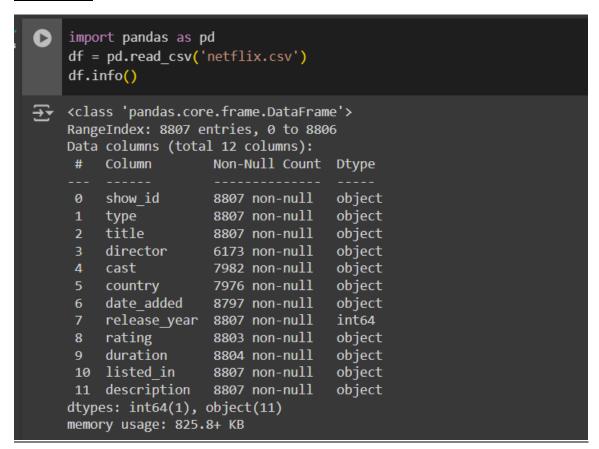
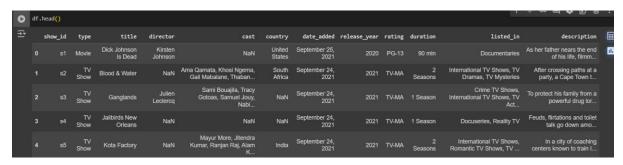
## Netflix - Data Exploration and Visualisation (Collab link: Netflix Data Analysis)

<u>Dataset:</u> It consists of movies and shows that are available in Netflix. Cast & Directors of those movies & shows and countries where the movies or shows produced in. Movies & Tv Shows released date and the date they got added into the Netflix. Genre & Ratings of them.

## **Dataset Info:**



#### Dataset:



From the dataset we can see that there are multiple values in a single cell, so we need to clean the data.

First, we will strip the data in the cells by using strip () method.

This will delete the trailing and leading spaces or commas.

Code to strip the required columns:

#### Code:

```
df['cast'] = df['cast'].str.strip(', ')
df['listed_in'] = df['listed_in'].str.strip(', ')
df['country'] = df['country'].str.strip(', ')
df['director'] = df['director'].str.strip(', ')
```

Now we dealt with leading and trailing spaces or special characters. We will split the multiple values in the cells using split () method.

## Code to split the cells:

```
df['cast'] = df['cast'].apply(lambda x: x.split(',') if pd.notna(x)
else x)
df['listed_in'] = df['listed_in'].apply(lambda x: x.split(',') if
pd.notna(x) else x)
df['director'] = df['director'].apply(lambda x: x.split(',') if
pd.notna(x) else x)
df['country'] = df['country'].apply(lambda x: x.split(',') if
pd.notna(x) else x)
```

We used apply () function to apply the split method to each row and we used notna () function to filter the Nan or Null values

```
df['cast'] = df['cast'].apply(lambda x: x.split(',') if pd.notna(x)
else x)
df['listed_in'] = df['listed_in'].apply(lambda x: x.split(',') if
pd.notna(x) else x)
df['director'] = df['director'].apply(lambda x: x.split(',') if
pd.notna(x) else x)
df['country'] = df['country'].apply(lambda x: x.split(',') if
pd.notna(x) else x)
```

Now this split method will create list of elements in each cell, we can use explode () function to create a separate row for each value in the list.

#### Code:

```
df = df.explode('cast')
df = df.explode('listed_in')
df = df.explode('country')
df = df.explode('director')
```

# Original Dataset shape:

```
df.shape

→ (8807, 12)
```

After using explode ():



We need to strip the data as after performing explode () there may be some trailing or leading spaces.

Code:

```
df['cast'] = df['cast'].str.strip(' ')
df['listed_in'] = df['listed_in'].str.strip(' ')
df['country'] = df['country'].str.strip(' ')
df['director'] = df['director'].str.strip(' ')
```

As we can see there are null values in few columns, we need to deal with Null values.



We will fill the null values using fillna () function.

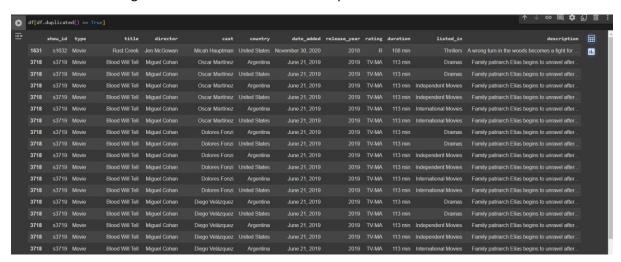
```
df['director'] = df['director'].fillna('Unknown Director')
df['cast'] = df['cast'].fillna('Unknown Cast')
df['country'] = df['country'].fillna('Unknown Country')
df['date_added'] = df['date_added'].fillna(0)
df['rating'] = df['rating'].fillna('Unknown rating')
```

## After performing Data Imputation



We can see that there are no null values anymore

From the below image we can see that there are 55 duplicated rows



We need to drop these rows to perform a precise analysis

Code:

```
df.drop_duplicates(inplace = True, keep = 'first')
```

We used inplace = True and keep = 'first' properties to keep the first duplicated row and make the changes permanent.

After performing Data Cleaning our dataset shape is



df.reset\_index(drop = True, inplace = True)

→ After performing explode operation rows got duplicated and explicit index is repeating so we need to reset the index to get the index in order. We are drop = True to drop the old index.

# **Exploratory Data Analysis**

Values counts of each category:

**Type Columns:** 

Code:

```
df.groupby(by = 'type')['title'].nunique()
```

## Output:

```
df.groupby(by = 'type')['title'].nunique() {

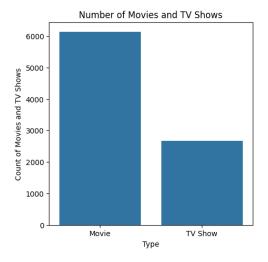
type
   Movie    6131
   TV Show    2676
   Name: title, dtype: int64
```

#### Visualization:

Code:

```
import matplotlib.pyplot as plt
import seaborn as sns

plt.figure(figsize = (5,5))
sns.barplot(x= df.groupby(by = 'type')['title'].nunique().index, y=
df.groupby(by = 'type')['title'].nunique().values)
plt.title('Number of Movies and TV Shows')
plt.xlabel('Type')
plt.ylabel('Count of Movies and TV Shows')
plt.show()
```



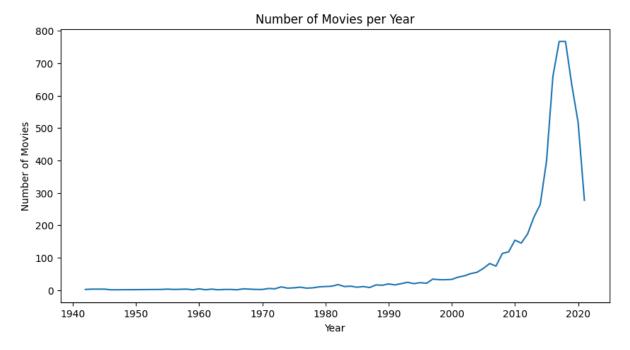
**Insights**: From the bar graph we can clearly derive that no of movies produced by Netflix are greater than TV shows.

# No of movies released per year

#### Code:

```
plt.figure(figsize = (10,5))
sns.lineplot(x = Movies.groupby(by
='release_year')['title'].nunique().index, y = Movies.groupby(by
='release_year')['title'].nunique().values)
plt.title('Number of Movies per Year')
plt.xlabel('Year')
plt.ylabel('Number of Movies')
plt.show()
```

# **Graph: (Trend analysis)**

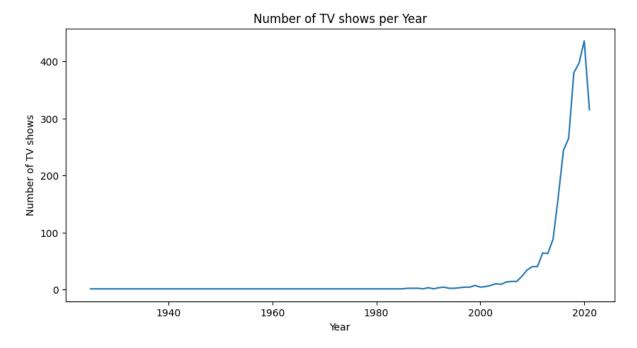


**Insights:** From the graph we can see the rapid increase in no of movies released per year.

## Number of TV Shows released per year

```
plt.figure(figsize = (10,5))
sns.lineplot(x = TV_shows.groupby(by
='release_year')['title'].nunique().index, y = TV_shows.groupby(by
='release_year')['title'].nunique().values)
plt.title('Number of TV shows per Year')
plt.xlabel('Year')
plt.ylabel('Number of TV shows')
plt.show()
```

# **Graph:**



**Insights:** From the graph we can derive that no of TV shows released per year increased rapidly every year.

# **Non-Graphical Data Analysis:**

Top 10 Directors who directed the most n.o of movies

# Code:

```
Movies.groupby(by =
  'director')['title'].nunique().sort_values(ascending =
  False).iloc[1:11]
```

# Output:

```
director
Rajiv Chilaka 22
Jan Suter 21
Raúl Campos 19
Suhas Kadav 16
Marcus Raboy 15
Jay Karas 15
Cathy Garcia-Molina 13
Martin Scorese 12
Jay Chapman 12
Youssef Chahine 12
Name: title, dtype: int64
```

**Insights**: From the output we can find the list of directors who directed the most movies. Rajiv Chilaka has directed most no of movies

Top 10 Directors who directed the most n.o of TV shows

#### Code:

```
TV_shows.groupby(by =
  'director')['title'].nunique().sort_values(ascending =
  False).iloc[1:11]
```

# **Output:**

Insights: Ken Burns and Alastair Fothergill are the directors who directed most TV shows.

Top 10 actors who have appeared in most movies

#### Movies:

Code:

```
Movies.groupby(by = 'cast')['title'].nunique().sort_values(ascending =
False).iloc[1:11]
```

# Output:

```
Movies.groupby(by = 'cast')['title'].nunique().sort_values(ascending = False).iloc[1:11]

→ cast

    Anupam Kher
                      42
                    35
   Shah Rukh Khan
   Naseeruddin Shah 32
   Om Puri
                      30
   Akshay Kumar
                      30
    Amitabh Bachchan 28
   Paresh Rawal
                      28
    Julie Tejwani
                      28
    Boman Irani
                      27
    Rupa Bhimani
    Name: title, dtype: int64
```

**Insights**: From the output we can see top 10 casted actors in movies.

**Recommendations**: Anupam Kher is the most casted actor, he must be popular so we can add collection of top actors in our platform so viewers can directly select his movies.

# Top 10 actors who have appeared in most movies

#### Code:

```
TV_shows.groupby(by = 'cast')['title'].nunique().sort_values(ascending
= False).iloc[1:11]
```

## **Output:**

**Insights:** From the output we can see top 10 casted actors in TV shows.

Recommendations: Takahiro Sakurai is the most casted actor, he must be popular so we can add collection of top actors in our platform so viewers can directly select his TV shows.

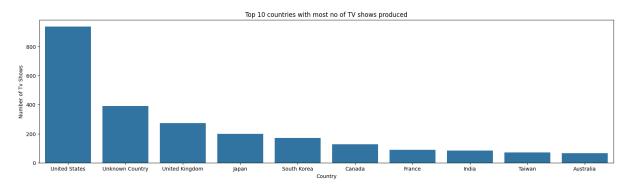
Top 10 countries with most n.o of Movies produced

#### Code:

```
Movies = df[df['type'] == 'Movie']
Movies.groupby(by = 'country')['title'].nunique().sort_values(ascending = False).head(10)
```

```
Movies.groupby(by = 'country')['title'].nunique().sort_values(ascending = False).head(10)
→ country
    United States
                    2752
    India
                      962
    United Kingdom
   Unknown Country
                      440
    Canada
                      319
    France
                     182
   Germany
    Spain
                     171
    Japan
                     119
                      114
    China
    Name: title, dtype: int64
```

## **Graphical Representation:**



#### Code:

```
plt.figure(figsize = (20,5))
sns.barplot(x = TV_shows.groupby(by =
'country')['title'].nunique().sort_values(ascending =
False).head(10).index, y = TV_shows.groupby(by =
'country')['title'].nunique().sort_values(ascending =
False).head(10).values)
plt.xlabel('Country')
plt.ylabel('Number of Tv Shows')
plt.title('Top 10 countries with most no of Tv shows produced')
plt.show()
```

Insights: From the graph we can clearly see that united states has produced most no of movies.

#### Top 10 Countries with most n.o of TV shows produced

```
Tv_Shows = df[df['type'] == 'TV Show']
Tv_Shows.groupby(by =
'country')['title'].nunique().sort_values(ascending = False).head(10)
Output:
```

```
[15] Tv_Shows = df[df['type'] == 'TV Show']

T = Tv_Shows.groupby(by = 'country')['title'].nunique().sort_values(ascending = False).head(10)

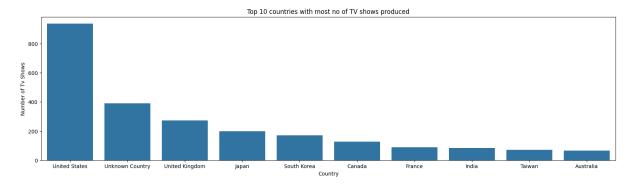
country
United States 938
Unknown Country 391
United Kingdom 272
Japan 199
South Korea 170
Canada 126
France 90
India 84
Taiwan 70
Australia 66
Name: title, dtype: int64
```

## **Graphical Analysis:**

#### Code:

```
plt.figure(figsize = (20,5))
sns.barplot(x = TV_shows.groupby(by =
'country')['title'].nunique().sort_values(ascending =
False).head(10).index, y = TV_shows.groupby(by =
'country')['title'].nunique().sort_values(ascending =
False).head(10).values)
plt.xlabel('Country')
plt.ylabel('Number of Tv Shows')
plt.title('Top 10 countries with most no of TV shows produced')
plt.show()
```

# Graph:



#### Week-wise movie count

# Code:

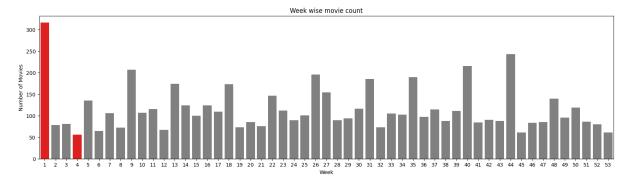
```
Movies.loc[:,'date_added'] = pd.to_datetime(Movies['date_added'],
format = 'mixed').dt.strftime('%Y-%m-%d')
Movies.loc[:,'Movies release week'] =
pd.to_datetime(Movies['date_added']).dt.isocalendar().week
Movies.groupby(by = 'Movies release
week')['title'].nunique().sort_values(ascending = False).head()
```

## Graphical representation of week-wise movie count

#### Code:

```
plt.figure(figsize = (20,5))
wwmd = pd.DataFrame(Movies.groupby(by = 'Movies release
week')['title'].nunique()).reset_index()
wwmd.rename(columns = {'title':'count'},inplace = True)
Movie_max_value = wwmd['count'].idxmax()
Movie_min_value = wwmd['count'].idxmin()
colors = ['gray' if i != Movie_max_value else 'red'for i in
range(len(wwmd))]
sns.barplot(x = Movies.groupby(by = 'Movies release
week')['title'].nunique().index, y = Movies.groupby(by = 'Movies
release week')['title'].nunique().values, palette = colors)
plt.xlabel('Week')
plt.ylabel('Number of Movies')
plt.title('Week wise movie count')
plt.show()
```

# Graph:



**Insights**: From the graph we can clearly see that most of no of movies were added to the Netflix during 1<sup>st</sup> and least no of movies were added during 4<sup>th</sup> week

#### **TV Shows**:

```
TV_shows.loc[:,'date_added'] = pd.to_datetime(TV_shows['date_added'],
format = 'mixed').dt.strftime('%Y-%m-%d')
TV_shows.loc[:,'TV show release week'] =
pd.to_datetime(TV_shows['date_added']).dt.isocalendar().week
TV_shows.groupby(by = 'TV show release
week')['title'].nunique().sort values(ascending = False).head()
```

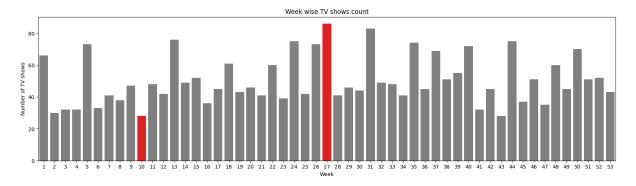
## Output:

#### **Graphical representation:**

#### Code:

```
plt.figure(figsize = (20,5))
wwtsd = pd.DataFrame(TV_shows.groupby(by = 'TV show release
week')['title'].nunique()).reset_index()
wwtsd.rename(columns = {'title':'count'},inplace = True)
TV_showmax_value = wwtsd['count'].idxmax()
TV_showmin_value = wwtsd['count'].idxmin()
colors = ['gray' if (x!= TV_showmax_value and x!= TV_showmin_value)
else 'red' for x in range(len(wwtsd))]
sns.barplot(x = wwtsd['TV show release week'], y = wwtsd['count'],
palette = colors)
plt.xlabel('Week')
plt.ylabel('Number of TV shows')
plt.title('Week wise TV shows count')
plt.show()
```

# Graph:



**Insights:** From the graph we can clearly see that most of no of TV shows were added to the Netflix during 27<sup>th</sup> week and least no of TV shows were added during 10<sup>th</sup> week

Find which is the best month to release the Tv-show or the movie. Do the analysis separately for Tv-shows and Movies

#### **Movies**

```
Movies.loc[:,'Movies release month'] =
pd.to_datetime(Movies['date_added']).dt.month
```

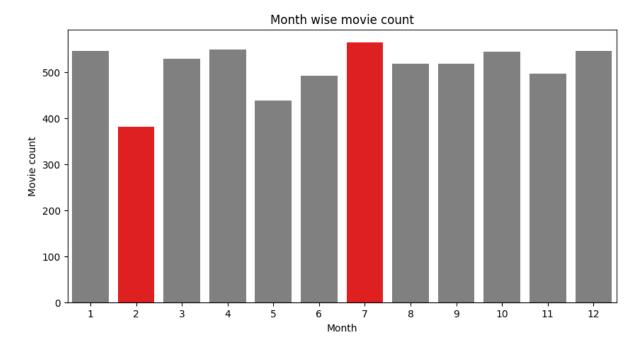
```
Movies.groupby(by = 'Movies release month')['title'].nunique().sort_values(ascending = False)
```

# Output:

## **Graphical Representation:**

```
plt.figure(figsize = (10,5))
mwmd = pd.DataFrame(Movies.groupby(by = 'Movies release
month')['title'].nunique()).reset_index()
mwmd.rename(columns = {'title':'count'},inplace = True)
Movie_month_max_value = mwmd['count'].idxmax()
Movie_month_min_value = mwmd['count'].idxmin()
colors = ['gray' if x != Movie_month_max_value and x!=
Movie_month_min_value else 'red' for x in range(len(mwmd))]
sns.barplot(x = Movies.groupby(by = 'Movies release
month')['title'].nunique().index, y = Movies.groupby(by = 'Movies
release month')['title'].nunique().values, palette = colors)
plt.title('Month wise movie count')
plt.xlabel('Month')
plt.ylabel('Movie count')
plt.show()
```

## **Graph:**



**Insights:** From the graph we can clearly see that most of no of movies were added to the Netflix during 7<sup>th</sup> week and least no of movies were added during 2<sup>nd</sup> week.

**Recommendations:** As many movies are getting added during 7<sup>th</sup> month, we need to increase our marketing at least 1 or 2 weeks before this month to attract more customers to take our subscription.

- 2. As least no of movies are getting during 2<sup>nd</sup> month, we need to understand the reason behind it. If it's an unavoidable reason then we need to try to bring popular movies during 2<sup>nd</sup> month to increase the subscription count and to keep our existing customers.
- 3. If we are not able to get the bid for popular movies then we need to offer 3months/6months at a discount price to maintain our viewership.

#### TV Shows:

#### Code:

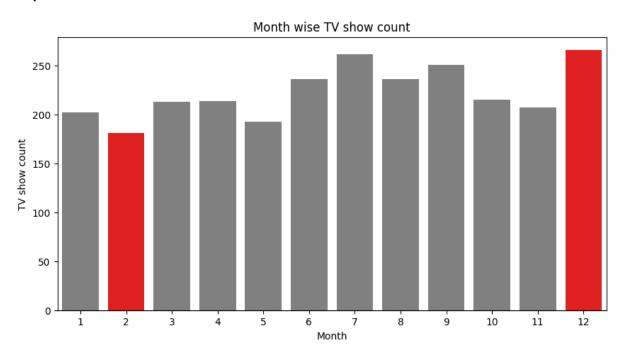
```
TV_shows.loc[:,'TV show release month'] =
pd.to_datetime(TV_shows['date_added']).dt.month
TV_shows.groupby(by = 'TV show release
month')['title'].nunique().sort_values(ascending = False)
```

Graphical representation:

#### Code:

```
plt.figure(figsize = (10,5))
mwtsd = pd.DataFrame(TV_shows.groupby(by = 'TV show release
month')['title'].nunique()).reset_index()
mwtsd.rename(columns = {'title':'count'},inplace = True)
TV_month_max_value = mwtsd['count'].idxmax()
TV_month_min_value = mwtsd['count'].idxmin()
colors = ['gray' if x != TV_month_max_value and x!= TV_month_min_value
else 'red' for x in range(len(mwtsd))]
sns.barplot(x= TV_shows.groupby(by = 'TV show release
month')['title'].nunique().index, y = TV_shows.groupby(by = 'TV show
release month')['title'].nunique().values, palette = colors,legend =
False)
plt.title('Month wise TV show count')
plt.xlabel('Month')
plt.ylabel('TV show count')
plt.show()
```

#### Graph:



**Insights:** From the graph we can clearly see that most of no of movies were added to the Netflix during 12<sup>th</sup> week and least no of movies were added during 2<sup>nd</sup> week.

# Top 10 most produced Movie genres

#### Code:

```
Movies.groupby(by =
'listed_in')['title'].nunique().sort_values(ascending =
False).iloc[0:10]
```

## Output:

```
listed_in
International Movies
                            2752
Dramas
                            2427
Comedies
                            1674
Documentaries
                            869
Action & Adventure
                            859
Independent Movies
Children & Family Movies 641
Romantic Movies
                            616
Thrillers
Music & Musicals
                             375
Name: title, dtype: int64
```

**Insights**: From the output we can derive that "international movies" are the most produced movies.

**Recommendations**: International movies are most produced movies, first we need to check whether these movies are giving us profits, if not we need to check genre is bringing the profits to focus on those genres.

#### Top 10 most produced TV show genres

## Code:

```
TV_shows.groupby(by =
'listed_in')['title'].nunique().sort_values(ascending =
False).iloc[0:10]
```

```
→ listed_in

    International TV Shows
    TV Dramas
    TV Comedies
                             581
   Crime TV Shows
                             470
                             451
    Docuseries
                             395
    Romantic TV Shows
                             370
    Reality TV
   British TV Shows
                             253
    Anime Series
                             176
    Name: title, dtype: int64
```

**Insights:** From the output we can derive that International TV shows are the most produced TV shows.

#### Time taken for Movies and TV shows to enter Netflix after getting released

#### Code:

```
df['date_added'] = pd.to_datetime(df['date_added'], errors = 'coerce')
df['release_year'] = pd.to_datetime(df['release_year'], errors =
'coerce') + pd.offsets.YearBegin(0)
df['release_year'] =
pd.to_datetime(df['release_year']).dt.strftime('%Y-%m-%d')
df['release_year'] = pd.to_datetime(df['release_year'])
df['Time_taken_to_enter_ott'] = df.apply(lambda x:
(pd.to_datetime(x['date_added']) - pd.to_datetime(x['release_year']))
if pd.notna(x['date_added']) and pd.notna(x['release_year']) else None
,axis=1)
df.head()
```

#### Output:



Average time taken for movie to enter Netflix

# Code:

```
Movie_Mean = df.drop_duplicates(subset='title', keep =
'first')['Time_taken_to_enter_ott'][df['type'] == 'Movie'].mean()
print(Movie_Mean)
```

# **Output:**

```
    ➤ Average time taken for a movie to enter Netflix after getting released
    D Movie_Mean = df.drop_duplicates(subset='title', keep = 'first')['Time_taken_to_enter_ott'][df['type'] == 'Movie'].mean() print(Movie_Mean)
    2273 days 03:25:02.593377920
```

#### Average time taken for TV show to enter Netflix

```
TV_shows_Mean = df.drop_duplicates(subset='title', keep =
'first')['Time_taken_to_enter_ott'][df['type'] == 'TV Show'].mean()
print(TV_shows_Mean)
```

## **Output:**

```
✓ Average time taken for a TV show to enter Netflix after getting released
TV_shows_Mean = df.drop_duplicates(subset='title', keep = 'first')['Time_taken_to_enter_ott'][df['type'] == 'TV Show'].mean() print(TV_shows_Mean)
999 days 21:05:43.522110160
```

**Top 5 most rated movies** 

# Code:

```
Movies.groupby(by='rating')['title'].nunique().sort_values(ascending =
False).iloc[0:5]
```

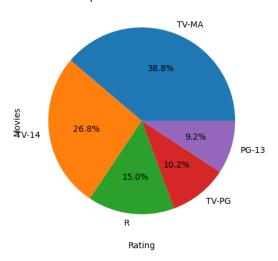
#### **Output:**

# **Graphical Representation:**

```
Movie_Ratings_plot =
pd.DataFrame(Movies.groupby(by='rating')['title'].nunique().sort_values
(ascending = False)).reset_index().head()
Movie_Ratings_plot.rename(columns = {'title':'count'},inplace = True)
plt.figure(figsize = (10,5))
plt.pie(Movie_Ratings_plot['count'], labels =
Movie_Ratings_plot['rating'], autopct='%1.1f%%')
plt.title('Top 5 most rated movies')
plt.xlabel('Rating')
plt.ylabel('Movies')
plt.show()
```

# Graph:

Top 5 most rated movies



# Insights:

From the graph we can clearly see that most movies in the Netflix are TV-MA rated (38.8%)

# Top 5 most rated TV shows

# Code:

```
TV_shows.groupby(by='rating')['title'].nunique().sort_values(ascending
= False)
```

## Output:

```
rating
TV-MA
                   1145
TV-14
                    733
TV-PG
                    323
TV-Y7
                    195
TV-Y
                    176
TV-G
                     94
NR
Unknown rating
                      2
TV-Y7-FV
Name: title, dtype: int64
```

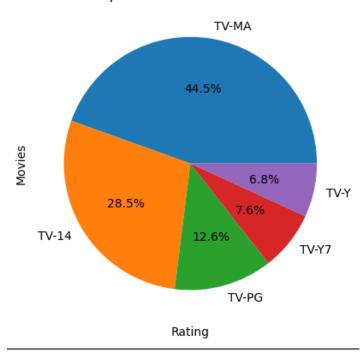
# **Graphical Representation:**

```
TV_shows_Ratings_plot =
pd.DataFrame(TV_shows.groupby(by='rating')['title'].nunique().sort_valu
es(ascending = False)).reset_index().head()
TV_shows_Ratings_plot.rename(columns = {'title':'count'},inplace =
True)
```

```
plt.pie(TV_shows_Ratings_plot['count'], labels =
TV_shows_Ratings_plot['rating'], autopct='%1.1f%%')
plt.title('Top 5 most rated TV shows')
plt.xlabel('Rating')
plt.ylabel('Movies')
plt.show()
```

Graph:

Top 5 most rated TV shows



<u>Insights:</u> From the graph we can clearly see that most TV shows in Netflix are TV-MA rated.

<u>Recommendations</u>: From the graph we can see most TV shows are TV-MA rated, if these rated movies bringing us more customers, we need to focus more marketing them by ads, if not we need to either decrease the count of bringing the most popular ones in these genres.