

The recent pandemic led to the usage of OTT a lot, here I have tried to do an content analysis of Netflix, this analysis focuses on what content Netflix targets and what is their target audience. I have also done a comparison with their direct competetion Amazon Prime.

NETFLIX DATASET: (https://www.kaggle.com/datasets/shivamb/netflix-shows)).

AMAZON PRIME DATASET: (https://www.kaggle.com/datasets/shivamb/amazon-prime-movies-and-tv-shows (https://www.kaggle.com/datasets/shivamb/amazon-prime-movies-and-tv-shows))

```
In [2]: #libraries
   import pandas as pd
   import numpy as np
   import seaborn as sns
   import matplotlib.pyplot as plt
%matplotlib inline
```

The cleaning of Datasets is done seperately, Netflix followed by Amazon Prime to avoid confusion to readers.

```
In [3]: #reading the csv file
data = pd.read_csv(r'netflix_titles.csv')
```

The data set contains 8807 values spread across 12 Columns

In [5]: #EDA OF THE DATASET
 data.head(5)

Out[5]:		show_id	type	title	director	cast	country	date_added	release_year	rating	dura
	0	s 1	Movie	Dick Johnson Is Dead	Kirsten Johnson	NaN	United States	September 25, 2021	2020	PG- 13	90
	1	s2	TV Show	Blood & Water	NaN	Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban	South Africa	September 24, 2021	2021	TV- MA	Sea
	2	s3	TV Show	Ganglands	Julien Leclercq	Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabi	NaN	September 24, 2021	2021	TV- MA	Sea
	3	s4	TV Show	Jailbirds New Orleans	NaN	NaN	NaN	September 24, 2021	2021	TV- MA	Sea
	4	s 5	TV Show	Kota Factory	NaN	Mayur More, Jitendra Kumar, Ranjan Raj, Alam K	India	September 24, 2021	2021	TV- MA	Sea

In [6]: #shape of the dataset
data.shape

Out[6]: (8807, 12)

In [7]: #total values present in the dataset i.e the size of the dataset data.size

Out[7]: 105684

In [6]: #info of the dataset that shows the count of values in the dataset and the dataty
data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8807 entries, 0 to 8806
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype				
0	show_id	8807 non-null	object				
1	type	8807 non-null	object				
2	title	8807 non-null	object				
3	director	6173 non-null	object				
4	cast	7982 non-null	object				
5	country	7976 non-null	object				
6	date_added	8797 non-null	object				
7	release_year	8807 non-null	int64				
8	rating	8803 non-null	object				
9	duration	8804 non-null	object				
10	listed_in	8807 non-null	object				
11	description	8807 non-null	object				
dtvn	dtynes: int64(1), object(11)						

dtypes: int64(1), object(11)
memory usage: 825.8+ KB

```
In [8]: #name of the columns present in the dataset
data.columns
```

In [59]: #cleaning of data #to check duplicate values x = data.duplicated().value_counts() print(x)

#there are no duplicates found in this dataset, if any:
data.drop_duplicates()

False 8807 dtype: int64

	71									
Out[59]:		show_id	type	title	director	cast	country	date_added	release_year	rati
	0	s 1	Movie	Dick Johnson Is Dead	Kirsten Johnson	NA	United States	September 25, 2021	2020	F
	1	s2	TV Show	Blood & Water	NA	Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban	South Africa	September 24, 2021	2021	I
	2	s3	TV Show	Ganglands	Julien Leclercq	Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabi	NA	September 24, 2021	2021	I
	3	s4	TV Show	Jailbirds New Orleans	NA	NA	NA	September 24, 2021	2021	- I
	4	s5	TV Show	Kota Factory	NA	Mayur More, Jitendra Kumar, Ranjan Raj, Alam K	India	September 24, 2021	2021	-
	8802	s8803	Movie	Zodiac	David Fincher	Mark Ruffalo, Jake Gyllenhaal, Robert Downey J	United States	November 20, 2019	2007	
	8803	s8804	TV Show	Zombie Dumb	NA	NA	NA	July 1, 2019	2018	TV-

