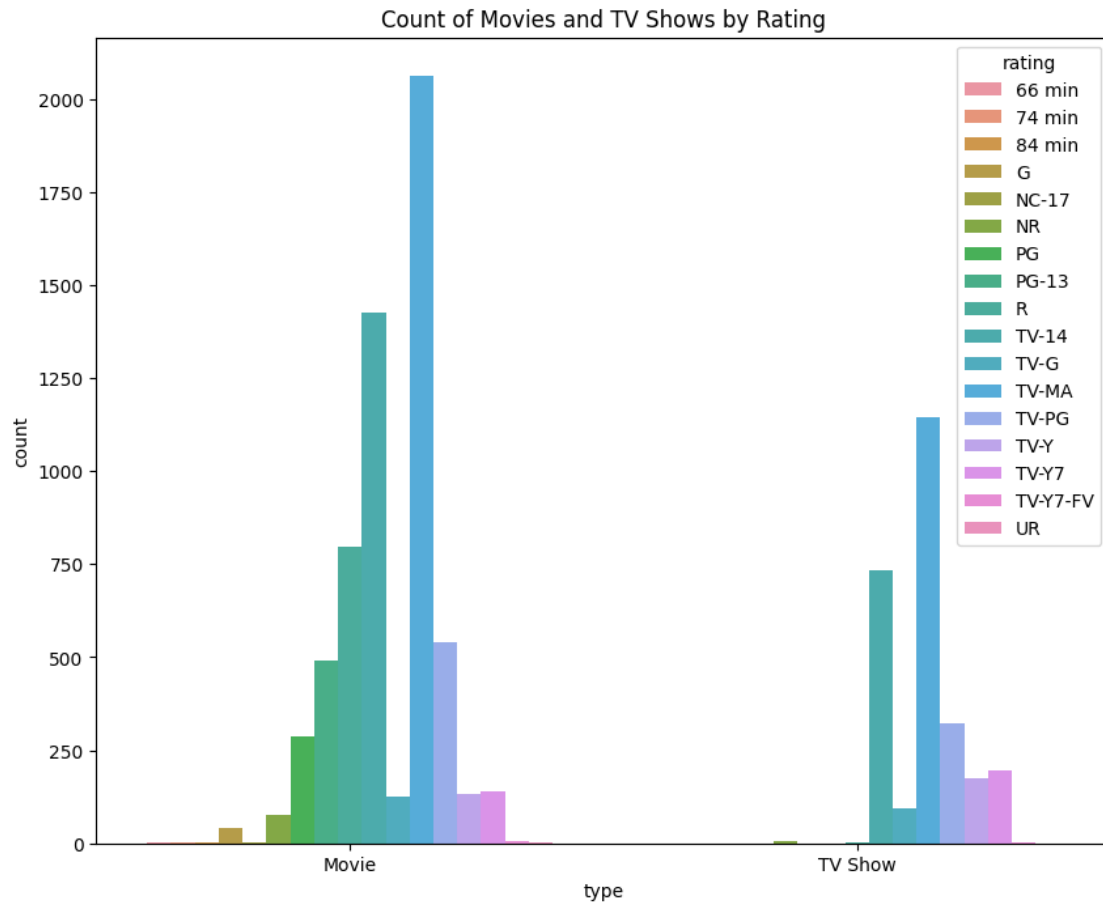


In this example, we use a box plot to visualize the distribution of the 'Release\_year' variable and identify potential outliers. Outliers are often represented as individual data points beyond the “whiskers” of the box plot.

Keep in mind that outlier treatment is context-dependent, and you may choose to remove outliers, transform the data, or apply other methods based on your specific analysis goals.

Remember that the treatment of missing values and outliers should be done carefully and in a way that is appropriate for your analysis and doesn't introduce bias or distort the data unnecessarily.

```
[84]: plt.figure(figsize=(10, 8))
sns.countplot(x = 'type', hue='rating', data = data_copy)
plt.title('Count of Movies and TV Shows by Rating')
plt.show()
```



