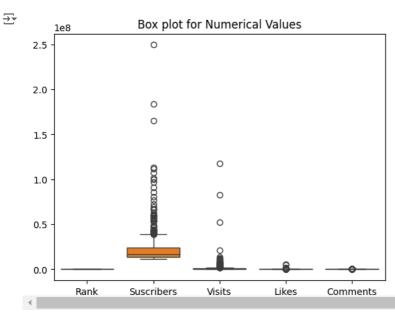
```
from google.colab import files
import pandas as pd
# Prompt the file upload dialog
uploaded = files.upload()
# List the uploaded files and read the CSV file
for filename in uploaded.keys():
   print(f'User uploaded file "{filename}" with length {len(uploaded[filename])} bytes')
    # Read the uploaded file
   df = pd.read_csv(filename)
     Choose Files No file chosen
                                      Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to
     enable.
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
# Loading the data
df = pd.read_csv(r"youtubers_df.csv")
# Data exploration
print(df.info())
<pr
     RangeIndex: 1000 entries, 0 to 999
     Data columns (total 9 columns):
      #
         Column
                     Non-Null Count Dtype
      0
         Rank
                      1000 non-null
                                      int64
      1
         Username
                      1000 non-null
                                      object
         Categories 694 non-null
                                      object
         Suscribers 1000 non-null
                                     float64
                      1000 non-null
         Country
                                      obiect
                      1000 non-null
         Visits
                                      int64
                      1000 non-null
                                      int64
         Likes
                     1000 non-null
                                      int64
         Comments
      8
         Links
                      1000 non-null
                                      object
     dtypes: float64(1), int64(4), object(4)
     memory usage: 70.4+ KB
     None
print(df.describe())
₹
                   Rank
                           Suscribers
                                             Visits
                                                            Likes
                                                                        Comments
     count
           1000.000000
                        1.000000e+03
                                      1.000000e+03
                                                     1.000000e+03
                                                                     1000.000000
             500.500000
                        2.189440e+07
                                      1.209446e+06
                                                     5.363259e+04
                                                                     1288.768000
     mean
             288.819436
                        1.682775e+07
                                       5.229942e+06
                                                     2.580457e+05
                                                                     6778.188308
     std
              1.000000
                        1.170000e+07
                                      0.000000e+00
                                                     0.000000e+00
                                                                        0.000000
     min
     25%
             250.750000
                        1.380000e+07
                                       3.197500e+04
                                                     4.717500e+02
                                                                        2.000000
     50%
             500.500000
                        1.675000e+07
                                       1.744500e+05
                                                     3.500000e+03
                                                                       67.000000
     75%
             750.250000
                         2.370000e+07
                                       8.654750e+05
                                                                      472.000000
                                                     2.865000e+04
           1000.000000
                        2.495000e+08
                                      1.174000e+08
                                                     5.300000e+06 154000.000000
     max
df.shape
→ (1000, 9)
print(df.isnull().sum())
     Rank
                     0
→
     Username
                     0
     Categories
                   306
     Suscribers
                    a
     Country
                     a
     Visits
                     0
     Likes
                     0
     Comments
                     0
     Links
     dtype: int64
#Drop Rows with missing values
df = df.dropna()
```

df.isnull().sum()

```
₹
                0
        Rank
                0
     Username
                0
     Categories 0
     Suscribers 0
      Country
                0
       Visits
                0
       Likes
                0
     Comments 0
       Links
                0
```

dtype: int64

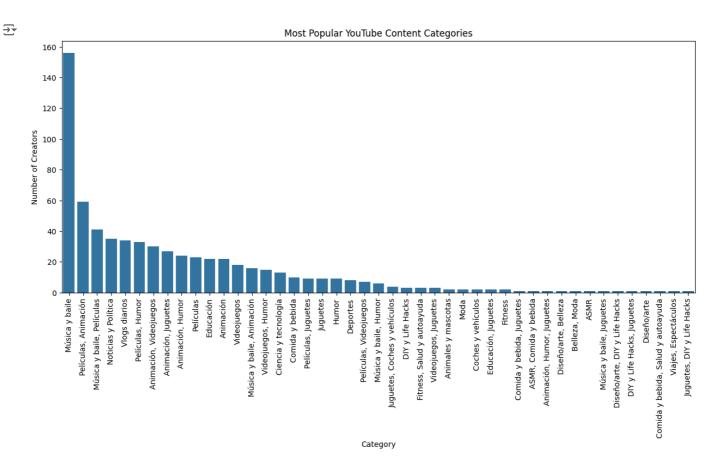
```
# Check for outliers using boxplots (For numerical values)
sns.boxplot(data=df)
plt.title("Box plot for Numerical Values")
plt.show()
```



```
# Function to remove outliers using z-score method
from scipy.stats import zscore
def remove_outliers_zscore(dataframe, columns, threshold=3):
    df_no_outliers = dataframe.copy()
    for column in columns:
        z_scores = zscore(df_no_outliers[column])
        df_no_outliers = df_no_outliers[(z_scores < threshold)) & (z_scores > -threshold)]
    return df_no_outliers
# Remove outliers
columns_to_clean = ['Suscribers', 'Visits', 'Likes', 'Rank']
df = remove_outliers_zscore(df, columns_to_clean)
df.head()
```

