```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
df=pd.read_csv('/content/drive/MyDrive/Colab Notebooks/original_netflix.csv')
df.head(10)
 #For stacking Cast columns with title
constraint=df['cast'].apply(lambda x: str(x).split(', ')).tolist()
df_new=pd.DataFrame(constraint,index=df['title'])
df_new=df_new.stack().reset_index()
df_new=df_new[['title',0]]
df_new.columns=['title','cast']
df_new['cast'].replace(to_replace="nan",
                  value='Unknown',inplace=True)
 #For stacking director columns with title
df.head()
c=df['director'].apply(lambda x: str(x).split(', ')).tolist()
df1=pd.DataFrame(c,index=df['title'])
df1=df1.stack().reset_index()
df1=df1[['title',0]]
df1.columns=['title','director']
df1['director'].replace(to_replace="nan",
                  value='Unknown',inplace=True)
 #filling the nan values with the most occuring elements in the director column
  #For stacking Genre columns with title
d=df['listed_in'].apply(lambda x: str(x).split(', ')).tolist()
df2=pd.DataFrame(d,index=df['title'])
df2=df2.stack().reset_index()
df2=df2[['title',0]]
df2.columns=['title','listed_in']
df2['listed_in'].isnull().sum() #checking null values
#For stacking Country columns with title
e=df['country'].apply(lambda x: str(x).split(', ')).tolist()
df3=pd.DataFrame(e,index=df['title'])
df3=df3.stack().reset_index()
df3=df3[['title',0]]
df3.columns=['title','country']
df3['country'].replace(to_replace="nan",
                  value=df3['country'].mode()[0],inplace=True)
 #Merging all the data frame into a single one
d=df_new.merge(df1,how='inner',on='title')
e=d.merge(df2,how='inner',on='title')
df updated=e.merge(df3,how='inner',on='title')
df_updated.head(10)
#Now merging data frame with the original data frame
df_final=df_updated.merge(df,on='title')
df_final.drop(['director_y','cast_y','country_y','listed_in_y'],axis=1,inplace=True)
df_final.rename(columns={'director_x': 'director',
                                        'cast_x': 'cast',
                                        'country_x': 'country','listed_in_x':'listed_in'},inplace=True)
df_final.shape #checking the new size after cleaning, transforming and merging
#removing null values from the updated data frame
df_final.loc[df_final['rating'].isnull()] #checking null values in rating column
\label{eq:df_final} $$ df_final['rating']. mode()[0], inplace=True) $$ filling null values with mode $$ df_final['rating']. $$ d
df_final.loc[df_final['date_added'].isnull()] #filling null values in date added column with mode
df_final['date_added'].fillna(df_final['date_added'].mode()[0], inplace=True) #filling null values with mode
df_final['duration'].fillna(df_final['duration'].mode()[0], inplace=True) #filling null values with mode
#Performing Exploratory Data Analysis on the final data set after cleaning
df_final.describe()
#From the above command we can conclude that maximum percentage of the movies released in the year 2019
df_pop=df_final.groupby(['cast'],as_index=False)['title'].nunique().sort_values(by='title',ascending=False)#from this command Anupam Kher is
```

df_pop

```
₽
                                       1
                                             th
                        cast title
     34214
                    Unknown
                                825
     2833
                Anupam Kher
                                 43
     30489
              Shah Rukh Khan
                                 35
                 Julie Tejwani
     16697
                                 33
     24215 Naseeruddin Shah
                                 32
     14221
                   Jamie Lee
                                  1
     14219
                 Jamie Kenna
     14218
                  Jamie Kaler
     14217
               Jamie Johnston
                                  1
     36439
                  Şọpé Dìrísù
    36440 rows × 2 columns
```

#Which Type is more popular:Tv shows or Movies?
df_type=df_final.groupby(['type'],as_index=False)['type'].value_counts().sort_values(by='type',ascending=False)
df_type

```
type count1 TV Show 561480 Movie 145843
```

ax = sns.countplot(x='rating', data=df_final,hue='type')
plt.figure(figsize=(20, 12))

ax.set_xticklabels(ax.get_xticklabels(), rotation=45) # Adjust the rotation angle (45) as needed
nlt show()

#Movie rating based on type(tv show or movies): from this we can conclude that TV-MA rating for movies type is highest

```
#Analysis of actors/directors of different types of shows/movies
df_final.head()
#Analysis of actors/directors of different types of shows/movies for eg: getting all the movies where cast is 'Shahrukh khan'
df_pop=df_final[['cast','title']]
df_pop[df_pop['cast']=='Shah Rukh Khan']
```

	cast	title
2491	Shah Rukh Khan	Anjaam
2492	Shah Rukh Khan	Anjaam
2493	Shah Rukh Khan	Anjaam
7327	Shah Rukh Khan	Chennai Express
7328	Shah Rukh Khan	Chennai Express
181970	Shah Rukh Khan	Shakti: The Power
181971	Shah Rukh Khan	Shakti: The Power
197883	Shah Rukh Khan	Trimurti
197884	Shah Rukh Khan	Trimurti
197885	Shah Rukh Khan	Trimurti