In this example, we're using a **count plot** to visualize the relationship between the content type ('Type') and the content rating ('Rating'). The hue parameter is used to differentiate the counts for different content ratings within each content type.

```
[117]: data_copy.head()
```

```
[117]:  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

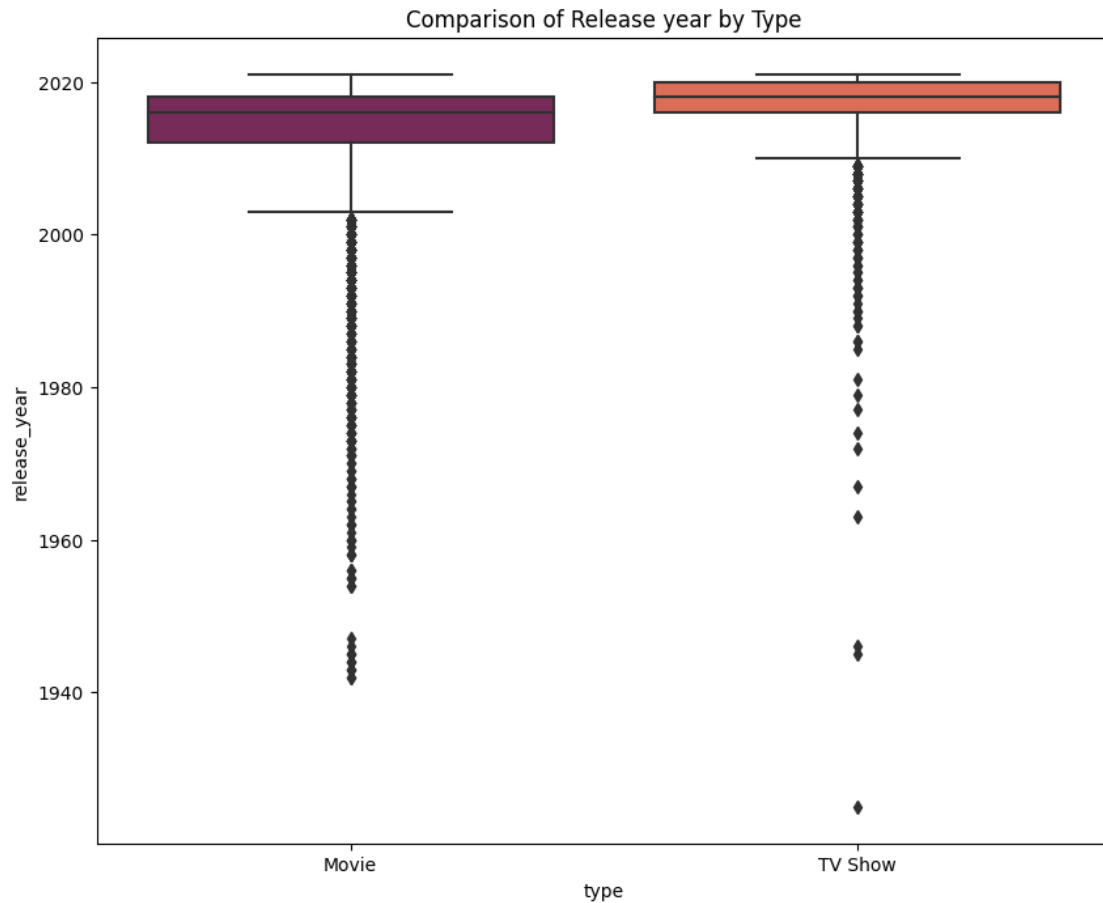
        cast    country \
0      NaN  United States
1  Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban...  South Africa
2  Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabi...    NaN
3      NaN    NaN
4  Mayur More, Jitendra Kumar, Ranjan Raj, Alam K...    India
```

	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...

```
[118]: plt.figure(figsize=(10, 8))
sns.boxplot(x = 'type', y='release_year', data = data_copy, palette='rocket')
plt.title('Comparison of Release year by Type')
plt.show()
```



In this example, we're using a box plot to compare the distribution of 'Release\_year' by 'Type.' This allows you to see how the release years differ between movies and TV shows.

```
[87]: pre_Data.head()
```

```
[87]:  show_id    type    title    director    country \
0      s1    Movie  Dick Johnson Is Dead  Kirsten Johnson  United States
1      s2  TV Show    Blood & Water    Unknown    South Africa
2      s2  TV Show    Blood & Water    Unknown    South Africa
3      s2  TV Show    Blood & Water    Unknown    South Africa
4      s2  TV Show    Blood & Water    Unknown    South Africa

    date_added  release_year  rating  duration \
0  2021-09-25         2020  PG-13    90 min
1  2021-09-24         2021  TV-MA  2 Seasons
2  2021-09-24         2021  TV-MA  2 Seasons
3  2021-09-24         2021  TV-MA  2 Seasons
4  2021-09-24         2021  TV-MA  2 Seasons
```

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

	cast_y	genre
0	NaN	Documentaries
1	Ama Qamata	International TV Shows
2	Ama Qamata	TV Dramas
3	Ama Qamata	TV Mysteries
4	Khosi Ngema	International TV Shows

```
[88]: d1=pre_Data[pre_Data["type"]=="Movie"]
constraint=d1['duration'].apply(lambda x: str(x).split(' ')).tolist()
data_new=pd.DataFrame(data=constraint,index=d1["title"])
data_new=pd.DataFrame(data_new)
data_new.reset_index(inplace=True)
data_new=data_new[["title",0]]
data_new
data_new.columns=['title','mins']
data_new
```

```
[88]:
```

	title	mins
0	Dick Johnson Is Dead	90
1	My Little Pony: A New Generation	91
2	My Little Pony: A New Generation	91
3	My Little Pony: A New Generation	91
4	My Little Pony: A New Generation	91
...	...	...
101687	Zubaan	111
101688	Zubaan	111
101689	Zubaan	111
101690	Zubaan	111
101691	Zubaan	111

[101692 rows x 2 columns]

```
[89]: dd=pre_Data.merge(data_new,on="title",how="left")
```

```
[90]: dd
```

```
[90]:
```

	show_id	type	title	director	\
0	s1	Movie	Dick Johnson Is Dead	Kirsten Johnson	
1	s2	TV Show	Blood & Water	Unknown	
2	s2	TV Show	Blood & Water	Unknown	



3	s2	TV Show	Blood & Water	Unknown
4	s2	TV Show	Blood & Water	Unknown
...	...	...	...	...
2482587	s8807	Movie	Zubaan	Mozez Singh
2482588	s8807	Movie	Zubaan	Mozez Singh
2482589	s8807	Movie	Zubaan	Mozez Singh
2482590	s8807	Movie	Zubaan	Mozez Singh
2482591	s8807	Movie	Zubaan	Mozez Singh

	country	date_added	release_year	rating	duration	\
0	United States	2021-09-25	2020	PG-13	90 min	
1	South Africa	2021-09-24	2021	TV-MA	2 Seasons	
2	South Africa	2021-09-24	2021	TV-MA	2 Seasons	
3	South Africa	2021-09-24	2021	TV-MA	2 Seasons	
4	South Africa	2021-09-24	2021	TV-MA	2 Seasons	
...	...	...	...	...	...	
2482587	India	2019-03-02	2015	TV-14	111 min	
2482588	India	2019-03-02	2015	TV-14	111 min	
2482589	India	2019-03-02	2015	TV-14	111 min	
2482590	India	2019-03-02	2015	TV-14	111 min	
2482591	India	2019-03-02	2015	TV-14	111 min	