₹	Rank		Username	Categories	Suscribers	Country	Visits	Likes	Comments	L	
	39	40	JuegaGerman	Películas, Animación	48600000.0	México	2000000	117100	3000	http://youtube.com/channel/UCYiGq8XF7YQD00x7	
	40	41	BillieEilish	Música y baile	48600000.0	Estados Unidos	208100	27200	476	http://youtube.com/channel/UCiGm_E4ZwYSHV3bc	
	42	43	sonymusicindiaVEVO	Música y baile	47500000.0	India	36600	664	28	http://youtube.com/channel/UC3MLnJtqc_phABE	
	4									→	

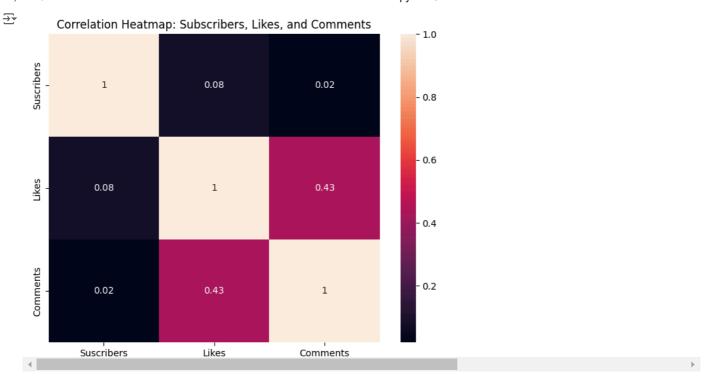
```
#To identify the most popular categories
category_counts = df['Categories'].value_counts()
plt.figure(figsize=(15,6))
sns.barplot(x=category_counts.index, y=category_counts.values)
plt.xticks(rotation=90)
plt.title("Most Popular YouTube Content Categories")
plt.xlabel("Category")
plt.ylabel("Number of Creators")
plt.show()
```



```
#The Correlation between Subscribers, Likes, and Comments
corr_likes = np.corrcoef(df['Suscribers'], df['Likes']) [0,1]
corr_comments = np.corrcoef(df['Suscribers'], df['Comments']) [0,1]
print(f'Correlation b/w Subscribers and Likes: {corr_likes}')
print(f'Correlation b/w Subscribers and Comments: {corr_comments}')
```

Correlation b/w Subscribers and Likes: 0.07958997301642444
Correlation b/w Subscribers and Comments: 0.01990529146597779

```
#The Correlation between Subscribers, Likes, and Comments
correlation = df[['Suscribers', 'Likes', 'Comments']].corr()
plt.figure(figsize=(8,6))
sns.heatmap(correlation, annot=True)
plt.title("Correlation Heatmap: Subscribers, Likes, and Comments")
plt.show()
```



#Audience Study

#To count the number of creators in each combination
category_country_counts = df.groupby(['Categories', 'Country'])['Categories'].count().reset_index(name='Count')
print(category_country_counts)

∑ *			Cat	tegories	(Country	Count
	0			ASMR	Estados	Unidos	1
	1	ASMR,	Comida	/ bebida	Estados	Unidos	1
	2		Ar	nimación	Arg	gentina	1
	3		Ar	nimación		Brasil	3
	4		Ar	nimación	Estados	Unidos	4
	163		Vlogs	diarios		India	12
	164		Vlogs	diarios	Ind	donesia	1
	165		Vlogs	diarios	Pa	akistán	1
	166		Vlogs	diarios	1	Γurquía	2
	167		Vlogs	diarios	l	Jnknown	7

[168 rows x 3 columns]

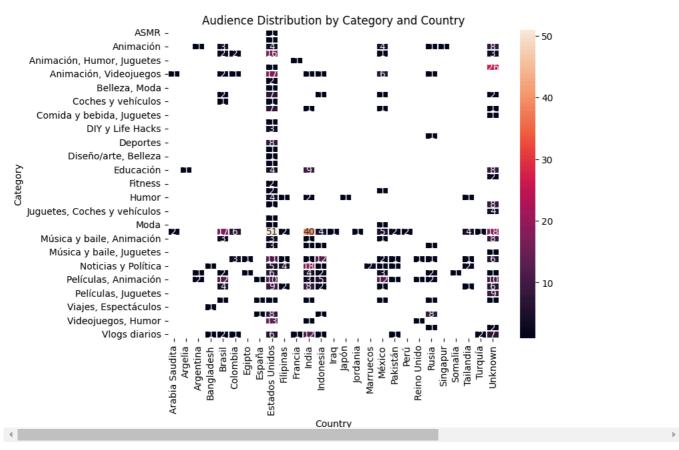
country_visit_count = df.groupby('Country')['Visits'].sum().reset_index()
country_visit_count

	Country	Visits
0	Arabia Saudita	3474500
1	Argelia	333500
2	Argentina	6371400
3	Bangladesh	100700
4	Brasil	18643600
5	Colombia	6256400
6	Egipto	305400
7	España	1984300
8	Estados Unidos	207221500
9	Filipinas	8914600
10	Francia	5308000
11	India	35783000
12	Indonesia	11556900
13	Iraq	103600
14	Japón	2100000
15	Jordania	267600
16	Marruecos	12000
17	México	31365000
18	Pakistán	1969600
19	Perú	2219400
20	Reino Unido	9250700
21	Rusia	23299600
22	Singapur	26400
23	Somalia	1900000
24	Tailandia	2553800
25	Turquía	1604900
26	Unknown	47211700

#Create a pivot table for better visualization
pivot_table = category_country_counts.pivot(index='Categories',columns='Country', values='Country')

```
#Heatmap for streamers audiences by country and category
plt.figure(figsize=(8,6))
sns.heatmap(pivot_table, annot=True, linewidths=0.3)
plt.title("Audience Distribution by Category and Country")
plt.xlabel("Country")
plt.ylabel("Category")
plt.show()
```





```
#Benchmarking
#Calculate average values for each metric
avg_subscribers = df['Suscribers'].mean()
avg_visits = df['Visits'].mean()
avg_likes = df['Likes'].mean()
avg_comments = df['Comments'].mean()
print(f'Average Subscribers: {avg_subscribers}')
print(f'Average Visits: {avg_visits}')
print(f'Average Likes: {avg_likes}')
print(f'Average Comments: {avg_comments}')
```

Average Subscribers: 20573867.069486406
 Average Visits: 649755.4380664653
 Average Likes: 24143.44259818731
 Average Comments: 1016.6510574018126

Identify streamers with above-average performance
above_avg_streamers = df[(df['Suscribers'] > avg_subscribers) &(df['Visits'] > avg_visits) &(df['Likes'] > avg_likes) &(df['Comments'] :
print(above_avg_streamers)

