- ♣ Netflix, Inc. is an American technology and media services provider and production company headquartered in Los Gatos, California. Netflix was founded in 1997 by Reed Hastings and Marc Randolph in Scotts Valley, California. The company's primary business is its subscription-based streaming service, which offers online streaming of a library of films and television series, including those produced in-house.
- Netflix is one of the most popular media and video streaming platforms. They have over 8000 movies or tv shows in about 45 languages available on their platform. They have over 247.15 M Subscribers globally as of Oct,2023.
- ♣ Netflix is one of the most popular digital streaming media service providers today. Netflix provides streaming services for movies and tv shows from various countries in the world.
- ♣ The particular business case focuses on the Netflix show data and provides insightful information on 8807 shows
- ♣ Analysing the data and generating insights helps Netflix decide which type of shows/movies to produce and how to grow the business.
- ♣ Netflix data analysis will be performed using several python libraries like numpy,pandas,matplotlib,seaborn.
- ♣ Data analysis using Python libraries is widely popular among the Data Scientists and Data Analysts.
- ♣ This tabular dataset consists of listings of all the movies and tv shows available on Netflix, along with details such as - cast, directors, ratings, release year, duration, etc.
- ♣ The following data is available in a single csv file :
 - Show ID: The ID of the show
 - Type: Identifier A Movie or TV Show
 - Title: Title of the Movie / Tv Show
 - Director: Director of the Movie
 - Cast: Actors involved in the movie/show
 - Country: Country where the movie/show was produced
 - Date added: Date it was added on Netflix
 - Release year: Actual Release year of the movie/show

• Rating: TV Rating of the movie/show

• Duration: Total Duration - in minutes or number of seasons

• Listed in: Genre

• Description: The summary description

The questions under line of analysis, Google Colab Notebook commands along with a screenshot of the output are submitted below along with the valuable insights that I drew from my analysis and recommendations regarding the company's growth and profit perspective are submitted below.

BASIC ANALYSIS:

Downloading the .csv file and uploading it into Google Colab Notebook

Loading The Dataset:



Import Libraries:

Importing the libraries, we need.

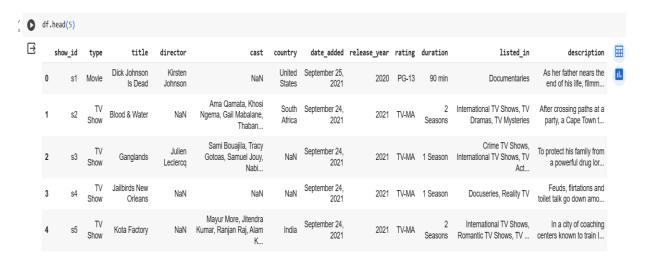
import numpy as np

import pandas as pd

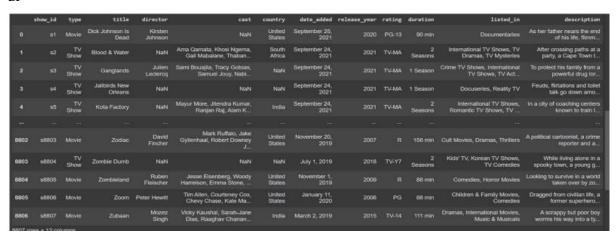
import matplotlib.pyplot as plt

import seaborn as sns

Let's check the first 5 data:



df



The dataset contains over 8807 titles,12 descriptions. It is a typical movie/TV shows dataframe. We can also see that there are NaN values in some columns.

2: Observations on the shape of data, data types of all the attributes, conversion of categorical attributes to 'category' (If required), missing value detection, statistical summary.

To get All attributes df.columns:

The Shape of the Data:

```
df.shape - o/p is (8807,14) (8807, 14)
```

Columns of the dataset: (df.columns)

The Dimension of the dataframe and Data types of all the attributes:

df.ndim and df.info()

Based on the output obtained, it can be seen that the majority of columns have a total of **8807** non-null values, but there are **8** columns that have a varying number of non-null values. In addition, it can be seen that **13** columns have an **object** data type and **1** column has a **int64** data type. Out of which 2 columns were added by me for unnesting of cast data and country data i.e., str_split and cast_split(12 & 13 columns).