	description	Month_added	\
0	As her father nears the end of his life, filmm...	September	
1	After crossing paths at a party, a Cape Town t...	September	
2	After crossing paths at a party, a Cape Town t...	September	
3	After crossing paths at a party, a Cape Town t...	September	
4	After crossing paths at a party, a Cape Town t...	September	
...	...	...	
2482587	A scrappy but poor boy worms his way into a ty...	March	
2482588	A scrappy but poor boy worms his way into a ty...	March	
2482589	A scrappy but poor boy worms his way into a ty...	March	
2482590	A scrappy but poor boy worms his way into a ty...	March	
2482591	A scrappy but poor boy worms his way into a ty...	March	

	Day_added	cast_y	genre	mins
0	Saturday	NaN	Documentaries	90
1	Friday	Ama Qamata	International TV Shows	NaN
2	Friday	Ama Qamata	TV Dramas	NaN
3	Friday	Ama Qamata	TV Mysteries	NaN
4	Friday	Khosi Ngema	International TV Shows	NaN
...	...	...	...	...
2482587	Saturday	Chittaranjan Tripathy	Music & Musicals	111
2482588	Saturday	Chittaranjan Tripathy	Music & Musicals	111
2482589	Saturday	Chittaranjan Tripathy	Music & Musicals	111
2482590	Saturday	Chittaranjan Tripathy	Music & Musicals	111
2482591	Saturday	Chittaranjan Tripathy	Music & Musicals	111

[2482592 rows x 15 columns]

```
[91]: d1=data[data["type"]=="TV Show"]
constraint=d1['duration'].apply(lambda x: str(x).split(' ')).tolist()
data_new=pd.DataFrame(data=constraint,index=d1["title"])
data_new=pd.DataFrame(data_new)
data_new.reset_index(inplace=True)
data_new=data_new[["title",0]]
data_new
data_new.columns=['title','season']
data_new
```

```
[91]:
```

	title	season
0	Blood & Water	2
1	Ganglands	1
2	Jailbirds New Orleans	1
3	Kota Factory	2
4	Midnight Mass	1
...	...	...
2671	Yu-Gi-Oh! Arc-V	2
2672	Yunus Emre	2
2673	Zak Storm	3
2674	Zindagi Gulzar Hai	1
2675	Zombie Dumb	2

[2676 rows x 2 columns]

```
[92]: pre_Data1=dd.merge(data_new,on="title",how="left")
```

```
[93]: pre_Data1.head()
```

```
[93]:
```

	show_id	type	title	director	country	\
0	s1	Movie	Dick Johnson Is Dead	Kirsten Johnson	United States	
1	s2	TV Show	Blood & Water	Unknown	South Africa	
2	s2	TV Show	Blood & Water	Unknown	South Africa	
3	s2	TV Show	Blood & Water	Unknown	South Africa	
4	s2	TV Show	Blood & Water	Unknown	South Africa	

	date_added	release_year	rating	duration	\
0	2021-09-25	2020	PG-13	90 min	
1	2021-09-24	2021	TV-MA	2 Seasons	
2	2021-09-24	2021	TV-MA	2 Seasons	
3	2021-09-24	2021	TV-MA	2 Seasons	
4	2021-09-24	2021	TV-MA	2 Seasons	

	description	Month_added	Day_added	\
--	-------------	-------------	-----------	---

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

	cast_y	genre	mins	season
0	NaN	Documentaries	90	NaN
1	Ama Qamata	International TV Shows	NaN	2
2	Ama Qamata	TV Dramas	NaN	2
3	Ama Qamata	TV Mysteries	NaN	2
4	Khosi Ngema	International TV Shows	NaN	2

```
[94]: pre_Data1["mins"]=pre_Data1["mins"].astype(float)
pre_Data1["mins"].fillna(pre_Data1["mins"].mean()).astype(int)
```

```
[94]: 0          90
      1         109
      2         109
      3         109
      4         109
      ...
      2482587    111
      2482588    111
      2482589    111
      2482590    111
      2482591    111
      Name: mins, Length: 2482592, dtype: int64
```

```
[95]: pre_Data1["season"]
```

```
[95]: 0          NaN
      1           2
      2           2
      3           2
      4           2
      ...
      2482587    NaN
      2482588    NaN
      2482589    NaN
      2482590    NaN
      2482591    NaN
      Name: season, Length: 2482592, dtype: object
```

```
[96]: pre_Data1["season"]=pre_Data1["season"].astype(float)
```

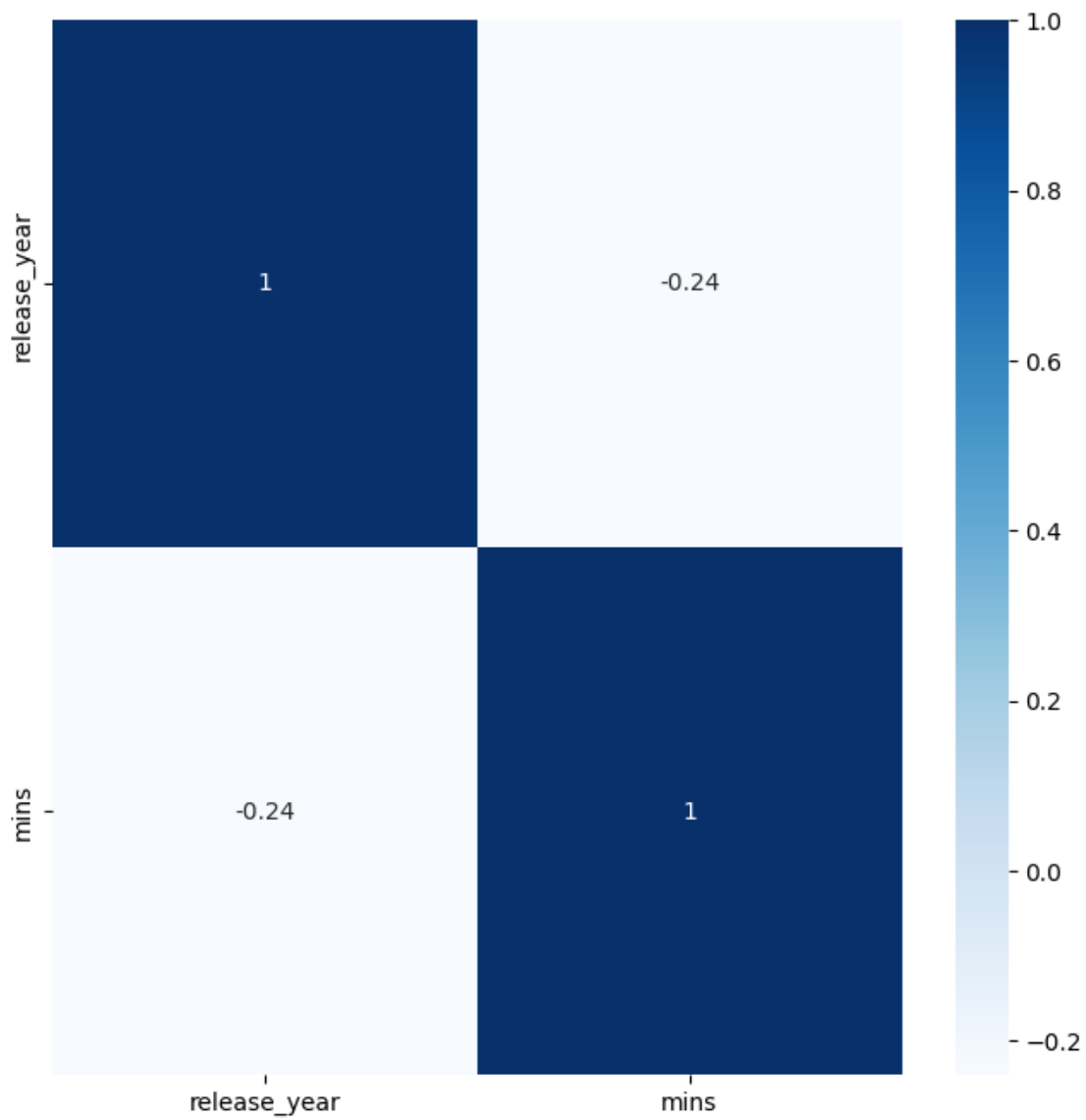
```
[97]: pre_Data1["season"].fillna(pre_Data1["season"].mean()).astype(int)
```

```
[97]: 0      1
      1      2
      2      2
      3      2
      4      2
      ..
      2482587  1
      2482588  1
      2482589  1
      2482590  1
      2482591  1
      Name: season, Length: 2482592, dtype: int64
```

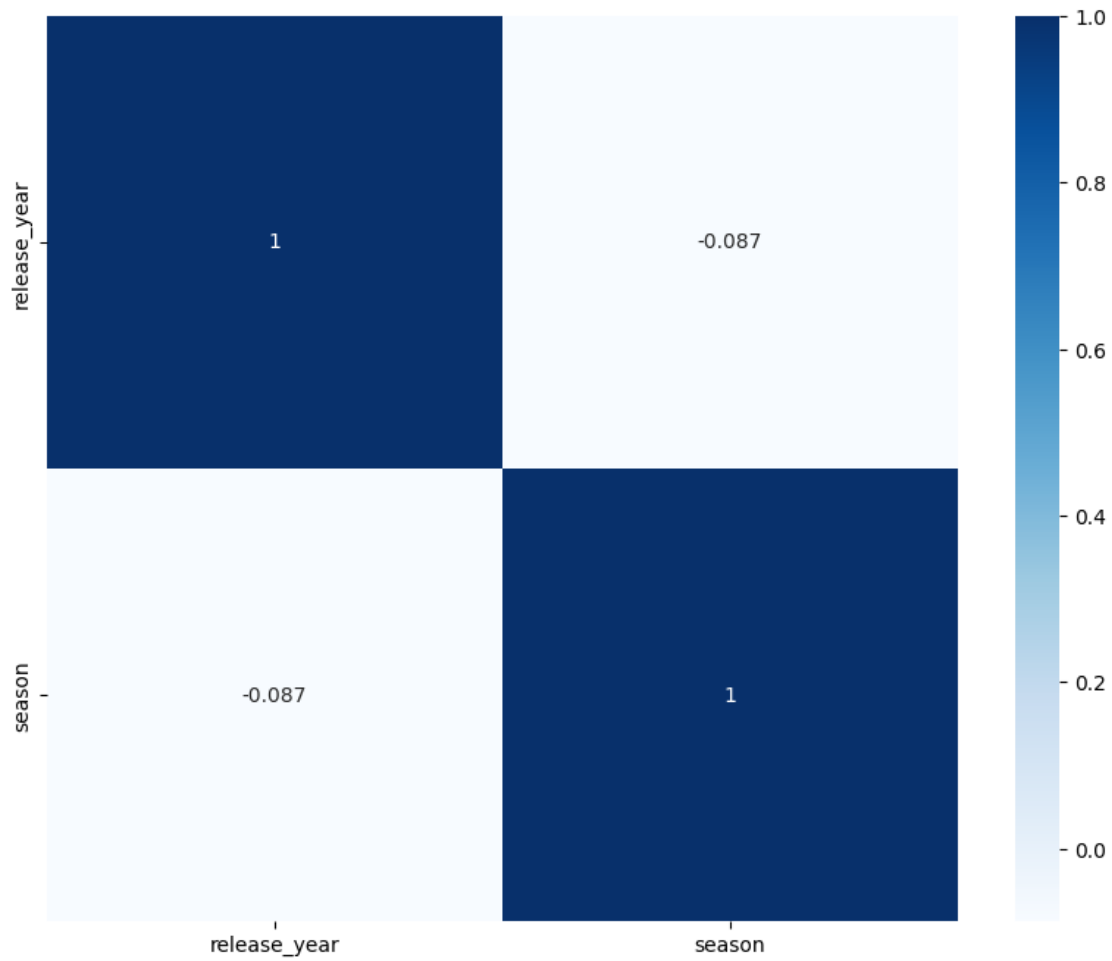
**Correlation analysis** is typically used for numerical or continuous variables to measure the strength and direction of the linear relationship between them. In your Netflix dataset, there are limited numerical variables, so correlation analysis may not be suitable. However, if you want to analyze any numerical variables for correlations, you can use methods like Pearson's correlation coefficient.

