```
In [2]: import pandas as pd
import numpy as np
from numpy import nan, NaN,NAN
from matplotlib import pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings("ignore")
In [3]: netflix_df=pd.read_csv("netflix.csv")
netflix_df
df=netflix_df.copy()
df
```

| Out[3]: | 5 | show_id | type | title | director | cast | country | date_added | release_year | rating | duration | listed_in | description |
|---------|------|---------|------------|-----------------------------|--------------------|--|------------------|-----------------------|--------------|-----------|--------------|--|---|
| | 0 | s1 | Movie | Dick Johnson Is Dead | Kirsten Johnson | NaN | United States | September 25, 2021 | 2020 | PG-13 | 90 min | Documentaries | As her father nears the end of his life, filmm |
| | 1 | s2 | TV Show | Blood & Water | NaN | Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban | South Africa | September 24, 2021 | 2021 | TV- MA | 2 Seasons | International TV Shows, TV Dramas, TV Mysteries | After crossing paths at a party, a Cape Town t |
| | 2 | s3 | TV Show | Ganglands | Julien Leclercq | Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabi | NaN | September 24, 2021 | 2021 | TV- MA | 1 Season | Crime TV Shows, International TV Shows, TV Act | To protect his family from a powerful drug lor |
| | 3 | s4 | TV Show | Jailbirds New Orleans | NaN | NaN | NaN | September 24, 2021 | 2021 | TV- MA | 1 Season | Docuseries, Reality TV | Feuds, flirtations and toilet talk go down amo |
| | 4 | s5 | TV Show | Kota Factory | NaN | Mayur More, Jitendra Kumar, Ranjan Raj, Alam K | India | September 24, 2021 | 2021 | TV- MA | 2 Seasons | International TV Shows, Romantic TV Shows, TV | In a city of coaching centers known to train I |
| | ••• | | | | | | | | | | | | |
| | 8802 | s8803 | Movie | Zodiac | David Fincher | Mark Ruffalo, Jake Gyllenhaal, Robert Downey J | United States | November 20, 2019 | 2007 | R | 158 min | Cult Movies, Dramas, Thrillers | A political cartoonist, a crime reporter and a |

| | show_id | type | title | director | cast | country | date_added | release_year | rating | duration | listed_in | description |
|------|---------|------------|----------------|--------------------|---|------------------|---------------------|--------------|--------|--------------|---|--|
| 8803 | s8804 | TV Show | Zombie Dumb | NaN | NaN | NaN | July 1, 2019 | 2018 | TV-Y7 | 2 Seasons | Kids' TV, Korean TV Shows, TV Comedies | While living alone in a spooky town, a young g |
| 8804 | s8805 | Movie | Zombieland | Ruben Fleischer | Jesse Eisenberg, Woody Harrelson, Emma Stone, | United States | November 1, 2019 | 2009 | R | 88 min | Comedies, Horror Movies | Looking to survive in a world taken over by zo |
| 8805 | s8806 | Movie | Zoom | Peter Hewitt | Tim Allen, Courteney Cox, Chevy Chase, Kate Ma | United States | January 11, 2020 | 2006 | PG | 88 min | Children & Family Movies, Comedies | Dragged from civilian life, a former superhero |
| 8806 | s8807 | Movie | Zubaan | Mozez Singh | Vicky Kaushal, Sarah- Jane Dias, Raaghav Chanan | India | March 2, 2019 | 2015 | TV-14 | 111 min | Dramas, International Movies, Music & Musicals | A scrappy but poor boy worms his way into a ty |

8807 rows × 12 columns

Missing Value Detection

In [4]: df.info()

```
<class 'pandas.core.frame.DataFrame'>
         RangeIndex: 8807 entries, 0 to 8806
         Data columns (total 12 columns):
             Column
                           Non-Null Count Dtype
             -----
                           -----
                           8807 non-null object
             show id
                           8807 non-null
          1
             type
                                          object
          2
             title
                           8807 non-null
                                          object
          3
                           6173 non-null
                                          object
             director
             cast
                           7982 non-null
                                          object
          4
          5
             country
                           7976 non-null
                                          object
             date added
                           8797 non-null
                                          object
         7
             release year 8807 non-null
                                          int64
             rating
                           8803 non-null
                                          object
             duration
                           8804 non-null
                                          object
          9
          10 listed in
                           8807 non-null
                                          object
         11 description 8807 non-null
                                          object
         dtypes: int64(1), object(11)
         memory usage: 825.8+ KB
        missing data=df.isna().sum().to frame()
In [33]:
         missing data.rename(columns={0:"Null Value Cnt"},inplace=True)
         missing data["Null Value Percentage"]=round(missing data["Null Value Cnt"]*100/len(df),2)
        missing data
In [34]:
```

| Out[34]: | | Null_Value_Cnt | Null_Value_Percentage |
|----------|--------------|----------------|-----------------------|
| | show_id | 0 | 0.00 |
| | type | 0 | 0.00 |
| | title | 0 | 0.00 |
| | director | 2634 | 29.91 |
| | cast | 825 | 9.37 |
| | country | 831 | 9.44 |
| | date_added | 10 | 0.11 |
| | release_year | 0 | 0.00 |
| | rating | 4 | 0.05 |
| | duration | 3 | 0.03 |
| | listed_in | 0 | 0.00 |
| | description | 0 | 0.00 |

Observation: There is almost 30% data missing from director column .Followed By the country and cast with over 9% of its data missing

In [35]: df.describe()

```
        count
        8807.000000

        mean
        2014.180198

        std
        8.819312

        min
        1925.000000

        25%
        2013.000000

        50%
        2017.000000

        75%
        2019.000000

        max
        2021.000000
```

In [9]: df.describe(include="object")

| Out[9]: | | show_id | type | title | director | cast | country | date_added | rating | duration | listed_in | description |
|---------|--------|---------|-------|-------------------------|------------------|-----------------------|------------------|--------------------|-----------|----------|------------------------------------|--|
| | count | 8807 | 8807 | 8807 | 6173 | 7982 | 7976 | 8797 | 8803 | 8804 | 8807 | 8807 |
| | unique | 8807 | 2 | 8807 | 4528 | 7692 | 748 | 1767 | 17 | 220 | 514 | 8775 |
| | top | s1 | Movie | Dick Johnson Is Dead | Rajiv Chilaka | David Attenborough | United States | January 1, 2020 | TV- MA | 1 Season | Dramas, International Movies | Paranormal activity at a lush, abandoned prope |
| | freq | 1 | 6131 | 1 | 19 | 19 | 2818 | 109 | 3207 | 1793 | 362 | 4 |

Missing values Treatment and transformation of some of the Basic metric columns

```
df["director"].replace(to_replace=NaN,value="X",inplace=True)
df["cast"].replace(to_replace=NaN,value="X",inplace=True)
df["Year"]=df["Year"].astype("int")
df["Mon"]=df["Mon"].astype("int")
```

