Netflix data analysis

Problem statement : Analyse the Netflix data and find the type of shows to produce and how to grow the business

```
In [1]:

import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

In [2]:

df = pd.read_csv('netflix.csv')

In [3]:

df.head()
```

Out[3]:

	show_id	type	title	director	cast	country	date_added	release_year	rating	•
0	s 1	Movie	Dick Johnson Is Dead	Kirsten Johnson	NaN	United States	September 25, 2021	2020	PG- 13	_
1	s2	TV Show	Blood & Water	NaN	Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban	South Africa	September 24, 2021	2021	TV- MA	
2	s3	TV Show	Ganglands	Julien Leclercq	Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabi	NaN	September 24, 2021	2021	TV- MA	
3	s4	TV Show	Jailbirds New Orleans	NaN	NaN	NaN	September 24, 2021	2021	TV- MA	
4	s 5	TV Show	Kota Factory	NaN	Mayur More, Jitendra Kumar, Ranjan Raj, Alam K	India	September 24, 2021	2021	TV- MA	
4									•	•

In [4]:

df.shape

Out[4]:
(8807, 12)

In [5]:

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
4+,,,,	oc. in+61(1)	obioc+(11)	

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

In [6]: ▶

df.describe(include='all')

Out[6]:

	show_id	type	title	director	cast	country	date_added	release_year	ra
count	8807	8807	8807	6173	7982	7976	8797	8807.000000	3
unique	8807	2	8807	4528	7692	748	1767	NaN	
top	s2060	Movie	Tremors: Shrieker Island	Rajiv Chilaka	David Attenborough	United States	January 1, 2020	NaN	
freq	1	6131	1	19	19	2818	109	NaN	3
mean	NaN	NaN	NaN	NaN	NaN	NaN	NaN	2014.180198	
std	NaN	NaN	NaN	NaN	NaN	NaN	NaN	8.819312	
min	NaN	NaN	NaN	NaN	NaN	NaN	NaN	1925.000000	
25%	NaN	NaN	NaN	NaN	NaN	NaN	NaN	2013.000000	
50%	NaN	NaN	NaN	NaN	NaN	NaN	NaN	2017.000000	
75%	NaN	NaN	NaN	NaN	NaN	NaN	NaN	2019.000000	
max	NaN	NaN	NaN	NaN	NaN	NaN	NaN	2021.000000	

In [8]:

df.isna().sum()

Out[8]:

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

TV Shows Vs Movies

```
In [9]: ▶
```

```
df['type'].value_counts()
```

Out[9]:

Movie 6131 TV Show 2676

Name: type, dtype: int64

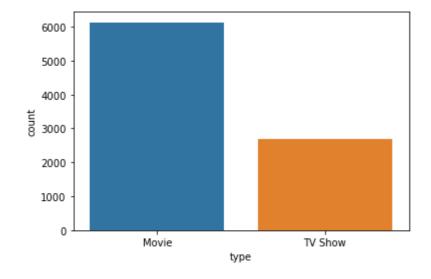
The type of show found to be more in number is Movie when compared to TV Shows.

```
In [13]: ▶
```

```
sns.countplot(data=df, x='type')
```

Out[13]:

<AxesSubplot:xlabel='type', ylabel='count'>



```
In [91]:
```

```
df['year_added'] = pd.to_datetime(df['date_added'])
```

```
In [92]:
```

df['year_added'] = df['year_added'].dt.year

In [40]: ▶

df.head()

Out[40]:

	show_id	type	title	director	cast	country	date_added	release_year	rating
0	s 1	Movie	Dick Johnson Is Dead	Kirsten Johnson	NaN	United States	September 25, 2021	2020	PG- 13
1	s2	TV Show	Blood & Water	NaN	Ama Qamata, Khosi Ngema, Gail Mabalane, Thaban	South Africa	September 24, 2021	2021	TV- MA
2	s3	TV Show	Ganglands	Julien Leclercq	Sami Bouajila, Tracy Gotoas, Samuel Jouy, Nabi	NaN	September 24, 2021	2021	TV- MA
3	s4	TV Show	Jailbirds New Orleans	NaN	NaN	NaN	September 24, 2021	2021	TV- MA
4	s 5	TV Show	Kota Factory	NaN	Mayur More, Jitendra Kumar, Ranjan Raj, Alam K	India	September 24, 2021	2021	TV- MA
4									•

In [42]:
▶

```
df.groupby(by='year_added')['type'].value_counts()
```

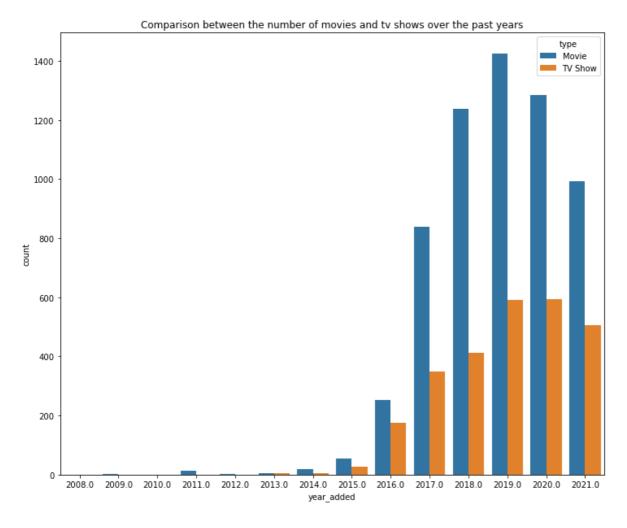
Out[42]:

year_added	type	
2008.0	Movie	1
	TV Show	1
2009.0	Movie	2
2010.0	Movie	1
2011.0	Movie	13
2012.0	Movie	3
2013.0	Movie	6
	TV Show	5
2014.0	Movie	19
	TV Show	5
2015.0	Movie	56
	TV Show	26
2016.0	Movie	253
	TV Show	176
2017.0	Movie	839
	TV Show	349
2018.0	Movie	1237
	TV Show	412
2019.0	Movie	1424
	TV Show	592
2020.0	Movie	1284
	TV Show	595
2021.0	Movie	993
	TV Show	505

Name: type, dtype: int64

In [41]: ▶

```
plt.figure(figsize=(12,10))
sns.countplot(x='year_added', hue='type', data=df)
plt.title('Comparison between the number of movies and tv shows over the past years')
plt.show()
```



In [43]:

Over the past years, the highest number of movies and tv shows were added to Netflix in the year 2019. Even though the tv shows are having increased popularity these days, movies are still higher in number.

Type of content available in different countries

```
df['country'].value_counts().sort_values(ascending=False)
Out[43]:
United States
                                            2818
                                             972
India
United Kingdom
                                             419
                                             245
Japan
South Korea
                                             199
United Kingdom, China
                                                1
United States, Germany, Mexico
                                                1
United States, Australia, Mexico
                                                1
                                                1
France, Belgium, Luxembourg, Cambodia,
Denmark, France, United States, Sweden
                                                1
Name: country, Length: 748, dtype: int64
In [44]:
                                                                                              H
df['country'].nunique()
Out[44]:
748
Among the 748 countries, let's consider the top countries from above analysis - United States, India, United
Kingdom, Japan, South Korea
In [46]:
                                                                                              H
top_countries = df[(df['country']=='United States') | (df['country']=='India') | (df['count
                   (df['country']=='Japan') | (df['country']=='South Korea')]
In [49]:
                                                                                              H
top_countries['country'].value_counts()
```

Name: country, dtype: int64

2818

972

419 245

199

localhost:8888/notebooks/Documents/Scaler Advanced/Netflix data Case study.ipynb#

Out[49]:

India

Japan

United States

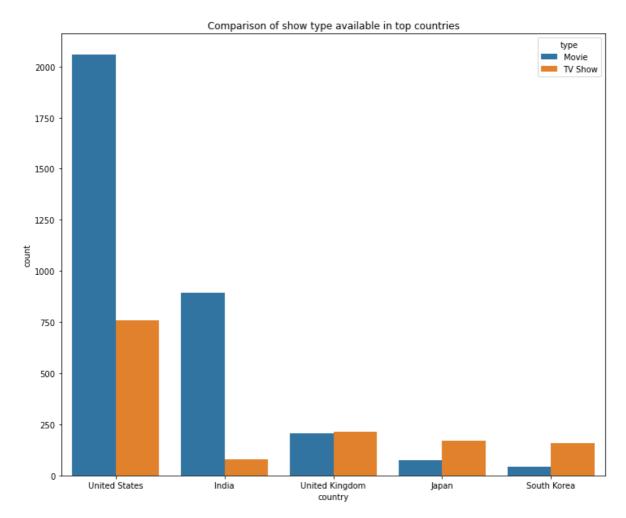
United Kingdom

South Korea

M

In [52]:

```
plt.figure(figsize=(12,10))
sns.countplot(x='country', hue='type', data=top_countries)
plt.title('Comparison of show type available in top countries')
plt.show()
```



United States has the highest no. of movies and tv shows and India has the second highest. Rest three countries have lesser in number when compared to United states and India. Also, these 3 countries, UK, Japan and South Korea is found to be more focused on TV shows than movies.

Content update analysis

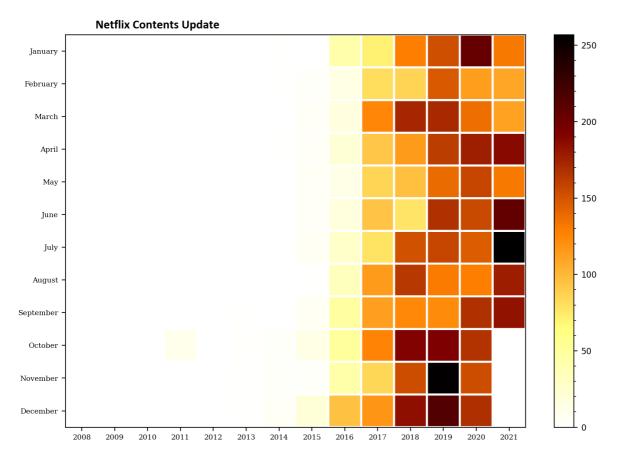
In [55]:

H

```
date_update = df[['date_added']].dropna()
date_update['year'] = date_update['date_added'].apply(lambda x : x.split(', ')[-1])
date_update['month'] = date_update['date_added'].apply(lambda x : x.lstrip().split(' ')[0])

months = ['January', 'February', 'March', 'April', 'May', 'June', 'July', 'August', 'Septem content_update = date_update.groupby('year')['month'].value_counts().unstack().fillna(0)[mo plt.figure(figsize=(10, 7), dpi=200)
plt.pcolor(content_update, cmap='afmhot_r', edgecolors='white', linewidths=2)
plt.xticks(np.arange(0.5, len(content_update.columns), 1), content_update.columns, fontsize plt.yticks(np.arange(0.5, len(content_update.index), 1), content_update.index, fontsize=7,
plt.title('Netflix Contents Update', fontsize=12, fontfamily='calibri', fontweight='bold', cbar = plt.colorbar()

cbar.ax.tick_params(labelsize=8)
cbar.ax.minorticks_on()
plt.show()
```



17

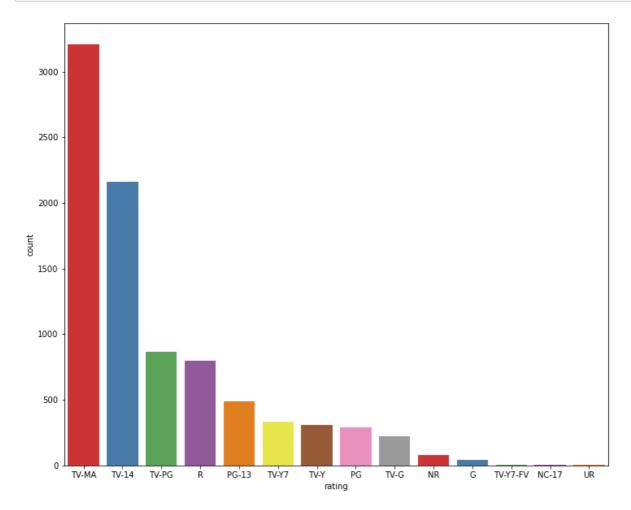
From 2018 to 2020, frequent content update is found from the month of October to January. But in 2021, more content update is seen starting from mid year.

Type of ratings

```
In [57]:
                                                                                               M
df['rating'].value_counts()
Out[57]:
TV-MA
            3207
TV-14
            2160
TV-PG
             863
              799
R
PG-13
             490
TV-Y7
             334
TV-Y
              307
PG
              287
TV-G
              220
NR
               80
G
               41
TV-Y7-FV
               6
NC-17
                3
               3
UR
               1
74 min
               1
66 min
84 min
               1
Name: rating, dtype: int64
In [58]:
                                                                                               H
df['rating'].nunique()
Out[58]:
```

In [61]: ▶

```
plt.figure(figsize=(12,10))
sns.countplot(x="rating", data=df, palette="Set1", order=df['rating'].value_counts().index[
plt.show()
```



Among the 14 types of ratings, TV-MA(TV Mature Audience Only) has the highest category followed by TV-14(Parents Strongly Cautioned) and TV-PG(Parental Guidance Suggested). This shows that Netflix has more content for Adults.

Ratings vs Show type

```
In [68]:

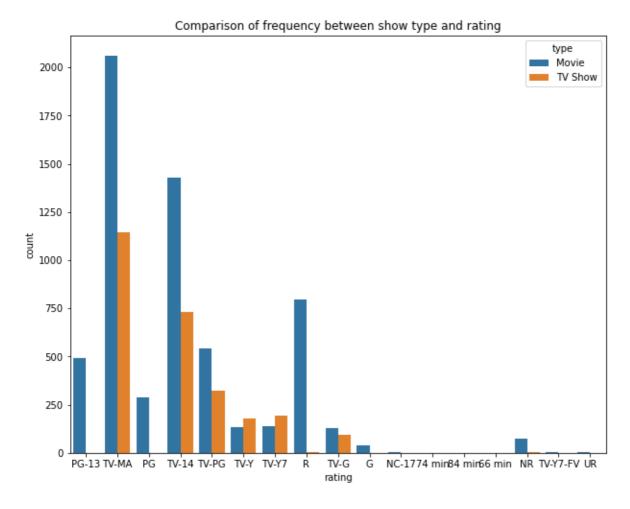
df.groupby(by='type')['rating'].value_counts()
```

Out[68]:

type	rating			
Movie	TV-MA	2062		
	TV-14	1427		
	R	797		
	TV-PG	540		
	PG-13	490		
	PG	287		
	TV-Y7	139		
	TV-Y	131		
	TV-G	126		
	NR	75		
	G	41		
	TV-Y7-FV	5		
	NC-17	3		
	UR	3		
	66 min	1		
	74 min	1		
	84 min	1		
TV Show	TV-MA	1145		
	TV-14	733		
	TV-PG	323		
	TV-Y7	195		
	TV-Y	176		
	TV-G	94		
	NR	5		
	R	2		
	TV-Y7-FV	1		
Name: rating, dtype: int64				

In [65]: ▶

```
plt.figure(figsize=(10,8))
sns.countplot(x='rating',hue='type',data=df)
plt.title('Comparison of frequency between show type and rating')
plt.show()
```



TV-MA, TV-14, TV-PG, TV-Y, TV-G only have both movie and tv shows. And the other ratings have only show type has movie. It says that Movies is the majority category.

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

As the cast, director and listed_in columns have nested data, let's unnest them

```
In [70]:

cast_constraint=df['cast'].apply(lambda x : str(x).split(', ')).tolist()
cast_df = pd.DataFrame(cast_constraint, index=df['title'])
cast_df = cast_df.stack()
cast_df = pd.DataFrame(cast_df)
cast_df.reset_index()
```

Out[70]:

	title	level_1	0
0	Dick Johnson Is Dead	0	nan
1	Blood & Water	0	Ama Qamata
2	Blood & Water	1	Khosi Ngema
3	Blood & Water	2	Gail Mabalane
4	Blood & Water	3	Thabang Molaba
64946	Zubaan	3	Manish Chaudhary
64947	Zubaan	4	Meghna Malik
64948	Zubaan	5	Malkeet Rauni
64949	Zubaan	6	Anita Shabdish
64950	Zubaan	7	Chittaranjan Tripathy

64951 rows × 3 columns

In [71]: ▶

```
director_constraint=df['director'].apply(lambda x : str(x).split(', ')).tolist()
director_df = pd.DataFrame(director_constraint, index=df['title'])
director_df = director_df.stack()
director_df = pd.DataFrame(director_df)
director_df.reset_index()
```

Out[71]:

0	level_1	title	
Kirsten Johnson	0	Dick Johnson Is Dead	0
nan	0	Blood & Water	1
Julien Leclercq	0	Ganglands	2
nan	0	Jailbirds New Orleans	3
nan	0	Kota Factory	4
David Fincher	0	Zodiac	9607
nan	0	Zombie Dumb	9608
Ruben Fleischer	0	Zombieland	9609
Peter Hewitt	0	Zoom	9610
Mozez Singh	0	Zubaan	9611

9612 rows × 3 columns

In [72]: ▶

```
listedin_constraint = df['listed_in'].apply(lambda x : str(x).split(', ')).tolist()
listedin_df = pd.DataFrame(listedin_constraint, index=df['title'])
listedin_df = listedin_df.stack()
listedin_df = pd.DataFrame(listedin_df)
listedin_df.reset_index()
```

Out[72]:

	title	level_1	0
0	Dick Johnson Is Dead	0	Documentaries
1	Blood & Water	0	International TV Shows
2	Blood & Water	1	TV Dramas
3	Blood & Water	2	TV Mysteries
4	Ganglands	0	Crime TV Shows
19318	Zoom	0	Children & Family Movies
19319	Zoom	1	Comedies
19320	Zubaan	0	Dramas
19321	Zubaan	1	International Movies
19322	Zubaan	2	Music & Musicals

19323 rows × 3 columns

```
In [73]:
```

```
cast_df = cast_df.reset_index()
cast_df.drop(columns='level_1', inplace=True)
director_df = director_df.reset_index()
director_df.drop(columns='level_1', inplace=True)
listedin_df = listedin_df.reset_index()
listedin_df.drop(columns='level_1', inplace=True)
```

```
In [77]:
```

```
from functools import reduce
merged_df = reduce(lambda x,y: pd.merge(x,y, on='title', how='inner'), [cast_df, director_d
```

Out[77]:

	title	0_x	0_у	0
0	Dick Johnson Is Dead	nan	Kirsten Johnson	Documentaries
1	Blood & Water	Ama Qamata	nan	International TV Shows
2	Blood & Water	Ama Qamata	nan	TV Dramas
3	Blood & Water	Ama Qamata	nan	TV Mysteries
4	Blood & Water	Khosi Ngema	nan	International TV Shows

```
In [78]:

merged_df.columns = ['title','cast','director','listed_in']
merged_df.head()
```

Out[78]:

	title	cast	director	listed_in
0	Dick Johnson Is Dead	nan	Kirsten Johnson	Documentaries
1	Blood & Water	Ama Qamata	nan	International TV Shows
2	Blood & Water	Ama Qamata	nan	TV Dramas
3	Blood & Water	Ama Qamata	nan	TV Mysteries
4	Blood & Water	Khosi Ngema	nan	International TV Shows

```
In [79]:
                                                                                            H
merged_df.shape
Out[79]:
(161216, 4)
In [80]:
                                                                                            H
netflix_df = reduce(lambda x,y: pd.merge(x,y, on='title', how='inner'), [merged_df, df])
In [82]:
                                                                                            M
netflix_df.drop(columns=['director_y','cast_y','listed_in_y','description'], inplace=True)
In [84]:
                                                                                            H
netflix_df['cast_x'].value_counts()
Out[84]:
nan
                     1733
Anupam Kher
                      120
                       99
Shah Rukh Khan
Naseeruddin Shah
                       98
```

```
Nicole Richie 1
Name: cast_x, Length: 36440, dtype: int64
```

89

1

1

1

1

Radhika Apte

Nanao

Denzel Whitaker

Grant McFarland

Jennifer L. Yen

Netflix has shows of the actors Anupam Kher, Shah Rukh Khan, Naseeruddin Shah and Radhika Apte more in number when compared other thousands of actors.

Caio Cobra 1
Matthew Cooke 1
Marc-Antoine Hélard 1
Jon Olb 1
David Batra 1

Name: director_x, Length: 4994, dtype: int64

Cathy Garcia-Molina, Youssef Chahine, Martin Scorsese, David Dhawan are the top directors

```
In [86]:
                                                                                              M
netflix_df['listed_in_x'].value_counts()
Sci-Fi & Fantasy
                                   2603
Anime Series
                                   2126
Documentaries
                                   1937
Spanish-Language TV Shows
                                   1853
TV Action & Adventure
                                  1796
British TV Shows
                                  1434
Sports Movies
                                  1275
TV Mysteries
                                  1138
                                  1086
Korean TV Shows
Classic Movies
                                  1085
Anime Features
                                   928
TV Sci-Fi & Fantasy
                                   866
                                    841
TV Horror
Cult Movies
                                    781
Docuseries
                                   753
Teen TV Shows
                                   742
LGBTQ Movies
                                   727
TV Thrillers
                                    650
Reality TV
                                    578
```

```
In [87]:

netflix df['listed in x'].nunique()
```

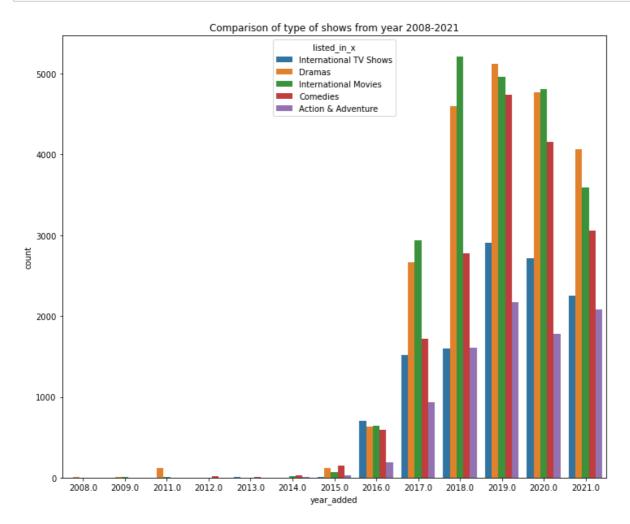
Out[87]:

42

Netflix has 42 types of shows. Among them Internation movies, dramas, comedies, International TV shows are the top categories.

```
In [99]: ▶
```

```
top_categories = netflix_df[(netflix_df['listed_in_x']=='International Movies') | (netflix_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_to_lin_
```



Above graph clearly shows that Netflix should produce more shows in the above five categories to gain more business.