We use the `.corr()` method to calculate the Pearson correlation coefficient between these two variables.

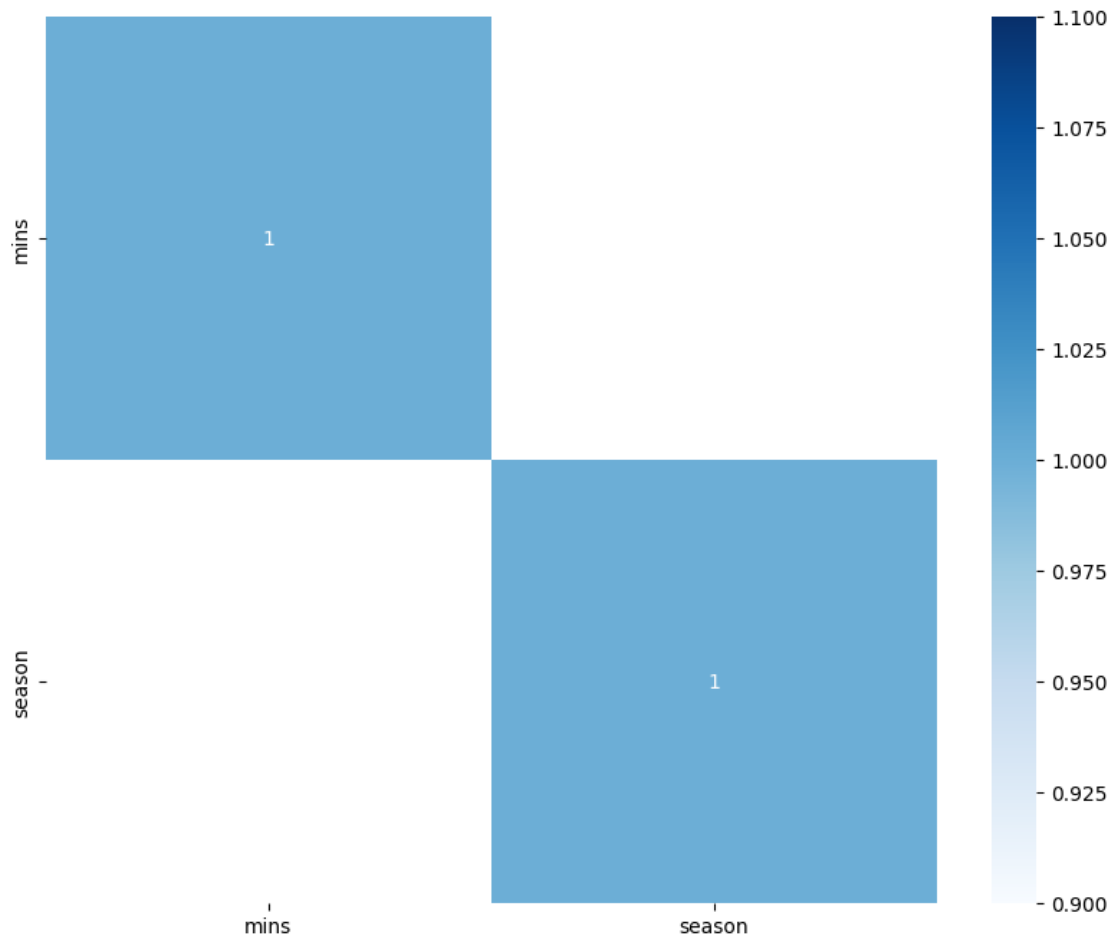
```
[112]: #Correlation - HeatMap
plt.figure(figsize=(8, 8))
heat=pre_Data1[["release_year", "mins"]]
sns.heatmap(heat.corr(), cmap="Blues", annot=True)
plt.show()
```



