1. Problem Statement

Netflix wants to increase its growth and wants to decide which type of shows/movies to produce. Also, find other strategies to push the growth.

- 2. The shape of data, data types of all the attributes, conversion of categorical attributes to 'category' (If required), missing value detection, statistical
 - 2.1 The shape of the data frame is 8807x12
 - 2.2 Used head function to observe all the columns

:												
	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, Cape Town t
2	s3	TV Show	Ganglands	Julien Leclercq	Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabl	NaN	September 24, 2021	2021	TV- MA	1 Season	Crime TV Shows, International TV Shows, TV Act	To protect hi 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, flirtation and tollet 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

2.3 Used describe function to a statistical summary of all the rows.

In [31]: df.describe(include='all') C:\Users\gokul\AppData\Local\Temp/ipykernel_25772/2884002236.py:1: FutureWarning: Treating datetime data as categorical rather than numeric in '.describe' is deprecated and will be removed in a future version of pandas. Specify 'datetime_is_numeric=True' to silence this warning and adopt the future behavior now.

df.describe(include='all') Out[31]: show id type cast country date added release year rating duration title director listed in duration number year added month added 8797 8797 8797.000000 8797 8797 4529 16 Dick Dramas. TV- 1 MA Season top s1 Movie Joh other unknown NaN Interna NaN NaN July Is Dead 6131 NaN NaN 2008-01-01 00:00:00 NaN 2021-09-25 00:00:00 NaN last NaN NaN 2014.183472 NaN NaN NaN 69.921792 2018.871888 std NaN NaN NaN NaN NaN NaN NaN 8 822191 NaN NaN 50 788599 1 574243 NaN NaN NaN NaN NaN NaN NaN 1925.000000 NaN NaN 1.000000 2008.000000 min NaN NaN NaN NaN 2017.000000 NaN 50% NaN NaN NaN NaN NaN NaN 88 000000 2019 000000 NaN NaN NaN NaN 2019.000000 NaN NaN 106.000000 2020.000000 NaN 2.4 Used dtypes function to observe the data types of each column

```
In [9]: df.dtypes
Out[9]: show_id
                      object
                    object
       type
       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
```

2.5 Used df.isna().sum()to calculate the number of null values in each column.

Note: I have done some data manipulation because of which the number of null values has reduced.

3. Non-Graphical Analysis and Missing Value & Outlier check

Each columns are check with function value_count().

```
In [10]: #Checking for irregularity in data-1
         df['type'].value_counts()
Out[10]: Movie
                    6131
         TV Show
                    2676
         Name: type, dtype: int64
In [11]: #Checking for irregularity in data-2
         df['release_year'].value_counts()
Out[11]: 2018
                 1147
         2017
                 1032
         2019
                 1030
         2020
                  953
         2016
                  902
         1959
         1925
                    1
         1961
                    1
         1947
                    1
         1966
         Name: release_year, Length: 74, dtype: int64
```

In type, release_year column no problem was found.



But in the rating column, 3 row has a duration in it. So that 3 rows were identified. In the sam e row, duration was missing.

```
In [14]: df.iat[5541,9]=df.iat[5541,8]
    df.iat[5794,9]=df.iat[5794,8]
    df.iat[5813,9]=df.iat[5813,8]
In [15]: df.iat[5541,8]="Unknown"
    df.iat[5794,8]="Unknown"
    df.iat[5813,8]="Unknown"
```

So, the duration was transferred from the rating column to the duration column and the rating was marked as unknown.

Again, the rating column was checked once again to ensure the removal of durations.

```
In [17]: df['duration_number']=df['duration'].apply(lambda x: int(x.split(" ")[0]))
df_7=df[df['type']=='Movie']
sns.set(rc = ('figure.figsize':(5,5)))
sns.boxplot(data=df_7,x='duration_number')
#Value can not be considered as outliers. These will be exeptional cases

Out[17]: <AxesSubplot:xlabel='duration_number'>

0 50 100 150 200 250 300
duration_number
```

To check the outliners in the duration columns box for Movies and TV shows separately. The plot showed many outliers. They can't to treated as outliers and removed since there is much value and has to be considered exceptional cases.



Same is the case with TV Shows. They can't to treated as outliers and removed since there is much value and has to be considered exceptional cases.

Value count of duration was checked for any irregularities and found none.

```
In [21]: #Handling null values

df["director"]=df["director"].fillna("other")

df["rating"]=df["rating"].fillna("unknown")

df["cast"]=df["cast"].fillna("unknown")

df["cully"]=df["country"].fillna("unknown")

df["release_year"]=df["release_year"].fillna(df["release_year"].median())

In [22]: #Dropping duplicate values

df=df.drop_duplicates()

In [23]: df.shape

Out[23]: (8807, 13)
```

All the null values were replaced with as shown above.

```
In [23]: df.shape
Out[23]: (8807, 13)

In [24]: # Removing unnecessary columns
    df.drop(columns=['description',],inplace=True)

In [25]: #Adding columns
    df["date_added"] = pd.to_datetime(df["date_added"])
    df["year_added"] = df["date_added"].dt.year
    df["month_added"] = df["date_added"].dt.month_name()
In [26]: #Droping row with null values in rating and duration(no. of row dropped only be 3 & 4)
    df.dropna(subset=["date_added"],inplace=True)
```

Description column was dropped, the data type of date_added was converted to datetime from object ,and add 2 additional columns for year_added & month_added.

Then it was checked whether all the null values were removed.

```
In [37]: df_country.country.value_counts().iloc[:10]
Out[37]: United States 3205
         India
         unknown
         United Kingdom 627
United States 479
         Canada
                            271
                            258
         Japan
         France
                            212
                       21.
181
         South Korea
          France
         Name: country, dtype: int64
In [38]: #United States is repeated
         df_country[df_country.country==" United States"]="United States"
```

During data pre-processing, it was found that the United States occurred twice because of extra white space in front. So, the same was corrected.

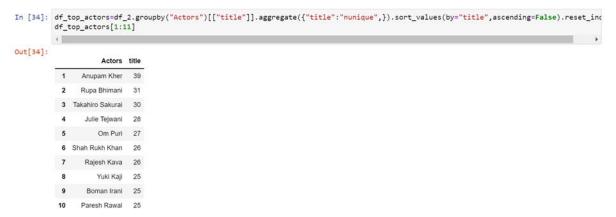
4. Prepossessing, Visual Analysis, and Insights

Cast and directors were unpacked and merged into a single data frame.

Count and listed in were unpacked and merged into a single data frame.

Note:

Every column was not merged into the same data frame due to the limitation of RAM.



The top 10 actors were identified.

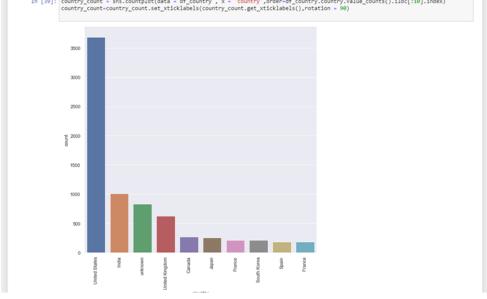
This table shows the top 10 actors who have acted in most contents. There is no large variation number of the content the actors have acted.



The top 10 directors were identified.

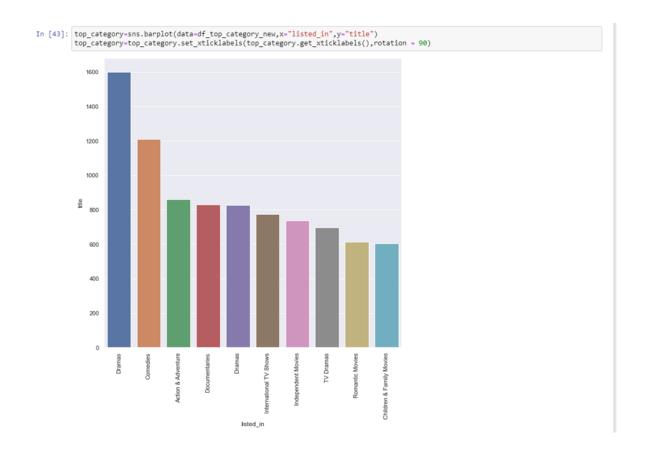
This table shows the top 10 actors who have acted in most contents. Here also there is no large variation number of the content directed by each director.





The country

for which the most content is added was identified. The table shows the content added for each country. The USA has the most content released for it followed by India and UK. This might be because Netflix was stated in the US and most of the content will be from Hollywood movies. Apart from the US rest of the countries are showing no large variations.

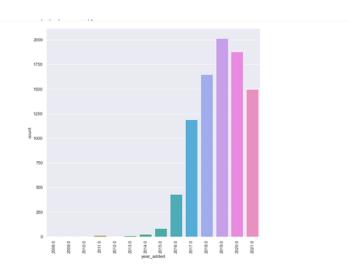


Most watched genres were identified. Here we can see dramas being watched followed by comedies, action & adventure, and documentaries. Dramas are watched far more than any other genre on Netflix.

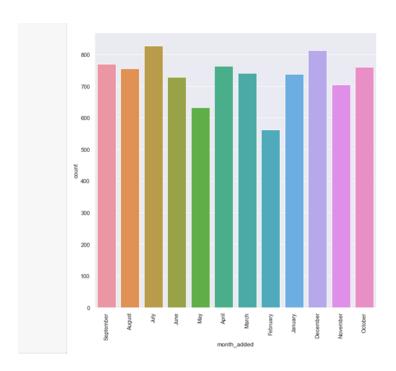


Most rated categories of content were identified. TV-14 and TV-MA have the highest number of content in Netflix. These are more Mature

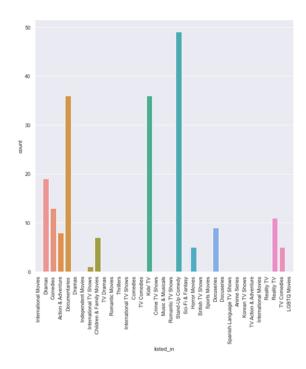
audiences and audiences requiring parental guidance & above 14 year of age.



