```
import numpy as np
In [2]:
        import pandas as pd
        import matplotlib.pyplot as plt
        import seaborn as sns
In [3]: !wget "https://d2beiqkhq929f0.cloudfront.net/public_assets/assets/000/000/940/origi
        --2024-07-24 05:32:09-- https://d2beiqkhq929f0.cloudfront.net/public_assets/asset
        s/000/000/940/original/netflix.csv
        Resolving d2beiqkhq929f0.cloudfront.net (d2beiqkhq929f0.cloudfront.net)... 18.160.
        146.106, 18.160.146.28, 18.160.146.45, ...
        Connecting to d2beiqkhq929f0.cloudfront.net (d2beiqkhq929f0.cloudfront.net) | 18.16
        0.146.106|:443... connected.
        HTTP request sent, awaiting response... 200 OK
        Length: 3399671 (3.2M) [text/plain]
        Saving to: 'netflix.csv'
        netflix.csv
                            100%[========>]
                                                       3.24M --.-KB/s
                                                                           in 0.09s
        2024-07-24 05:32:09 (37.5 MB/s) - 'netflix.csv' saved [3399671/3399671]
In [4]: df=pd.read_csv("netflix.csv")
        df
```

Out[4]: show id title director cast country date_added release_year rating type Dick Kirsten United September 0 Johnson Is NaN 2020 PG-13 s1 Movie Johnson 25, 2021 States Dead Ama Qamata, Khosi TV Blood & South September TV-1 2021 s2 NaN Ngema, Show Water Africa 24, 2021 MA Gail Mabalane, Thaban... Sami Bouajila, Tracy TV-Julien September TV 2 s3 2021 Ganglands Gotoas, NaN Show 24, 2021 MA Leclercq Samuel Jouy, Nabi... **Jailbirds** TV TV-September 3 New NaN NaN 2021 s4 NaN Show 24, 2021 MA Orleans Mayur More, Jitendra TV Kota September TV-2021 s5 NaN Kumar, India 24, 2021 Show Factory MA Ranjan Raj, Alam K... Mark Ruffalo, Jake David United November 8802 s8803 Zodiac Gyllenhaal, 2007 R Movie Fincher States 20, 2019 Robert Downey J... TV Zombie 8803 s8804 NaN NaN NaN July 1, 2019 2018 TV-Y7 Show Dumb Jesse Eisenberg, Ruben Woody United November 8804 s8805 Movie Zombieland 2009 R Fleischer Harrelson, States 1, 2019 Emma Stone, ... 8805 Tim Allen, PG s8806 Movie Zoom Peter United January 11, 2006 Courteney States 2020 Hewitt Cox, Chevy

	show_id	type	title	director	cast	country	date_added	release_year	rating		
					Chase, Kate Ma						
8806	s8807	Movie	Zubaan	Mozez Singh	Vicky Kaushal, Sarah- Jane Dias, Raaghav Chanan	India	March 2, 2019	2015	TV-14		
8807 ı	8807 rows × 12 columns										

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

```
Out[5]: release_year

count 8807.000000

mean 2014.180198

std 8.819312

min 1925.000000

25% 2013.000000

50% 2017.000000

75% 2019.000000

max 2021.000000
```

In [6]: df.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)	

memory usage: 825.8+ KB

```
In [7]: df.isna().sum()
```

```
show_id
                      0
Out[7]:
                     0
      type
      title
                     0
                 2634
      director
      cast
                   825
      country
                   831
      date_added
                   10
      release_year
                     4
      rating
      duration
                     3
      listed_in
      description
      dtype: int64
```

In [8]: #As the Null values are very few dropping them for better analysis
df.dropna(subset= ["date_added", "rating", "duration"], inplace=True)

```
In [9]: df1=df.loc[:,["title","director","cast","country","listed_in"]]
    df1.head()
```

listed_in	country	cast	director	title		Out[9]:	
Documentaries	United States	NaN	Kirsten Johnson	Dick Johnson Is Dead	0		
International TV Shows, TV Dramas, TV Mysteries	South Africa	Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban	NaN	Blood & Water	1		
Crime TV Shows, International TV Shows, TV Act	NaN	Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabi	Julien Leclercq	Ganglands	2		
Docuseries, Reality TV	NaN	NaN	NaN	Jailbirds New Orleans	3		
International TV Shows, Romantic TV Shows, TV	India	Mayur More, Jitendra Kumar, Ranjan Raj, Alam K	NaN	Kota Factory	4		