	show_id	type	title	director	cast	country	date_added	release_year	rati	
8804	s8805	Movie	Zombieland	Ruben Fleischer	Jesse Eisenberg, Woody Harrelson, Emma Stone,	United States	November 1, 2019	2009		
8805	s8806	Movie	Zoom	Peter Hewitt	Tim Allen, Courteney Cox, Chevy Chase, Kate Ma	United States	January 11, 2020	2006		
8806	s8807	Movie	Zubaan	Mozez Singh	Vicky Kaushal, Sarah- Jane Dias, Raaghav Chanan	India	March 2, 2019	2015	TV	
8807 rows × 12 columns										~
4									•	

In [10]: #now to check if there are any null values in the dataset nul = data.isnull().sum() print(nul)

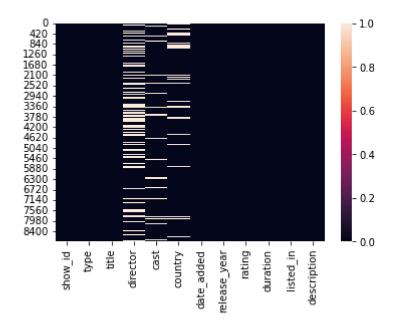
show_id	0
type	0
title	0
director	2634
cast	825
country	831
date_added	10
release_year	0
rating	4
duration	3
listed_in	0
description	0
dtype: int64	

I have created an heatmap to show where the null values are concentrated more, we can see that it is concentrated more in the director column.

In [64]: #heatmap to show where the null values are concentrated more
sns.heatmap(data.isnull())

#We can see that the null values are concentrated more in the director column

Out[64]: <AxesSubplot:>



My aim in any datset is to not remove any data unnecessarily, so we try to retain those null values and fill it with 'NA' indicating only those values are not available.

```
In [11]: data.fillna('NA',inplace = True)
         data.isnull().sum()
Out[11]: show id
                          0
         type
         title
                          0
         director
                          0
         cast
         country
                          0
         date_added
                          0
         release_year
         rating
                          0
         duration
                          0
         listed in
         description
         dtype: int64
In [12]: #To check how many values we actually don't have, significant number of values ar
         data['director'].str.contains('NA').sum()
Out[12]: 2634
```

I have saved the file to a CSV file, for the visualisation part I will be using PowerBi.

```
data.to csv('netedited.csv',index=False)
In [17]:
In [13]: data['listed_in'].head(10)
Out[13]: 0
                                                   Documentaries
         1
                International TV Shows, TV Dramas, TV Mysteries
              Crime TV Shows, International TV Shows, TV Act...
         2
         3
                                          Docuseries, Reality TV
              International TV Shows, Romantic TV Shows, TV ...
                              TV Dramas, TV Horror, TV Mysteries
         6
                                        Children & Family Movies
         7
               Dramas, Independent Movies, International Movies
                                    British TV Shows, Reality TV
         8
                                                Comedies, Dramas
         Name: listed_in, dtype: object
```

If you observe the listed_in(genre) column you can see one TV show can be put under multiple genre, so here I am creating a new dataframe and spliting those genres seperately and mapping them to their show id so that we do not create a data mess of it.

```
In [10]: #Listing unpacking and making a dataframe

df_listing = pd.DataFrame()

df_listing['listed_in'] = data['listed_in'].str.split(', ').explode()

df_listing['show_id'] = data['show_id']

df_listing['title'] = data['title']

#to check the count of each category

xyz = df_listing['listed_in'].value_counts()

#print(xyz)

print(df_listing['listed_in'].nunique())

#saving it to a csv file

df_listing.to_csv('df_listing.csv',index = False)
```

42

It is the same case with the country column so I have followed similar steps and created a new dataframe for it.

```
In [14]: #country column unpacking and making a dataframe
    df_country = pd.DataFrame()
    df_country['country'] = data['country'].str.split(',').explode()
    df_country['show_id'] = data['show_id']
    df_country['title'] = data['title']
    countt = df_country['country'].value_counts().head(10)
    print(countt)
    df_country['country'].nunique()

#Saving it to CSV file
    df_country.to_csv('df_country.csv')
United States 3211
```

```
India
                   1008
                   831
NA
United Kingdom
                   628
United States
                   479
Canada
                   271
Japan
                    259
France
                    212
South Korea
                   211
 France
                    181
Name: country, dtype: int64
```

Out[14]: 198

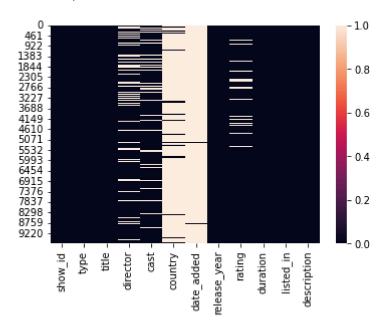
In this duration table we can observe that Movies and TV Shows are in the same column i.e it is in min and seasons, so I have created to seprate dataframes respectively to remove 'min' and 'season' and convert it to numerical type for easy calculation.

```
In [17]: #creating duration table for movies
         duration_df = pd.DataFrame()
         duration_df['show_id'] = data['show_id']
         duration df['duration'] = data['duration'].replace(['Season 1','Season 2','Season'])
         duration_df['duration'] = duration_df['duration'].str.split(" ").str.get(0)
         duration_df.dropna(inplace = True)
         duration_df['duration']
         #saving it as a CSV file
         duration_df.to_csv("dmovies.csv")
Out[17]: 0
                  90
                  91
         6
         7
                  125
         9
                 104
         12
                 127
                 . . .
         8801
                  96
         8802
                 158
         8804
                  88
                  88
         8805
         8806
                 111
         Name: duration, Length: 6140, dtype: object
In [43]: #creating duration table for TVShow
         tv df = pd.DataFrame()
         tv df['Seasons'] = data['duration'].str.split(' ').str.get(0)
         #I have replaced those 3 values with average seasons of the available dataset
         tv_df['Seasons'].replace("NA",np.nan,inplace = True)
         tv df["Seasons"].fillna("1.8",inplace = True)
         tv_df['Seasons'] = pd.to_numeric(tv_df['Seasons'])
         tv df['Seasons'] = tv df[(tv df["Seasons"]>=1) & (tv df['Seasons']<=17)]
         tv df['show id'] = data['show id']
         tv df.dropna(inplace = True)
         tv_df.count()
         tv_df.to_csv("tvduration.csv")
Out[43]: Seasons
                    2702
         show id
                    2702
         dtype: int64
         AMAZON PRIME DATASET AND DATA CLEANING
In [44]: | data_am = pd.read_csv('amazon_prime_titles.csv')
```

```
In [45]: data_am.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 9668 entries, 0 to 9667
         Data columns (total 12 columns):
          #
              Column
                            Non-Null Count Dtype
              -----
                             -----
          0
              show_id
                            9668 non-null
                                            object
          1
                            9668 non-null
                                            object
              type
          2
              title
                            9668 non-null
                                            object
          3
              director
                            7586 non-null
                                            object
          4
              cast
                            8435 non-null
                                            object
          5
              country
                            672 non-null
                                            object
          6
              date added
                            155 non-null
                                            object
          7
                                            int64
              release_year 9668 non-null
          8
              rating
                            9331 non-null
                                            object
          9
              duration
                            9668 non-null
                                            object
          10 listed_in
                            9668 non-null
                                            object
          11 description
                            9668 non-null
                                            object
         dtypes: int64(1), object(11)
         memory usage: 906.5+ KB
         #to check duplicate values
In [47]:
         data_am.duplicated().value_counts()
Out[47]: False
                  9668
         dtype: int64
         #To check the number of NULL values
In [48]:
         data am.isnull().sum()
Out[48]: show id
                            0
         type
                            0
         title
                            0
         director
                         2082
         cast
                         1233
         country
                         8996
                         9513
         date added
         release_year
                            0
         rating
                          337
         duration
                            0
         listed in
                            0
         description
                            0
         dtype: int64
```

```
In [51]: sns.heatmap(data_am.isnull())
```