Statistical Summary Before Data Cleaning:

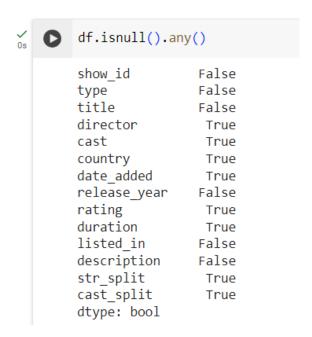
df.describe()

os	[31]	df.describe()		
	release_year			
		count	8807.000000	
		mean	2014.180198	
		std	8.819312	
		min	1925.000000	
		25%	2013.000000	
		50%	2017.000000	
		75%	2019.000000	
		max	2021.000000	

Missing Value Detection Data Profiling & Cleaning:

Data Cleaning means the process of identifying incorrect, incomplete, inaccurate, irrelevant, or missing pieces of data and then modifying, replacing, or deleting them as needed. Data Cleansing is considered as the basic element of Data Science.

df.isnull().any()



#Number of null values per column

df.isna().sum() / df.isnull ().sum()

0	df.isnull().	sum()	
	show_id type title director cast country date_added release_year rating duration listed_in description str_split cast_split dtype: int64	0 0 2634 825 831 10 0 4 3 0 0 831 825	

From the output above, we know that there are 8 columns with null values. 'director' column has 2,634 null values, 'cast' has 825 null values, 'country' has 831 null values, 'date_added' has 10 null values, 'rating' has 4 null values, 'duration' has 3 null values, 'str_split'(column for unnesting countries) has 831 null values, and 'cast_split'(column added for unnesting cast) has 825 null values.

The null values in total are 5963.

```
os df.isnull().sum().sum()
5963
```

Checking for duplicate Values: There are no duplicate values in the data.

df.duplicated()

Data Cleansing: Imputation is a treatment method for filling of missing values by using certain techniques. Can use mean, mode, or use predictive modelling. In this case study, we will discuss the use of the fillna function from Pandas for this imputation. Drop rows containing missing values using the dropna function from Pandas

#Imputing missing values

Panda's fillna()/replace functions can be used to fill in missing values in a dataset. I replaced the null values in the 'country' and "str_split" column with cast Unknown Country, I also replaced the null values in the 'cast' column and "cast_split" column with Unknown Cast, and replaced the null values in the 'director' column with Unknown Director(as per Solution Approach Sheet).

```
df["director"].replace(np.nan,"Unknown Director",inplace=True)
     df["cast"].fillna("Unknown Cast",inplace=True)
    df["country"].replace(np.nan,"Unknown Country",inplace=True)
    df["cast_split"].fillna("Unknown Cast",inplace=True)
    df["str_split"].replace(np.nan,"Unknown Country",inplace=True)
    df.isnull().sum()
    show_id
                     0
    type
    title
                     0
    director
    cast
                     0
     country
    date_added
    release_year
    rating
    duration
    listed_in
                     0
    description
                     0
     str_split
                     0
     cast_split
    dtype: int64
```

** The rating column has 3 time(in minutes) datatype as shown below.

```
df["rating"].value_counts()
 TV-MA
              3207
 TV-14
              2160
 TV-PG
               863
 PG-13
               490
 TV-Y7
               334
 TV-Y
               307
               287
 PG
 TV-G
               220
 NR
                80
 TV-Y7-FV
 NC-17
 UR
 74 min
                 1
 84 min
                 1
 66 min
                 1
 Name: rating, dtype: int64
```

Hence I have dropped those three rows as the number is quite low.

```
df.drop(df[df["rating"]=="66 min"].index,inplace=True)
df.rating.value_counts()
<ipython-input-103-c66b98757d85>:1: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: <a href="https://pandas.pydata.org/pardf.drop">https://pandas.pydata.org/pardf.drop</a>(df[df["rating"]=="66 min"].index,inplace=True)
TV-MA
               3205
TV-14
               2157
TV-PG
                 799
PG-13
                 490
TV-Y
                 306
PG
                 287
TV-G
NR
                  41
TV-Y7-FV
NC-17
Name: rating, dtype: int64
```

#Dropping missing values:

Columns 'date_added', 'rating', 'duration' have very less missing values. Therefore, I will remove any rows that have missing values in those columns.

```
df=df.dropna()
0s
         df.isnull().sum()
         show id
         type
title
                            0
                            0
         director
                            0
         cast
                            0
         country
                            0
         date_added
                            0
         release_year
         rating
                            0
         duration
                            0
         listed in
                            0
         description
                            0
         str_split
cast_split
                            0
         dtype: int64
```

Now let's check for the total null values:

```
os df.isnull().sum().sum()

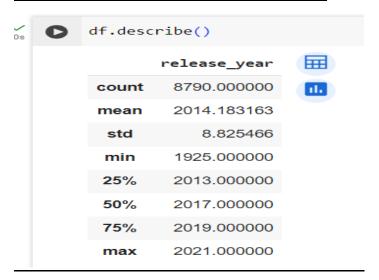
0
```

The total null values are 0 after data cleaning

#Dataset information: The following is the dataset information after cleaning the data.

```
df.info()
     <class 'pandas.core.frame.DataFrame'>
     Int64Index: 8790 entries, 0 to 8806
     Data columns (total 14 columns):
                         Non-Null Count
          show_id
                         8790 non-null
                                              object
                         8790 non-null
8790 non-null
          type
                                              object
          title
                                              object
                          8790 non-null
          director
                                              object
                          8790 non-null
          cast
                                              object
          country 8790 non-null date_added 8790 non-null
                                              object
                                              object
          release_year 8790 non-null
                                              int64
         duration 8790 non-null 8790 non-null 8790 non-null 8790 non-null
                                              object
                                              object
      10
                                              object
      11 description 8790 non-null
                                              object
      12 str_split
13 cast_split
                         8790 non-null
8790 non-null
                                              object
                                              object
     dtypes: int64(1), object(13)
     memory usage: 1.0+ MB
```

Statistical Summary After Data Cleaning:



What does 'good' look like?

- 1. Find the counts of each categorical variable both using graphical and nongraphical analysis.
- **a. For Non-graphical Analysis:** I found out the value_counts() for all the categorical data which shown as below:

```
1) df["title"].value counts()
```

```
Dick Johnson Is Dead 1
Cooked 1
My Beautiful Broken Brain 1
Pee-wee's Big Holiday 1
Netflix Presents: The Characters 1
Sleepless Society: Insomnia 1
Palazuelos mi rey 1
Narcos: Mexico 1
Love Is Blind 1
Zubaan 1
Name: title, Length: 8790, dtype: int64
```

2) df["director"].value counts()

```
Rajiv Chilaka
                                  19
Raúl Campos, Jan Suter
                                  18
Marcus Raboy
                                  16
Suhas Kadav
                                  16
Jay Karas
                                  14
Raymie Muzquiz, Stu Livingston
Joe Menendez
                                   1
Eric Bross
                                   1
Will Eisenberg
                                   1
Mozez Singh
Name: director, Length: 4528, dtype: int64
```

3) df["rating"].value_counts()

```
TV-MA
         3205
TV-14
          2157
TV-PG
PG-13
TV-Y7
            333
TV-Y
PG
            287
TV-G
            220
NR
             79
             41
TV-Y7-FV
             6
NC-17
              3
Name: rating, dtype: int64
```

4) df.release_year.value_counts().sort_index(ascend ing=False)

```
→ 2021
           592
   2020
          953
   2019
         1030
   2018
          1146
   2017
         1030
   1945
   1944
             3
   1943
   1942
   1925
   Name: release_year, Length: 74, dtype: int64
 5) df["type"].value counts()
      Movie
                   6131
                   2676
      TV Show
      Name: type, dtype: int64
 6) df.date added.value counts()
    January 1, 2020
                         109
    November 1, 2019
                         89
    March 1, 2018
                          75
    December 31, 2019
                          74
    October 1, 2018
                          71
    December 4, 2016
                           1
    November 21, 2016
    November 19, 2016
    November 17, 2016
    January 11, 2020
    Name: date added, Length: 1767, dtype: int64
```

```
7) df.duration.value counts()
```

```
1 Season
           1793
2 Seasons
            425
3 Seasons
            199
             152
90 min
94 min
             146
16 min
186 min
193 min
189 min
191 min
Name: duration, Length: 220, dtype: int64
```

```
8) df.listed_in.value_counts()
```

```
Dramas, International Movies
                                                      362
                                                      359
Documentaries
Stand-Up Comedy
                                                      334
Comedies, Dramas, International Movies
                                                      274
Dramas, Independent Movies, International Movies
                                                      252
Kids' TV, TV Action & Adventure, TV Dramas
                                                        1
TV Comedies, TV Dramas, TV Horror
                                                        1
Children & Family Movies, Comedies, LGBTQ Movies
                                                        1
Kids' TV, Spanish-Language TV Shows, Teen TV Shows
                                                       1
Cult Movies, Dramas, Thrillers
Name: listed in, Length: 514, dtype: int64
```

9) df.show_id.value_counts()

```
s1 1
s5875 1
s5869 1
s5870 1
s5871 1
...
s2931 1
s2930 1
s2929 1
s2928 1
s8807 1
Name: show_id, Length: 8807, dtype: int64
```

```
10) df["str_split"]=df.country.str.split(", ")
df.explode("str_split").str_split.value_counts()
```

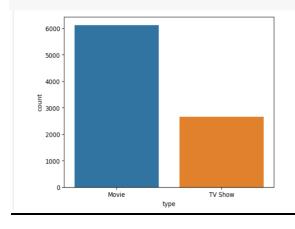
```
United States
                  3689
India
                  1046
United Kingdom
                   393
France
Bermuda
                      1
Ecuador
Armenia
                      1
Mongolia
Montenegro
                     1
Name: str_split, Length: 127, dtype: int64
```

```
11) df["cast_split"]=df.cast.str.split(", ")
df.explode("cast_split").cast_split.value_counts()
```

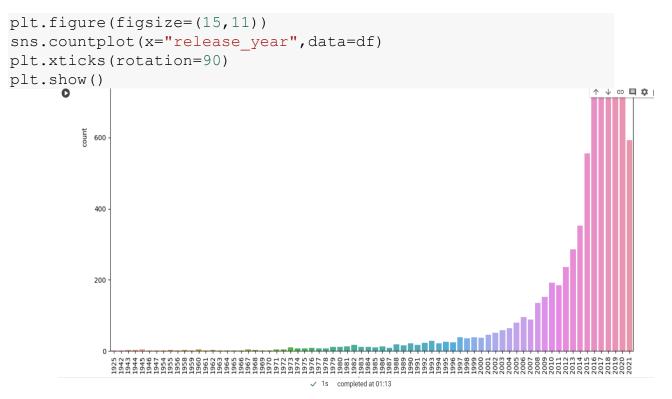
```
Anupam Kher
                            43
Shah Rukh Khan
                            35
Julie Tejwani
                            33
Naseeruddin Shah
                            32
Takahiro Sakurai
                            32
Maryam Zaree
                            1
Melanie Straub
                             1
Gabriela Maria Schmeide
Helena Zengel
Chittaranjan Tripathy
                             1
Name: cast split, Length: 36439, dtype: int64
```

b. For graphical analysis:

```
_sns.countplot(x="type", data=df)
plt.show()
```



```
plt.figure(figsize=(12,6))
sns.countplot(x="rating", data=df, hue="type")
plt.show()
      2000
                                                                                  Movie
                                                                                  TV Show
      1750
      1500
      1250
       750
      500
      250
                TV-MA
                                                                   NC-17
                                                                             TV-Y7-FV
           PG-13
                            TV-14
                                                        TV-G
                                               rating
```



2. Comparison of tv shows vs. movies.

a) Find the number of movies produced in each country and pick the top 10 countries.

```
df movies = df[df["type"] == "Movie"]
df movies = df movies.assign(country =
df movies["country"].str.split(", ")).explode("country")
movie count
=df movies["country"].value counts().sort values(ascending=False)
top 10 = movie count.head(10).to frame().reset index()
top 10.columns = ["country", "movie count"]
print(top 10)
              country movie_count
                             2492
     0 United States
     1
                              942
                India
     2 United Kingdom
                               475
               Canada
                               295
               France
     5
             Germany
                               168
               Spain
                               156
                 Japan
                               108
                China
             Hong Kong
```

b. Find the number of Tv-Shows produced in each country and pick the top 10 countries.

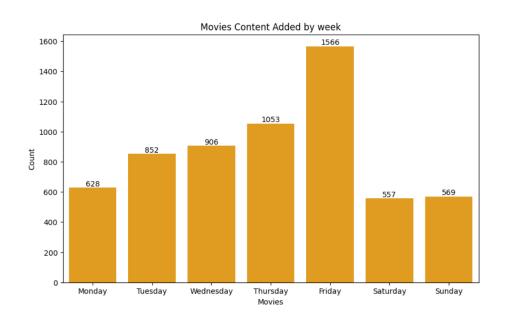
```
df tv shows = df[df["type"] == "TV Show"]
df tv shows = df tv shows.assign(country =
df_tv_shows["country"].str.split(", ")).explode("country")
tv shows count
=df tv shows["country"].value counts().sort values(ascending=False)
top 10 = tv shows count.head(10).to frame().reset index()
top 10.columns = ["country", "tv show count"]
print(top 10)
                country tv show count
         United States
                                   782
      1 United Kingdom
                                   232
                  Japan
                                   193
           South Korea
      3
                                   168
      4
                 Canada
                                   119
                 France
                                    76
                 Taiwan
                                    70
      6
      7
                  India
                                    65
              Australia
      8
                                    58
                 Mexico
```

- 3. What is the best time to launch a TV show?
- a) Find which is the best week to release the Tv-show or the movie. Do the analysis separately for Tv-shows and Movies

For Movies: Best Day of the Week to release a movie is Friday

```
df["week added"]=pd.to datetime(df["date added"]).dt.day name(
week order=
['Monday','Tuesday','Wednesday','Thursday','Friday','Saturday'
, 'Sunday']
weekly show count=df["week added"].value counts().loc[week ord
df movies = df[df['type'] == 'Movie']
movies count =
df movies['week added'].value counts().loc[week order]
movies count
       Monday
                       628
       Tuesday
                       852
       Wednesday
                       906
       Thursday
                      1053
       Friday
                      1566
       Saturday
                       557
       Sunday
                       569
       Name: week added, dtype: int64
```

```
plt.figure(figsize=(10,6))
bar_plot = sns.barplot(x=movies_count.index, y=movies_count.values,
color="orange")
plt.xlabel('Movies')
plt.ylabel('Count')
plt.title('Movies Content Added by week')
for index, value in enumerate(movies_count.values):
    bar_plot.text(index, value, str(value), ha='center', va='bottom')
plt.show()
```



For TV SHOWS: Best Day to add is Friday

```
df["week_added"]=pd.to_datetime(df["date_added"]).dt.day_name()
week_order=
['Monday','Tuesday','Wednesday','Thursday','Friday','Saturday'
,'Sunday']
weekly_show_count=df["week_added"].value_counts().loc[week_order]
df_tv_shows = df[df['type'] == 'TV Show']
tv_shows_count =
df_tv_shows['week_added'].value_counts().loc[week_order]
tv_shows_count
```

```
tv_shows_count
[9]
    Monday
                   223
    Tuesday
                   345
    Wednesday
                   382
     Thursday
                   343
     Friday
                   932
    Saturday
                   259
    Sunday
                   182
    Name: week added, dtype: int64
```

plt.figure(figsize=(10,5)) bar plot = sns.barplot(x=tv shows count.index, y=tv shows count.values, color="cyan") plt.xlabel('TV Shows') plt.ylabel('Count') plt.title('TV Shows Added in the week') for index, value in enumerate (tv shows count.values): bar plot.text(index, value, str(value), ha='center', va='bottom') plt.show() TV Shows Added in the week 800 600 400 382 345 343 259 200 182 Tuesday Monday Wednesday Thursday Friday Saturday Sunday

b) Find which is the best month to release the Tv-show or the movie.

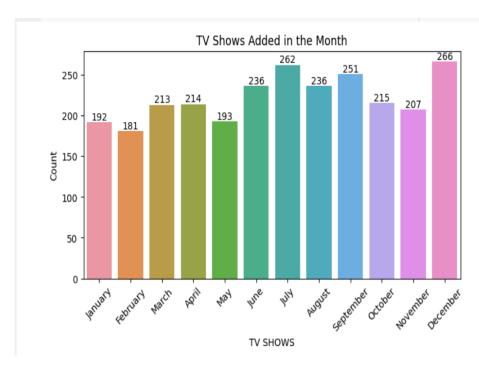
Do the analysis separately for Tv-shows and Movies

For TV SHOWS: BEST MONTH is December

```
tv_show=df[df["type"] == "TV Show"]
df["month_added"] = pd.to_datetime(df["date_added"]).dt.month_name()
month_order= ['January', 'February', 'March', 'April', 'May', 'June',
'July', 'August', 'September', 'October', 'November', 'December']
monthly_tv_show_count =
df_tv_shows['month_added'].value_counts().loc[month_order]
monthly_tv_show_count
plt.figure(figsize=(8,4))
bar_plot = sns.barplot(x=monthly_tv_show_count.index,
y=monthly_tv_show_count.values)
plt.xlabel('TV SHOWS')
plt.ylabel('Count')
```

```
plt.title('TV Shows Added in the Month')
plt.xticks(rotation=45)
for index, value in enumerate(monthly_tv_show_count.values):
    bar_plot.text(index, value, str(value), ha='center', va='bottom')
plt.show()
```

```
[36] tv_shows_count
        January
        February
                      181
        March
                      213
        April
                      214
        May
                      193
                      236
        June
        July
                      262
        August
                      236
        September
                      251
        October
                      215
        November
                      207
                      266
        December
        Name: month_added, dtype: int64
```



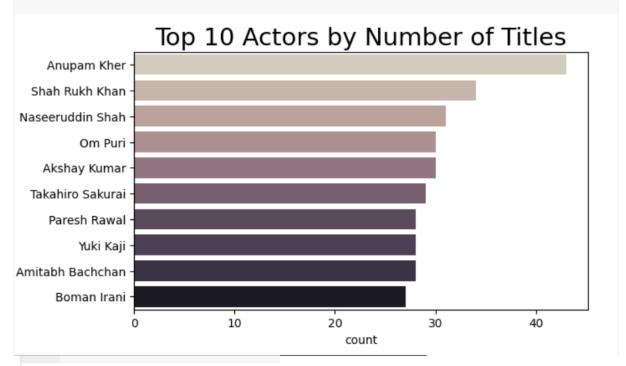
FOR MOVIES: Best Month is July

```
df["month_added"]=pd.to_datetime(df["date_added"]).dt.month_name()
month_order= ['January', 'February', 'March', 'April', 'May', 'June',
'July','August', 'September', 'October', 'November', 'December']
```

```
movies count = df movies['month added'].value counts().loc[month order]
movies count
plt.figure(figsize=(9,6))
bar plot = sns.barplot(x=movies count.index, y=movies count.values)
plt.xlabel('Movies')
plt.ylabel('Count')
plt.title('Movies Added in the Month')
plt.xticks(rotation=45)
for index, value in enumerate(movies count.values):
    bar plot.text(index, value, str(value), ha='center', va='bottom')
plt.show()
                 546
    January
    February
                 382
    March
                 529
    April
                 550
    May
                439
    June
    July
                 565
    August
                 519
    September
                519
    October
                545
    November
                498
    December
                 547
    Name: month_added, dtype: int64
                     Movies Added in the Month
                          492
                      439
  400
          382
300
  200
  100
```

4. Analysis of actors/directors of different types of shows/movies. a. Identify the top 10 directors who have appeared in most movies or TV shows.

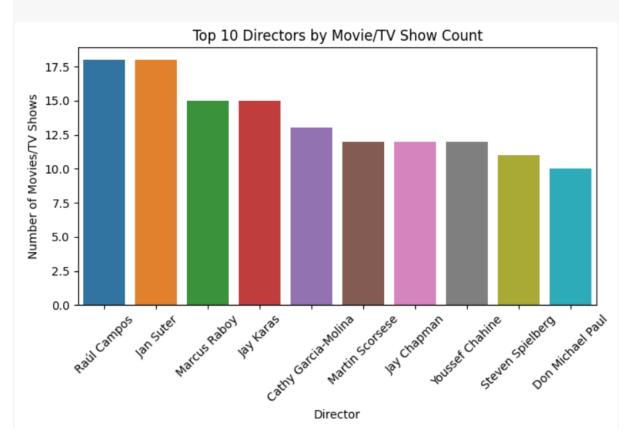
```
plt.figure(figsize=(10,8))
df_cast = df[df.cast != 'Unknown
Cast'].set_index('title').cast.str.split(', ',
expand=True).stack().reset_index(level=1, drop=True)
sns.countplot(y = df_cast, order=df_cast.value_counts().index[:10],
palette='magma_r', saturation=.2)
plt.title('Top 10 Actors by Number of Titles', fontsize=21);
plt.show()
```



Anupam Kher	43
Shah Rukh Khan	34
Naseeruddin Shah	31
Om Puri	30
Akshay Kumar	30
Takahiro Sakurai	29
Paresh Rawal	28
Yuki Kaji	28
Amitabh Bachchan	28
Boman Irani	27

b. Identify the top 10 directors who have appeared in most movies or TV shows.

```
df_director = df['director'].str.split(',', expand=True).stack()
df_director = df_director.reset_index(level=1,
drop=True).to_frame('director')
df_director['show_id'] = df['show_id']
director_counts = df_director['director'].value_counts()[1:]
top_10_directors = director_counts.head(10)
plt.figure(figsize=(9, 5))
bar_plot = sns.barplot(x=top_10_directors.index,
y=top_10_directors.values)
plt.xlabel('Director')
plt.ylabel('Number of Movies/TV Shows')
plt.title('Top 10 Directors by Movie/TV Show Count')
plt.xticks(rotation=45)
plt.show()
```



```
top_10_directors

Raúl Campos 18
Jan Suter 18
Marcus Raboy 15
Jay Karas 15
Cathy Garcia-Molina 13
Martin Scorsese 12
Jay Chapman 12
Youssef Chahine 12
Steven Spielberg 11
Don Michael Paul 10
Name: director, dtype: int64
```

5. Which genre movies are more popular or produced more

```
from wordcloud import WordCloud
text= " ".join(str(each) for each in df.listed_in)
wordcloud=WordCloud(max_words=200,background_color="black").generate(te
xt)
colormaps = ['plasma', 'Purples','inferno','PuRd', 'Blues',
'BuGn','cividis', 'PuRd','YlGn','Greens',
'YlOrBr', 'BuGn','YlOrRd', 'Oranges', 'RdPu',
    'YlGnBu', 'PuBuGn', 'OrRd','magma','Reds','BuGn' ]
plt.figure(figsize=(8,4))
plt.figure(figsize=(10,8))
plt.imshow(wordcloud,interpolation="Bilinear")
plt.title("More Popular Movies or More Produced Genres",fontsize=25)
plt.axis("off")
plt.show()
```

<Figure size 800x400 with 0 Axes>

More Popular Movies or More Produced Genres



From the graph, we know that International Movies take the first place, followed by dramas and comedies.

6) Find After how many days the movie will be added to Netflix after the release of the movie (you can consider the recent past data)

```
df["year_added"]=pd.to_datetime(df.date_added).dt.year
df["year_added"].dropna().astype(int)

df_movies = df[df['type'] == 'Movie']
df_movies ['year_added'] = pd.to_datetime(df_movies.date_added).dt.year
df_tv_shows = df[df['type'] == 'TV Show']
df_tv_shows.loc[:,"year_added"]=pd.to_datetime(df_tv_shows.date_added).
dt.year
year_df= df.loc[:,
'year_added'].value_counts().to_frame().reset_index().rename(columns={"index": "year",
"year_added":"count"})
year_df = year_df[year_df.year != 2020]
print(year df)
```

```
year count
0 2019
        1722
2 2018 1404
 2021 1056
4 2017
         989
5 2016
         358
6 2015
          70
7
  2014
          20
8 2011
          13
9 2013
           11
10 2012
          3
11 2009
           2
12 2008
           1
13 2010
cincethan innut ara all
```

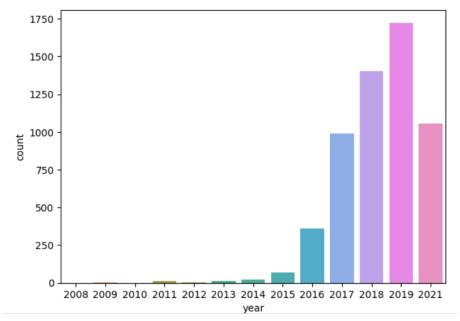
```
movies_year_df
=df_movies.year_added.value_counts().to_frame().reset_index().rename(co
lumns={"index":
"year", "year_added":"count"})
movies_year_df = movies_year_df[movies_year_df != 2020]
movies_year_df
```

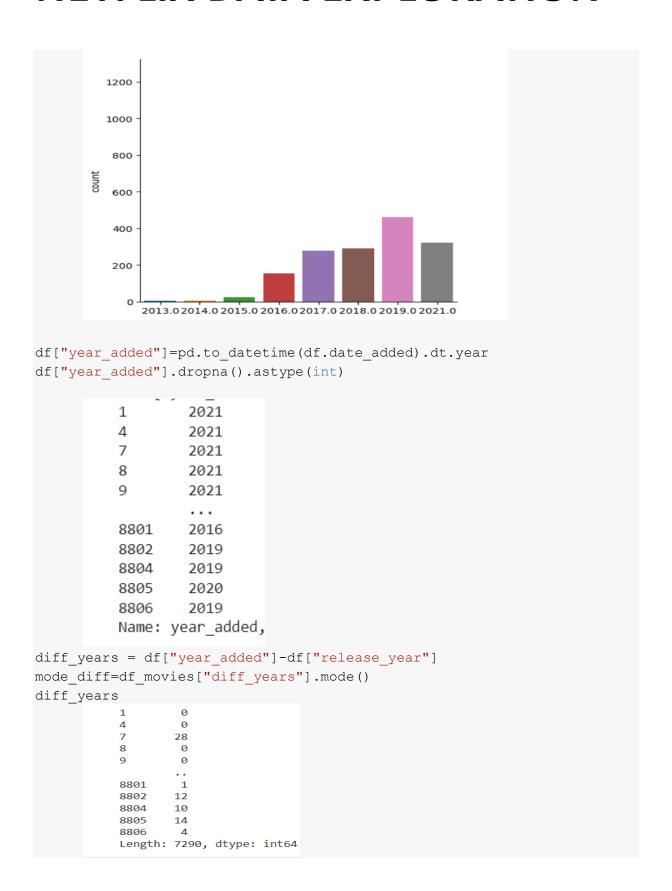
```
year count
              0 2019.0 1261
    NaN
2 2018.0 1114
3 2021.0
4 2017.0 710
5 2016.0 205
6 2015.0 47
7 2014 0
8 2011.0 13
9 2013.0
10 2012.0 3
11 2009 0
            2
12 2008.0
13 2010.0
```

```
shows_year_df =
df_tv_shows.year_added.value_counts().to_frame().reset_index().rename(c
olumns={"index":
"year", "year_added":"count"})
shows_year_df = shows_year_df[shows_year_df != 2020]
shows_year_df
```

	year	count	
0	NaN	476	11.
1	2019.0	461	
2	2021.0	321	
3	2018.0	290	
4	2017.0	279	
5	2016.0	153	
6	2015.0	23	
7	2014.0	5	
8	2013.0	5	

```
fig, ax = plt.subplots(figsize=(7, 5))
sns.barplot(data=year_df, x='year', y='count')
sns.displot(data=movies_year_df, x='year', y='count')
sns.barplot (data=shows_year_df, x='year', y='count')
ax.set_xticks(np.arange(2008, 2020, 1),rotation=90)
plt.title("Total content added across all years (up to 2019)")
plt.legend(['Total','Movie','TV Show'])
plt.ylabel("Releases")
plt.xlabel("Year")
plt.show()
```





6) Find After how many days the movie will be added to Netflix after the release of the movie (you can consider the recent past data)

```
df.loc[df["type"] == "Movie", "date_added"] =
pd.to_datetime(df["date_added"])
# Calculate the difference between the release_year and
year_added columns
df.loc[df["type"] == "Movie", "diff_years"] =
df["release_year"] - df["year_added"]
# Find the mode of the difference
mode_diff = df_movies["diff_years"].mode()
diff_years
df_movies_added =
df.groupby('diff_years').size().reset_index(name='movies_added')
df_movies_added[:10]
```

扁 diff_years movies_added 0 -1 days +23:59:59.999999925 ıl. **1** -1 days +23:59:59.999999927 2 -1 days +23:59:59.999999928 3 -1 days +23:59:59.999999999 4 -1 days +23:59:59.999999930 5 -1 days +23:59:59.999999934 2 2 6 -1 days +23:59:59.999999935 2 7 -1 days +23:59:59.999999936 8 -1 days +23:59:59.999999937 **9** -1 days +23:59:59.99999938 2

Conclusions:

- Country that produces the largest number of content titles on Netflix is the **United States** with **2,000**++ content titles production.
- The genre with the largest number of content titles is **International Movies** with **1,700**++ content.
- The number of content titles on Netflix continued to **increase** from **2012** to **2019**.
- The actor with the largest number of content titles on Netflix is **Anupam Kher followed by Shah Rukh Khan**.
- The total number of movies are **6131** and the total number of TV shows are **2676** of the total content.
- The Director with the largest number of content titles on Netflix is **Jan Suter and Raul Campos** who has directed **17++** number of content titles on Netflix.
- The best day to release new TV Shows or Movies is **FRIDAY.**
- The best month to release new TV Shows is **December** and for new movies is **July**.
- The most popular Genres produced are International Movies and TV Shows.
- Most of the TV shows with shorter duration are most watched ones.

As the streaming industry evolves, understanding these patterns and trends becomes increasingly essential for navigating the dynamic landscape of Netflix and its vast library.

Our data analysis journey showcased the power of data in unravelling the mysteries of Netflix's content landscape, providing valuable insights for viewers and content creators.

RECOMMENDATIONS:

- Netflix has to focus on TV Shows also because there are people who will like to see to shows rather than movies
- By approaching the top directors we can plan some more movies/tv shows in order to increase the popularity
- •Not only reaching top director we can also meet the director with less no of movies with quality content and having high rating as there may be some financial issues or any other issue which can be sorted out in order to get good content on Netflix
- We have seen most no of international movies genre so it is needed to give priority to other geners like horror, comedy..etc and also to concentrate mostly on Internation Movies Genre to get more subscribers
- In TV Shows we may focus on thriller genre which will be helpful for having more no of seasons

- ❖ Most of the movies released in OTT are in a year 2019 so we need to go on increasing this value in order to attract people by showing that getting subscription is useful as Netflix is releasing more movies per year
- Mainly the release of films or tv shows in Netflix should focus on the festival holidays, year end and week ends which is to be mainly focussed
- Some movies can be released directly into Netflix platform to gain more popularity which has some positive talk which may help in improving subscriptions
- Should focus on a actor who has immense following and make use of it by doing a TV Shows or web series.
- Advertisement between the content should be very less and movies released should be increased to drag customers for subscription.

Link to python Colab Notebook:

https://colab.research.google.com/drive/10h6xm1Nl_1i ZiIPxQIN3eN4badARnwyj?usp=drive_link