In [37]: **df**

| Out[37]: | 5 | show_id | type | title | director | cast | country | date_added | release_year | rating | duration | listed_in | description | Date | Мо |
|----------|-----|---------|------------|-----------------------------|--------------------|--|------------------|-----------------------|--------------|-----------|--------------|--|---|----------------|--------|
| | 0 | s1 | Movie | Dick Johnson Is Dead | Kirsten Johnson | Х | United States | September 25, 2021 | 2020 | PG-13 | 90 min | Documentaries | As her father nears the end of his life, filmm | 2021- 09-25 | Septem |
| | 1 | s2 | TV Show | Blood & Water | Х | Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban | South Africa | September 24, 2021 | 2021 | TV- MA | 2 Seasons | International TV Shows, TV Dramas, TV Mysteries | After crossing paths at a party, a Cape Town t | | Septen |
| | 2 | s3 | TV Show | Ganglands | Julien Leclercq | Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabi | NaN | September 24, 2021 | 2021 | TV- MA | 1 Season | Crime TV Shows, International TV Shows, TV Act | To protect his family from a powerful drug lor | 2021- 09-24 | Septen |
| | 3 | s4 | TV Show | Jailbirds New Orleans | Х | Х | NaN | September 24, 2021 | 2021 | TV- MA | 1 Season | Docuseries, Reality TV | Feuds, flirtations and toilet talk go down amo | | Septen |
| | 4 | s5 | TV Show | Kota Factory | Х | Mayur More, Jitendra Kumar, Ranjan Raj, Alam K | India | September 24, 2021 | 2021 | TV- MA | 2 Seasons | International TV Shows, Romantic TV Shows, TV | In a city of coaching centers known to train l | 2021- 09-24 | Septen |
| | ••• | | | | | | | | | | | | | ••• | |

| | show_id | type | title | director | cast | country | date_added | release_year | rating | duration | listed_in | description | Date | Мо |
|------|---------|------------|----------------|--------------------|--|------------------|----------------------|--------------|--------|--------------|---|--|----------------|-------|
| 8802 | s8803 | Movie | Zodiac | David Fincher | Mark Ruffalo, Jake Gyllenhaal, Robert Downey J | United States | November 20, 2019 | 2007 | R | 158 min | Cult Movies, Dramas, Thrillers | A political cartoonist, a crime reporter and a | 2019- 11-20 | Novem |
| 8803 | s8804 | TV Show | Zombie Dumb | Х | Х | NaN | July 1, 2019 | 2018 | TV-Y7 | 2 Seasons | Kids' TV, Korean TV Shows, TV Comedies | While living alone in a spooky town, a young g | 2019- 07-01 | |
| 8804 | s8805 | Movie | Zombieland | Ruben Fleischer | Jesse Eisenberg, Woody Harrelson, Emma Stone, | United States | November 1, 2019 | 2009 | R | 88 min | Comedies, Horror Movies | Looking to survive in a world taken over by zo | 2019- 11-01 | Noven |
| 8805 | s8806 | Movie | Zoom | Peter Hewitt | Tim Allen, Courteney Cox, Chevy Chase, Kate Ma | United States | January 11, 2020 | 2006 | PG | 88 min | Children & Family Movies, Comedies | Dragged from civilian life, a former superhero | 2020- 01-11 | Janı |
| 8806 | s8807 | Movie | Zubaan | Mozez Singh | Vicky Kaushal, Sarah-Jane Dias, Raaghav Chanan | India | March 2, 2019 | 2015 | TV-14 | 111 min | Dramas, International Movies, Music & Musicals | A scrappy but poor boy worms his way into a ty | 2019- 03-02 | Mi |

8807 rows × 16 columns

Analyse the amount of content added in the platform Yearly and Monthly

```
In [38]: df.groupby(["Year","type"])["type"].count()
         Year type
Out[38]:
              TV Show
                           10
         2008 Movie
                            1
               TV Show
                            1
         2009 Movie
                            2
         2010 Movie
                            1
         2011 Movie
                           13
         2012 Movie
                            3
         2013 Movie
                            6
              TV Show
                            5
         2014 Movie
                           19
               TV Show
                            5
         2015 Movie
                           56
               TV Show
                           26
         2016 Movie
                          253
               TV Show
                          176
         2017 Movie
                          839
               TV Show
                          349
         2018 Movie
                         1237
               TV Show
                          412
         2019 Movie
                         1424
               TV Show
                          592
         2020 Movie
                         1284
                          595
               TV Show
         2021 Movie
                          993
              TV Show
                          505
         Name: type, dtype: int64
In [39]: trend_year=df.groupby(["Year","type"])["type"].count().to_frame()
         trend_year.rename(columns={"type":"count"},inplace=True)
         trend year.reset index(inplace=True)
         trend_year.loc[trend_year["Year"]==0]
         trend year.drop(0,axis=0,inplace=True)
         trend year
```

| Out[39]: | | Year | type | count |
|----------|----|------|---------|-------|
| | 1 | 2008 | Movie | 1 |
| | 2 | 2008 | TV Show | 1 |
| | 3 | 2009 | Movie | 2 |
| | 4 | 2010 | Movie | 1 |
| | 5 | 2011 | Movie | 13 |
| | 6 | 2012 | Movie | 3 |
| | 7 | 2013 | Movie | 6 |
| | 8 | 2013 | TV Show | 5 |
| | 9 | 2014 | Movie | 19 |
| | 10 | 2014 | TV Show | 5 |
| | 11 | 2015 | Movie | 56 |
| | 12 | 2015 | TV Show | 26 |
| | 13 | 2016 | Movie | 253 |
| | 14 | 2016 | TV Show | 176 |
| | 15 | 2017 | Movie | 839 |
| | 16 | 2017 | TV Show | 349 |
| | 17 | 2018 | Movie | 1237 |
| | 18 | 2018 | TV Show | 412 |
| | | | | |

2019

2020

2021

2019 TV Show

2020 TV Show

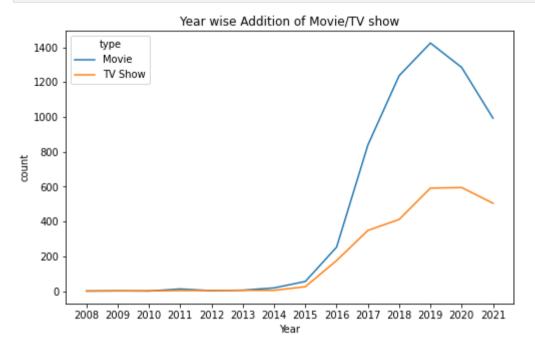
2021 TV Show

Movie

Movie

Movie

```
In [41]: plt.rcParams["figure.figsize"] = (8,5)
plt.xticks(np.arange(2008,2022,1))
plt.title("Year wise Addition of Movie/TV show")
sns.lineplot(x="Year",y="count",data=trend_year,hue="type")
plt.show()
```

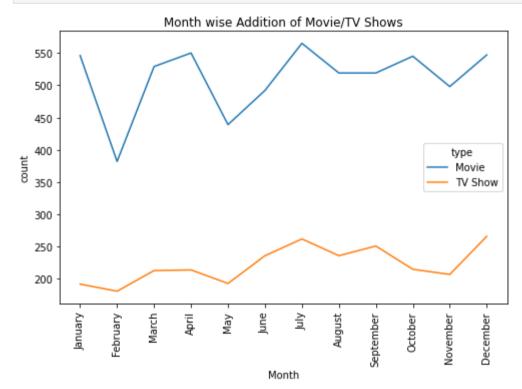


Observation: There is an Exponential increase in the number of Movies/TV shows added from 2015-2019. But past two years this has dropped. Most possible reason is Covid.

```
In [42]: trend_mon=df.groupby(["Mon","Month","type"])["type"].count().to_frame()
    trend_mon.rename(columns={"type":"count"},inplace=True)
    trend_mon.reset_index(inplace=True)
    trend_mon
```

| Out[42]: | | Mon | Month | type | coun |
|----------|----|-----|-----------|---------|------|
| | 0 | 1 | January | Movie | 540 |
| | 1 | 1 | January | TV Show | 192 |
| | 2 | 2 | February | Movie | 382 |
| | 3 | 2 | February | TV Show | 18 |
| | 4 | 3 | March | Movie | 529 |
| | 5 | 3 | March | TV Show | 213 |
| | 6 | 4 | April | Movie | 550 |
| | 7 | 4 | April | TV Show | 214 |
| | 8 | 5 | May | Movie | 439 |
| | 9 | 5 | May | TV Show | 193 |
| | 10 | 6 | June | Movie | 492 |
| | 11 | 6 | June | TV Show | 236 |
| | 12 | 7 | July | Movie | 56 |
| | 13 | 7 | July | TV Show | 262 |
| | 14 | 8 | August | Movie | 519 |
| | 15 | 8 | August | TV Show | 236 |
| | 16 | 9 | September | Movie | 519 |
| | 17 | 9 | September | TV Show | 25 |
| | 18 | 10 | October | Movie | 54! |
| | 19 | 10 | October | TV Show | 21! |
| | 20 | 11 | November | Movie | 498 |
| | 21 | 11 | November | TV Show | 20 |
| | 22 | 12 | December | Movie | 547 |
| | 23 | 12 | December | TV Show | 260 |

```
In [500... plt.rcParams["figure.figsize"] = (8,5)
    plt.xticks(rotation=90)
    plt.title("Month wise Addition of Movie/TV Shows")
    sns.lineplot(x="Month",y="count",data=trend_mon,hue="type")
    plt.show()
```



Observation: Both movie and TV show follows almost a similar trend where July has most number of movie/TV shows added to Netflix and February the least

Nested Column Unpacking(Director, Country, Listed In)

```
director_df=pd.DataFrame(director_lst,index=df["show_id"])
director_df=director_df.stack()

director_df=pd.DataFrame(director_df)
director_df.reset_index(inplace=True)
director_df=director_df[["show_id",0]]
director_df.columns=["show_id","director"]
director_df=director_df.merge(df[["show_id","title","type"]])
director_df.describe(include="object")
```

title

type

Out[44]: show_id director

count 9612 9612

 count
 9612
 9612
 9612
 9612

 unique
 8807
 4994
 8807
 2

 top
 s5888
 X
 Walt Disney Animation Studios Short Films Coll...
 Movie

freq 13 2634 13 6854

In [65]: dir_con

| Out[65]: | | show_id | director | title | type | country |
|----------|------|---------|-----------------|----------------------|-------|----------------|
| | 0 | s1 | Kirsten Johnson | Dick Johnson Is Dead | Movie | United States |
| | 1 | s8 | Haile Gerima | Sankofa | Movie | United States |
| | 2 | s8 | Haile Gerima | Sankofa | Movie | Ghana |
| | 3 | s8 | Haile Gerima | Sankofa | Movie | Burkina Faso |
| | 4 | s8 | Haile Gerima | Sankofa | Movie | United Kingdom |
| | ••• | | | | | |
| | 8465 | s8802 | Majid Al Ansari | Zinzana | Movie | Jordan |
| | 8466 | s8803 | David Fincher | Zodiac | Movie | United States |
| | 8467 | s8805 | Ruben Fleischer | Zombieland | Movie | United States |
| | 8468 | s8806 | Peter Hewitt | Zoom | Movie | United States |
| | 8469 | s8807 | Mozez Singh | Zubaan | Movie | India |

8470 rows × 5 columns

```
United States
                             3689
Out[67]:
          India
                            1046
          United Kingdom
                              804
          Canada
                              445
          France
                              393
          Japan
                              318
                              232
          Spain
                              231
          South Korea
          Germany
                              226
          Mexico
                              169
          Name: country, dtype: int64
```

Observation- 127 Unique Countries (Missing Excluded) .The top 10 countries are lised above based on the amount of content

Analyse content added in the platform based on their Genre

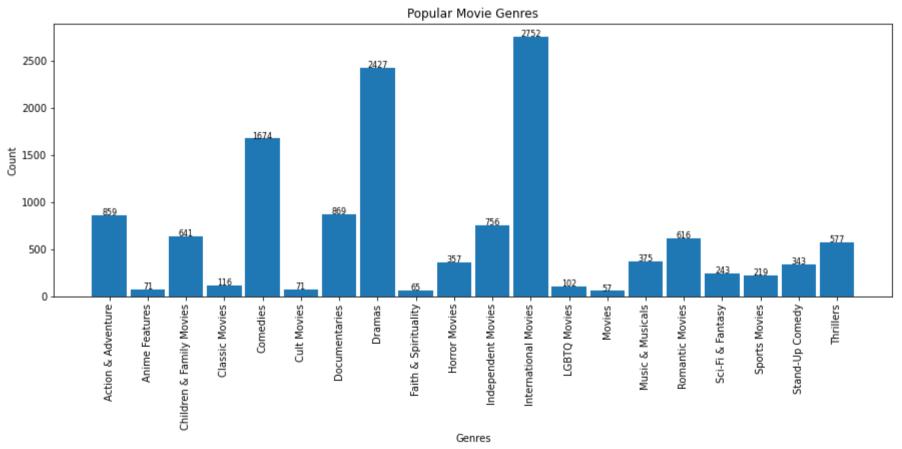
```
In [144... x_movie x_tv
```

Out[144]: listed count

| type | | |
|---------|------------------------------|------|
| TV Show | Anime Series | 176 |
| TV Show | British TV Shows | 253 |
| TV Show | Classic & Cult TV | 28 |
| TV Show | Crime TV Shows | 470 |
| TV Show | Docuseries | 395 |
| TV Show | International TV Shows | 1351 |
| TV Show | Kids' TV | 451 |
| TV Show | Korean TV Shows | 151 |
| TV Show | Reality TV | 255 |
| TV Show | Romantic TV Shows | 370 |
| TV Show | Science & Nature TV | 92 |
| TV Show | Spanish-Language TV Shows | 174 |
| TV Show | Stand-Up Comedy & Talk Shows | 56 |
| TV Show | TV Action & Adventure | 168 |
| TV Show | TV Comedies | 581 |
| TV Show | TV Dramas | 763 |
| TV Show | TV Horror | 75 |
| TV Show | TV Mysteries | 98 |
| TV Show | TV Sci-Fi & Fantasy | 84 |
| TV Show | TV Shows | 16 |
| TV Show | TV Thrillers | 57 |
| TV Show | Teen TV Shows | 69 |
| | | |

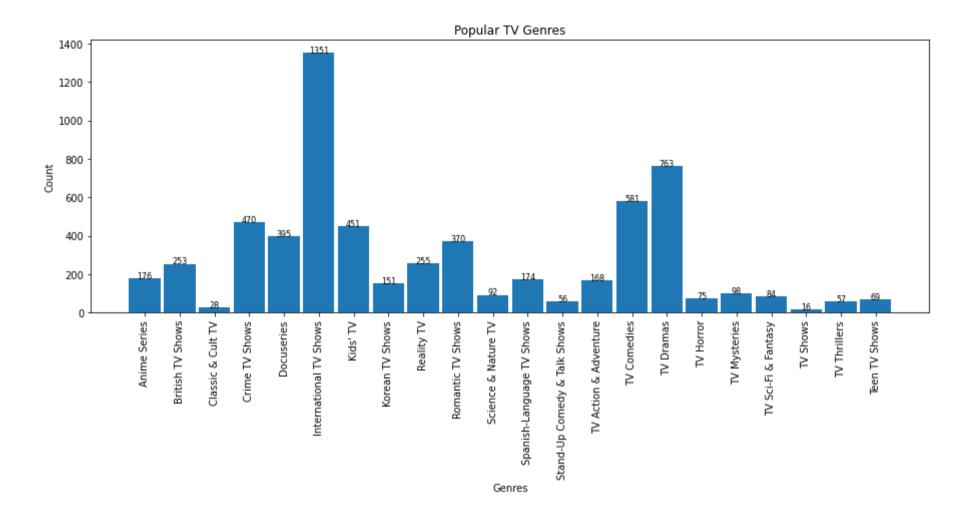
```
In [150... x_movie_val1=x_movie["listed"].to_list()
    x_movie_val2=x_movie["count"].to_list()

In [201... plt.rcParams["figure.figsize"] = (15,5)
    plt.bar(x_movie_val1,x_movie_val2,width=.9)
    plt.xticks(x_movie_val1,rotation=90)
    plt.title("Popular Movie Genres")
    plt.xlabel("Genres")
    plt.ylabel("Count")
    for idx,val in enumerate(x_movie_val2):
        plt.text(idx,val+.2,str(val),fontsize=8,ha="center")
    plt.show()
```



Observation: Popular Movie Genres-

International, Drama, Comedy, Documentries, Action and Adventure.



Observation: Popular TV Genres-Drama, International, Comedy, Crime, Kids TV

```
In [64]: lis_con=listed_df.merge(country_df[["country","show_id"]])
    lis_con=lis_con.groupby(["country","listed"])["listed"].count().to_frame()
    lis_con.rename(columns={"listed":"count"},inplace=True)
    lis_con.reset_index(inplace=True)
    to_del=lis_con.loc[lis_con["country"]==''].index
    lis_con.drop(to_del,axis=0,inplace=True)
    lis_con.reset_index(drop=True,inplace=True)
    lis_con.sort_values("count",ascending=False,inplace=True)
```

```
lis_con_fil=lis_con.loc[lis_con["count"]>50]
lis_con_fil.reset_index(drop=True,inplace=False).head(20)
```

| Out[64]: | | country | listed | count |
|----------|----|----------------|--------------------------|-------|
| | 0 | India | International Movies | 864 |
| | 1 | United States | Dramas | 835 |
| | 2 | United States | Comedies | 680 |
| | 3 | India | Dramas | 662 |
| | 4 | United States | Documentaries | 511 |
| | 5 | United States | Action & Adventure | 404 |
| | 6 | United States | Children & Family Movies | 390 |
| | 7 | United States | Independent Movies | 390 |
| | 8 | India | Comedies | 323 |
| | 9 | United States | Thrillers | 292 |
| | 10 | United States | TV Comedies | 258 |
| | 11 | United States | TV Dramas | 232 |
| | 12 | United Kingdom | British TV Shows | 225 |
| | 13 | United States | Romantic Movies | 225 |
| | 14 | United States | Stand-Up Comedy | 216 |
| | 15 | United States | Kids' TV | 214 |
| | 16 | France | International Movies | 207 |
| | 17 | United States | Horror Movies | 201 |
| | 18 | United Kingdom | Dramas | 196 |
| | 19 | United States | Docuseries | 192 |
| | | | | |

Observation - Above frame shows some of the top genres preffered in

countries

Analyse data wrt Country and Director

```
idx=country df.loc[country df["country"]=="nan"].index
In [53]:
          country df.drop(idx,axis=0,inplace=True)
In [54]: ixd=director_df.loc[director_df["director"]=="X"].index
          director df.drop(ixd,axis=0,inplace=True)
          director df.reset index(drop=True,inplace=True)
In [55]:
          director df
Out[55]:
                 show_id
                                director
                                                                 title
                                                                         type
              0
                      s1 Kirsten Johnson
                                                   Dick Johnson Is Dead
                                                                        Movie
                           Julien Leclercq
                                                           Ganglands TV Show
              1
              2
                           Mike Flanagan
                                                        Midnight Mass TV Show
                           Robert Cullen My Little Pony: A New Generation
              3
                                                                        Movie
              4
                           José Luis Ucha My Little Pony: A New Generation
                                                                        Movie
          6973
                   s8802
                          Majid Al Ansari
                                                              Zinzana
                                                                        Movie
          6974
                   s8803
                           David Fincher
                                                               Zodiac
                                                                        Movie
          6975
                   s8805
                         Ruben Fleischer
                                                          Zombieland
                                                                        Movie
          6976
                   s8806
                             Peter Hewitt
                                                               Zoom
                                                                        Movie
          6977
                   s8807
                            Mozez Singh
                                                              Zubaan
                                                                        Movie
          6978 rows × 4 columns
          director_df.value_counts("director").head(20)
```

director Out[190]: Rajiv Chilaka 22 Jan Suter 21 Raúl Campos 19 Suhas Kadav 16 Marcus Raboy 16 Jay Karas 15 Cathy Garcia-Molina 13 Jay Chapman 12 Youssef Chahine 12 Martin Scorsese 12 Steven Spielberg 11 Don Michael Paul 10 Shannon Hartman 9 Anurag Kashyap 9 9 David Dhawan Yılmaz Erdoğan 9 Quentin Tarantino Fernando Ayllón Troy Miller Hakan Algül 8 dtype: int64

Observation-Top 20 Directors are listed above along with the number of Movie/TV shows directed by them

```
In [56]: country_df.reset_index(drop=True,inplace=True)
    country_df
```

| Out[56]: | | show_id | country | title | type |
|----------|-------|---------|---------------|----------------------|---------|
| | 0 | s1 | United States | Dick Johnson Is Dead | Movie |
| | 1 | s2 | South Africa | Blood & Water | TV Show |
| | 2 | s5 | India | Kota Factory | TV Show |
| | 3 | s8 | United States | Sankofa | Movie |
| | 4 | s8 | Ghana | Sankofa | Movie |
| | | | | | |
| | 10009 | s8802 | Jordan | Zinzana | Movie |
| | 10010 | s8803 | United States | Zodiac | Movie |
| | 10011 | s8805 | United States | Zombieland | Movie |
| | 10012 | s8806 | United States | Zoom | Movie |
| | 10013 | s8807 | India | Zubaan | Movie |

10014 rows × 4 columns

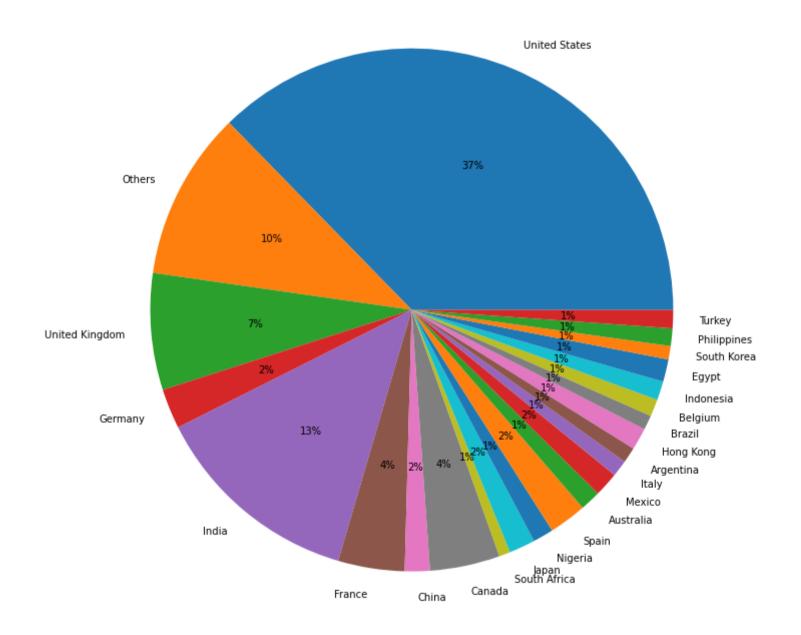
Out[206]:

| | country | director | count |
|----|----------------|--------------------------|-------|
| 0 | Canada | Justin G. Dyck | 8 |
| 1 | Canada | John Paul Tremblay | 5 |
| 2 | Canada | Mike Clattenburg | 5 |
| 3 | France | Thierry Donard | 5 |
| 4 | France | Youssef Chahine | 4 |
| 5 | France | Florent Bodin | 3 |
| 6 | India | Anurag Kashyap | 9 |
| 7 | India | David Dhawan | 9 |
| 8 | India | Umesh Mehra | 8 |
| 9 | Japan | Toshiya Shinohara | 7 |
| 10 | Japan | Masahiko Murata | 5 |
| 11 | Japan | Hiroyuki Seshita | 4 |
| 12 | Mexico | Jan Suter | 12 |
| 13 | Mexico | Raúl Campos | 10 |
| 14 | Mexico | Alex Díaz | 3 |
| 15 | South Korea | Bong Joon Ho | 2 |
| 16 | South Korea | Jung-ah Im | 2 |
| 17 | South Korea | Mark A.Z. Dippé | 2 |
| 18 | Spain | Fernando González Molina | 4 |
| 19 | Spain | Hernán Zin | 4 |
| 20 | Spain | Alexis Morante | 3 |
| 21 | United Kingdom | Alastair Fothergill | 4 |
| 22 | United Kingdom | Edward Cotterill | 4 |
| 23 | United Kingdom | Blair Simmons | 3 |

| | country | director | count |
|----|-----------------|-----------------------|-------|
| 24 | United Kingdom, | Farah Nabulsi | 1 |
| 25 | United Kingdom, | Orlando von Einsiedel | 1 |
| 26 | United States | Jay Karas | 15 |
| 27 | United States | Marcus Raboy | 15 |
| 28 | United States | Jay Chapman | 12 |
| 29 | United States, | Madeleine Gavin | 1 |

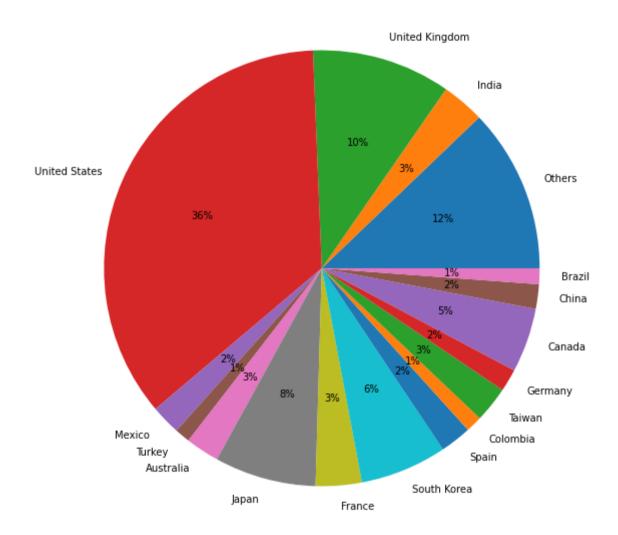
Recommendation-Top 3 directors country wise are shown above for eg:India-Anurag Kashyap,David Dhawan,Priyadarshan etc US-Jay Karas,Marcus Raboy and so on.Hence adding more of their work can fetch more viewers in the respective countries

```
country df movie=country df.loc[country df["type"]=="Movie"]
In [68]:
         country df tv=country df.loc[country df["type"]=="TV Show"]
In [70]: country_df_movie
         country list mv=country df movie["country"].to list()
In [444... from collections import Counter
         count dict=Counter(country list mv)
         dict toplot=Counter()
         for key in count dict.keys():
             if count dict[key]<50:</pre>
                  dict toplot['Others']+=count dict[key]
              else:
                  dict_toplot[key]=count_dict[key]
         print(dict toplot)
         list country=list(dict toplot.keys())
         list country count=list(dict toplot.values())
```



Observation-Most number of movies in Netflix comes from countries like US ,India,UK .

```
country df tv
In [78]:
          country list tv=country df tv["country"].to list()
In [79]: from collections import Counter
         count dict tv=Counter(country list tv)
         dict toplot tv=Counter()
         for key in count dict tv.keys():
              if count dict tv[key]<30:</pre>
                 dict toplot tv['Others']+=count dict tv[key]
              else:
                  dict toplot tv[key]=count dict tv[key]
         print(dict toplot tv)
         list country tv=list(dict toplot tv.keys())
         list country count tv=list(dict toplot tv.values())
         Counter({'United States': 938, 'Others': 321, 'United Kingdom': 272, 'Japan': 199, 'South Korea': 170, 'Canada': 126, 'France':
         90, 'India': 84, 'Taiwan': 70, 'Australia': 66, 'Spain': 61, 'Mexico': 58, 'China': 48, 'Germany': 44, 'Colombia': 32, 'Brazil':
         31, 'Turkey': 30})
In [80]: plt.rcParams["figure.figsize"] = (15,10)
         data=list country count tv
         labels=list country tv
         plt.pie(data,labels=labels,autopct='%.0f%%')
         plt.title("Percentage TV shows from different Countries")
         plt.show()
```



Most number of TV shows in Netflix comes from countries like US ,UK,Japan,South Korea,Canada .

Analyse data wrt Country and Cast

```
cast lst=df["cast"].apply(lambda x:str(x).split(", ")).to list()
In [212...
In [82]:
         cast df=pd.DataFrame(cast lst,index=df["show id"])
          cast df=cast df.stack()
         cast df=pd.DataFrame(cast df)
         cast df.reset index(inplace=True)
         cast df=cast df[["show id",0]]
         cast df.columns=["show id","cast"]
         cast_df=cast_df.merge(df[["show_id","title","type"]])
         ind=cast df.loc[cast df["cast"]=='X'].index
         cast df.drop(ind,axis=0,inplace=True)
         cast df.describe(include="object")
Out[82]:
                 show_id
                                                 type
                                cast
                                           title
                   64126
           count
                               64126
                                          64126 64126
                    7982
                               36439
                                           7982
          unique
                   s3775 Anupam Kher Black Mirror Movie
            top
                                  43
                      50
            freq
                                             50 44475
In [83]: cast_df
```

| Out[83]: | show_id | | cast | title | type |
|----------|---------|-------|-----------------------|---------------|---------|
| | 1 | s2 | Ama Qamata | Blood & Water | TV Show |
| | 2 | s2 | Khosi Ngema | Blood & Water | TV Show |
| | 3 | s2 | Gail Mabalane | Blood & Water | TV Show |
| | 4 | s2 | Thabang Molaba | Blood & Water | TV Show |
| | 5 | s2 | Dillon Windvogel | Blood & Water | TV Show |
| | ••• | | | | |
| | 64946 | s8807 | Manish Chaudhary | Zubaan | Movie |
| | 64947 | s8807 | Meghna Malik | Zubaan | Movie |
| | 64948 | s8807 | Malkeet Rauni | Zubaan | Movie |
| | 64949 | s8807 | Anita Shabdish | Zubaan | Movie |
| | 64950 | s8807 | Chittaranjan Tripathy | Zubaan | Movie |

64126 rows × 4 columns

In [93]: country_df

| Out[93]: | | show_id | country | title | type |
|----------|-------|---------|---------------|----------------------|---------|
| | 0 | s1 | United States | Dick Johnson Is Dead | Movie |
| | 1 | s2 | South Africa | Blood & Water | TV Show |
| | 2 | s5 | India | Kota Factory | TV Show |
| | 3 | s8 | United States | Sankofa | Movie |
| | 4 | s8 | Ghana | Sankofa | Movie |
| | ••• | | | | |
| | 10009 | s8802 | Jordan | Zinzana | Movie |
| | 10010 | s8803 | United States | Zodiac | Movie |
| | 10011 | s8805 | United States | Zombieland | Movie |
| | 10012 | s8806 | United States | Zoom | Movie |
| | 10013 | s8807 | India | Zubaan | Movie |