Out[51]: <AxesSubplot:>



```
In [52]: #Filling the null values with NA
data_am.fillna("NA",inplace = True)
data_am.isnull().sum()
```

```
Out[52]: show_id
                           0
          type
                           0
          title
                           0
                           0
          director
                           0
          cast
          country
                           0
          date_added
                           0
          release_year
                           0
                           0
          rating
          duration
                           0
          listed_in
                           0
          description
                           0
          dtype: int64
```

```
In [53]: #genre table
         lista_df = pd.DataFrame()
         lista df['genre'] = data_am['listed_in'].str.split(', ').explode()
         lista_df['show_id'] = data_am['show_id']
         #grouping it accordiing to show_id
         groupd = lista_df.groupby('show_id')
         groupd['genre'].value_counts()
         lista_df['genre'].nunique()
         #CSV Genre Table
         #Lista_df.to_csv('amazongenre.csv')
         #sql: select show_id,count(genre) from lista_df group by show_id
Out[53]: 31
In [55]: |#Duration in min
         dur df = pd.DataFrame()
         #data am[data am['duration'].str.contains('Seasons')]
         #Keeping the dataset to min and season differently
         dur_df['dmin'] = data_am['duration'].replace(['11 Seasons','12 Seasons','14 Seasons']
         dur df.dropna(inplace = True)
         dur_df[dur_df['dmin'].str.contains('Season')]
         #Now removing the min in the table and convert it into numeric type
         dur df['dmin'] = dur df['dmin'].str.split(' ').str.get(0)
         dur df['dmin'] = pd.to numeric(dur df['dmin'])
         dur df.dtypes
         dur df['show id'] = data am['show id']
         dur df['dmin'].head()
         #creating a CSV file for the duration_min:
         #dur df.to csv('amazonmovie.csv')
Out[55]: 0
              113
              110
         1
         2
               74
         3
               69
               45
```

Name: dmin, dtype: int64

In [56]: #Creating a timeframe table for TV Show for seasons

dur tv['Seasons'] = pd.to numeric(dur tv['Seasons'])

dur tv = pd.DataFrame()

```
#What I have done here is, we know that seasons are between the number 1 to 29, s
         dur_tv['Seasons'] = dur_tv[(dur_tv['Seasons']>=1) & (dur_tv['Seasons']<=29)]</pre>
         dur_tv['show_id'] = data_am['show_id']
         #to reference if the code has matchded the correct show id
         dur_tv
         #We can see it has created NaN values to the empty spaces so we drop those values
         dur_tv.dropna(inplace = True)
         #These are the number of TV Shows that are present
         dur_tv.count()
Out[56]: Seasons
                    2204
         show id
                    2204
         dtype: int64
In [57]: #Country Table
         countrydf = pd.DataFrame()
         #filling up the null values in the dataset
         data am['country'].fillna("NoData",inplace = True)
         data am['country'].isnull().unique()
         #seperating the values
         countrydf['country'] = data_am['country'].str.split(',').explode()
         countrydf['country'].nunique()
         countrydf['show id'] = data am['show id']
         countrydf.head()
```

dur_tv['Seasons'] = data_am['duration'].str.split(' ').str.get(0)

Out[57]:

	country	show_id
0	Canada	s1
1	India	s2
2	United States	s3
3	United States	s4
4	United Kingdom	s5

I have done the Visualisation part of this dataset using PowerBi, I will insert images of the report and I have attached the file of the report at the start.

