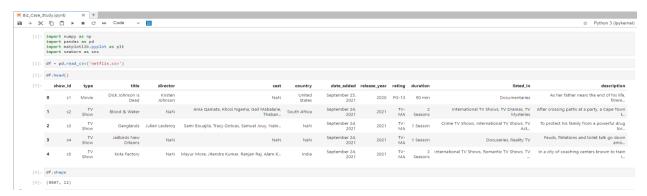
Business Case: Netflix - Data Exploration and Visualization

After loading the file, we can see as below,



From the file 'netflix.csv' we can see it have some issue regarding the data,

1. 'Nan' values are present in the director, cast, country etc. columns.



2. Some columns like cast, listed in, country, director etc. have nested data like,



To analyze the data, we need to clean the raw data first. We will do the below operations,

Operation 01: Un-nesting the columns,

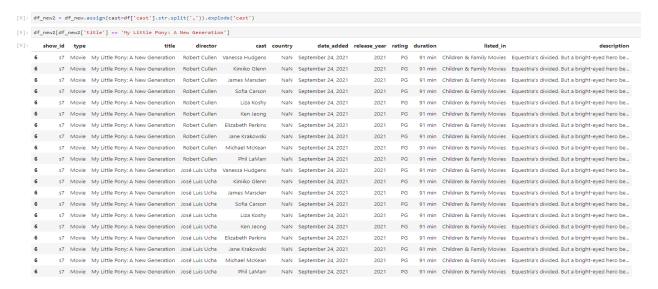
i. Director:



Here we can see for this movie two director present, after un-nesting the column we are getting as below,



ii. Cast:



iii. Country:

Same operation we have done for country also and we can see the row count increases.

iv. Listed in:

My final data frame has now 202065 number of rows.

Let's trim the whitespaces,

```
[25]: df_final['country'] = df_final['country'].str.strip()
df_final['director'] = df_final['director'].str.strip()
df_final['cast'] = df_final['cast'].str.strip()
df_final['listed_in'] = df_final['listed_in'].str.strip()
```

Operation 02:

Now we need to remove the 'NaN' values from the data frame.

```
[14]: df_final.isna().any()
[14]: show_id
                 False
                 False
     title
                 False
     director
                  True
                  True
     cast
                  True
     country
     date_added
                   True
     release_year False
     rating
                  True
     duration
                  True
     listed_in False
     description False
     dtype: bool
```

From above we can see some columns have Nan values, we need to handle those. As all the columns are categorical columns hence filled with 'Unknown Column Name',

```
[50]: df final['director'].fillna('Unknown Director', inplace = True)
      df_final['cast'].fillna('Unknown Cast', inplace = True)
      df_final['country'].fillna('Unknown Country', inplace = True)
      df_final['date_added'].fillna('January 1, 1900', inplace = True)
      df_final['rating'].fillna('Unknown Rating', inplace = True)
      df_final['duration'].fillna('Unknown Duration', inplace = True)
[16]: df_final.isna().any()
[16]: show_id
                   False
      type
                   False
     title
                   False
                   False
     director
                   False
     cast
     country
     date_added False
     release_year False
                 False
     rating
                   False
     duration
     listed_in
                   False
     description
                   False
     dtype: bool
```

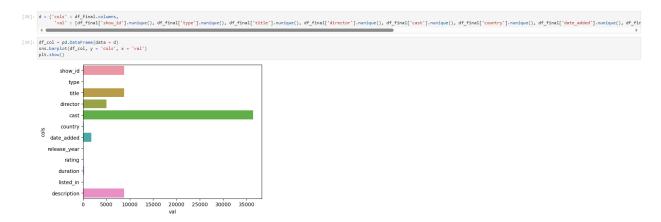
As we prepared the data, now let's analyze the data.

1. Find the counts of each categorical variable both using graphical and non-graphical analysis

We can see we have 12 columns and 202065 rows.

Below analysis shows the unique count present for each columns:

```
[26]: print('Show id count',df_final['show_id'].nunique())
      print('type count', df_final['type'].nunique())
      print('title count', df_final['title'].nunique())
      print('director count', df_final['director'].nunique())
      print('cast count', df_final['cast'].nunique())
      print('country count', df_final['country'].nunique())
      print('date added count', df_final['date_added'].nunique())
      print('release year count', df final['release year'].nunique())
      print('rating count', df_final['rating'].nunique())
      print('duration count', df final['duration'].nunique())
      print('listed in count', df_final['listed_in'].nunique())
      print('description count', df final['description'].nunique())
      Show id count 8807
      type count 2
      title count 8807
      director count 4994
      cast count 36440
      country count 124
      date added count 1768
      release year count 74
      rating count 18
      duration count 221
      listed in count 42
      description count 8775
```



Insights: From the above analysis we can see Netflix has rich amount of movies and tv shows. It have various shows from 1925 to 2018, among 124 countries.

2. Comparison of tv shows vs. movies.

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

```
[41]: grp_mv = df_final[df_final['type'] == 'Movie'].groupby('country')
      grp_tv = df_final[df_final['type'] == 'TV Show'].groupby('country')
[60]: grp_mv.nunique().sort_values(by = 'title', ascending = False)['title'].reset_index().head(11)
[60]:
                  country title
       0
              United States 2752
       1
                     India
                           962
       2
            United Kingdom
                            534
       3 Unknown Country
                            440
       4
                            319
                   Canada
       5
                    France
                            303
       6
                  Germany
                            182
       7
                    Spain
                            171
       8
                    Japan
                            119
                    China
                            114
                   Mexico
                            111
```

From above we can see top 10 countries if we ignore unknown country which produce movies.

```
[61]: grp_m = grp_mv.nunique().sort_values(by = 'title', ascending = False)['title'].reset_index().head(11)
        sns.barplot(grp_m, y = 'title', x = 'country')
        plt.xticks(rotation = 90)
        plt.show()
            2500
             2000
        을 1500
             1000
              500
                                                                        Germany -
                               India .
                                                                                Spain .
                                                                                                 China .
                       United States
                                       United Kingdom
                                               Unknown Country
                                                       Canada
                                                                France
                                                                                        Japan
                                                                                                         Mexico
                                                             country
```

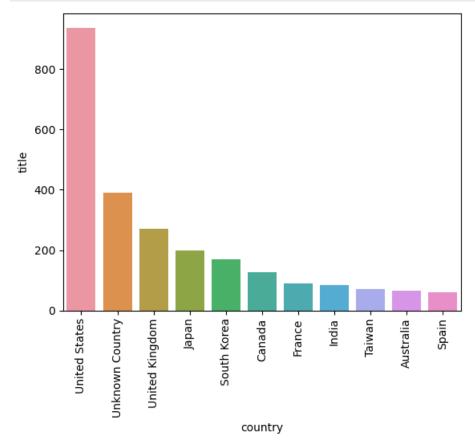
Insights: From the graph United States produce almost double movie than India which is in 2^{nd} place of top 10. The gap after 2^{nd} position is nominal.

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

```
[62]: grp_tv.nunique().sort_values(by = 'title', ascending = False)['title'].reset_index().head(11)
[62]:
                    country title
        0
               United States
                             391
           Unknown Country
        2
             United Kingdom
        3
                             199
                      Japan
                South Korea
                             170
                     Canada
        6
                     France
                               90
                       India
                               84
        8
                     Taiwan
                               70
                    Australia
                               66
       10
                      Spain
```

From above we can see top 10 countries if we ignore unknown country which produce Tv Shows.

```
[64]: grp_t = grp_tv.nunique().sort_values(by = 'title', ascending = False)['title'].reset_index().head(11)
sns.barplot(grp_t, y = 'title', x = 'country')
plt.xticks(rotation = 90)
plt.show()
```



Insights: From the graph United States produce almost thrice TV Shows than. The gap after 1st position is nominal.

Full Analysis Insight:

From both the cases we have seen United States produces most Movies and TV Shows in the world. We can say that United States have great focus on entertainment.

3. What is the best time to launch a TV show and Movies?

Before start this analysis lets create some required columns in the data frame,

<pre>2] df_final['date'] = pd.to_datetisme(df_final['date_added']) df_final['year'] = pd.to_datetisme(df_final['date_added']).dt.year df_final['nonth'] = pd.to_datetisme(df_final['date_added']).dt.month df_final['day'] = pd.to_datetisme(df_final['date_added']).dt.day df_final['weat'] = pd.to_datetisme(df_final['date_added']).dt.strftisme('%U') df_final['weat'] = pd.to_datetisme(df_final['date_added']).dt.strftisme('%U')</pre>																	
	show_id	type	title	director	cast	country	date_added	release_year	rating	duration	listed_in	description	date	year	month	day	week
0	s1	Movie	Dick Johnson Is Dead	Kirsten Johnson	Unknown Cast	United States	September 25, 2021	2020	PG-13	90 min	Documentaries	As her father nears the end of his life, filmm	2021-09-25	2021	9	25	38
1	s2	TV Show	Blood & Water	Unknown Director	Ama Qamata	South Africa	September 24, 2021	2021	TV-MA	2 Seasons	International TV Shows	After crossing paths at a party, a Cape Town $t_{\rm s}$	2021-09-24	2021	9	24	38
1	s2	TV Show	Blood & Water	Unknown Director	Ama Qamata	South Africa	September 24, 2021	2021	TV-MA	2 Seasons	TV Dramas	After crossing paths at a party, a Cape Town t	2021-09-24	2021	9	24	38
1	\$2	TV Show	Blood & Water	Unknown Director	Ama Qamata	South Africa	September 24, 2021	2021	TV-MA	2 Seasons	TV Mysteries	After crossing paths at a party, a Cape Town t	2021-09-24	2021	9	24	38
	-2	TV Chau	Diamel St. Minter	Universe Diseases	Vhasi Nassas	Courtle Africa	Contombox 24, 2021	2021	TIV NAA	2.5	International TV Chause	After receipe paths at a party a Cons Town t	2021 00 24	2021		24	20

a. Find which is the best week to release the TV-show or the movie. Let's do the analysis separately for TV-shows and Movies.

Movies:

First week, 39th and 26th week of every year is best time to release the Movies.

TV Shows:

For TV Show we can see 39th, 31st, 26th Week of the year is the best time.

Full Analysis Insight:

From the above analysis we can say that 39th week and 26th week of every year is best for Movie and TV Show release.

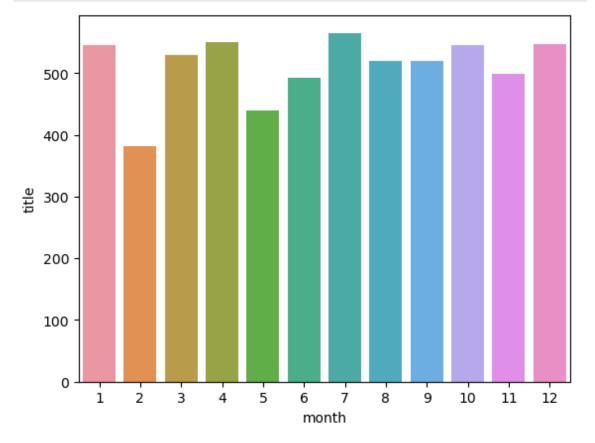
b. Find which is the best month to release the TV-show or the movie. Let's do the analysis separately for TV-shows and Movies.

Movies:

```
[67]: grp_month_mv = df_final[df_final['type'] == 'Movie'].groupby('month')
grp_mm = grp_month_mv.nunique().sort_values(by = 'title', ascending = False)['title'].reset_index()
grp_mm.head()
```

```
[67]: month title
0 7 565
1 4 550
2 12 547
3 1 546
4 10 545
```

```
[68]: sns.barplot(grp_mm, x = 'month', y = 'title')
plt.show()
```



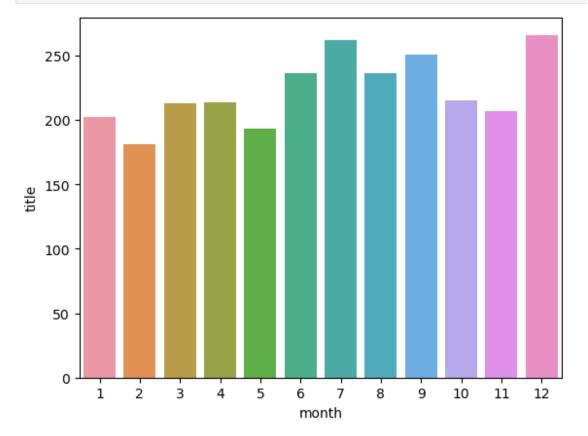
Here we can see the best month is July for Movies.

TV Show:

```
[69]: grp_month_tv = df_final[df_final['type'] == 'TV Show'].groupby('month')
grp_mt = grp_month_tv.nunique().sort_values(by = 'title', ascending = False)['title'].reset_index()
grp_mt.head()
[69]: month title
```

9]:		month	title
	0	12	266
	1	7	262
	2	9	251
	3	6	236
	4	8	236

```
[70]: sns.barplot(grp_mt, x = 'month', y = 'title')
plt.show()
```



December, July are the best months for TV Shows.

Full Analysis Insight:

From the month and week analysis we can see that data are matching for both the week and month. Jan, July are the best time for both Movies and TV shows.

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



Here the above table represent the top 10 actors who are worked on most TV shows and Movies. Though we have 825 unknown actors are there as it is not provided in the data file. I have tried with Mode as it is a categorical column but it leads to wrong value.

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



Here the above table represent the top 10 actors who are worked on most TV shows and Movies. Though we have 2634 unknown directors are there as it is not provided in the data file. I have tried with Mode as it is a categorical column but it leads to wrong value.

5. Which genre movies are more popular or produced more

```
[75]: from wordcloud import WordCloud, STOPWORDS

text = df_final['listed_in'].values
wordcloud = WordCloud().generate(str(text))

plt.imshow(wordcloud)
plt.axis("off")
plt.show()
```



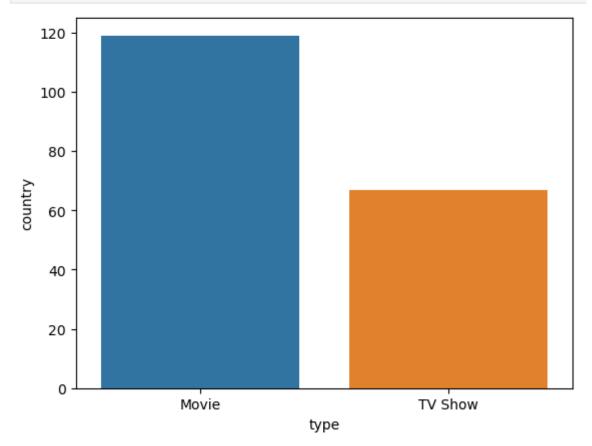
6. Find After how many days the movie will be added to Netflix after the release of the movie

From the past data we can observed that maximum movies are added in same year of release in the Netflix.

7. Type of contents in the countries:

From the below graph we can say that most of the countries preferred Movies over TV shows.

```
[77]: grp_t = df_final.groupby('type')
grp_tp = grp_t['country'].nunique().reset_index()
sns.barplot(grp_tp, x = 'type', y = 'country', )
plt.show()
```

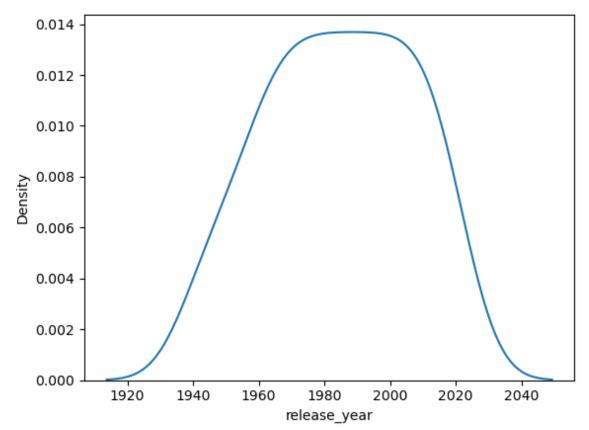


8. How has the number of movies released per year changed over the last 20-30 years?

```
[90]: grp_y = df_final[df_final['type'] == 'Movie'].groupby('release_year')
grp_year = grp_y['title'].nunique().sort_index(ascending = False).reset_index()
grp_year.head()
```

[90]:		release_year	title
	0	2021	277
	1	2020	517
	2	2019	633
	3	2018	767
	4	2017	767

```
[91]: sns.kdeplot(data = grp_year, x = 'release_year')
plt.show()
```



So the above KDE Plot is slightly right skewed, which depict that last 20 to 30 years had more movie release.

Final Insights:

- 1. Netflix has more movies than TV shows.
- 2. Netflix added movies and TV shows mostly within a year of their release date.
- 3. Netflix increases their content over last 20 to 30 years.
- 4. July month of every year is best time to release Movies and TV shows over the platform.
- 5. Maximum content is from United States.

Recommendations:

- 1. Netflix should add more TV shows.
- 2. Netflix should add content from more countries not only maximum from United States, other countries also have rich amount of contents, which lead Netflix to achieve more users.

3. Netflix should release more content on festival months depends on the countries.