10014 rows × 4 columns

```
In [94]: cast_con=cast_df.merge(country_df[["country","show_id"]])
    cast_con
```

| Out[94]: | | show_id | cast | title | type | country |
|----------|-------|---------|-----------------------|---------------|---------|--------------|
| | 0 | s2 | Ama Qamata | Blood & Water | TV Show | South Africa |
| | 1 | s2 | Khosi Ngema | Blood & Water | TV Show | South Africa |
| | 2 | s2 | Gail Mabalane | Blood & Water | TV Show | South Africa |
| | 3 | s2 | Thabang Molaba | Blood & Water | TV Show | South Africa |
| | 4 | s2 | Dillon Windvogel | Blood & Water | TV Show | South Africa |
| | ••• | | | | | |
| | 75800 | s8807 | Manish Chaudhary | Zubaan | Movie | India |
| | 75801 | s8807 | Meghna Malik | Zubaan | Movie | India |
| | 75802 | s8807 | Malkeet Rauni | Zubaan | Movie | India |
| | 75803 | s8807 | Anita Shabdish | Zubaan | Movie | India |
| | 75804 | s8807 | Chittaranjan Tripathy | Zubaan | Movie | India |

75805 rows × 5 columns

```
In [104... cast_cnt=cast_cnt.sort_values(["count","country"],ascending=[False,True])
    top_20=cast_cnt.head(20)
    top_20.reset_index(drop=True,inplace=True)
    top_20
```

| 0 1 [404] | | _ | | |
|-----------|----|---------------|-------------------|-------|
| Out[104]: | | country | cast | count |
| | 0 | India | Anupam Kher | 40 |
| | 1 | India | Shah Rukh Khan | 34 |
| | 2 | India | Naseeruddin Shah | 31 |
| | 3 | India | Akshay Kumar | 29 |
| | 4 | India | Om Puri | 29 |
| | 5 | Japan | Takahiro Sakurai | 29 |
| | 6 | India | Amitabh Bachchan | 28 |
| | 7 | India | Paresh Rawal | 28 |
| | 8 | Japan | Yuki Kaji | 28 |
| | 9 | India | Boman Irani | 27 |
| | 10 | India | Kareena Kapoor | 25 |
| | 11 | Japan | Daisuke Ono | 22 |
| | 12 | United States | Samuel L. Jackson | 22 |
| | 13 | United States | Tara Strong | 22 |
| | 14 | India | Ajay Devgn | 21 |
| | 15 | United States | Fred Tatasciore | 21 |
| | 16 | India | Salman Khan | 20 |
| | 17 | United States | Adam Sandler | 20 |
| | 18 | India | Kay Kay Menon | 19 |
| | | | | |

India Nawazuddin Siddiqui

19

Observation -The above shows the top 20 Popular actors in the platform

```
In [97]: cast_cnt=cast_con.groupby(["country","cast"])["cast"].count().to_frame()
    cast_cnt.rename(columns={"cast":"count"},inplace=True)
```

```
cast_cnt.reset_index(["cast","country"],inplace=True)
to_del=cast_cnt.loc[cast_cnt["country"]==''].index
cast_cnt.drop(to_del,axis=0,inplace=True)

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

cast_cnt=cast_cnt.sort_values(["count","country"],ascending=[False,True])
cast_cnt=cast_cnt.loc[cast_cnt["count"]>10]

top_cast=(cast_cnt.groupby("country").head(3)).sort_values("country")
top_cast.reset_index(drop=True,inplace=True)
top_cast
```

| Out[97]: | | country | cast | count |
|----------|----|----------------|---------------------|-------|
| | 0 | Canada | Ashleigh Ball | 12 |
| | 1 | Canada | John Paul Tremblay | 14 |
| | 2 | Canada | Robb Wells | 14 |
| | 3 | Egypt | Hassan Hosny | 13 |
| | 4 | Egypt | Ahmed Helmy | 13 |
| | 5 | India | Anupam Kher | 40 |
| | 6 | India | Shah Rukh Khan | 34 |
| | 7 | India | Naseeruddin Shah | 31 |
| | 8 | Japan | Takahiro Sakurai | 29 |
| | 9 | Japan | Yuki Kaji | 28 |
| | 10 | Japan | Daisuke Ono | 22 |
| | 11 | Nigeria | Richard Mofe-Damijo | 11 |
| | 12 | Nigeria | Blossom Chukwujekwu | 12 |
| | 13 | Nigeria | Tina Mba | 11 |
| | 14 | Philippines | Kathryn Bernardo | 11 |
| | 15 | Turkey | Demet Akbağ | 13 |
| | 16 | United Kingdom | David Attenborough | 17 |
| | 17 | United Kingdom | Michael Palin | 14 |
| | 18 | United Kingdom | John Cleese | 16 |
| | 19 | United States | Fred Tatasciore | 21 |
| | 20 | United States | Tara Strong | 22 |
| | | | | |

United States

Samuel L. Jackson

Recommendation-Top 3 Popular Actors in Netflix in each of the Countries

where the total count of their movie/TV show is greater than 10.Hence adding more content of these actors will fetch more viewers from their respective countries

Analyse data wrt Rating

Source https://help.netflix.com/en/node/2064/us

```
In [105... df.insert(9,"target aud",'')
In [668... rate idx=df.loc[df["rating"].isin(["PG","G",'TV-Y','TV-Y7','TV-G','TV-PG'])].index
          rate idx
          Int64Index([ 6, 13, 22, 23, 26, 34, 37, 39, 40, 41,
Out[668]:
                     8786, 8787, 8789, 8793, 8795, 8796, 8797, 8800, 8803, 8805],
                    dtype='int64', length=2052)
In [669... df.loc[rate idx,["target aud"]]='Kids'
In [670... rate_idx=df.loc[df["rating"].isin(["PG-13","TV-14"])].index
          rate idx
                             8, 9, 18, 20, 21, 24, 25, 27, 28,
          Int64Index([ 0,
Out[670]:
                     8766, 8767, 8770, 8771, 8772, 8774, 8782, 8794, 8799, 8806],
                    dtype='int64', length=2650)
         df.loc[rate idx,["target aud"]]='Teens'
          df
```

| Out[671]: _ | | show_id | type | title | director | cast | country | date_added | release_year | rating | target_aud | target_aug | duration | listed_in | d |
|-------------|-----|---------|------------|-----------------------------|--------------------|--|------------------|-----------------------|--------------|-----------|------------|------------|--------------|--|----|
| | 0 | s1 | Movie | Dick Johnson Is Dead | Kirsten Johnson | Х | United States | September 25, 2021 | 2020 | PG-13 | Teens | | 90 min | Documentaries | fa |
| | 1 | s2 | TV Show | Blood & Water | Х | Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban | South Africa | September 24, 2021 | 2021 | TV- MA | | | 2 Seasons | International TV Shows, TV Dramas, TV Mysteries | |
| | 2 | s3 | TV Show | Ganglands | Julien Leclercq | Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabi | NaN | September 24, 2021 | 2021 | TV- MA | | | 1 Season | Crime TV Shows, International TV Shows, TV Act | |
| | 3 | s4 | TV Show | Jailbirds New Orleans | X | Х | NaN | September 24, 2021 | 2021 | TV- MA | | | 1 Season | Docuseries, Reality TV | |
| | 4 | s5 | TV Show | Kota Factory | X | Mayur More, Jitendra Kumar, Ranjan Raj, Alam K | India | September 24, 2021 | 2021 | TV- MA | | | 2 Seasons | International TV Shows, Romantic TV Shows, TV | |
| | ••• | | | | | | | | | | | | | | |