$\overline{\Rightarrow}$		Rank	Username	Categories	Suscribers	١
	14	15	BTS	Música y baile	76500000.0	
	26	27	dudeperfect	Videojuegos	59700000.0	
	34	35	TaylorSwift	Música y baile	54100000.0	
	37	38	ArianaGrande	Música y baile	52900000.0	
	39	40	JuegaGerman	Películas, Animación	48600000.0	
	58	59	Mikecrack	Películas, Animación	43400000.0	
	62	63	KimberlyLoaiza	Música y baile	42100000.0	
	70	71	JessNoLimit	Películas, Animación	39600000.0	
	96	97	TotalGaming093	Películas, Videojuegos	36300000.0	
	100	101	markiplier	Animación, Videojuegos	35500000.0	
	109	110	SSSniperWolf	Animación, Humor	34200000.0	
	145	146	jacksepticeye	Animación, Humor	30400000.0	
	171	172	SandeepSeminars	Vlogs diarios	28000000.0	
	177	178	DanTDM	Animación, Videojuegos	27800000.0	
	179	180	brentrivera	Videojuegos, Humor	27600000.0	
	180	181	NichLmao	Vlogs diarios	27500000.0	
	195	196	nickiminaj	Música y baile	26100000.0	
	202	203	VanossGaming	Animación, Videojuegos	25900000.0	
	206	207	AlejoIgoa	Animación	25700000.0	
	207	208	ZHCYT	Diseño/arte, DIY y Life Hacks	25700000.0	
	234	235	rug	Videojuegos, Humor	24300000.0	
	241	242	juandediospantojaa	Música y baile, Películas	24000000.0	
	243	244	JamesCharles	Belleza, Moda	23900000.0	
	272	273	AmiRodrigueZZ	Animación, Humor	22900000.0	
	278	279	StokesTwins	Videojuegos, Humor	22700000.0	
	281	282	SSundee	Animación, Videojuegos	22700000.0	
	285	286	BenAzelart	Videojuegos, Humor	22500000.0	

```
302
           303
                         royaltyfam
                                                              Humor
                                                                     21900000.0
     304
           305
                           infinite
                                                Videojuegos, Humor
                                                                     21700000.0
     315
           316
                    lyricalemonade
                                         Música y baile, Animación
                                                                     21100000.0
           319
                         kurzgesagt
                                                          Educación
                                                                     21100000.0
     319
           320
                                                     Música y baile 21100000.0
                            Visits
                                     Likes Comments \
                 Country
     14
                            969700 180300
                                                 7400
                   India
                           5300000
                                     156500
                                                 4200
     26
          Estados Unidos
     34
          Estados Unidos
                            4300000
                                     300400
                                                15000
     37
          Estados Unidos
                           1100000
                                      85800
                                                 3800
     39
                  México
                           2000000
                                     117100
                                                 3000
     58
                  México
                            2200000
                                     183400
                                                 1800
     62
                  México
                            5300000
                                     271300
                                                 16000
     70
               Indonesia
                            1300000
                                                 1600
     96
                   India
                            1500000
                                     129400
                                                 4900
     100
         Estados Unidos
                           2100000
                                     126500
                                                 3800
                                                 2100
          Estados Unidos
                            1200000
                                      34600
     109
     145
         Estados Unidos
                           1600000
                                      83400
                                                 2300
                           1200000
                                      58500
                                                 4000
     171
                   India
          Estados Unidos
     177
                            3500000
                                     285000
                                                 52500
     179
          Estados Unidos
                            6400000
                                     154100
                                                 5000
     180
          Estados Unidos
                            1500000
                                      85800
                                                 1600
          Estados Unidos
                            1600000
                                      98300
                                                 7600
     195
     202
          Estados Unidos
                            1300000
                                      56500
                                                 1100
               Argentina
                            5700000
                                     208400
     207
          Estados Unidos
                            2600000
                                     127300
                                                 2200
                            3200000
                                      85300
                                                 5100
     234
          Estados Unidos
     241
                  México
                            3000000
                                     133200
                                                 3600
         Estados Unidos
                            964500
                                      62300
                                                 1100
     243
\hbox{\#Display information about the top-performing streamers}\\
top_performing_streamers = above_avg_streamers.sort_values(by=['Suscribers'],ascending=True)
print("Top Performing Streamers:")
print(top_performing_streamers[['Username', 'Suscribers', 'Visits', 'Likes','Comments']])
→ Top Performing Streamers:
                    Username Suscribers
                                             Visits
                                                       Likes
                                                             Comments
     319
                       romeo 21100000.0
                                            3200000
                                                       53900
                                                                  1600
     315
              lyricalemonade
                              21100000.0
                                            2800000
                                                      127300
                                                                  5800
                              21100000.0
                                            4900000
                                                      253500
                                                                 14000
     318
                  kurzgesagt
     304
                    infinite
                               21700000.0
                                             884800
                                                       45700
                                                                  1400