108 rows × 2 columns

plt.show()

#who is the most popular actor-director pair?
df_acd=df_final.groupby(['cast','director']).size().reset_index(name='popularity').sort_values(by='popularity',ascending=False)
df_acd.head(10)

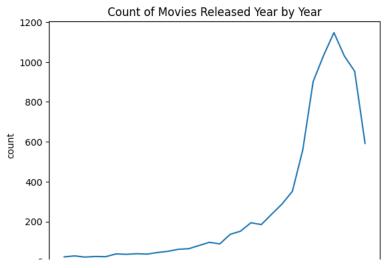
#Top 10 popular actor director pairs are given below:

	cast	director	popularity
59162	Unknown	Unknown	738
13508	David Attenborough	Unknown	72
55848	Takahiro Sakurai	Unknown	54
61867	Yuki Kaji	Unknown	43
28907	Jun Fukuyama	Unknown	38
61834	Yuichi Nakamura	Unknown	38
28993	Junichi Suwabe	Unknown	37
29930	Kate Harbour	Unknown	37
1186	Ai Kayano	Unknown	37
12529	Daisuke Ono	Unknown	36

```
#How has the number of movies released per year changed over the last 20-30 years?
df_yr=df_final.groupby('release_year')['title'].nunique().reset_index(name='count').sort_values(by='release_year')
df_yr
#Lets analyze the trend over the last 30 years?
dfyr1=df_yr.tail(30)

#Plotting the trend using line chart
plt.plot(dfyr1['release_year'], dfyr1['count'])
plt.xlabel('release_year')
plt.ylabel('count')
plt.title('Count of Movies Released Year by Year')
```

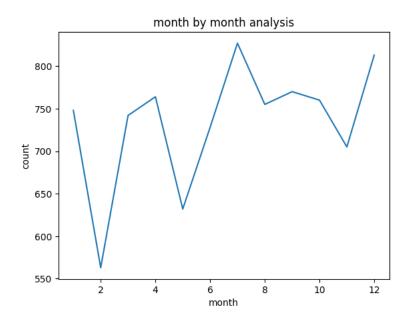
#From the below graph we can analyze that the movies released year by year is increasing.



```
#What is the best time to launch a TV show or a movie?
df_final['date_added']=pd.to_datetime(df_final['date_added'])
get_month = lambda x: x.month
df_final['Month'] = df_final['date_added'].apply(get_month)
df_t=df_final.groupby('Month')['title'].nunique().reset_index(name='count').sort_values(by='Month')

plt.plot(df_t['Month'], df_t['count'])
plt.xlabel('month')
plt.ylabel('count')
plt.title('month by month analysis')
plt.show()
```

#From this analysis we can conclude that the best release time to launch a movie or a tv show is between september to december(estimated)



#Understanding what content is available in different countries?
df_c=df_final.groupby(['country'])['listed_in'].unique().reset_index(name='Content Available')
df_c.head(10)

```
country
                                                       Content Available
       0
                         [International TV Shows, TV Dramas, Dramas, In...
       1 Afghanistan
                                      [Documentaries, International Movies]
       2
               Albania
                                              [Dramas, International Movies]
       3
               Algeria
                         [Dramas, Independent Movies, International Mov...
               Angola
                                  [Action & Adventure, International Movies]
                                      #let us find what type of content is available in India:
df_c[df_c['country']=='India'].values
      array([['India',
                array(['International TV Shows', 'Romantic TV Shows', 'TV Comedies',
                        'Comedies', 'International Movies', 'Romantic Movies', 'Thrillers', 
"Kids' TV", 'TV Dramas', 'TV Sci-Fi & Fantasy', 'Docuseries', 
'Action & Adventure', 'Dramas', 'Independent Movies',
                         'Horror Movies', 'Music & Musicals', 'Sci-Fi & Fantasy',
                         'TV Shows', 'Children & Family Movies', 'Reality TV',
                         'Documentaries', 'Sports Movies', 'Stand-Up Comedy', 'British TV Shows', 'Cult Movies', 'TV Action & Adventure',
                         'TV Horror', 'Crime TV Shows', 'Stand-Up Comedy & Talk Shows',
                         'Classic Movies', 'TV Mysteries', 'Teen TV Shows', 'TV Thrillers', 'LGBTQ Movies', 'Faith & Spirituality'], dtype=object)
             dtype=object)
#Who is the most popular director in india?
df d=df final[df final['country']=='India']
df_d.groupby(['director'],as_index=False)['title'].nunique().sort_values(by='title',ascending=False)
```

director title 684 Unknown 85 151 David Dhawan 9 80 Anurag Kashyap 9 Umesh Mehra 682 8 7 168 Dibakar Banerjee ... 341 Musthafa 1 95 Ashim Ahluwalia 343 N. Chandra Arvind Swamy 94 368 Nikhil Nagesh Bhat 1