| | $show_id$ | type | title | director | cast | country | ${\sf date_added}$ | release_year | rating | target_aud | target_aug | duration | listed_in | d |
|------|-----------|------------|----------------|--------------------|--|------------------|----------------------|--------------|--------|------------|------------|--------------|---|-------------------|
| 8802 | s8803 | Movie | Zodiac | David Fincher | Mark Ruffalo, Jake Gyllenhaal, Robert Downey J | United States | November 20, 2019 | 2007 | R | | | 158 min | Cult Movies, Dramas, Thrillers | |
| 8803 | s8804 | TV Show | Zombie Dumb | Х | Х | NaN | July 1, 2019 | 2018 | TV-Y7 | Kids | | 2 Seasons | Kids' TV, Korean TV Shows, TV Comedies | ٧ |
| 8804 | s8805 | Movie | Zombieland | Ruben Fleischer | Jesse Eisenberg, Woody Harrelson, Emma Stone, | United States | November 1, 2019 | 2009 | R | | | 88 min | Comedies, Horror Movies | I S W O' |
| 8805 | s8806 | Movie | Zoom | Peter Hewitt | Tim Allen, Courteney Cox, Chevy Chase, Kate Ma | United States | January 11, 2020 | 2006 | PG | Kids | | 88 min | Children & Family Movies, Comedies | (SI |
| 8806 | s8807 | Movie | Zubaan | Mozez Singh | Vicky Kaushal, Sarah-Jane Dias, Raaghav Chanan | India | March 2, 2019 | 2015 | TV-14 | Teens | | 111 min | Dramas, International Movies, Music & Musicals | k hi |

8807 rows × 18 columns

In [672...

rate_idx=df.loc[df["rating"].isin(["R","TV-MA","NC-17"])].index
rate_idx

```
Int64Index([
                                     3,
                                           4, 5,
                                                       7, 10, 11, 12, 14,
Out[672]:
                      8762, 8765, 8768, 8769, 8788, 8791, 8798, 8801, 8802, 8804],
                     dtype='int64', length=4009)
 In [673... df.loc[rate idx,["target aud"]]='Adults'
In [679... df.loc[~df["target aud"].isin(["Kids","Adults","Teens"])]["rating"].unique()
          array(['74 min', '84 min', '66 min', 'NR', nan, 'TV-Y7-FV', 'UR'],
Out[679]:
                dtype=object)
          rate_idx=df.loc[df["rating"].isin(["NR","UR"])].index
          rate idx
Out[681]:
 In [682... df.loc[rate idx,["target aud"]]='Adults'
          rate idx=df.loc[df["rating"].isin(['TV-Y7-FV'])].index
 In [685...
          rate idx
          df.loc[rate idx,["target aud"]]='Teens'
          rate idx=df.loc[df["rating"].isin(['74 min', '84 min', '66 min', np.nan])].index
 In [686...
          rate idx
          df.loc[rate_idx,["target_aud"]]='Not Available'
          rating df=country df.merge(df[["target aud", "show id"]])
 In [690...
 In [691... rating df
```

| Out[691]: | | show_id | country | title | type | target_aud |
|-----------|-------|---------|---------------|----------------------|---------|------------|
| | 0 | s1 | United States | Dick Johnson Is Dead | Movie | Teens |
| | 1 | s2 | South Africa | Blood & Water | TV Show | Adults |
| | 2 | s5 | India | Kota Factory | TV Show | Adults |
| | 3 | s8 | United States | Sankofa | Movie | Adults |
| | 4 | s8 | Ghana | Sankofa | Movie | Adults |
| | ••• | | | | | |
| | 10009 | s8802 | Jordan | Zinzana | Movie | Adults |
| | 10010 | s8803 | United States | Zodiac | Movie | Adults |
| | 10011 | s8805 | United States | Zombieland | Movie | Adults |
| | 10012 | s8806 | United States | Zoom | Movie | Kids |

India

Zubaan

Movie

10014 rows × 5 columns

s8807

10013

```
rating_cnt=rating_df.groupby(["country","target_aud"])["target_aud"].count().to_frame()
rating_cnt.rename(columns={"target_aud":"count"},inplace=True)
rating_cnt.reset_index(inplace=True)
to_del=rating_cnt.loc[rating_cnt["country"]==''].index
rating_cnt.drop(to_del,axis=0,inplace=True)
rating_cnt.reset_index(drop=True,inplace=True)
rating_cnt
```

Teens

Out[709]:

| | country | target_aud | count |
|-----|--------------|------------|-------|
| 0 | Afghanistan | Adults | 1 |
| 1 | Albania | Adults | 1 |
| 2 | Algeria | Adults | 2 |
| 3 | Algeria | Teens | 1 |
| 4 | Angola | Adults | 1 |
| ••• | | | |
| 266 | West Germany | Adults | 3 |
| 267 | West Germany | Kids | 1 |
| 268 | West Germany | Teens | 1 |
| 269 | Zimbabwe | Adults | 2 |
| 270 | Zimbabwe | Kids | 1 |

271 rows × 3 columns

```
In [836...
    top_country=rating_cnt.groupby("country")["count"].sum().sort_values(ascending=False).head(20)
    top_country=top_country.reset_index()["country"].to_list()
    top_country
    top_country_df=rating_cnt.loc[rating_cnt["country"].isin(top_country)]
    top_country_df.reset_index(drop=False,inplace=True)
    top_country_df.drop("index",axis=1,inplace=True)
    to_del=top_country_df.loc[top_country_df["target_aud"]=='Not Available'].index
    top_country_df.drop(to_del,axis=0,inplace=True)
    top_country_df
```

Out[836]:

| | country | target_aud | count |
|----|-----------|------------|-------|
| 0 | Argentina | Adults | 71 |
| 1 | Argentina | Kids | 10 |
| 2 | Argentina | Teens | 10 |
| 3 | Australia | Adults | 68 |
| 4 | Australia | Kids | 59 |
| 6 | Australia | Teens | 32 |
| 7 | Brazil | Adults | 64 |
| 8 | Brazil | Kids | 20 |
| 9 | Brazil | Teens | 13 |
| 10 | Canada | Adults | 192 |
| 11 | Canada | Kids | 171 |
| 12 | Canada | Teens | 82 |
| 13 | China | Adults | 56 |
| 14 | China | Kids | 34 |
| 15 | China | Teens | 72 |
| 16 | Egypt | Adults | 33 |
| 17 | Egypt | Kids | 7 |
| 18 | Egypt | Teens | 77 |
| 19 | France | Adults | 227 |
| 20 | France | Kids | 83 |
| 21 | France | Teens | 83 |
| 22 | Germany | Adults | 124 |
| 23 | Germany | Kids | 44 |
| 24 | Germany | Teens | 58 |

| | country | target_aud | count |
|----|-------------|------------|-------|
| 25 | Hong Kong | Adults | 54 |
| 26 | Hong Kong | Kids | 7 |
| 27 | Hong Kong | Teens | 44 |
| 28 | India | Adults | 278 |
| 29 | India | Kids | 184 |
| 30 | India | Teens | 584 |
| 31 | Indonesia | Adults | 21 |
| 32 | Indonesia | Kids | 30 |
| 33 | Indonesia | Teens | 39 |
| 34 | Italy | Adults | 57 |
| 35 | Italy | Kids | 18 |
| 37 | Italy | Teens | 24 |
| 38 | Japan | Adults | 110 |
| 39 | Japan | Kids | 99 |
| 41 | Japan | Teens | 108 |
| 42 | Mexico | Adults | 120 |
| 43 | Mexico | Kids | 23 |
| 44 | Mexico | Teens | 26 |
| 45 | Nigeria | Adults | 45 |
| 46 | Nigeria | Kids | 11 |
| 47 | Nigeria | Teens | 47 |
| 48 | South Korea | Adults | 98 |
| 49 | South Korea | Kids | 46 |
| 50 | South Korea | Teens | 87 |

| | country | target_aud | count |
|----|----------------|------------|-------|
| 51 | Spain | Adults | 185 |
| 52 | Spain | Kids | 24 |
| 53 | Spain | Teens | 23 |
| 54 | Turkey | Adults | 71 |
| 55 | Turkey | Kids | 10 |
| 56 | Turkey | Teens | 32 |
| 57 | United Kingdom | Adults | 409 |
| 58 | United Kingdom | Kids | 208 |
| 59 | United Kingdom | Teens | 187 |
| 60 | United States | Adults | 1805 |
| 61 | United States | Kids | 949 |
| 63 | United States | Teens | 932 |

```
In [837... def cal_percent(x):
    ans=x*100/x.sum()
    return round(ans,2)

In [838... top_country_df["percent"]=top_country_df.groupby("country")["count"].transform(cal_percent)

In [840... heatmap_df=top_country_df.pivot_table(index=["country"],columns="target_aud",values="percent")
    heatmap_df=heatmap_df[["Kids","Teens","Adults"]]
    heatmap_df
```

Kids Teens Adults target aud country Argentina 10.99 10.99 78.02 Australia 37.11 20.13 42.77 20.62 13.40 65.98 Brazil **Canada** 38.43 18.43 43.15 **China** 20.99 44.44 34.57 **Egypt** 5.98 65.81 28.21 **France** 21.12 21.12 57.76 **Germany** 19.47 25.66 54.87 **Hong Kong** 6.67 41.90 51.43 **India** 17.59 55.83 26.58 43.33 Indonesia 33.33 23.33 **Italy** 18.18 24.24 57.58 34.07 34.70 **Japan** 31.23 **Mexico** 13.61 15.38 71.01 Nigeria 10.68 45.63 43.69 South Korea 19.91 37.66 42.42 **Spain** 10.34 9.91 79.74 Turkey 8.85 28.32 62.83 United Kingdom 25.87 23.26 50.87 **United States** 25.75 25.28 48.97

Out[840]:

```
In [851... sns.heatmap(heatmap_df,annot=True,cmap="Blues",vmin=0,vmax=100)
    plt.xlabel('')
    plt.ylabel('')
    plt.title("Country Vs Percentage Content in Netflix for each Age Group")
```

plt.show()

Country Vs Percentage Content in Netflix for each Age Group

| | Country Vs F | ercentage Content in Netflix for each | 1 Age Group | 100 |
|------------------|--------------|---------------------------------------|-------------|------|
| Argentina - | 11 | 11 | 78 | 100 |
| Australia - | 37 | 20 | 43 | |
| Brazil - | 21 | 13 | 66 | |
| Canada - | 38 | 18 | 43 | |
| China - | 21 | 44 | 35 | - 80 |
| Egypt - | 6 | 66 | 28 | |
| France - | 21 | 21 | 58 | |
| Germany - | 19 | 26 | 55 | - 60 |
| Hong Kong - | 6.7 | 42 | 51 | - 60 |
| India - | 18 | 56 | 27 | |
| Indonesia - | 33 | 43 | 23 | |
| ltaly - | 18 | 24 | 58 | - 40 |
| Japan - | 31 | 34 | 35 | 40 |
| Mexico - | 14 | 15 | 71 | |
| Nigeria - | 11 | 46 | 44 | |
| South Korea - | 20 | 38 | 42 | - 20 |
| Spain - | 10 | 9.9 | 80 | 20 |
| Turkey - | 8.8 | 28 | 63 | |
| United Kingdom - | 26 | 23 | 51 | |
| United States - | | 25 | 49 | -0 |
| | Kids | Teens | Adults | -0 |

Recommendation-Most of the Asian countries like India, Indonesia, Hongkong, China targets on Teen Audiences, while most of the Western and European countries targets Adult audiences. So adding content belonging to those categories for respective countries can fetch more audience

Recommendation -It is also observed that countries like Spain, Mexico, Argentina have very less content added for the Teen audiences compared to other age groups. Hence adding more content for them can improve the business for Netflix in those countries.

Recommendation-Countries like Turkey, Hongkong, Egypt have less content added for Kids compared to other age groups. Hence adding more content in for them as well can improve the business in those countries. For instance country like Egypt has a higher density of population in 5-14 Age group (Source: https://www.populationpyramid.net/egypt/2020/)

```
In [144...
con_show=country_df.groupby("title")["title"].count().to_frame()
con_show.rename(columns={"title":"No:of country"},inplace=True)
con_show.reset_index(inplace=True)
to_show=con_show.sort_values(["No:of country"],ascending=False).head(20)
to_show.reset_index(drop=True,inplace=True)

to_show
```

| Out[144]: | | title | No:of country |
|-----------|----|--|---------------|
| | 0 | Barbecue | 12 |
| | 1 | The Look of Silence | 10 |
| | 2 | The Professor and the Madman | 8 |
| | 3 | Shaun the Sheep | 8 |
| | 4 | Domino | 7 |
| | 5 | The Take | 7 |
| | 6 | The Congress | 7 |
| | 7 | Arctic Dogs | 7 |
| | 8 | The Breadwinner | 7 |
| | 9 | Nymphomaniac: Volume II | 6 |
| | 10 | The Command | 6 |
| | 11 | Sankofa | 6 |
| | 12 | Stop at Nothing: The Lance Armstrong Story | 6 |
| | 13 | Another Forever | 6 |
| | 14 | Ultimate Beastmaster | 6 |
| | 15 | The Danish Girl | 6 |
| | 16 | A Sort of Family | 6 |
| | 17 | Wadjda | 6 |
| | 18 | Frozen Planet: The Epic Journey | 6 |

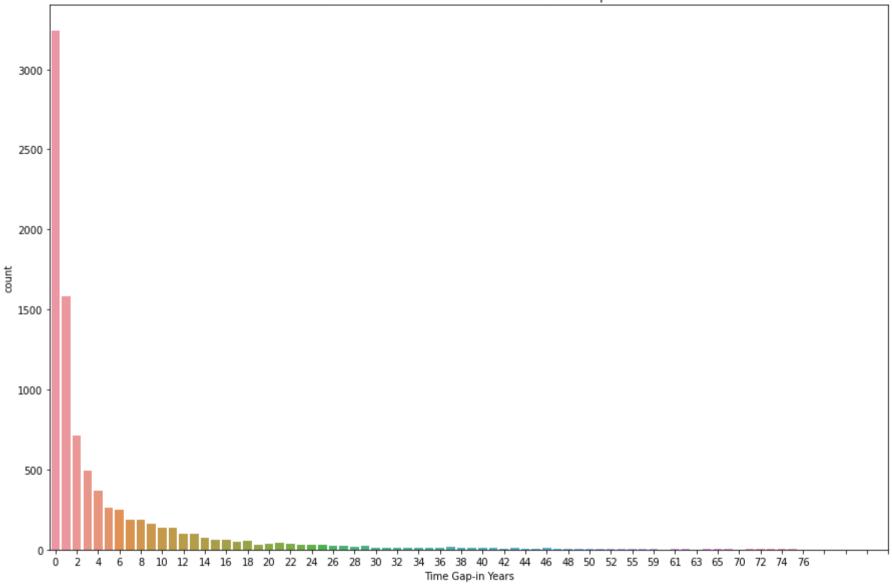
Beyond Skyline

19

Recommnebdation-The above shows the movies and the number of their producing countries. Adding content which was produced by multiple countries can improve business

Analyse wrt to time gap

Time gap :difference between the release year and year the content was added in the platform



Recommendation-Adding newer content in the platform favours more audience. The above plot proves the same.