                  royaltyfam 21900000.0
                                            4700000
                                                       67000
                                                                  6600
     285
                  BenAzelart
                              22500000.0
                                            3700000
                                                       44900
                                                                  2700
                     SSundee
                              22700000.0
                                            1700000
                                                       59800
                                                                  1800
                                                     235000
     278
                              22700000.0
                                           11700000
                                                                 10000
                 StokesTwins
               AmiRodrigueZZ
     272
                              22900000.0
                                            4300000
                                                     294400
                                                                  1300
                JamesCharles
     243
                              23900000.0
                                             964500
                                                       62300
                                                                  1100
     241
          juandediospantojaa
                              24000000.0
                                            3000000
                                                     133200
                                                                  3600
                         rug
     234
                              24300000.0
                                            3200000
                                                       85300
                                                                  5100
     207
                       ZHCYT
                              25700000.0
                                            2600000
                                                     127300
                                                                  2200
     206
                              25700000.0
                                            5700000
                                                     208400
                                                                  1700
                   AlejoIgoa
     202
                                            1300000
                                                       56500
                                                                  1100
                VanossGaming
                              25900000.0
     195
                  nickiminaj
                              26100000.0
                                            1600000
                                                       98300
                                                                  7600
                              27500000.0
                                            1500000
                                                       85800
                                                                  1600
     180
                    NichLmao
     179
                                            6400000
                                                     154100
                                                                  5000
                              27600000.0
                 brentrivera
                                            3500000
                              27800000.0
                                                     285000
                                                                 52500
     177
                      DanTDM
             SandeepSeminars
                                            1200000
                                                                  4000
     171
                              28000000.0
                                                       58500
                                            1600000
     145
               jacksepticeye
                              30400000.0
                                                       83400
                                                                  2300
     109
                SSSniperWolf
                               34200000.0
                                            1200000
                                                       34600
                                                                  2100
     100
                  markiplier
                              35500000.0
                                            2100000
                                                     126500
                                                                  3800
     96
              TotalGaming093
                              36300000.0
                                            1500000
                                                     129400
                                                                  4900
                 JessNoLimit
                                                      73500
     70
                              39600000.0
                                            1300000
                                                                  1600
     62
              KimberlyLoaiza
                               42100000.0
                                            5300000
                                                      271300
                                                                 16000
                   Mikecrack
                                            2200000
                                                     183400
     58
                              43400000.0
                                                                  1800
     39
                              48600000.0
                                            2000000
                                                     117100
                                                                  3000
                 JuegaGerman
     37
                ArianaGrande
                               52900000.0
                                            1100000
                                                       85800
                                                                  3800
                                                     300400
                                                                 15000
     34
                 TavlorSwift
                              54100000.0
                                            4300000
                               59700000.0
                                                     156500
     26
                 dudeperfect
                                            5300000
                                                                  4200
                              76500000.0
     14
                         BTS
                                             969700 180300
                                                                  7400
#Distribution of Categories
#To calculate category counts
category_counts = df['Categories'].value_counts()
print(category_counts)
    Categories
                                            156
     Música y baile
     Películas, Animación
                                             59
     Música y baile, Películas
                                             41
     Noticias y Política
                                             35
     Vlogs diarios
                                             34
```

Películas, Humor Animación, Videojuegos

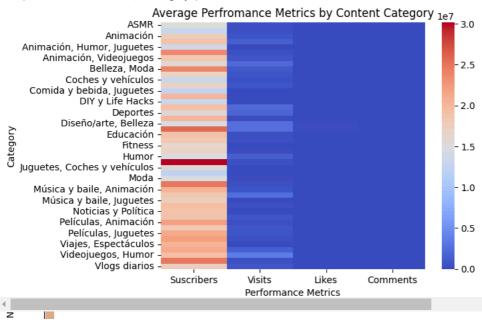
Animación, Juguetes

```
Animación, Humor
                                       24
Películas
                                       23
Educación
                                       22
Animación
                                       22
Videojuegos
                                       18
Música y baile, Animación
                                       16
Videojuegos, Humor
                                       15
Ciencia y tecnología
                                       13
Comida y bebida
                                       10
Películas, Juguetes
                                       9
Juguetes
Humor
                                       9
Deportes
                                       8
Películas, Videojuegos
Música y baile, Humor
Juguetes, Coches y vehículos
DIY y Life Hacks
Fitness, Salud y autoayuda
                                       3
Videojuegos, Juguetes
                                       3
Animales y mascotas
Moda
Coches y vehículos
Educación, Juguetes
                                       2
Fitness
Comida y bebida, Juguetes
ASMR, Comida y bebida
Animación, Humor, Juguetes
Diseño/arte, Belleza
                                       1
Belleza, Moda
ASMR
Música y baile, Juguetes
Diseño/arte, DIY y Life Hacks
                                       1
                                       1
DIY y Life Hacks, Juguetes
Diseño/arte
Comida y bebida, Salud y autoayuda
Viajes, Espectáculos
                                       1
Juguetes, DIY y Life Hacks
Name: count, dtype: int64
```

```
#Create a bar plot to visualize the number of streamers in each category
plt.figure(figsize=(10,8))
sns.barplot(x=category_counts.index, y=category_counts.values, palette="flare")
plt.title("Streamers in each content category")
plt.xlabel("Category")
10
plt.ylabel("Number of streamers")
plt.xticks(rotation=90)
```

```
<ipython-input-48-441f3c4bace1>:3: FutureWarning:
     Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `x` variable to `hue` and set
       sns.barplot(x=category_counts.index, y=category_counts.values, palette="flare")
     ([0,
       1,
       2,
       3,
       4,
       8,
       9,
       10.
       11.
       12.
       13,
       14,
       15,
       16,
       17,
       18,
       19,
       20.
       21.
       22.
       23,
       24,
       25,
       26,
       27,
       29,
       30.
       31.
       32,
       33,
       34,
       35,
       36,
       37,
       38,
       39,
       40,
       41,
       42,
       43.
       44],
      Text(0, 0, 'Música y baile'),
Text(1, 0, 'Películas, Animación'),
Text(2, 0, 'Música y baile, Películas'),
       Text(3, 0, 'Noticias y Política'),
       Text(4, 0, 'Vlogs diarios'),
Text(5, 0, 'Películas, Humor'),
       Text(6, 0, 'Animación, Videojuegos'),
#Identify the category with the highest number of streamers
max_category = category_counts.idxmax()
max_count = category_counts.max()
12
print(f"The category with the highest number of streamers is '{max_category}'_with '{max_count}' streamers")
The category with the highest number of streamers is 'Música y baile'_with '156' streamers
                                                                                                                                                 Text(17 0 'Películas Juguetes')
#Calculate average metrics by category
ambc= df.groupby('Categories')[['Suscribers', 'Visits', 'Likes', 'Comments']].mean()
#Visualize the average performance metrics by category
sns.heatmap(ambc, cmap="coolwarm", fmt=".2f")
plt.title("Average Perfromance Metrics by Content Category")
plt.xlabel("Performance Metrics")
plt.ylabel("Category")
```

Text(50.22222222222, 0.5, 'Category')

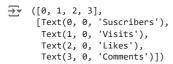


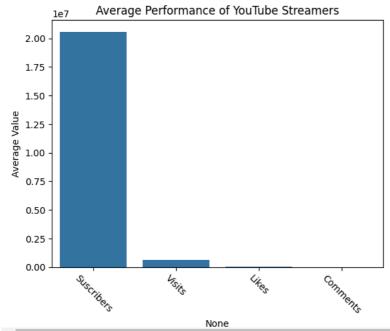
#Performance Metrics

To caculate the average metrics
average_metrics= df[['Suscribers', 'Visits', 'Likes', 'Comments']].mean()
print(average_metrics)

Suscribers 2.057387e+07
Visits 6.497554e+05
Likes 2.414344e+04
Comments 1.016651e+03
dtype: float64

#Barplot of average metrics
sns.barplot(x=average_metrics.index, y=average_metrics.values)
plt.title("Average Performance of YouTube Streamers")
plt.ylabel("Average Value")
plt.xticks(rotation=-45)





model = LinearRegression()

```
#For patterns in the metrics
sns.boxplot(data=df[['Suscribers', 'Visits', 'Likes', 'Comments']],palette="Blues")
plt.title("Box Plots of Performance")
plt.ylabel("Values")
plt.xticks(rotation=-45)
([0, 1, 2, 3],

[Text(0, 0, 'Suscribers'),

Text(1, 0, 'Visits'),

Text(2, 0, 'Likes'),

Text(3, 0, 'Comments')])
                                 Box Plots of Performance
                     0
                     000
          7
          6
          5
          4
          3
          2
          1
          0
#Content Recommendations
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import mean_absolute_error, mean_squared_error, r2_score
from sklearn.preprocessing import LabelEncoder
from sklearn.linear_model import LinearRegression
df['user_id']=range(1,len(df['Username'])+1)
x=df[['Rank', 'Visits', 'Comments']]
y=df['user_id']
x_train,x_test, y_train, y_test= train_test_split(x,y,test_size=0.3)
```