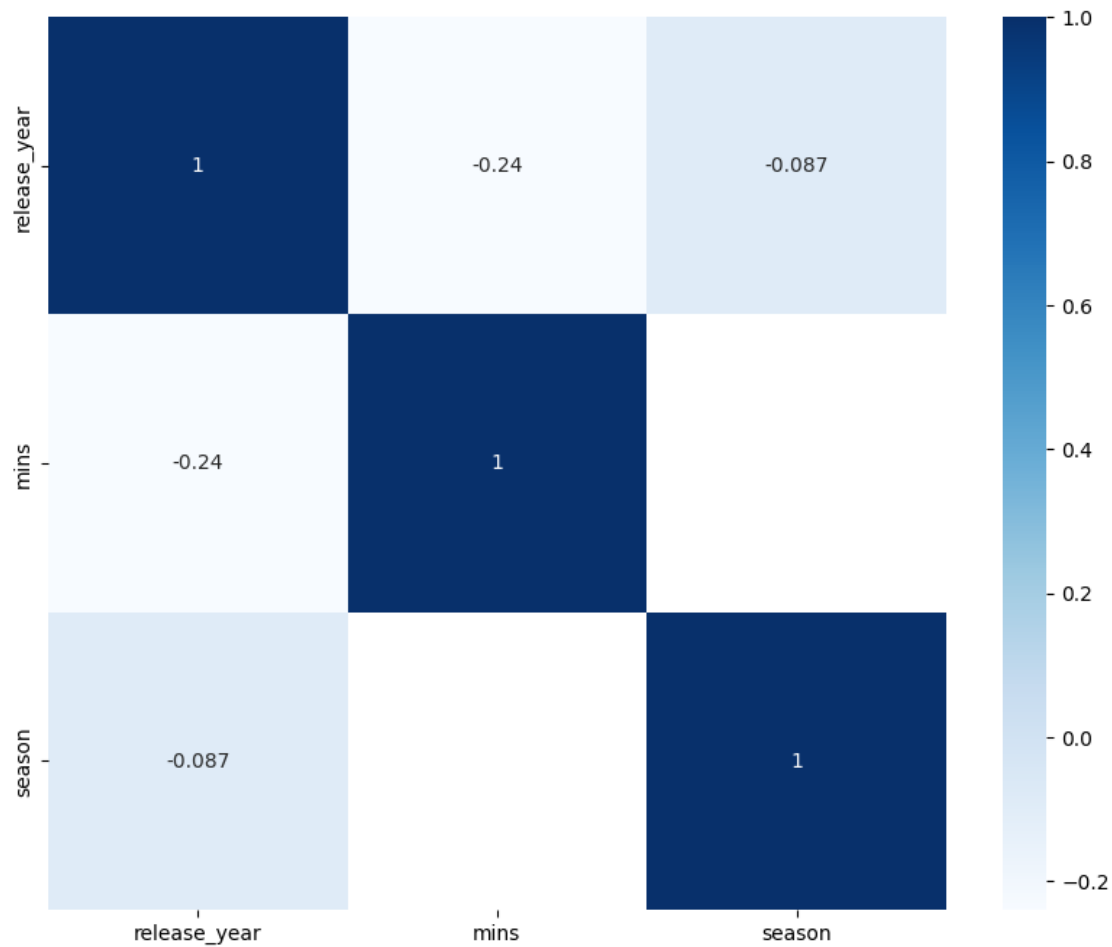
```
[107]: plt.figure(figsize=(8, 8))
heat=pre_Data1[["release_year", "season"]]
sns.heatmap(heat.corr(), cmap="Blues", annot=True)
plt.show()
```



```
[108]: plt.figure(figsize=(8, 8))
heat=pre_Data1[["mins", "season"]]
sns.heatmap(heat.corr(), cmap="Blues", annot=True)
plt.show()
```



```
[109]: plt.figure(figsize=(8, 8))
#heat=pre_Data1[["mins","season"]]
sns.heatmap(pre_Data1.corr(),cmap="Blues",annot=True)
plt.show()
```



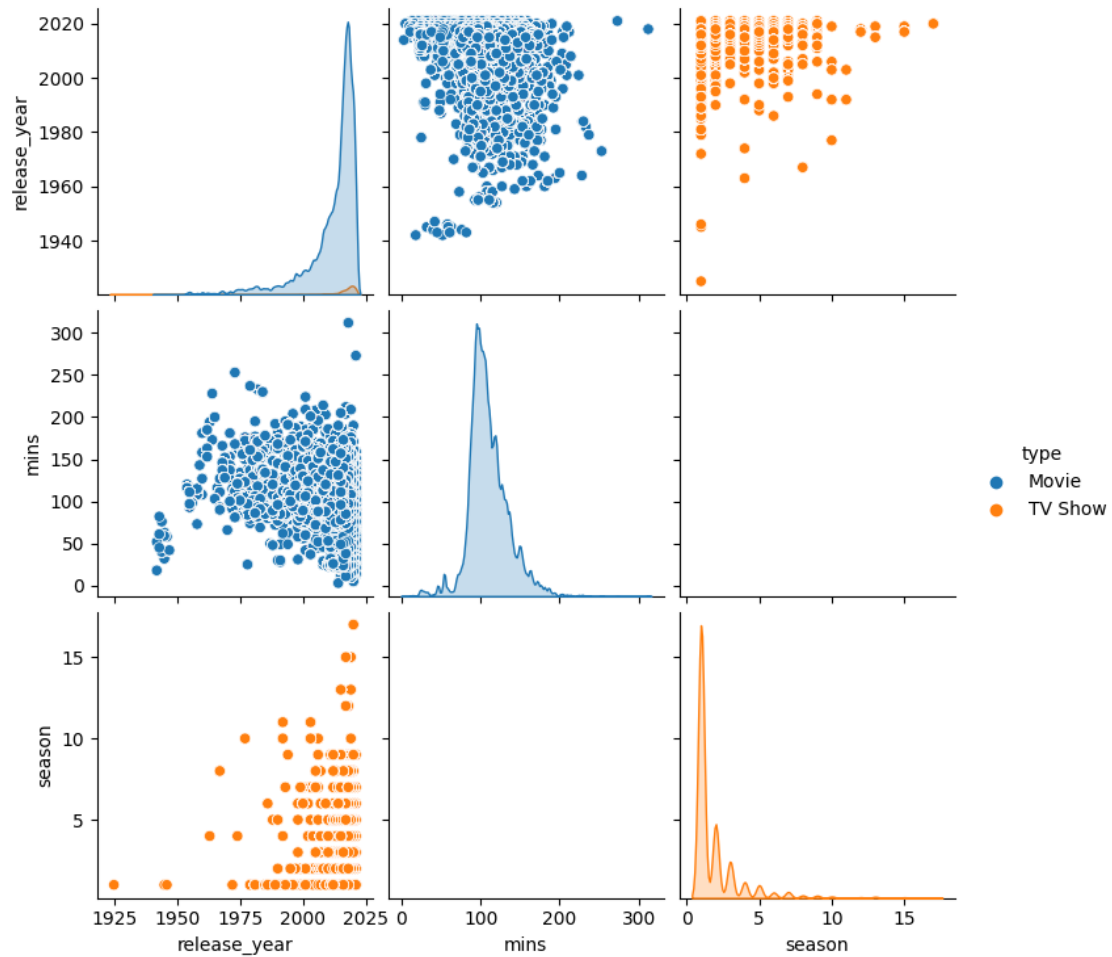
A **pair plot** is a powerful tool for visualizing the relationships between multiple numerical variables in a dataset. Since your Netflix dataset primarily consists of categorical variables and only a few numerical attributes, creating a pair plot may not be very informative.

