

Project Title	Netflix Data: Cleaning, Analysis and Visualization
Tools	Python, ML, SQL, Excel
Domain	Data Analyst & Data scientist
Project Difficulties level	intermediate

Dataset : Dataset is available in the given link. You can download it at your convenience.

Click here to download data set

About Dataset

Netflix is a popular streaming service that offers a vast catalog of movies, TV shows, and original contents. This dataset is a cleaned version of the original version which can be found here. The data consist of contents added to Netflix from 2008 to 2021. The oldest content is as old as 1925 and the newest as 2021. This dataset will be cleaned with PostgreSQL and visualized with Tableau. The purpose of this dataset is to test my data cleaning and visualization skills. The cleaned data can be found below and the Tableau dashboard can be found here .

Data Cleaning

We are going to:

- 1. Treat the Nulls
- 2. Treat the duplicates
- 3. Populate missing rows
- 4. Drop unneeded columns
- 5. Split columns

Extra steps and more explanation on the process will be explained through the code comments

Example: You can get the basic idea how you can create a project from here

Netflix Data: Cleaning, Analysis, and Visualization (Beginner ML Project)

This project involves loading, cleaning, analyzing, and visualizing data from a Netflix dataset. We'll use Python libraries like Pandas, Matplotlib, and Seaborn to work through the project. The goal is to explore the dataset, derive insights, and prepare for potential machine learning tasks.

Step 1: Import Required Libraries

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from wordcloud import WordCloud
```

Step 2: Load the Dataset

```
Assume we have a dataset named netflix_titles.csv.
# Load the dataset
data = pd.read_csv('netflix_titles.csv')
# Display the first few rows of the dataset
print(data.head())
Step 3: Data Cleaning
Identify and handle missing data, correct data types, and drop duplicates.
# Check for missing values
print(data.isnull().sum())
# Drop duplicates if any
data.drop_duplicates(inplace=True)
# Drop rows with missing critical information
data.dropna(subset=['director',
                                       'cast'. 'country'].
inplace=True)
# Convert 'date_added' to datetime
data['date_added'] = pd.to_datetime(data['date_added'])
```

```
# Show data types to confirm changes
print(data.dtypes)
Step 4: Exploratory Data Analysis (EDA)
1. Content Type Distribution (Movies vs. TV Shows)
# Count the number of Movies and TV Shows
type_counts = data['type'].value_counts()
# Plot the distribution
plt.figure(figsize=(8, 6))
                                  y=type_counts.values,
sns.barplot(x=type_counts.index,
palette='Set2')
plt.title('Distribution of Content by Type')
plt.xlabel('Type')
plt.ylabel('Count')
plt.show()
2. Most Common Genres
# Split the 'listed_in' column and count genres
data['genres'] = data['listed_in'].apply(lambda x: x.split(',
'))
all_genres = sum(data['genres'], [])
genre_counts = pd.Series(all_genres).value_counts().head(10)
```

```
# Plot the most common genres
plt.figure(figsize=(10, 6))
sns.barplot(x=genre_counts.values, y=genre_counts.index,
palette='Set3')
plt.title('Most Common Genres on Netflix')
plt.xlabel('Count')
plt.ylabel('Genre')
plt.show()
3. Content Added Over Time
# Extract year and month from 'date_added'
data['year_added'] = data['date_added'].dt.year
data['month_added'] = data['date_added'].dt.month
# Plot content added over the years
plt.figure(figsize=(12, 6))
sns.countplot(x='year_added', data=data, palette='coolwarm')
plt.title('Content Added Over Time')
plt.xlabel('Year')
plt.ylabel('Count')
plt.xticks(rotation=45)
plt.show()
```

4. Top 10 Directors with the Most Titles

```
# Count titles by director
top_directors = data['director'].value_counts().head(10)
# Plot top directors
plt.figure(figsize=(10, 6))
sns.barplot(x=top_directors.values, y=top_directors.index,
palette='Blues_d')
plt.title('Top 10 Directors with the Most Titles')
plt.xlabel('Number of Titles')
plt.ylabel('Director')
plt.show()
5. Word Cloud of Movie Titles
# Generate word cloud
movie_titles = data[data['type'] == 'Movie']['title']
                        WordCloud(width=800, height=400,
wordcloud
background_color='black').generate(' '.join(movie_titles))
# Plot word cloud
plt.figure(figsize=(10, 6))
plt.imshow(wordcloud, interpolation='bilinear')
plt.axis('off')
plt.show()
```

Step 5: Conclusion and Insights

In this project, we:

- 1. **Cleaned the data** by handling missing values, removing duplicates, and converting data types.
- 2. **Explored the data** through various visualizations such as bar plots and word clouds.
- 3. **Analyzed content trends** over time, identified popular genres, and highlighted top directors.

Step 6: Next Steps

- 1. **Feature Engineering**: Create new features, such as counting the number of genres per movie or extracting the duration in minutes.
- 2. **Machine Learning**: Use the cleaned and processed data to build models for recommendations or trend predictions.
- 3. **Advanced Visualization**: Use interactive plots or dashboards for more detailed analysis.

This project is a foundational exercise that introduces essential data analysis techniques, paving the way for more advanced projects.

Sample code

import pandas as pd
import numpy as np

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

Importing data from csv and getting info about data.

In [2]:

data=pd.read_csv("/kaggle/input/netflix-data-cleaning-analysisand-visualization/netflix1.csv") data.head()

Out[2]:

	sho w_i d	typ e	title	direct or	cou ntry	date_ added	releas e_year	rati ng	dura tion	listed_in
C) s1	Mo vie	Dick Johnson Is Dead	Kirste n Johns on	Unit ed Stat es	9/25/2 021	2020	P G- 13	90 min	Documentari es
	s3	TV Sh ow	Ganglan ds	Julien Lecle rcq	Fran ce	9/24/2 021	2021	TV -M A	1 Sea son	Crime TV Shows, International TV Shows, TV Act

2	s6	TV Sh ow	Midnight Mass	Mike Flana gan	Unit ed Stat es	9/24/2 021	2021	TV -M A	1 Sea son	TV Dramas, TV Horror, TV Mysteries
3	s14	Mo vie	Confessi ons of an Invisible Girl	Bruno Garot ti	Braz il	9/22/2 021	2021	TV -P G	91 min	Children & Family Movies, Comedies
4	s8	Mo vie	Sankofa	Haile Geri ma	Unit ed Stat es	9/24/2 021	1993	TV -M A	125 min	Dramas, Independent Movies, International Movies

In [3]:

data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8790 entries, 0 to 8789

Data columns (total 10 columns):

#	Column	Non-Null Count	Dtype
0	show_id	8790 non-null	object
1	type	8790 non-null	object
2	title	8790 non-null	object
3	director	8790 non-null	object
4	country	8790 non-null	object

```
5
     date_added 8790 non-null
                                        object
     release_year 8790 non-null
                                        int64
 6
     rating 8790 non-null duration 8790 non-null listed_in 8790 non-null
 7
                                        object
     rating
 8
                                        object
 9
                                       object
dtypes: int64(1), object(9)
memory usage: 686.8+ KB
                                                                 In [4]:
data.shape
                                                                 Out[4]:
(8790, 10)
                                                                 In [5]:
data=data.drop_duplicates()
Content distribution on Netflix.
                                                                 In [6]:
data['type'].value_counts()
                                                                 Out[6]:
type
Movie
            6126
```

```
TV Show 2664

Name: count, dtype: int64

In [7]:

freq=data['type'].value_counts()

fig, axes=plt.subplots(1,2, figsize=(8, 4))

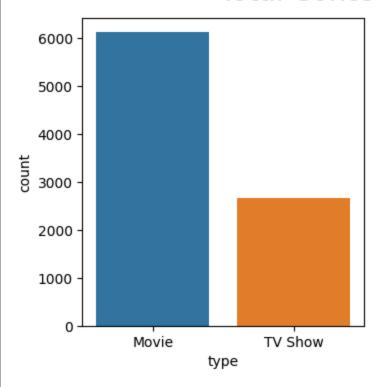
sns.countplot(data, x=data['type'], ax=axes[0])
plt.pie(freq, labels=['Movie', 'TV Show'], autopct='%.0f%%')

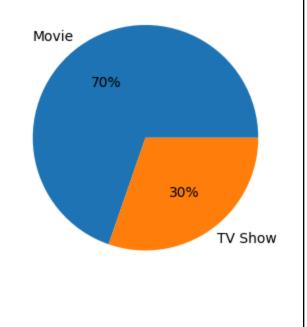
plt.suptitle('Total Content on Netflix', fontsize=20)

Out[7]:

Text(0.5, 0.98, 'Total Content on Netflix')
```

Total Content on Netflix





In [8]:

data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8790 entries, 0 to 8789
Data columns (total 10 columns):

#	Column	Non-Null Count	Dtype
0	show_id	8790 non-null	object
1	type	8790 non-null	object
2	title	8790 non-null	object
3	director	8790 non-null	object
4	country	8790 non-null	object
5	date_added	8790 non-null	object
6	release_year	8790 non-null	int64
7	rating	8790 non-null	object

```
duration 8790 non-null
                                   object
 8
     listed_in 8790 non-null
                                  object
 9
dtypes: int64(1), object(9)
memory usage: 686.8+ KB
Visual representation of rating frequency of movies and TV Shows on Netflix.
                                                          In [9]:
data['rating'].value_counts()
                                                          Out[9]:
rating
TV-MA
           3205
TV-14
            2157
            861
TV-PG
R
            799
PG-13
            490
TV-Y7
             333
TV-Y
             306
PG
             287
TV-G
             220
             79
NR
G
             41
TV-Y7-FV
              6
NC-17
               3
               3
UR
Name: count, dtype: int64
                                                         In [10]:
ratings=data['rating'].value_counts().reset_index().sort_values
(by='count', ascending=False)
```

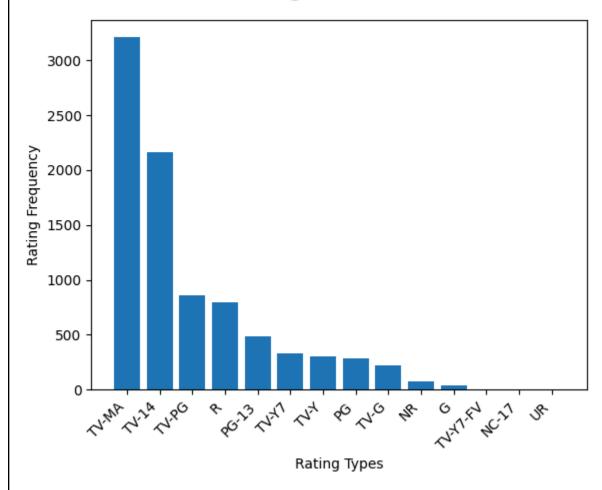
```
plt.bar(ratings['rating'], ratings['count'])
plt.xticks(rotation=45, ha='right')
plt.xlabel("Rating Types")
plt.ylabel("Rating Frequency")

plt.suptitle('Rating on Netflix', fontsize=20)

Out[10]:
```

Text(0.5, 0.98, 'Rating on Netflix')

Rating on Netflix

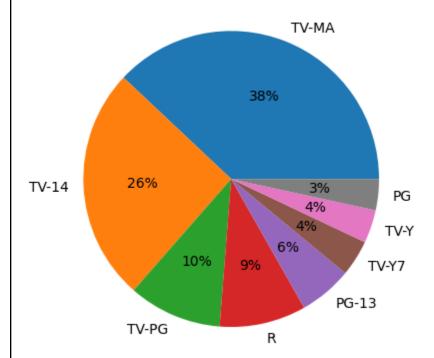


```
In [11]:
plt.pie(ratings['count'][:8], labels=ratings['rating'][:8],
autopct='%.0f%%')
plt.suptitle('Rating on Netflix', fontsize=20)
```

Out[11]:

Text(0.5, 0.98, 'Rating on Netflix')

Rating on Netflix



Converting date_added column to datetime.

In [12]:

lets convert column date_added to datetime.
data['date_added']=pd.to_datetime(data['date_added'])

In [13]:

data.describe()

Out[13]:

	date_added	release _year
cou	8790	8790.00 0000
me an	2019-05-17 21:44:01.638225408	2014.18 3163
min	2008-01-01 00:00:00	1925.00 0000
25 %	2018-04-06 00:00:00	2013.00 0000

50 %	2019-07-03 00:00:00	2017.00 0000
75 %	2020-08-19 18:00:00	2019.00 0000
ma x	2021-09-25 00:00:00	2021.00 0000
std	NaN	8.82546 6

In [14]:

data['country'].value_counts()

Out[14]:

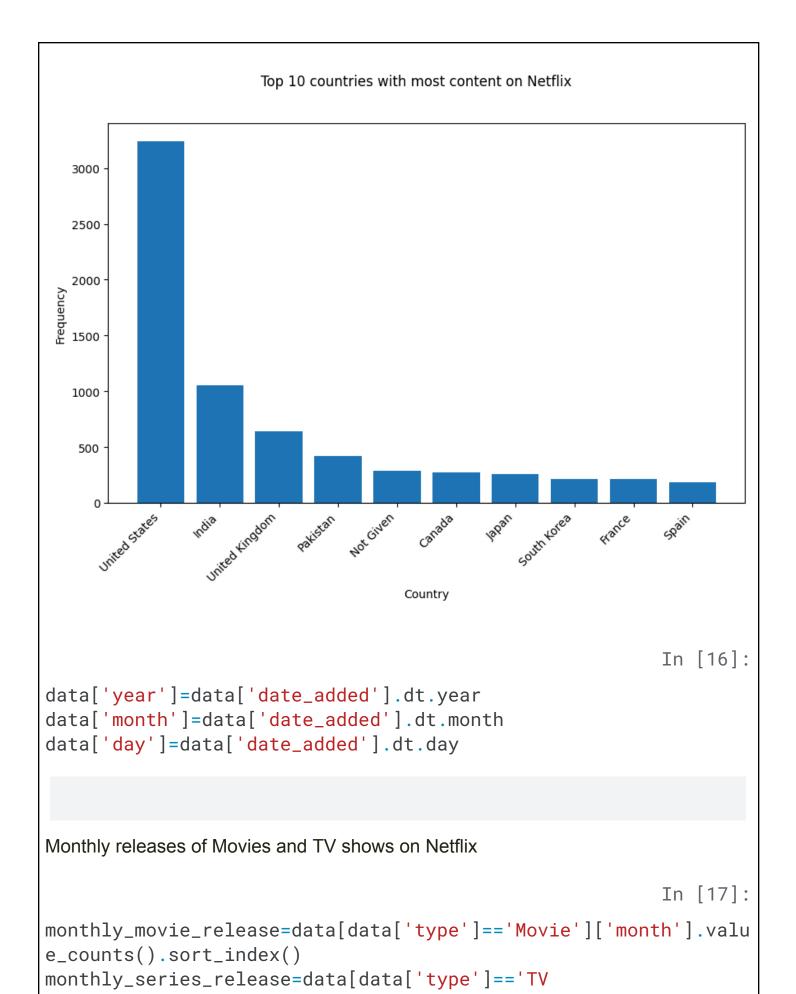
country

Zimbabwe

United States 3240 India 1057 United Kingdom 638 Pakistan 421 Not Given 287 Iran 1 West Germany 1 Greece

1

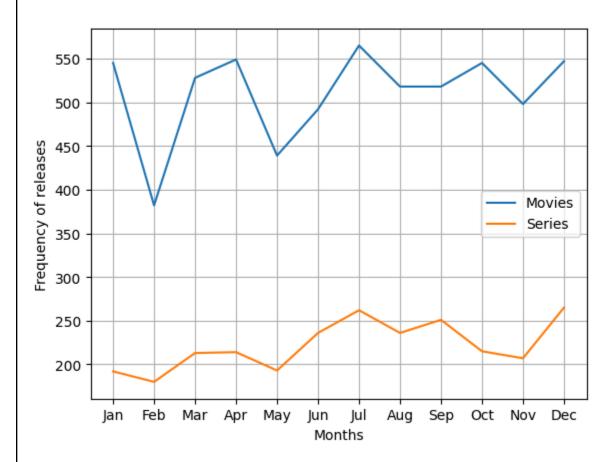
1



```
Show']['month'].value_counts().sort_index()

plt.plot(monthly_movie_release.index,
monthly_movie_release.values, label='Movies')
plt.plot(monthly_series_release.index,
monthly_series_release.values, label='Series')
plt.xlabel("Months")
plt.ylabel("Frequency of releases")
plt.ylabel("Frequency of releases")
plt.xticks(range(1, 13), ['Jan', 'Feb', 'Mar', 'Apr', 'May',
'Jun', 'Jul', 'Aug', 'Sep', 'Oct', 'Nov', 'Dec'])
plt.legend()
plt.grid(True)
plt.suptitle("Monthly releases of Movies and TV shows on
Netflix")
plt.show()
```





Yearly releases of Movies and TV Shows on Netflix

```
In [18]:
yearly_movie_releases=data[data['type']=='Movie']['year'].value
_counts().sort_index()
yearly_series_releases=data[data['type']=='TV
Show']['year'].value_counts().sort_index()

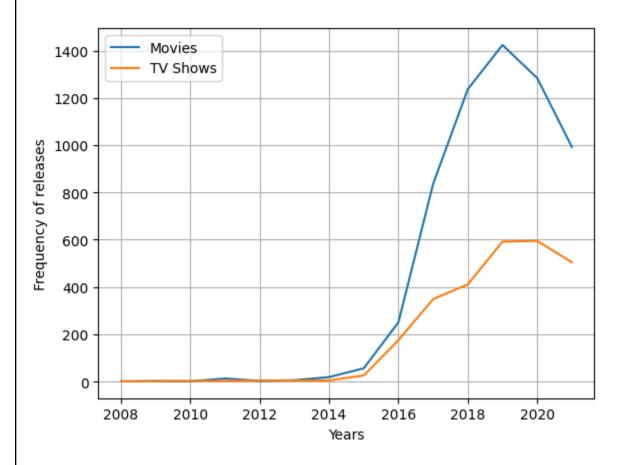
plt.plot(yearly_movie_releases.index,
yearly_movie_releases.values, label='Movies')
plt.plot(yearly_series_releases.index,
yearly_series_releases.values, label='TV Shows')
plt.xlabel("Years")
plt.ylabel("Frequency of releases")
plt.grid(True)
plt.suptitle("Yearly releases of Movies and TV Shows on
```

```
Netflix")
plt.legend()
```

Out[18]:

<matplotlib.legend.Legend at 0x7a14cb8327a0>

Yearly releases of Movies and TV Shows on Netflix



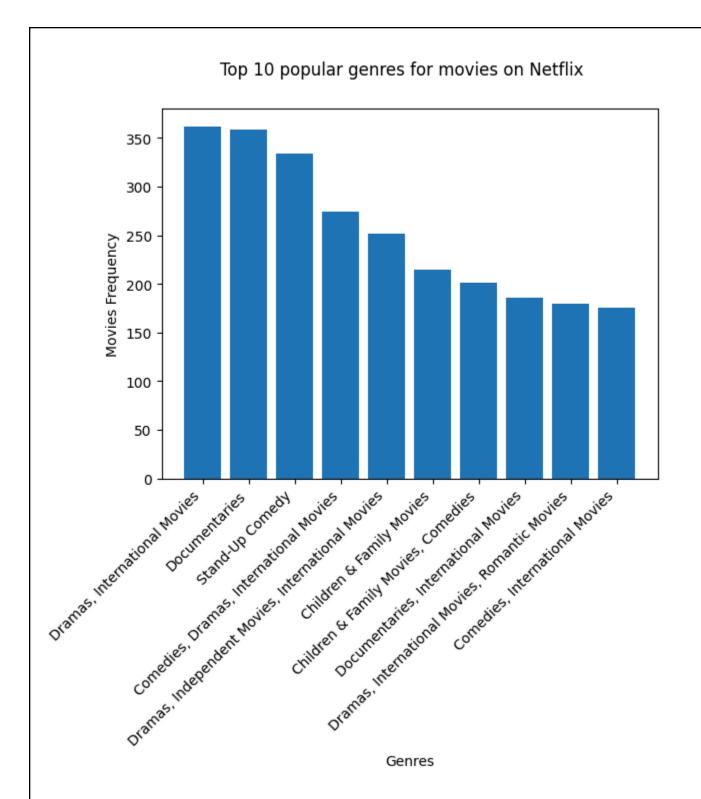
Top 10 popular movie genres

```
In [19]:
by("listed
```

```
popular_movie_genre=data[data['type']=='Movie'].groupby("listed
_in").size().sort_values(ascending=False)[:10]
popular_series_genre=data[data['type']=='TV
```

```
Show'].groupby("listed_in").size().sort_values(ascending=False)
[:10]

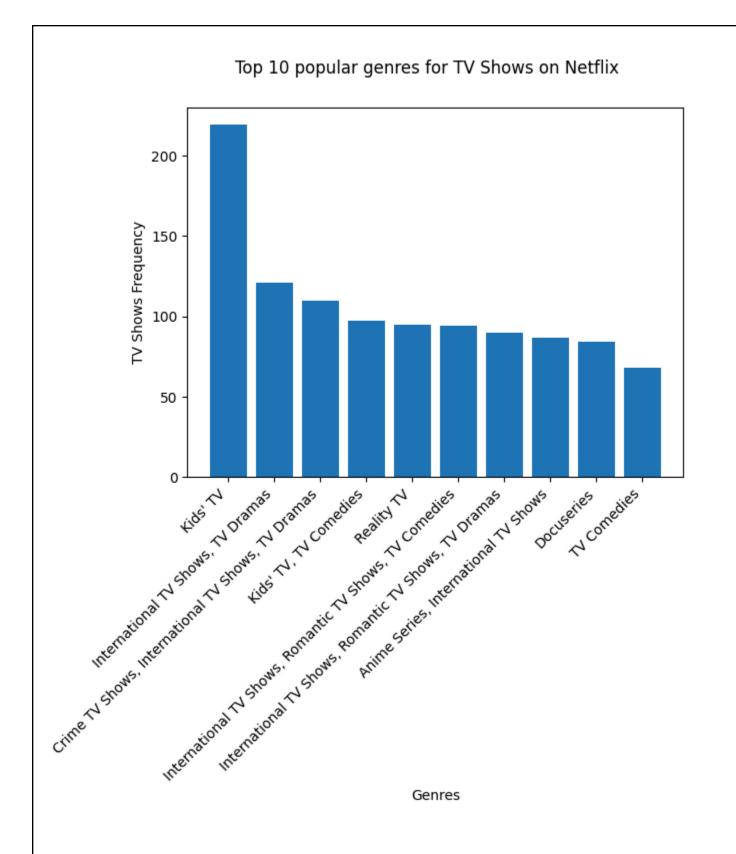
plt.bar(popular_movie_genre.index, popular_movie_genre.values)
plt.xticks(rotation=45, ha='right')
plt.xlabel("Genres")
plt.ylabel("Movies Frequency")
plt.suptitle("Top 10 popular genres for movies on Netflix")
plt.show()
```



Top 10 TV Shows genres

```
In [20]:
plt.bar(popular_series_genre.index,
popular_series_genre.values)
plt.xticks(rotation=45, ha='right')
```

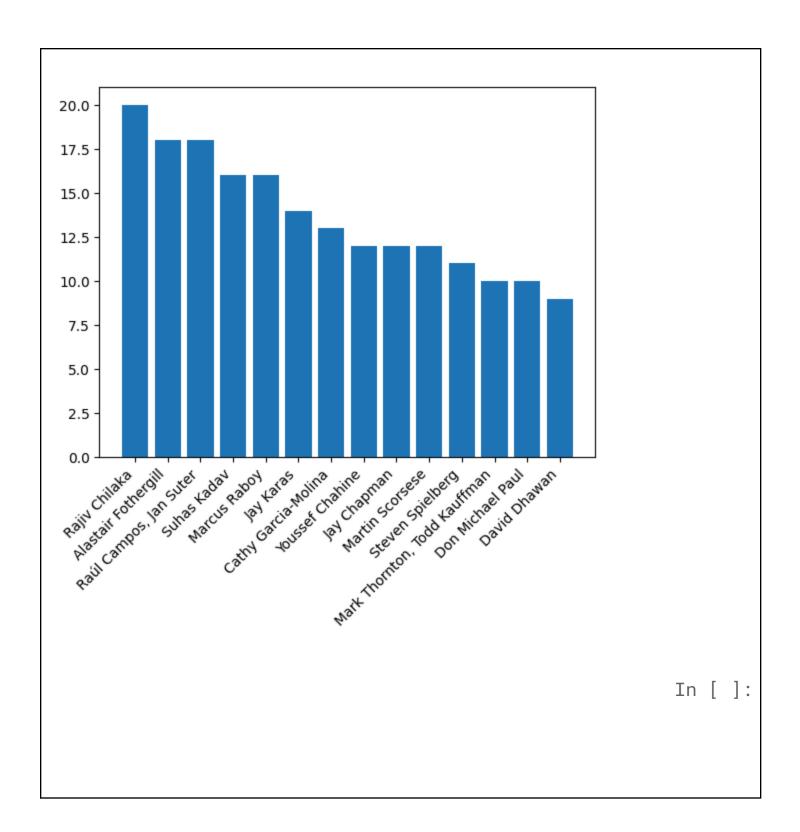
```
plt.xlabel("Genres")
plt.ylabel("TV Shows Frequency")
plt.suptitle("Top 10 popular genres for TV Shows on Netflix")
plt.show()
```



Top 15 directors across Netflix with hoigh frequency of movies and shows.

In [21]:
directors=data['director'].value_counts().reset_index().sort_va

```
lues(by='count', ascending=False)[1:15]
plt.bar(directors['director'], directors['count'])
plt.xticks(rotation=45, ha='right')
                                                             Out[21]:
([0, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 11, 12, 13],
 [Text(0, 0, 'Rajiv Chilaka'),
  Text(1, 0, 'Alastair Fothergill'),
  Text(2, 0, 'Raúl Campos, Jan Suter'),
  Text(3, 0, 'Suhas Kadav'),
  Text(4, 0, 'Marcus Raboy'),
Text(5, 0, 'Jay Karas'),
  Text(6, 0, 'Cathy Garcia-Molina'),
  Text(7, 0, 'Youssef Chahine'),
  Text(8, 0, 'Jay Chapman'),
  Text(9, 0, 'Martin Scorsese'),
  Text(10, 0, 'Steven Spielberg'),
  Text(11, 0, 'Mark Thornton, Todd Kauffman'),
  Text(12, 0, 'Don Michael Paul'),
  Text(13, 0, 'David Dhawan')])
```



Reference link