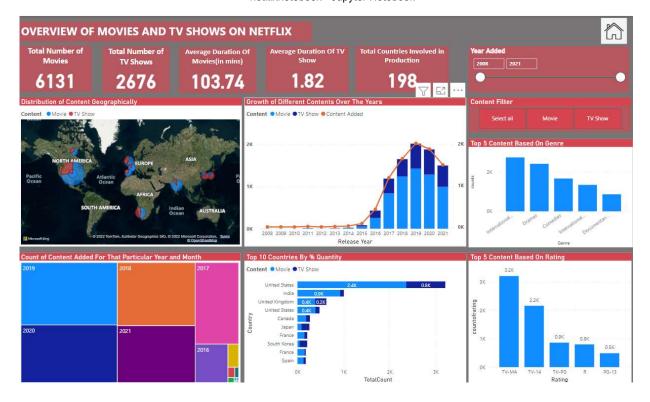
Since Netflix debuted in 1997, the streaming market started to soar in 2015 and 2016 with tremendous growth as they were gradually released in international countries. Netflix Movies and TV Shows Analytics dashboards detail the growth of international digital contents on Netflix over the years. It helps users understand and study the development of the streaming services ecosystem that is associated with each content/genre/maturity rating type/IMDB rating, segmented based on different factors.

Introduction Page: This Page was designed for users to navigate easily through the sections



Overview Of Data:

- 1)Out of the total available data, we can clearly see that Netflix has a lot more Movies(6131) in it's list compared to the TV Shows(2676).
- 2)There has been a change in approch to this from Netflix from the year 2016, it is slowly shifting it's attention to add TV Shows, the pandemic was an added advantage to their change, so we can see there is a slight rise in the number of TV shows added from 2018 2020.
- 3)Since Netflix is based off USA, the number of content producers is significantly more there a total of 3211, which is followed by India.
- 4)Netflix has focused on the International Movie genre to attract crowd all over the world, focus is to increase their target audience seeing the number of International Movies are more in it's content list.
- 5)The average duration of a Movie in Netflix is 103.74 minutes and where as a TV show averages 1.82, this indicates that Netflix understands that people don't have time to sit for long hours.



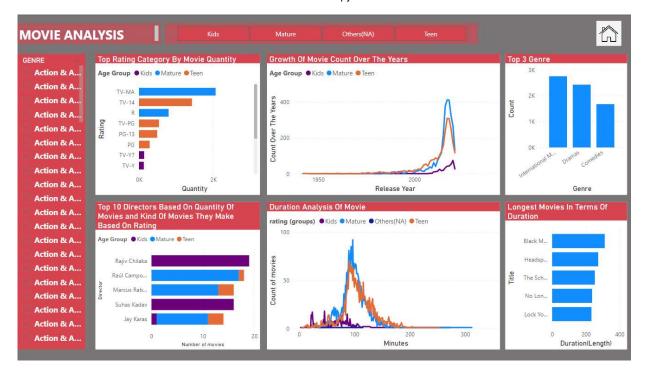
MOVIE ANALYSIS:

Note: I have grouped the ratings as Kids, Teens, Mature and the Unavailable data as Others.

Kids: G,TV-G,TV-Y,TV-Y7,TV-Y7-FV

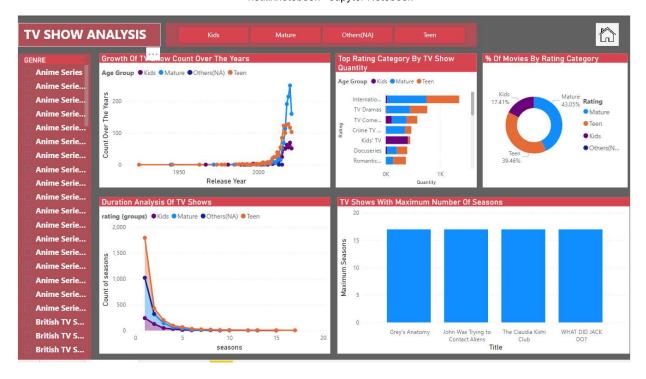
Teen: PG, PG-13,TV-14,TV-PG Mature: R,NC-17,NR,TV-MA,UR