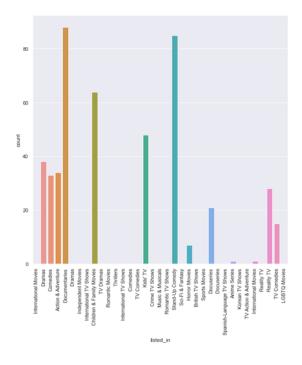
The year with the most added content was identified. It shows the number of content added in each year. Till 2019 the content added to Netflix was going up exponentially but in 2020 there was a decrease in the number of content added to Netflix compared to 2020 and further reduced to 2021.



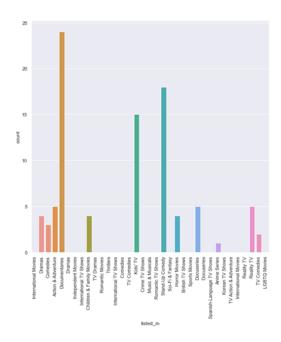
Which month has seen the most content added was identified. The month vs addition of content. Most contents are added in the month of Dec and Jul. And least content was added in the month of Feb and May.



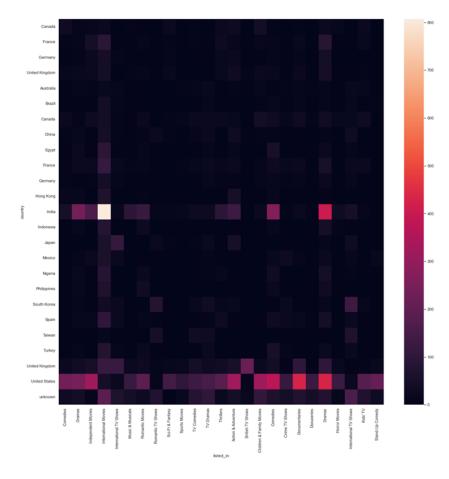
The most watched category in India was identified. The plot shows no. of content vs each Genre. Most uploaded content in India is Stand-upcomedy.



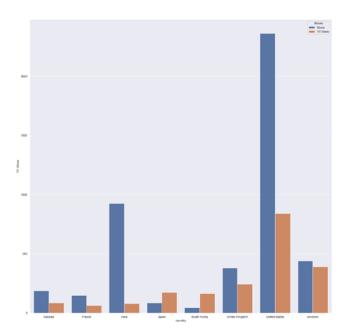
The most watched category in the USA was identified. The plot shows no. of content vs each Genre. Most uploaded content in India is documentaries followed by Stand-up-comedy.



The most watched category in the UK was identified. The plot shows no. of content vs each Genre. Most uploaded content in India is documentaries followed by Stand-up-comedy.

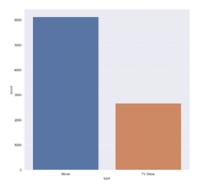


Heatmap was plotted to identify the popular genre in the top most released countries. The heat map can be used to identify which is the most popular content in each country. Light color shows more popular genres and darker the plot shows least watched.



Movies v/s TV show release in each country(counties having at least 200 content added) was identified. The plot shows counties v/s movies/TV shows released. Movies are generally more favored than TV shows except in Japan and South Korea.

5.5. Business Insight



More movies are being added than TV-Shows if you check the overall picture. Movies are generally more favored than TV shows except in Japan and South Korea.

The USA has the most content released for it followed by India and UK. This might be because Netflix was stated in the US and most of the content will be from Hollywood movies. Apart from the US rest of the countries are showing no large variations.

Dramas are watched far more than any other genre on Netflix.

Actors are not producing a large impact on the success of content as there is no actor who has acted in more the 32-content compared to the dataset size of more than 8000.

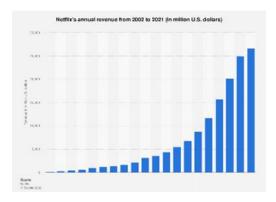
These are more Mature audiences and audiences requiring parental guidance & above 14 years of age.

Even though India has large population, penetration is comparatively less. This can be seen from the number of content release in US v/s India.

Till 2019 the content added to Netflix was going up exponentially by 2020 there was a decrease in the number of contents added to Netflix compared to 2020 and further reduced to 2021.

Most contents are added in the month of Dec and Jul. And least content was added in the month of Feb and May.

More content is being watched by people who are more than 14 years of age.



Source: https://www.statista.com/statistics/272545/annual-revenue-of-netflix/

More content may not produce more growth. Even the revenue was continuously growing. The number of content added reduced in 2020 and 2021.

5. 6. Recommendations

- a. More quality content should be produced using the best directors such as Rajiv Chilaka, Jan Suter, Raðl Campos, etc.
- b. More content should be released in the holiday season.
- c. Movies should be more preferred than TV-Shows(Expect in Japan and South Korea)
- d. Country wise target should be made while making content by referring to the heatmap given above. As a whole, more drama content should be produced
- e. Releasing of new content should be kept on the weekend for more engagement.
- f. New actors can be tried out.
- g. India's audience should be more targeted.
- h. More content should be made for adolescents and adults.