What does 'good' look like

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

Non-Graphical analysis

df1=df1.explode("country")
df1=df1.explode("listed_in")

In [13]: df1

Out[13]:

	title	director	cast	country	listed_in
0	Dick Johnson Is Dead	Kirsten Johnson	NaN	United States	Documentaries
1	Blood & Water	NaN	Ama Qamata	South Africa	International TV Shows
1	Blood & Water	NaN	Ama Qamata	South Africa	TV Dramas
1	Blood & Water	NaN	Ama Qamata	South Africa	TV Mysteries
1	Blood & Water	NaN	Khosi Ngema	South Africa	International TV Shows
•••					
8806	Zubaan	Mozez Singh	Anita Shabdish	India	International Movies
8806	Zubaan	Mozez Singh	Anita Shabdish	India	Music & Musicals
8806	Zubaan	Mozez Singh	Chittaranjan Tripathy	India	Dramas
8806	Zubaan	Mozez Singh	Chittaranjan Tripathy	India	International Movies
8806	Zubaan	Mozez Singh	Chittaranjan Tripathy	India	Music & Musicals

201837 rows × 5 columns

```
In [14]: df.drop(columns=["director","cast","country","listed_in"],inplace=True)
In [15]: df_final=df.merge(df1,on="title")
    df_final.head()
```

Out[15]:		show_id	type	title	date_added	release_year	rating	duration	description	director	
	0	s1	Movie	Dick Johnson Is Dead	September 25, 2021	2020	PG-13	90 min	As her father nears the end of his life, filmm	Kirsten Johnson	
	1	s2	TV Show	Blood & Water	September 24, 2021	2021	TV- MA	2 Seasons	After crossing paths at a party, a Cape Town t	NaN	Qar
	2	s2	TV Show	Blood & Water	September 24, 2021	2021	TV- MA	2 Seasons	After crossing paths at a party, a Cape Town t	NaN	Qar
	3	s2	TV Show	Blood & Water	September 24, 2021	2021	TV- MA	2 Seasons	After crossing paths at a party, a Cape Town t	NaN	Qar
	4	s2	TV Show	Blood & Water	September 24, 2021	2021	TV- MA	2 Seasons	After crossing paths at a party, a Cape Town t	NaN	K Ng
4											•
In [16]:			_		f date_adde	ed column	['date	added!]	ennons-lo	pence!)	

```
In [16]: # converting data type of date_added column
    df_final['date_added']=pd.to_datetime(df_final['date_added'], errors='coerce')
    df_final['month_added'] = df_final['date_added'].dt.month
    df_final['year_added'] = df_final['date_added'].dt.year
    df_final['week_added'] = df_final['date_added'].dt.isocalendar().week
    df_final
```

Out[16]:		show_id	type	title	date_added	release_year	rating	duration	description	directo
	0	s1	Movie	Dick Johnson Is Dead	2021-09-25	2020	PG-13	90 min	As her father nears the end of his life, filmm	Kirster Johnsor
	1	s2	TV Show	Blood & Water	2021-09-24	2021	TV- MA	2 Seasons	After crossing paths at a party, a Cape Town t	NaN
	2	s2	TV Show	Blood & Water	2021-09-24	2021	TV- MA	2 Seasons	After crossing paths at a party, a Cape Town t	NaN
	3	s2	TV Show	Blood & Water	2021-09-24	2021	TV- MA	2 Seasons	After crossing paths at a party, a Cape Town t	NaN
	4	s2	TV Show	Blood & Water	2021-09-24	2021	TV- MA	2 Seasons	After crossing paths at a party, a Cape Town t	NaN
	201832	s8807	Movie	Zubaan	2019-03-02	2015	TV-14	111 min	A scrappy but poor boy worms his way into a ty	Moze: Singł
	201833	s8807	Movie	Zubaan	2019-03-02	2015	TV-14	111 min	A scrappy but poor boy worms his way into a ty	Moze: Singl
	201834	s8807	Movie	Zubaan	2019-03-02	2015	TV-14	111 min	A scrappy but poor boy worms his way into a ty	Moze: Singł
	201835	s8807	Movie	Zubaan	2019-03-02	2015	TV-14	111 min	A scrappy but poor boy worms his way into a ty	Moze: Singł
	201836	s8807	Movie	Zubaan	2019-03-02	2015	TV-14	111 min	A scrappy but poor boy worms his way into a ty	Moze: Singł

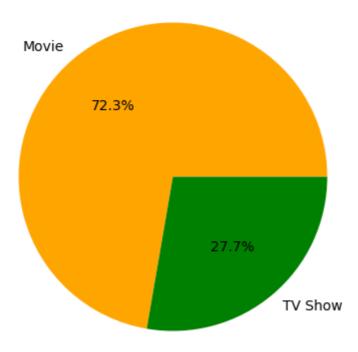
201837 rows × 15 columns

In	[17]:	df_final.	fillna(value={"	director":	'Unknown Dir	rector"	,"country	":"Unknown	Country"	,"(
In	[18]:	df_final.	head()								
Ou ⁻	t[18]:	show_id	type	title	date_added	release_year	rating	duration	description	director	
		0 s1	Movie	Dick Johnson Is Dead	2021-09-25	2020	PG-13	90 min	As her father nears the end of his life, filmm	Kirsten Johnson	Ur
		1 s2	TV Show	Blood & Water	2021-09-24	2021	TV- MA	2 Seasons	After crossing paths at a party, a Cape Town t	Unknown Director	C
		2 s2	TV Show	Blood & Water	2021-09-24	2021	TV- MA	2 Seasons	After crossing paths at a party, a Cape Town t	Unknown Director	C
		3 s2	TV Show	Blood & Water	2021-09-24	2021	TV- MA	2 Seasons	After crossing paths at a party, a Cape Town t	Unknown Director	C
		4 s2	TV Show	Blood & Water	2021-09-24	2021	TV- MA	2 Seasons	After crossing paths at a party, a Cape Town t	Unknown Director	
4											•

Graphical analysis

```
In [19]: mylabel = df_final['type'].value_counts()
label = mylabel.index
sizes = mylabel.values
plt.figure(figsize=(10,5))
plt.pie(sizes, labels=label, autopct='%1.1f%%', colors=['orange', 'green'])
plt.title('Movies vs TV Shows')
plt.show()
```

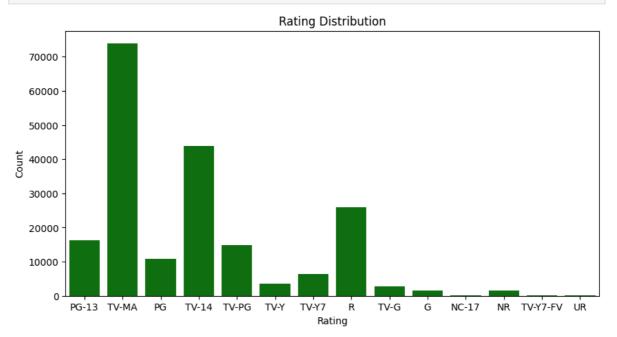
Movies vs TV Shows



Insight -

- Movies make up 72.3% of the content, indicating a strong focus on feature films over TV shows.
- TV shows constitute only 27.7% of the content, highlighting an opportunity for growth in serialized content.

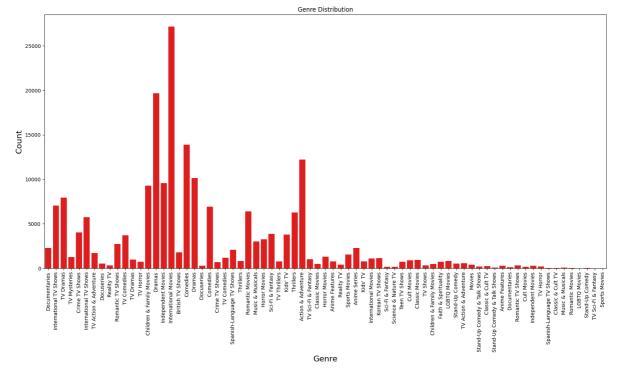
```
In [20]: #Rating count distribution
  plt.figure(figsize=(10,5))
  sns.countplot(data=df_final,x="rating",color="green")
  plt.xlabel("Rating")
  plt.ylabel("Count")
  plt.title("Rating Distribution")
  plt.show()
```



Insight -

- TV-MA content has the highest count, indicating strong viewer preference for maturerated content.
- Ratings like TV-Y, TV-Y7, and G have significantly lower counts, suggesting less emphasis on family-friendly content.

```
In [21]: #Different genres using countplot
    fig = plt.figure(figsize = (20,20))
    plt.subplot(2,1,1)
    plt.xticks(rotation = 90)
    sns.countplot(data = df_final, x = 'listed_in',color="red")
    plt.xlabel("Genre",fontsize=16)
    plt.ylabel("Count",fontsize=16)
    plt.title("Genre Distribution")
    plt.show()
```



Insight -

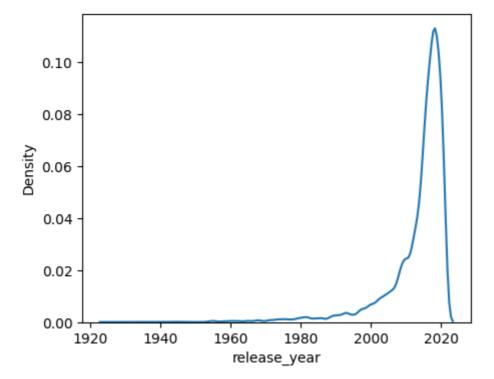
- Dramas, international TV shows, and children & family movies are among the most popular genres.
- Genres like science & nature TV, teen TV shows, and anime movies have lower counts, indicating potential areas for content expansion

```
In [22]: # getting top 10 directors
# dropping of the unknown rating to get better visualisation, and for that creating
df_new= df_final.loc[df_final['director']!='Unknown Director']
df_new.groupby('director')['title'].nunique().sort_values(ascending = False)[0:11]
```

director title 0 Rajiv Chilaka 22 1 Jan Suter 18 2 Raúl Campos 18 3 Suhas Kadav 16 4 Marcus Raboy 16 5 Jay Karas 15 Cathy Garcia-Molina 13 7 Jay Chapman 12 8 12 Martin Scorsese Youssef Chahine 12 10 Steven Spielberg 11

Out[22]:

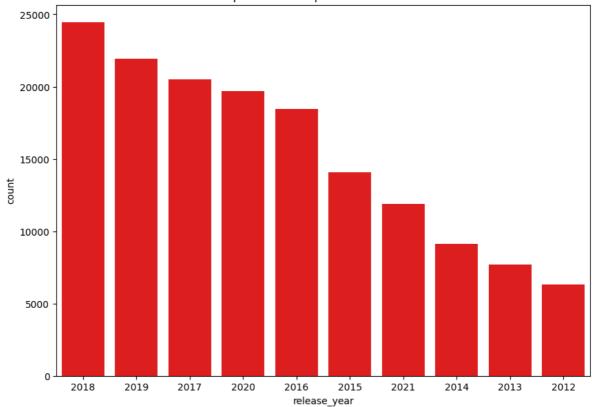
```
In [29]: # Movies and TV shows released trend analysis
  plt.figure(figsize=(5,4))
  sns.kdeplot(df_final['release_year'])
  plt.show()
```



```
In [24]: #Count of Movies,TV Shows releases,which year has most releases on netflix
plt.figure(figsize=(10,7))
sns.countplot(data=df_final, x='release_year',order=df_final['release_year'].value_
plt.title('Top 10 Most Frequent Release Years')
```

Out[24]: Text(0.5, 1.0, 'Top 10 Most Frequent Release Years')





```
In [25]: df_final["country"]=df_final["country"].str.strip()
```

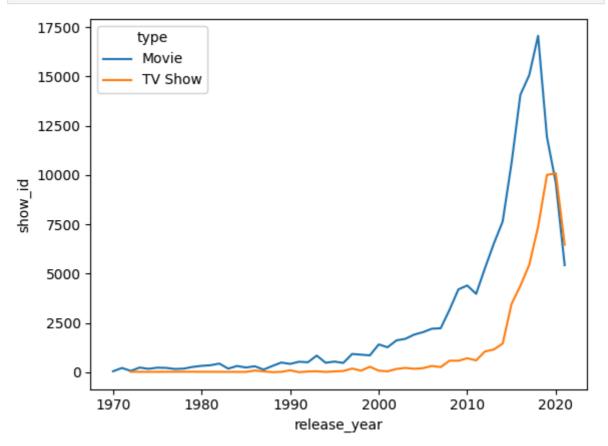
2. Comparison of TV Shows vs. Movies.

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

```
#Movies count by Various Countries
In [26]:
          df_co=df_final.loc[df_final["country"]!="Unknown Country"]
          df_co[(df_co['type'] == 'Movie')]['country'].value_counts().head(11)
         country
Out[26]:
         United States
                            45814
         India
                            21411
         United Kingdom
                             8580
         France
                             6607
         Canada
                             5738
         Japan
                             3525
         Spain
                             3469
         Germany
                             3427
         China
                             2377
         Nigeria
                             2236
         Hong Kong
                             2205
         Name: count, dtype: int64
         #TV Shows count by Various Countries
In [27]:
          df_co=df_final.loc[df_final["country"]!="Unknown Country"]
          df_co[(df_co['type'] == 'TV Show')]['country'].value_counts().head(11)
```

```
country
Out[27]:
          United States
                             13449
          Japan
                              5074
          United Kingdom
                              4358
          South Korea
                              3754
          Canada
                              2177
          Mexico
                              2018
          Spain
                              1846
          Taiwan
                              1719
          France
                              1647
          India
                              1403
          Colombia
                              1284
          Name: count, dtype: int64
```

```
In [28]: #Movies Vs TV Shows release trend analysis
s = df_final.groupby(['release_year','type'])['show_id'].count()
s = s.reset_index()
s = s[s['release_year'] >= 1970]
sns.lineplot(data = s, x = 'release_year', y = 'show_id',hue='type')
plt.show()
```



Insight -

• The graph shows a sharp increase in the release of both movies and TV shows from 2000 onwards, peaking around 2018.

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

a. Find which is the best week to release the Tv-show or the movie. Do the analysis separately for

Tv-shows and Movies

```
best_week = df_final.groupby('week_added')['type'].value_counts()
         best week
        week_added type
Out[]:
                    Movie
                               8456
                    TV Show
                               938
                    Movie
                                1618
                    TV Show
                                585
                    Movie
                                2031
                    TV Show
        51
                               1173
                    Movie
                               1840
                    TV Show
                               1111
        53
                    Movie
                               1413
                    TV Show
                               1038
        Name: count, Length: 106, dtype: int64
```

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

```
best_month = df_final.groupby('type')['month_added'].value_counts()
         best month
                  month_added
         type
Out[]:
                  7.0
         Movie
                                  15075
                  1.0
                                  13945
                  10.0
                                  13541
                  9.0
                                  13219
                  12.0
                                  12768
                  4.0
                                  12537
                  8.0
                                  11923
                  6.0
                                  11616
                  3.0
                                  11500
                  11.0
                                  11065
                  5.0
                                   9579
                  2.0
                                   9137
         TV Show 12.0
                                   5297
                  7.0
                                   5129
                  8.0
                                   5029
                  6.0
                                   4959
                  9.0
                                   4818
                  4.0
                                   4460
                  11.0
                                   4428
                  3.0
                                   4201
                  10.0
                                   4199
                  5.0
                                   4111
                  1.0
                                   3941
                  2.0
                                   3772
         Name: count, dtype: int64
```

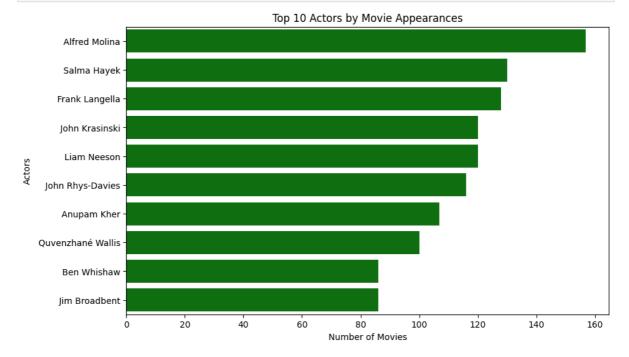
4. Analysis of actors/directors of different types of shows/movies.

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

```
In [31]: top_actors = df_final[df_final['type'] == 'Movie']['cast'].value_counts()[1:11].res
top_actors
```

Out[31]:		cast	count
	0	Alfred Molina	157
	1	Salma Hayek	130
	2	Frank Langella	128
	3	John Krasinski	120
	4	Liam Neeson	120
	5	John Rhys-Davies	116
	6	Anupam Kher	107
	7	Quvenzhané Wallis	100
	8	Ben Whishaw	86
	9	Jim Broadbent	86

```
In [33]: plt.figure(figsize=(10, 6))
    sns.barplot(x='count', y='cast', data=top_actors, color='green')
    plt.title('Top 10 Actors by Movie Appearances')
    plt.xlabel('Number of Movies')
    plt.ylabel('Actors')
    plt.show()
```

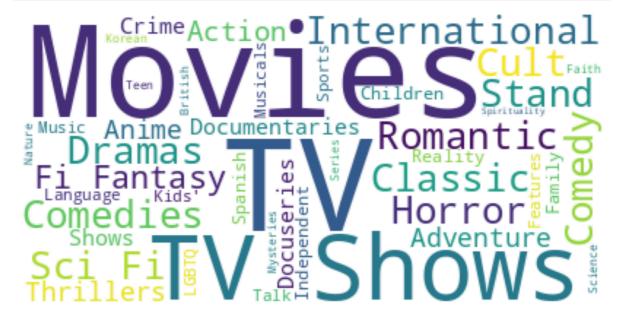


```
In [34]: df_new= df_final.loc[df_final['director']!='Unknown Director']
df_new.groupby('director')['title'].nunique().sort_values(ascending = False)[0:11]
```

Out[34]:		director	title
	0	Rajiv Chilaka	22
	1	Jan Suter	18
	2	Raúl Campos	18
	3	Suhas Kadav	16
	4	Marcus Raboy	16
	5	Jay Karas	15
	6	Cathy Garcia-Molina	13
	7	Jay Chapman	12
	8	Martin Scorsese	12
	9	Youssef Chahine	12
	10	Steven Spielberg	11

5. Which genre movies are more popular or produced more

```
In [ ]: a = df_final['listed_in'].unique()
    from wordcloud import WordCloud
    word = WordCloud(background_color = 'white').generate(' '.join(a))
    plt.figure(figsize = (10,8))
    plt.imshow(word)
    plt.axis("off")
    plt.show()
```



Insight -

The word cloud shows "Movies," "TV Shows," and "International" as dominant interests.
 For a balance of genres, try "Sci-Fi Thrillers" for excitement and "Comedies" for a lighter mood.

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

In	[41]:	df_f df_f	<pre># calculating the difference of release year and date added df_final['release_date'] = pd.to_datetime(df_final['release_year'].astype(str) + '- df_final['date_added'] = pd.to_datetime(df_final['date_added']) df_final['days_to_netflix'] = (df_final['date_added'] - df_final['release_date']).c</pre>										
In	[43]:	df_f	inal.h	ead()									
Ou	t[43]:	sl	how_id	type	title	date_added	release_year	rating	duration	description directo			
		0	s1	Movie	Dick Johnson Is Dead	2021-09-25	2020	PG-13	90 min	As her father nears the end of his life, filmm	Kirsten Johnson	Ur	
		1	s2	TV Show	Blood & Water	2021-09-24	2021	TV- MA	2 Seasons	After crossing paths at a party, a Cape Town t	Unknown Director	C	
		2	s2	TV Show	Blood & Water	2021-09-24	2021	TV- MA	2 Seasons	After crossing paths at a party, a Cape Town t	Unknown Director	C	
		3	s2	TV Show	Blood & Water	2021-09-24	2021	TV- MA	2 Seasons	After crossing paths at a party, a Cape Town t	Unknown Director	C	
		4	s2	TV Show	Blood & Water	2021-09-24	2021	TV- MA	2 Seasons	After crossing paths at a party, a Cape Town t	Unknown Director		
4												•	

Insight -

 Negative days in this context occur when the date_added is earlier than the release_date. This means the movie was added to Netflix before it was officially released.

Recommendations --

• Continue producing and acquiring TV-MA content to cater to the strong demand among mature audiences.

- Invest in creating more TV-Y, TV-Y7, and G-rated content to attract and retain families and younger viewers.
- Continue producing and acquiring content in high-demand genres like dramas and international TV shows to maintain strong viewership.
- Invest in expanding content in underrepresented genres to attract diverse audience segments and fill gaps in the content library.
- Increase offerings in underrepresented genres like documentary, independent films, and foreign cinema to attract a broader audience.
- Given the popularity of true crime documentaries and mystery thrillers, prioritize production in these genres to maintain high viewer interest and retention.