- 1)Netflix focuses largerly on the Mature group category, it shows that Netflix is targeting the group of 18-40+. It shows that there is a large section of people in this age group who are addicted to watching Movies.
- 2)The line graph also clearly indicates there is a significantly higher number of Movies that is being added of the maturity content in recent years.
- 3)Director Rajiv Chilika has a total of 19 Movies which is available in Netflix making him the highest.
- 4)Top 3 Genre's are International Movie, Dramas and Comedy.
- 5)The longest movie is Black Mirror which is 312 minutes long. The average movie duration in Netflix is concentrated in the region of 100-120 minutes.



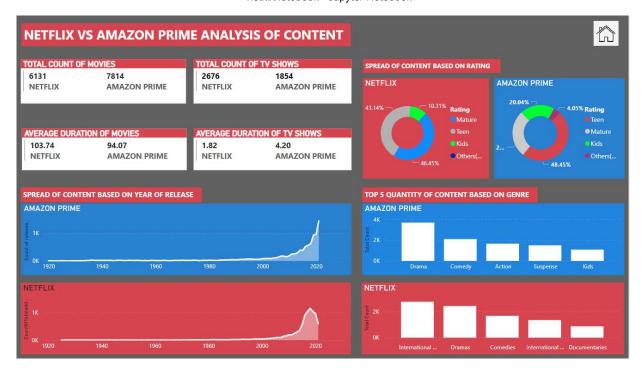
TV SHOW ANALYSIS:

- 1)The TV Shows have a good balance of Content in them focusing on all kinds of Groups, Netflix has been trying to expand it's wings to all categories and there is a significant improvement towards that.
- 2)The duration of a TV Show might be a subject to disscussion since people tend to get attracted to the characters having short duration TV Show might not exactly be the right way to go about it, the average being close to 2. Netflix might want to look into that.
- 3)The longest TV Show Netflix has is Grey's Anatomy, John Was Trying To Contact Aliens, The Claudia Kishi Club and What did Jack Do?
- 4)As I mentioned the percentage of division is good Mature(43.05%) toping the list compared to Kids(17.41%) and Teen(39.46)



NETFLIX VS AMAZON PRIME CONTENT ANALYSIS:

- 1)Netflix has a total content of 8807, Amazon Prime has 9668 which includes both TV Show and Movies. Amazon holding a slight edge in this.
- 2)Netflix has an average duration of 103.74 min compared to 94.07 min of Amazon Prime, the difference comes in the TV Show duration, Amazon prefers shows which are longer averaging 4.20, this is a plus point considering people get attached to the shows.
- 3)We can see a dip in content added from the year 2020 to 2021 in Netflix where has the graph of Amazon has not dipped and continues to grow.
- 4)The major difference we can see is the target audience, Netflix prefers to target the Mature audience(18+) whereas Amazon targets the Teen group(7-18), very interesting approach taken by Amazon to target this section, I feel teens might get distracted and not focus on their career or studies given the lack of maturity at that age.
- 5)Netflix has stated previously targets International genre whereas Amazon targets the Drama genre, the one common genre between the two is comedy.



Note:

Although the dashboard provides a fundamental understanding of the content ecosystem on Netflix, the analytics could go much deeper if I am able to access users' data(ex: user subscription, user experience, user behavior, etc), which is usually not open to public due to confidentiality. As a next step, if I can break through the limitations of data, I may be able to provide more insights to help connect content usability with user data to discuss more in-depth insights for potential business growth opportunities.

Thank You for reading, I hope you liked it.

C Varun