736 rows × 2 columns

```
#Identifying the Top 5 genres of content on Netflix
df_g=df_final.groupby(['listed_in']).size().reset_index(name='popularity').sort_values(by='popularity',ascending=False)
df_g.head(5)
#From the below table,Dramas,International Movies,comedies,Tv shows,Action and Adevnture are the most popular genre across the world
```

```
listed_in popularity

12 Dramas 29775

16 International Movies 28211

df_final.describe()
```

	release_year	Month
count	201991.000000	201991.000000
mean	2013.452891	6.631909
std	9.003933	3.444674
min	1925.000000	1.000000
25%	2012.000000	4.000000
50%	2016.000000	7.000000
75%	2019.000000	10.000000
max	2021.000000	12.000000

```
#Outlier Check
y=df_final['date_added'].dt.year
sns.boxplot(y)
#In the below graph, can clearly see that values below 2015 are acting as outliers.
```

```
Traceback (most recent call last)
    <ipython-input-2-a7270dcc8954> in <cell line: 2>()
          1 #Outlier Check
     ----> 2 y=df_final['date_added'].dt.year
         3 sns.boxplot(y)
          4 #In the below graph, can clearly see that values below 2015 are acting as outliers.
    NameError: name 'df final' is not defined
     SEARCH STACK OVERFLOW
#Comments on the range of attributes:
df_final.shape #(the final data set after transforming and cleaning has 201991 rows and 13 rows)
     (201991, 13)
df final.info()
#Most of the attributes are of object type except date_added and release_year with no null entries.All the null values are detected and recti
     <class 'pandas.core.frame.DataFrame'>
    Int64Index: 201991 entries, 0 to 201990
    Data columns (total 13 columns):
     # Column
                     Non-Null Count Dtype
     ___
     0 title
                      201991 non-null object
                       201991 non-null object
     1
        cast
         director
                       201991 non-null object
         listed_in
                      201991 non-null object
         country
                      201991 non-null object
                      201991 non-null object
         show_id
                       201991 non-null object
         type
         date_added
                       201991 non-null datetime64[ns]
         release_year 201991 non-null int64
     8
         rating
                       201991 non-null object
     10 duration
                       201991 non-null object
```

11 description 201991 non-null object 12 Month 201991 non-null int64 dtypes: datetime64[ns](1), int64(2), object(10) memory usage: 29.6+ MB df_final.size #total no of elements in the dataframe 2625883 df_final.ndim #the datset is of 2 dimensional(Dataframe) 2

#Business insights:

By performing EDA of the final dataset we can analyze various trends in our data in every aspect and came out with the following:

- 1.From the statistical analysis we can analyze that 75% of the movies released in netflix by the year 2019.
- 2.The first movie was released in the year 1925 and Anupam Kher is the most popular actor having 43 Titles.
- 3. Movies gor maximum rating of type 'TV-MA' and Tv shows of the same type
- 4.After analysing movies count year by year we can see that production of movies is increasing year by year in netflix.
- 5. Anakysing month by month the peak time for any movie release is between september to december.
- 6. The most popular actor in india is found out to be david Dhawan with movies count of 9.
- 7. The most popular genres in Netflix is found out to be Dramas, International Movies, comedies, Tv shows, Action and Adventure.
- 8.From the outlier Check using Box plot the movies released before 2015 are founded out to be outliers.

'\nBy performing EDA of the final dataset we can analyze various trends in our data in every aspect an d came out with the following:\n1.From the statistical analysis we can analyze that 75% of the movies released in netflix by the year 2019.\n2.The first movie was released in the year 1925 and Anupam Kher is the most popular actor having 43 Titles.\n3.Movies gor maximum rating of type 'TV-MA' and $\overline{\text{Tv}}$ shows of the same type \n 4.After analysing movies count year by year we can see that production of movies is increasing year by year in netflix.\n5.Anakysing month by month the peak time for any movie release is between sentember to december \n6. The most nonular actor in india is found out to be david Dhawan with

#Recommendations:

- 1.From the above analysis the content of movies or tv series with high rating should be focused more
- 2. The major issue of the dataset is max of Nan values in each column so if we use the mode operator our insights and data prediction may go w
- 3. The peak time for any series released in between September to December... so the movies releasing between these time interval should be focu
- 4. Some of the contents of netlix are not available in many contries...to make people aware of the present scenario of the world more no of mo
- 5.The movie or TV show that got a rating of TV-MA should be broadcasted to a wider audience.

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