However, if you still want to create a pair plot for the available numerical attributes (e.g., 'release\_year', 'mins', 'season'), you can use the Seaborn library. Here's an example:

```
[111]: sns.pairplot(data=pre_Data1, hue="type")
```

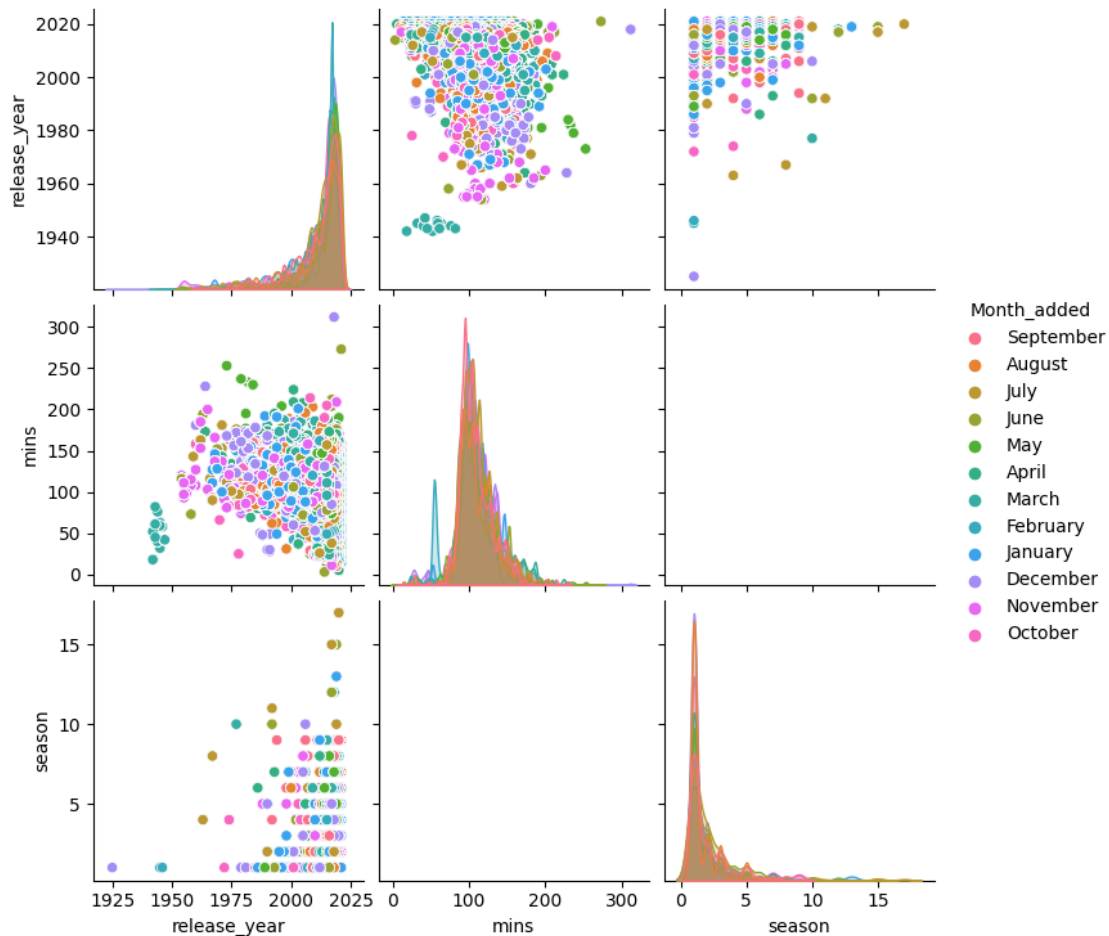
```
[111]: <seaborn.axisgrid.PairGrid at 0x7adb13e1d480>
```





```
[113]: sns.pairplot(data=pre_Data1,hue="Month_added")
```

```
[113]: <seaborn.axisgrid.PairGrid at 0x7adb13e1f3d0>
```



In this example, we're using Seaborn's pairplot to create a pair plot for the 'Release\_year', 'mins' and 'season' variables. However, since the dataset mainly contains categorical variables, this pair plot won't be very informative.

**Pair plots** are most useful when you have multiple numerical variables that you want to visualize together to explore relationships and correlations. If you have other numerical attributes in your dataset that you'd like to include in the pair plot, you can expand the numeric\_attributes DataFrame to include those variables. Otherwise, consider using other visualization techniques more suitable for your dataset's characteristics.

To check for missing values in your dataset, you can use the `isnull()` function followed by the `sum()` function to count the missing values in each column. Here's how you can do it:

```
[99]: # missing values
missing_values = data_copy.isnull().sum()
print(f"Number of Missing Values: {missing_values}")
```

```
Number of Missing Values: 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

```

This code will print the number of missing values in each column of your DataFrame. You can decide whether you want to treat missing values by filling them, dropping rows, or other methods based on the context and the specific attributes with missing data.

```

[100]: def missing_data(df):
        miss_val=df.isnull().sum()

        miss_per=miss_val/len(df)*100
        miss_val_table=pd.concat([miss_val,miss_per,df.dtypes],axis=1)
        miss_val_table=miss_val_table.rename(columns={0:"Missing_value_count",1:
        ↪"Missing_value_percentage(%)",2:"Data Type"})
        return miss_val_table

```

```

[101]: # Percertange for missing values
        missing_data(data_copy)

```

```

[101]:

```

	Missing_value_count	Missing_value_percentage(%)	Data Type
show_id	0	0.000000	object
type	0	0.000000	category
title	0	0.000000	object
director	2634	29.908028	object
cast	825	9.367549	object
country	831	9.435676	category
date_added	10	0.113546	object
release_year	0	0.000000	int64
rating	4	0.045418	category
duration	3	0.034064	object
listed_in	0	0.000000	category
description	0	0.000000	object

```

[112]:

```

```

[ ]:

```

```

[ ]:

```

### 0.0.5 Comments on the Range of Attributes :

- The dataset consists of both categorical and numerical attributes, such as type(categorical), rating(categorical), country(categorical), listed\_in(categorical) and release\_year(numerical) and duration(numerical) represents the durations of content
- Categorical attributes like 'Country' 'cast' 'directors' has wide range of unique values
- The date\_added attribute represents the date when content is added to netflix and is in datetime format

### 0.0.6 Comments on the Distribution of variables and relationships between them:

- The distribution of content ratings ('Rating') shows that 'TV-MA' is the most common rating, followed by 'TV-14' and 'TV-PG.'
- There is a broad range of release years ('Release\_year'), spanning from earlier years to recent years.
- The distribution of the 'Duration' variable varies between movies (in minutes) and TV shows (in terms of the number of seasons).
- Relationships between variables can be explored further, such as examining the distribution of content ratings for movies versus TV shows.

### 0.0.7 Comments for Each Univariate and Bivariate Plot:

- In the count plot of 'Type' vs. 'Rating,' it's evident that 'TV-MA' content is the most common for both movies and TV shows. 'TV-14' and 'TV-PG' are also prevalent ratings for both types of content.
- The box plot comparing 'Release\_year' and 'Type' shows that movies have a broader range of release years compared to TV shows.
- The distribution plot (histogram) of 'Release\_year' provides insights into the distribution of content over the years, with peaks in different time periods.
- The box plot for 'Rating' helps visualize the distribution and spread of content ratings. It shows that 'TV-MA' has a wide range, 'TV-14' is also diverse, while 'TV-PG' is comparatively more consistent in terms of content.
- The count plot of 'Type' and 'Rating' helps to understand the distribution of content ratings for movies and TV shows. It shows that 'TV-MA' is common in both, but the distribution of other ratings may differ between the two types.
- The box plot comparing 'Release\_year' and 'Type' reveals that movies tend to have more diverse release years, while TV shows have a relatively narrower range, with some potential outliers.
- The count plot comparing 'Type' and 'Country' shows the distribution of content types by country, helping to identify which countries contribute more to Netflix's library.
- These insights provide a general understanding of the dataset's attributes and their distributions. Further analysis and domain-specific questions can lead to more detailed and actionable insights for decision-making or content recommendation systems on Netflix.

### 0.0.8 Content Distribution and Popularity:

- The dataset reveals that 'TV-MA' is a popular content rating on Netflix, which suggests that mature and edgier content may resonate with the audience. 'TV-14' and 'TV-PG' content

ratings also have a significant presence, indicating a diverse audience with varying age groups. ### Content Library Diversity:

- Netflix's content library is diverse, with content originating from various countries. This diversity reflects Netflix's commitment to offering a wide range of international content. ### Content Release Trends:
- The distribution of content release years shows that Netflix has content spanning across several decades, suggesting a focus on offering both classic and recent content. ### Movie and TV Show Distribution:
- The analysis indicates that Netflix's library includes both movies and TV shows. Movies are often more diverse in terms of their release years, while TV shows may have a more consistent distribution of release years. ### Influential Directors and Actors:
- The dataset can be used to identify influential directors and actors/actresses whose work is available on Netflix. Insights about the popularity of specific talent can inform casting and acquisition decisions. ### Content Recommendation and Personalization:
- Understanding user preferences based on content type, rating, and genre can aid in improving content recommendation algorithms and personalization to enhance user engagement. ### Content Localization:
- The dataset provides insights into the popularity of content from different countries. Netflix can use this information to prioritize content localization efforts. ### Content Acquisition Strategy:
- By analyzing the release years and types of content, Netflix can refine its acquisition strategy, targeting specific genres or eras that align with audience preferences. ### Content Licensing and Expiration:
- The 'Date\_added' attribute can help manage content licensing and expiration dates, ensuring that popular titles remain available to subscribers. ### Content Quality and User Engagement:
- The dataset can be used to identify patterns in content quality, helping Netflix invest in content that engages users and attracts new subscribers.

[ ]: