```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

Problem Statement

To enhance Netflix's revenue, a key strategy is to increase user subscriptions. This can be achieved by analyzing the existing Netflix dataset to identify patterns, preferences, and gaps in the current content offerings. By understanding what drives user engagement and satisfaction, we can make data-driven decisions to attract and retain more subscribers.

Basic Metrics

```
In [2]: data = pd.read_csv(r'E:\NextStep\Scaler\Business-Case-Study\Netflix\Dataset\netf
In [3]: data.head()
```

Out[3]:		show_id	type	title	director	cast	country	date_added	release_year
	0	s1	Movie	Dick Johnson Is Dead	Kirsten Johnson	NaN	United States	September 25, 2021	2020
	1	s2	TV Show	Blood & Water	NaN	Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban	South Africa	September 24, 2021	2021
	2	s3	TV Show	Ganglands	Julien Leclercq	Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabi	NaN	September 24, 2021	2021
	3	s 4	TV Show	Jailbirds New Orleans	NaN	NaN	NaN	September 24, 2021	2021
	4	s5	TV Show	Kota Factory	NaN	Mayur More, Jitendra Kumar, Ranjan Raj, Alam K	India	September 24, 2021	2021
	4								•
In [4]:	data	a.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
dtype	es: int64(1),	object(11)	

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

Inference:

- No. of rows: 8807, No. of columns: 12
- All columns except release_year are of string type

```
In [5]: data.describe()
```

 count
 8807.000000

 mean
 2014.180198

 std
 8.819312

 min
 1925.000000

 25%
 2013.000000

 50%
 2017.000000

 75%
 2019.000000

 max
 2021.000000

Inference:

- Release year:
 - Only Release year is numerical variable and others are all categorical variable
 - Dataset contains movies from 1925 to 2021
- 50% of Movies/shows present in netflix are released after 2017

```
In [6]: data.describe(include = 'object')
```

Out[6]:		show_id	type	title	director	cast	country	date_added	rating
	count	8807	8807	8807	6173	7982	7976	8797	8803
	unique	8807	2	8807	4528	7692	748	1767	17
	top	s1	Movie	Dick Johnson Is Dead	Rajiv Chilaka	David Attenborough	United States	January 1, 2020	TV- MA
	freq	1	6131	1	19	19	2818	109	3207
	4								>

Inference:

- Type: There are 2 types of videos: Movies and Shows
- Country: It needs further investigation as there are only 190 countries are recognised
- date_added: Movies are added mostly together on same days

Inference:

- The first movie added to Netflix was "To and From New York," released in 2006.
- The oldest movie currently on Netflix is "First Women Filmmakers," released in 1925

Missing Value Detection

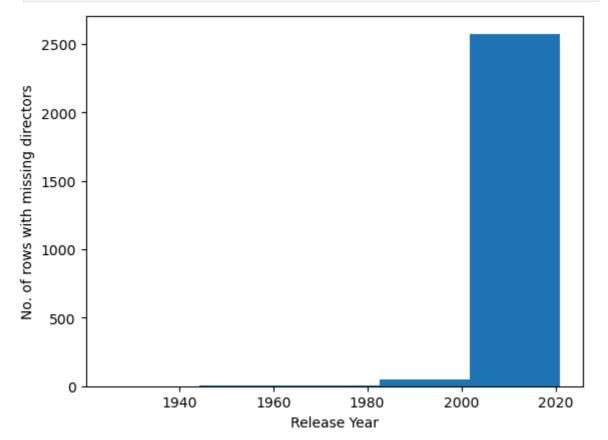
```
In [8]: data.isna().sum()
Out[8]: show_id
                           0
                           0
        type
        title
                           0
                       2634
        director
                        825
        cast
        country
                         831
                          10
        date_added
        release_year
        rating
        duration
        listed in
                           0
        description
        dtype: int64
```

Inference:

- show_id, type, title, release_year, listed_in, description doesn't contain missing values
- date_added, rating and duration: Very few samples with missing values

 director: Almost 30% samples are having null values, needs further investigation

```
In [9]: cnt, bins, _ = plt.hist(data[data['director'].isna()]['release_year'], bins = 5)
    plt.xlabel('Release Year')
    plt.ylabel('No. of rows with missing directors')
    plt.show()
```



Inference: Samples with movies/shows released after 2000 are having
missing director names

Non Graphical Analysis of Columns

In [11]:	<pre>data.nunique()</pre>				
Out[11]:	show_id	8807			
	type	2			
	title	8807			
	director	4528			
	cast	7692			
	country	748			
	date_added	1767			
	release_year	74			
	rating	17			
	duration	220			
	listed_in	514			
	description	8775			
	dtype: int64				
	Inference:				

- show_id, title contains all unique values that makes sense
- director: All unique values with some missing values
- description: Interestingly there are some duplicates in description that needs to be explored

Detailed Analysis of Cast

```
In [12]: data['cast_list'] = data['cast'].str.split(',')
         data_cast_exp = data[['show_id', 'type', 'cast_list']].explode('cast_list')
         data_cast_exp['cast_list'] = data_cast_exp['cast_list'].str.strip()
         data_cast_exp.nunique()
Out[12]: show_id
                     8807
         type
         cast_list
                      36439
         dtype: int64
         Inference: 36439 Actors names where listed in the cast, individually
         analysis everyone is infeasible
In [13]: data_cast_exp['cast_list'].value_counts()
Out[13]: cast_list
                                   43
         Anupam Kher
                                   35
         Shah Rukh Khan
         Julie Tejwani
                                    33
         Naseeruddin Shah
                                   32
         Takahiro Sakurai
                                   32
                                    . .
         Maryam Zaree
                                    1
         Melanie Straub
         Gabriela Maria Schmeide
                                    1
         Helena Zengel
         Chittaranjan Tripathy
                                    1
         Name: count, Length: 36439, dtype: int64
         Inference: Anupam Kher's and Shah rukh khan's movies and shows are
         popular (most) in netflix
```

Detailed Analysis of Country

```
data_cntr_exp['country_list'].value_counts()
In [15]:
Out[15]: country_list
         United States
                          3690
         India
                          1046
         United Kingdom
                          806
         Canada
                           445
         France
                           393
         Ecuador
         Armenia
                             1
         Mongolia
                             1
                             1
         Bahamas
         Montenegro
         Name: count, Length: 123, dtype: int64
         Inference: Netflix has listed the majority of the movies and shows from
```

United States

Detailed Analysis of Genre (Listed in)

```
In [16]: data['genre'] = data['listed_in'].str.split(',')
         data_exp_gen = data[['show_id','type', 'genre']].explode('genre')
         data_exp_gen['genre'] = data_exp_gen['genre'].str.strip()
         data_exp_gen['genre'].nunique()
Out[16]: 42
         Inference: Movies and TV shows are listed under 42 different type of
         genres
In [17]: data_exp_gen['genre'].unique()
Out[17]: array(['Documentaries', 'International TV Shows', 'TV Dramas',
                 'TV Mysteries', 'Crime TV Shows', 'TV Action & Adventure',
                 'Docuseries', 'Reality TV', 'Romantic TV Shows', 'TV Comedies',
                 'TV Horror', 'Children & Family Movies', 'Dramas',
                 'Independent Movies', 'International Movies', 'British TV Shows',
                 'Comedies', 'Spanish-Language TV Shows', 'Thrillers',
                 'Romantic Movies', 'Music & Musicals', 'Horror Movies',
                 'Sci-Fi & Fantasy', 'TV Thrillers', "Kids' TV",
                 'Action & Adventure', 'TV Sci-Fi & Fantasy', 'Classic Movies',
                 'Anime Features', 'Sports Movies', 'Anime Series',
                 'Korean TV Shows', 'Science & Nature TV', 'Teen TV Shows',
                 'Cult Movies', 'TV Shows', 'Faith & Spirituality', 'LGBTQ Movies',
                 'Stand-Up Comedy', 'Movies', 'Stand-Up Comedy & Talk Shows',
                 'Classic & Cult TV'], dtype=object)
```

Inference: There are common genre between movies and TV shows such as Documentaries for movies and docuseries for tv shows. In further analysis, combining such names to a common name for both movies and shows, helps to understand the genre better

```
data_exp_gen['genre_comb'] = data_exp_gen['genre'].str.lower()
In [18]:
         data_exp_gen['genre_comb'] = data_exp_gen['genre_comb'].str.replace('tv', '').st
```

data_exp_gen['genre_comb'] = data_exp_gen['genre_comb'].str.replace('docuseries'

- Removed 'Movies' and 'TV' text from genre
- Merged Docuseries and documentaries

```
In [19]: print('No. of genre:', data_exp_gen['genre_comb'].nunique())
    print(data_exp_gen['genre_comb'].unique())

No. of genre: 30
    ['documentaries' 'international' 'dramas' 'mysteries' 'crime'
        'action & adventure' 'reality' 'romantic' 'comedies' 'horror'
        'children & family' 'independent' 'british' 'spanish-language'
        'thrillers' 'music & musicals' 'sci-fi & fantasy' "kids'" 'classic'
        'anime' 'sports' 'korean' 'science & nature' 'teen' 'cult' ''
        'faith & spirituality' 'lgbtq' 'stand-up comedy' 'classic & cult']

Inference: Genre of movies and tv shows are combined to 30 from 42 and
        the list is shown above
```

Detailed Analysis of Rating

```
In [20]: data['rating'].nunique()
Out[20]: 17
In [21]: data['rating'].value_counts()
Out[21]: rating
         TV-MA
                     3207
         TV-14
                     2160
         TV-PG
                     863
                     799
         PG-13
                     490
         TV-Y7
                      334
         TV-Y
                     307
         PG
                     287
         TV-G
                      220
         NR
                       80
                       41
         TV-Y7-FV
                       6
         NC-17
                        3
         UR
         74 min
         84 min
                        1
         Name: count, dtype: int64
```

Inference:

- 17 unique rating categories exist
- 74min, 84min, 66min are seems to be outliers
- Predominant of movies and tv shows are not suitable for children under 14 (Recommendation 2)

Detailed Analysis of Duration

```
In [22]: data[data['type'] == 'Movie']['duration'].unique(), data[data['type'] != 'Movie'
Out[22]: (array(['90 min', '91 min', '125 min', '104 min', '127 min', '67 min',
                    '94 min', '161 min', '61 min', '166 min', '147 min', '103 min',
                   '97 min', '106 min', '111 min', '110 min', '105 min', '96 min', '124 min', '116 min', '98 min', '23 min', '115 min', '122 min',
                   '99 min', '88 min', '100 min', '102 min', '93 min', '95 min',
                   '85 min', '83 min', '113 min', '13 min', '182 min', '48 min',
                   '145 min', '87 min', '92 min', '80 min', '117 min', '128 min',
                   '119 min', '143 min', '114 min', '118 min', '108 min', '63 min',
                   '121 min', '142 min', '154 min', '120 min', '82 min', '109 min',
                   '101 min', '86 min', '229 min', '76 min', '89 min', '156 min', '112 min', '107 min', '129 min', '135 min', '136 min', '165 min',
                   '150 min', '133 min', '70 min', '84 min', '140 min', '78 min',
                   '64 min', '59 min', '139 min', '69 min', '148 min', '189 min',
                   '141 min', '130 min', '138 min', '81 min', '132 min', '123 min',
                   '65 min', '68 min', '66 min', '62 min', '74 min', '131 min',
                   '39 min', '46 min', '38 min', '126 min', '155 min', '159 min',
                   '137 min', '12 min', '273 min', '36 min', '34 min', '77 min',
                   '60 min', '49 min', '58 min', '72 min', '204 min', '212 min', '25 min', '73 min', '29 min', '47 min', '32 min', '35 min',
                   '71 min', '149 min', '33 min', '15 min', '54 min', '224 min',
                   '162 min', '37 min', '75 min', '79 min', '55 min', '158 min',
                   '164 min', '173 min', '181 min', '185 min', '21 min', '24 min',
                   '51 min', '151 min', '42 min', '22 min', '134 min', '177 min',
                   '52 min', '14 min', '53 min', '8 min', '57 min', '28 min',
                   '50 min', '9 min', '26 min', '45 min', '171 min', '27 min',
                   '44 min', '146 min', '20 min', '157 min', '17 min', '203 min',
                   '41 min', '30 min', '194 min', '233 min', '237 min', '230 min',
                   '195 min', '253 min', '152 min', '190 min', '160 min', '208 min',
                   '180 min', '144 min', '5 min', '174 min', '170 min', '192 min',
                   '209 min', '187 min', '172 min', '16 min', '186 min', '11 min',
                   '193 min', '176 min', '56 min', '169 min', '40 min', '10 min',
                   '3 min', '168 min', '312 min', '153 min', '214 min', '31 min',
                   '163 min', '19 min', nan, '179 min', '43 min', '200 min',
                   '196 min', '167 min', '178 min', '228 min', '18 min', '205 min',
                   '201 min', '191 min'], dtype=object),
            array(['2 Seasons', '1 Season', '9 Seasons', '4 Seasons', '5 Seasons',
                   '3 Seasons', '6 Seasons', '7 Seasons', '10 Seasons', '8 Seasons',
                   '17 Seasons', '13 Seasons', '15 Seasons', '12 Seasons',
                   '11 Seasons'], dtype=object))
```

Inference:

- Duration of Movies: Represented in minutes
- Duration of TV Shows: Represented in seasons

```
In [23]: dur_movie = data[data['type'] == 'Movie']
    dur_movie['dur'] = dur_movie['duration'].str.split()
    dur_movie['dur'] = dur_movie['dur'].str[0]
    dur_movie['dur'] = dur_movie['dur'].astype('float64')
    dur_movie[~dur_movie['dur'].isna()]['dur'].describe()
```

```
C:\Users\Muthukumar\AppData\Local\Temp\ipykernel_27252\1707604282.py:2: SettingWi
thCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stabl
e/user_guide/indexing.html#returning-a-view-versus-a-copy
  dur_movie['dur'] = dur_movie['duration'].str.split()
C:\Users\Muthukumar\AppData\Local\Temp\ipykernel_27252\1707604282.py:3: SettingWi
thCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stabl
e/user_guide/indexing.html#returning-a-view-versus-a-copy
 dur_movie['dur'] = dur_movie['dur'].str[0]
C:\Users\Muthukumar\AppData\Local\Temp\ipykernel_27252\1707604282.py:4: SettingWi
thCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame.
Try using .loc[row_indexer,col_indexer] = value instead
See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stabl
e/user_guide/indexing.html#returning-a-view-versus-a-copy
 dur_movie['dur'] = dur_movie['dur'].astype('float64')
```

```
6128.000000
Out[23]: count
                    99.577187
         mean
         std
                    28.290593
         min
                     3.000000
         25%
                    87.000000
         50%
                    98.000000
         75%
                   114.000000
                   312.000000
         Name: dur, dtype: float64
```

Inference: Duration of Movies in netflix ranges from 3 minutes to 312
minutes

```
In [24]: dur_show = data[data['type'] != 'Movie']
   dur_show['dur'] = dur_show['duration'].str.split()
   dur_show['dur'] = dur_show['dur'].str[0]
   dur_show['dur'] = dur_show['dur'].astype('float64')
   dur_show[~dur_show['dur'].isna()]['dur'].describe()
```

```
C:\Users\Muthukumar\AppData\Local\Temp\ipykernel_27252\1543813386.py:2: SettingWi
         thCopyWarning:
         A value is trying to be set on a copy of a slice from a DataFrame.
         Try using .loc[row_indexer,col_indexer] = value instead
         See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stabl
         e/user_guide/indexing.html#returning-a-view-versus-a-copy
           dur_show['dur'] = dur_show['duration'].str.split()
         C:\Users\Muthukumar\AppData\Local\Temp\ipykernel_27252\1543813386.py:3: SettingWi
         thCopyWarning:
         A value is trying to be set on a copy of a slice from a DataFrame.
         Try using .loc[row_indexer,col_indexer] = value instead
         See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stabl
         e/user_guide/indexing.html#returning-a-view-versus-a-copy
           dur_show['dur'] = dur_show['dur'].str[0]
         C:\Users\Muthukumar\AppData\Local\Temp\ipykernel_27252\1543813386.py:4: SettingWi
         thCopyWarning:
         A value is trying to be set on a copy of a slice from a DataFrame.
         Try using .loc[row_indexer,col_indexer] = value instead
         See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stabl
         e/user_guide/indexing.html#returning-a-view-versus-a-copy
           dur_show['dur'] = dur_show['dur'].astype('float64')
                   2676.000000
Out[24]: count
                      1.764948
          mean
          std
                      1.582752
          min
                      1.000000
          25%
                      1.000000
          50%
                      1.000000
          75%
                      2.000000
                     17.000000
          Name: dur, dtype: float64
          Inference: Duration of TV shows in netflix ranges from 1 season to 17
          seasons
In [102...
          dur show['dur'].value counts()
Out[102...
          dur
          1.0
                  1793
          2.0
                   425
          3.0
                   199
          4.0
                    95
          5.0
                    65
          6.0
                    33
          7.0
                    23
                    17
          8.0
          9.0
                     9
                     7
          10.0
          13.0
                     3
                     2
          15.0
          12.0
                      2
          11.0
                      2
          17.0
                     1
          Name: count, dtype: int64
```

Inference: Only 17 TV shows on Netflix have more than 10 seasons.
(Recommendation 8)

Detailed Analysis of Description

```
In [25]: data_desc_dup = data[data['description'].duplicated(keep=False)]
    data_desc_dup.groupby('description')['title'].apply(lambda x: ','.join(x)).reset
```

Out[25]: description title

	description	title
0	A budding politician has devious plans to rise	Tughlaq Durbar,Tughlaq Durbar (Telugu)
1	A scheming matriarch plots to cut off her disa	Together For Eternity, Hum Saath-Saath Hain
2	A surly septuagenarian gets another chance at	Oh! Baby (Malayalam),Oh! Baby (Tamil),Oh! Baby
3	A young Han Solo tries to settle an old score	Solo: A Star Wars Story, Solo: A Star Wars Stor
4	After devastating terror attacks in Norway, a	22 July,22-Jul
5	After growing up enduring criticism from his f	Angu Vaikuntapurathu (Malayalam),Ala Vaikuntha
6	An affable, newly appointed college warden pro	Petta (Telugu Version),Petta
7	An aspiring musician battles age-old caste div	Sarvam Thaala Mayam (Telugu Version),Sarvam Th
8	As a blind librarian, dispirited cricketer and	Andhaghaaram, Andhakaaram
9	As a psychology professor faces Alzheimer's, h	February 9,Feb-09
10	As a series of murders hit close to home, a vi	Game Over (Hindi Version),Game Over (Tamil Ver
11	As a woman scours Hyderabad for her missing hu	Anaamika,Nee Enge En Anbe
12	Born into a small town controlled by the mafia	Sin senos sí hay paraíso,Sin Senos sí Hay Paraíso
13	Challenged to compose 100 songs before he can	99 Songs,99 Songs (Tamil),99 Songs (Telugu)
14	Cora has three sons and a daughter and she's a	Esperando la carroza,Esperando La Carroza
15	From controversial cartoonist to powerful Mumb	Thackeray (Hindi),Thackeray
16	Mistakenly accused of an attack on the Fourth	Naruto Shippuden the Movie: Blood Prison,Narut
17	Multiple women report their husbands as missin	Seven (Tamil),Seven (Telugu),7 (Seven)
18	On India's Independence Day, a zany mishap in	15 August,15-Aug
19	Paranormal activity at a lush, abandoned prope	Boomika,Boomika (Hindi),Boomika (Malayalam),Bo
20	Secrets bubble to the surface after a sensual	Consequences, Consequences

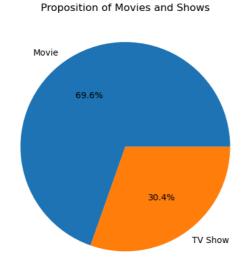
title	description	
InuYasha the Movie 3: Swords of an Honorable R	The Great Dog Demon beaqueathed one of the Thr	21
ChuChu TV Nursery Rhymes & Kids Songs (Hindi),	This educational series for tiny tots features	22
The Ollie & Moon Show,Ollie & Moon	Two quirky cats, Ollie and Moon, and their fri	23
Incredibles 2 (Spanish Version),The Incredibles 2	When Elastigirl gets recruited for a high- prof	24
Chashme Baddoor,Chashme Buddoor	When pretty new neighbor Seema falls for their	25
InuYasha the Movie 2: The Castle Beyond the Lo	With their biggest foe seemingly defeated, Inu	26

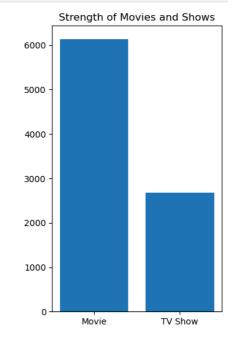
Inference: Duplicate descriptions are due to the presence of multi lingual movies or Shows

Graphical Analysis of Columns

Type: Movies Vs TV Shows

```
In [26]: plt.figure(figsize=(12,6))
    plt.subplot(1,2,1)
    plt.pie(data['type'].value_counts().reset_index()['count'], labels = list(data['
    plt.title('Proposition of Movies and Shows')
    plt.subplot(1,3,3)
    plt.bar(height = data['type'].value_counts().reset_index()['count'], x = list(data['type'].title('Strength of Movies and Shows')
    plt.show()
```

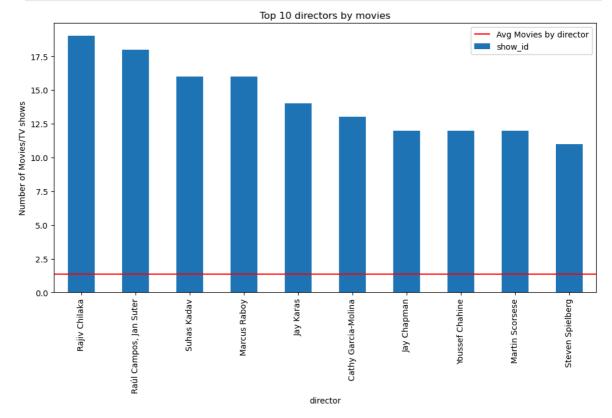




Inference: Movies dominate over TV shows, comprising 70% compared to 30%

Director: Top 10 directors Movies and TV shows

```
In [27]: plt.figure(figsize=(12,6))
   data_grp = data.groupby('director')['show_id'].count().sort_values(ascending=Fal
   data_grp.iloc[:10].plot(kind='bar')
   plt.title('Top 10 directors by movies')
   plt.ylabel('Number of Movies/TV shows')
   plt.hlines(data_grp.mean(), -1, 10, color = 'r', label='Avg Movies by director')
   plt.legend()
   plt.show()
```

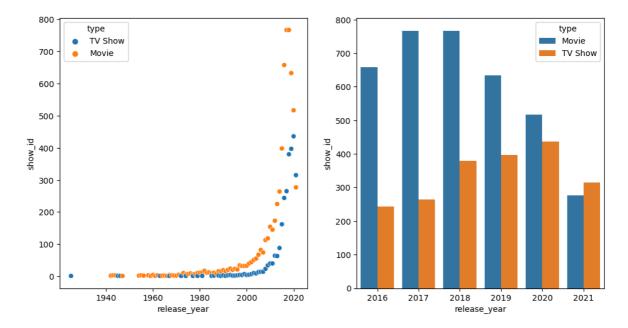


Inference: Among the movies and TV shows listed on Netflix, Rajiv
Chilaka stands out as the leading director with 18 movies

Release Year: Movies and TV Shows by Year

```
In [28]: # Movies and Shows by year
plt.figure(figsize=(12,6)).suptitle('Movies and Show by Release year')
data_grp = data.groupby(['release_year', 'type'])['show_id'].count().reset_index
plt.subplot(1,2,1)
sns.scatterplot(data = data_grp, x = 'release_year', y='show_id', hue = 'type')
plt.subplot(1,2,2)
sns.barplot(data = data_grp[data_grp['release_year'] > 2015], x = 'release_year'
plt.show()
```

Movies and Show by Release year



Inference:

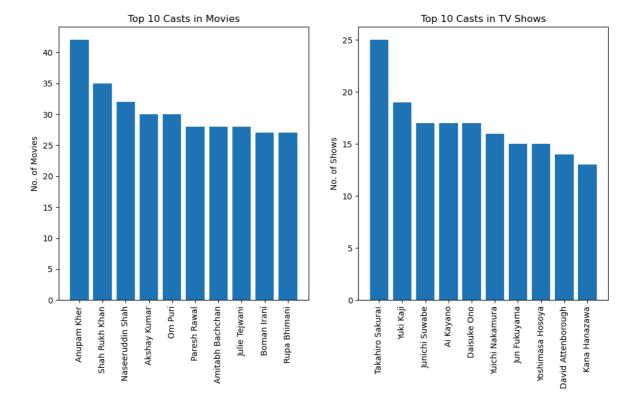
- Plot1: The majority of movies and shows available on Netflix were released after 2010 (Recommendation 1)
- Plot2: The strength of TV shows compared to movies is increasing towards the end, indicating a declining trend in the number of movies

Cast: Top 10 Actors in Movies and TV Shows

```
In [29]: data_cast_grp = data_cast_exp.groupby('type')['cast_list'].value_counts()

plt.figure(figsize=(12,6))
plt.subplot(1,2,1)
plt.bar(x = data_cast_grp['Movie'].index[:10], height = data_cast_grp['Movie'][:
plt.xticks(rotation = 90)
plt.ylabel('No. of Movies')
plt.title('Top 10 Casts in Movies')

plt.subplot(1,2,2)
plt.bar(x = data_cast_grp['TV Show'].index[:10], height = data_cast_grp['TV Show plt.xticks(rotation = 90)
plt.ylabel('No. of Shows')
plt.title('Top 10 Casts in TV Shows')
```

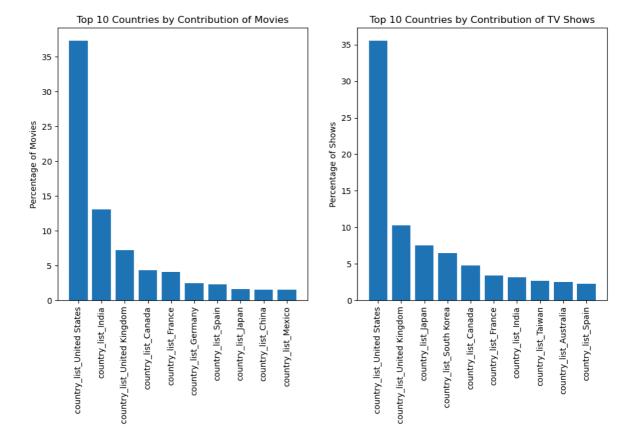


Inference:

- Cast in Movies:
 - The Top actors are predominantly from India
 - Anupam kher and Shahrukh khan are the leading actors
 - No Actors from South India are listed in top 10 (Recommendation 6)
- Cast in Shows:
 - The Top actors are predominantly from Japan and Korea

Country: Movies and TV Shows by Country

```
In [30]:
        data_country = pd.get_dummies(data_cntr_exp, columns =['country_list'])
In [31]:
         plt.figure(figsize=(12,6))
         data_temp = data_country[data_country['type'] == 'Movie']
         data_country_grp = (data_temp.iloc[:, 2:].sum() * 100/(data_temp.iloc[:, 2:].sum
         plt.subplot(1,2,1)
         plt.bar(x = data_country_grp.index[:10], height = data_country_grp[:10])
         plt.xticks(rotation = 90)
         plt.ylabel('Percentage of Movies')
         plt.title('Top 10 Countries by Contribution of Movies')
         data_temp = data_country[data_country['type'] != 'Movie']
         data_country_grp = (data_temp.iloc[:, 2:].sum() * 100/(data_temp.iloc[:, 2:].sum
         plt.subplot(1,2,2)
         plt.bar(x = data_country_grp.index[:10], height = data_country_grp[:10])
         plt.xticks(rotation = 90)
         plt.ylabel('Percentage of Shows')
         plt.title('Top 10 Countries by Contribution of TV Shows')
         plt.show()
```



Inference:

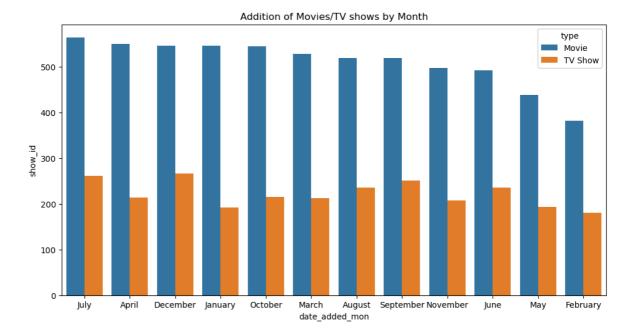
- 35% of shows and movies are from United States
- While India is the second largest contributor of movies, it ranks
 7th in the contribution of shows (Recommendation 7)
- Japan and Korean series are more popular than their movies

Date_Added:

```
In [32]: data['date_added'] = pd.to_datetime(data['date_added'].str.strip())
  data['date_added_mon'] = data['date_added'].dt.month_name()
  data['date_added_yr'] = data['date_added'].dt.year
  data['date_added_day'] = data['date_added'].dt.day
  data['date_added_weekday'] = data['date_added'].dt.day_name()
```

Addition of Movies by Month

```
In [33]: plt.figure(figsize=(12,6))
  data_grp = data.groupby(['date_added_mon', 'type'])['show_id'].count().reset_ind
  sns.barplot(data = data_grp, x = 'date_added_mon', y='show_id', hue = 'type')
  plt.title('Addition of Movies/TV shows by Month')
  plt.show()
```

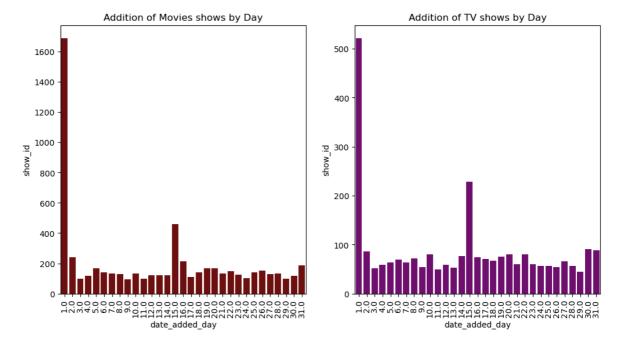


Inference:

- A large number of movies are added in July and April, while the majority of TV shows are added in July and September
- Fewer movies and TV shows are added in January and February (Recommendation 5)

Addition of Movies/TV Shows by day of month

```
In [34]:
    plt.figure(figsize=(12,6))
    data_grp = data.groupby(['date_added_day', 'type'])['show_id'].count().reset_ind
    plt.subplot(1,2,1)
    sns.barplot(data = data_grp[data_grp['type'] == 'Movie'], x = 'date_added_day',
    plt.title('Addition of Movies shows by Day')
    plt.xticks(rotation = 90)
    plt.subplot(1,2,2)
    sns.barplot(data = data_grp[data_grp['type'] != 'Movie'], x = 'date_added_day',
    plt.title('Addition of TV shows by Day')
    plt.xticks(rotation = 90)
    plt.show()
# Both plots clearly depicts that major addition of movies/tv shows on start of
```

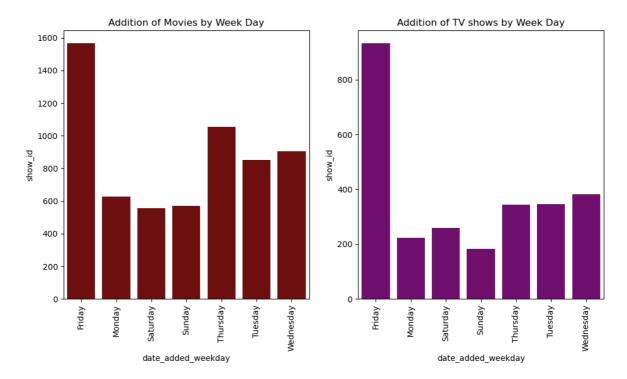


Inference: Both plots clearly depict that the majority of movies and TV
shows are added at the beginning and middle of the month

Addition of Movies/TV Shows by week day

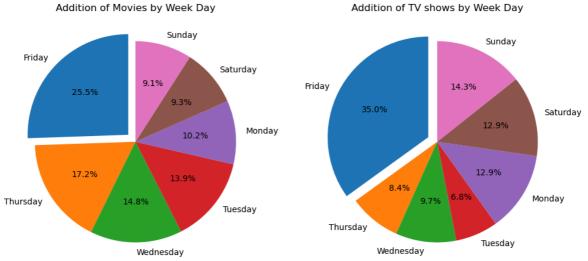
```
In [35]: plt.figure(figsize=(12,6))

data_grp = data.groupby(['date_added_weekday', 'type'])['show_id'].count().reset
   plt.subplot(1,2,1)
   sns.barplot(data = data_grp[data_grp['type'] == 'Movie'], x = 'date_added_weekda
   plt.title('Addition of Movies by Week Day')
   plt.xticks(rotation = 90)
   plt.subplot(1,2,2)
   sns.barplot(data = data_grp[data_grp['type'] != 'Movie'], x = 'date_added_weekda
   plt.title('Addition of TV shows by Week Day')
   plt.xticks(rotation = 90)
   plt.show()
```



Inference: It is clearly evident that the majority of movies and TV
shows are added on Fridays, targeting the weekend audience

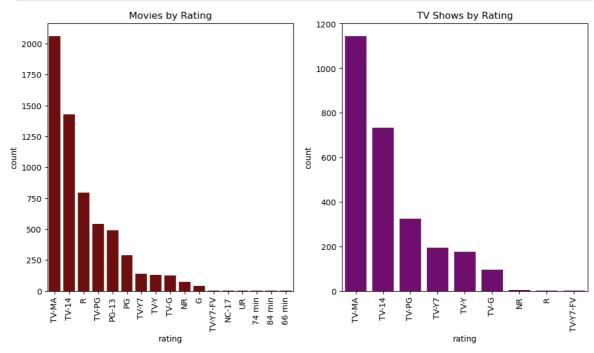




Inference: In the detailed view, TV shows are predominantly added on
Fridays, whereas movies are added on Thursdays and Fridays

Rating: Distribution of Movies and TV shows by Rating

```
In [37]: plt.figure(figsize=(12,6))
  plt.subplot(1,2,1)
  sns.countplot(data = data[data['type'] == 'Movie'], x = 'rating', color = 'maroc
  plt.title('Movies by Rating')
  plt.xticks(rotation = 90)
  plt.subplot(1,2,2)
  sns.countplot(data = data[data['type'] != 'Movie'], x = 'rating', color = 'purpl
  plt.title('TV Shows by Rating')
  plt.xticks(rotation = 90)
  plt.show()
```



Inference:

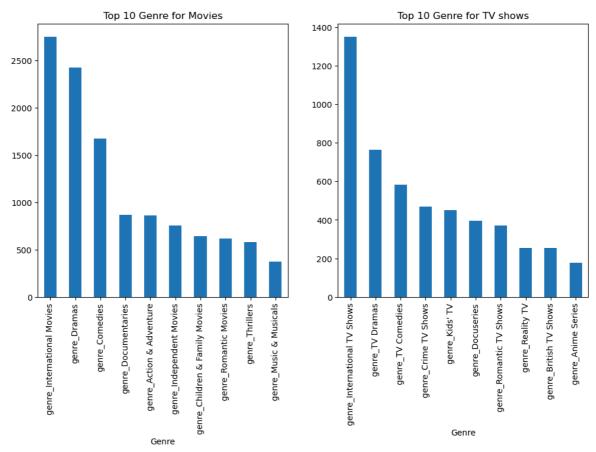
- The majority of movies on Netflix are rated TV-MA (Mature Audience) and TV-14 (Unsuitable for children under 14).
- Interestingly, 'R' rated TV shows are significantly fewer compared to 'R' rated movies, which typically contain adult content.
- Most of the movies and TV shows listed on Netflix are unsuitable for children. (Recommendation 2)

Listed in (Genre): Movies and TV Shows by Genre

```
In [38]: data_grp = pd.get_dummies(data_exp_gen, columns = ['genre'])
In [53]: plt.figure(figsize=(12,6))
   plt.subplot(1,2,1)
   test = data_grp[data_grp['type'] == 'Movie'].drop(['genre_comb', 'type'], axis=1
```

```
test.sum(axis = 0).sort_values(ascending = False).iloc[:10].plot(kind='bar')
plt.xlabel('Genre')
plt.title('Top 10 Genre for Movies')
plt.subplot(1,2,2)
test = data_grp[data_grp['type'] != 'Movie'].drop(['genre_comb', 'type'], axis=1
test.sum(axis = 0).sort_values(ascending = False).iloc[:10].plot(kind='bar')
plt.xlabel('Genre')
plt.title('Top 10 Genre for TV shows')
# Most of the shows are International Movies/TV shows
# Most of the movies belong to Dramas and comedies genre
```

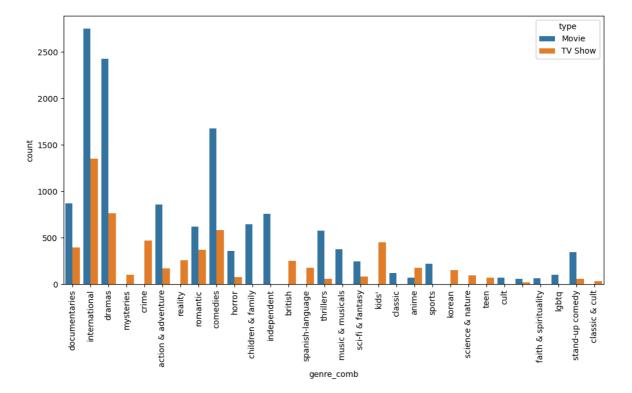
Out[53]: Text(0.5, 1.0, 'Top 10 Genre for TV shows')



Inference:

- Many of the shows on Netflix are categorized as International Movies/TV shows (non-US)
- The majority of movies on Netflix belong to the genres of Dramas and Comedies

```
In [54]: plt.figure(figsize=(12, 6))
    sns.countplot(data = data_exp_gen, x = 'genre_comb', hue = 'type')
    plt.xticks(rotation = 90);
```

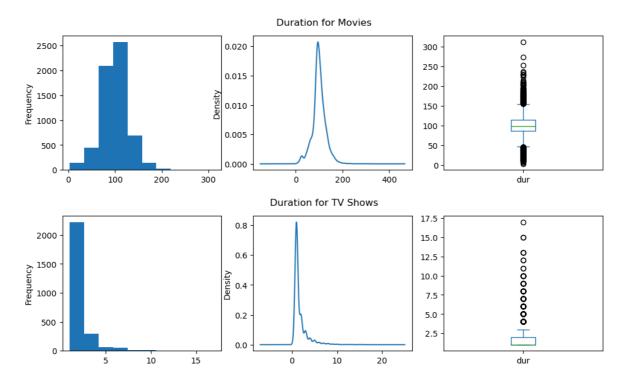


Inference:

- After combining the genre between Movies and TV shows
 - Both of them are categorized in Dramas and Comedies
- Horror, Scify & fantasy, anime movie are comparatively lesser (Recommendation 9)

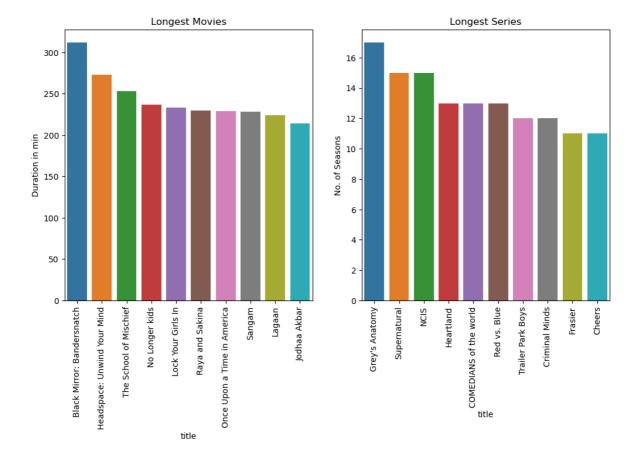
Duration

```
In [40]:
         plt.figure(figsize=(12,3)).suptitle('Duration for Movies')
         plt.subplot(1,3,1)
         dur_movie['dur'].plot(kind='hist')
         plt.subplot(1,3,2)
         dur_movie['dur'].plot(kind='kde')
         plt.subplot(1,3,3)
         dur_movie['dur'].plot(kind='box')
         #dur_show = data[data['type'] != 'Movie']['duration'].str.split()
         plt.figure(figsize=(12,3)).suptitle('Duration for TV Shows')
         plt.subplot(1,3,1)
         dur_show['dur'].plot(kind='hist')
         plt.subplot(1,3,2)
         dur_show['dur'].plot(kind='kde')
         plt.subplot(1,3,3)
         dur_show['dur'].plot(kind='box')
         plt.show()
```



```
In [41]: plt.figure(figsize=(12,6))
   plt.subplot(1,2,1)
   dur_movie = dur_movie.sort_values('dur', ascending = False)
   sns.barplot(data = dur_movie[:10], x = 'title', y = 'dur')
   plt.xticks(rotation = 90)
   plt.ylabel('Duration in min')
   plt.title('Longest Movies')

plt.subplot(1,2,2)
   dur_show = dur_show.sort_values('dur', ascending = False)
   sns.barplot(data = dur_show[:10], x = 'title', y = 'dur')
   plt.xticks(rotation = 90)
   plt.ylabel('No. of Seasons')
   plt.title('Longest Series')
   plt.show()
```



Release Year Vs Date Added Year

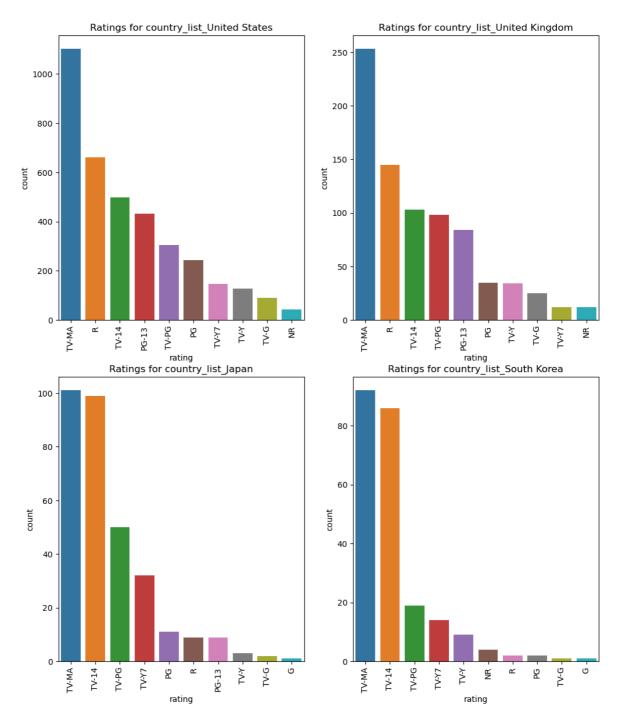
```
In [45]: temp_data = data[data['release_year'] > data['date_added_yr']]
In [105... temp_data[temp_data['type'] == 'Movie']['show_id'].nunique(), temp_data[temp_data[temp_data[temp_data]].
Out[105... (2, 12)
```

Inference: Only a few movies (2) and TV shows (12) have been released
directly on Netflix before hitting theaters (Recommendation 4)

Countries by rating

```
In [76]: data_country = data_country.merge(data[['show_id', 'rating']], on = 'show_id', h
    data_country_top = list(data_country_grp[].index[:4])

In [88]: plt.figure(figsize=(12,14))
    i = 1
    data_country_m = data_country#[data_country['type'] == 'Movie']
    for cntr in data_country_top:
        ratings_list = list(data_country_m[data_country_m[cntr] == True]['rating'].v
        plt.subplot(2,2,i)
        i = i + 1
        sns.countplot(data = data_country_m[data_country_m[cntr] == True], x = 'rati
        plt.title('Ratings for ' + cntr)
        plt.xticks(rotation = 90);
```



Business Insights and Recommendations

• Expand Classic Movie Selection:

■ Most movies on Netflix are released after 2010. Including more classic films will attract users from older age groups.

• Increase Animation Content:

■ The platform currently features movies predominantly for adult age groups. Adding more animated movies for kids will attract teenage users.

• Broaden Geographic Diversity:

■ Netflix hosts movies and TV shows from 123 countries. Adding content from the remaining ~70 countries will expand the

customer base in those regions.

• Increase Netflix Originals:

Only a few movies (2) have been released directly on Netflix before hitting theaters. Increasing this number could boost user engagement.

• Enhance January and February Lineup:

■ Fewer movies and TV shows are added in January and February.

Increasing content additions during these months will provide more options for users during this period.

• Include Influential South Indian Actors:

In South India, there is a high interest in movies and TV shows, yet none of the influential South Indian actors are listed among Netflix's top 10 actors. Including movies featuring these actors will increase the customer base in those regions.

Increase Indian TV Shows:

■ Despite India being the second-largest contributor of movies, it ranks seventh in terms of TV shows. Increasing the number of Indian TV shows will expand the customer base.

Add More Long-Running TV Shows:

Currently, only 17 TV shows on Netflix have more than 10 seasons. Increasing the number of shows with longer seasons can help extend user subscriptions.

• Diversify Genres:

■ The genres of movies and TV shows are not uniformly distributed. Horror and sci-fi fantasy movies are comparatively fewer. Increasing content in these genres will attract a broader audience to the platform.

Prepared By

Muthukumar G

In []: