App Rating Prediction

Objective: - Make a model with other information ab

```
In [74]: import numpy as np
         import pandas as pd

         import seaborn as sns import
         matplotlib.pyplot as plt

         from sklearn.linear_model import LinearRegression
         from sklearn.model_selection import
         train_test_split

         import os import warnings
         warnings.filterwarnings('ignore'
         )
```

```
In [75]: df = pd.read_csv('googleplaystore.csv')
```

```
In [76]: df.head()
```

Out[76]:

	App	Category	Rating	Reviews	Size	Installs	Type	Price	Con Ra
0	Photo Editor & Candy Camera & Grid & ScrapBook	ART_AND_DESI4G.N1 15919M 10,00F0r+ee 0Eve							
1	Coloring bookART_AND_DESI3G.N9 96714M 500,00F0r+ee 0Eve moana								
2	U Launcher Lite – FREE LiveART_AND_DESI4G.N7875180.7M5,000,0F0r0e+e 0Eve Cool Themes, Hide ...								
3	Sketch - Draw &ART_AND_DESI4G.N521564245M50,000,F0r0e0e+ 0 Paint								
4	Pixel Draw - Number Coloring Book	ArtART_AND_DESI4G.N3 9672.8M100,00F0r+ee 0Eve							

```
In [77]: print(f'Number of rows : {df.shape [0]}')
         print(f'Number of columns : { df.shape[1]}')

 Number of rows : 10841
 Number of columns : 13

In [78]: df.info()

 <class 'pandas.core.frame.DataFrame'>
 RangeIndex: 10841 entries, 0 to 10840
 Data columns (total 13 columns):
  #   Column          Non-Null Count  Dtype
 ---  ------          --------------  -----
 0   App             10841 non-null  object
 1   Category        10841 non-null  object
 2   Rating          9367 non-null   float64
 3   Reviews         10841 non-null  object
 4   Size            10841 non-null  object
 5   Installs        10841 non-null  object
 6   Type            10840 non-null  object
 7   Price           10841 non-null  object
 8   Content Rating  10840 non-null  object
 9   Genres          10841 non-null  object
 10  Last Updated    10841 non-null  object
 11  Current Ver     10833 non-null  object  12  Android Ver     10838 non-null
     object dtypes: float64(1), object(12) memory usage: 1.1+ MB

In [79]: df.duplicated().sum()

         print(f"DataFrame has {df.duplicated().sum()} duplicate values")

 DataFrame has 483 duplicate values

In [80]: df.drop_duplicates(inplace=True) print(f" Total duplicate

         values : {df.duplicated().sum()}")

  Total duplicate values : 0

In [81]: df.isnull().sum()

Out[81]: App                0 Category
         0
         Rating           1465
         Reviews             0
         Size                0
         Installs            0
         Type                1
         Price               0
         Content Rating      1
         Genres              0
         Last Updated        0
         Current Ver         8
         Android Ver         3

         dtype: int64
```

```
In [82]: ## Drop records with nulls in any of the columns.

         df.dropna(inplace=True)

In [83]: df.isnull().sum()

Out[83]: App              0 Category
         0
         Rating            0
         Reviews           0
         Size              0
         Installs          0
         Type              0
         Price             0
         Content Rating    0
         Genres            0
         Last Updated      0
         Current Ver       0
         Android Ver       0
         dtype: int64

In [84]: df.shape

Out[84]: (8886, 13)
In [85]: # # Variables seem to have incorrect type and inconsistent
formatting.
       # Size column has sizes in Kb as well as Mb.
       # To analyze, you'll need to convert these to numeric.
       # Extract the numeric value from the column and Multiply the value by


       def size_col_processing(x):
           x= str(x.lower())
       if 'm' in x:

               val=float(x.replace('m',
       ''))          val=val*1000
                   elif
       'k'in x:
               val=float(x.replace('k',''))

       else:
       val=0
       return val


In [86]: df['Size']=df['Size'].apply(size_col_processing)
         df.head()
```

Out[86]:

	App	Category	Rating	Reviews	Size	Installs	Type	Price	C
0	Photo Editor & Candy Camera & Grid & ScrapBook	ART_AND_DESI4G.N1	15919000.010,00F0r+ee						0E

1 Coloring bookART_AND_DESI3G.N9 96714000.0500,00F0r+ee 0E moana

U
Launcher
2 Lite –
FREE LiveART_AND_DESI4G.N7875108700.50,000,000+Free 0E Cool
Themes, Hide ...

Sketch -
3 Draw &ART_AND_DESI4G.N521564245000.050,000,F0r0e0e+ 0
Paint

4 Pixel Draw
- Number
ArtART_AND_DESI4G.N3 9672800.0100,00F0r+ee 0E

Coloring
Book

```
In
[[85]: df['Price']= df['Price'].apply(lambda x :str(x).replace('$','')if '$'
       df['Price']= df['Price'].apply(lambda x : float(x))
       df['Reviews']=pd.to_numeric(df['Reviews'], errors ='coerce')
In [[85]: df['Installs']=df['Installs'].apply(lambda x :
str(x).replace('+','')
      df['Installs']=df['Installs'].apply(lambda x : str(x).replace(',','')
      df['Installs']=df['Installs'].apply(lambda x : float(x))
In [89]: df.info()

 <class 'pandas.core.frame.DataFrame'>
 Int64Index:  8886  entries,  0  to  10840
 Data columns (total 13 columns):
  #   Column          Non-Null Count  Dtype
 ---  ------          --------------  -----
 0   App             8886 non-null   object
 1   Category        8886 non-null   object
 2   Rating          8886 non-null   float64
 3   Reviews         8886 non-null   int64
 4   Size            8886 non-null   float64
 5   Installs        8886 non-null   float64
 6   Type            8886 non-null   object
 7   Price           8886 non-null   float64
 8   Content Rating  8886 non-null   object
 9   Genres          8886 non-null   object
 10  Last Updated    8886 non-null   object
 11  Current Ver     8886 non-null   object  12  Android Ver     8886 non-null
     object dtypes: float64(4), int64(1), object(8) memory usage: 971.9+ KB

In [90]: df.describe()
```

Out[90]:

	Rating	Reviews	Size	Installs	Price
count	8886.00	80.080806000e+03	8886.00	80.080806000e+03	8886.00000
mean	4.18	7945.9730928e+19	000.65	15.961590061e+075	0.963526
std	0.52	2422.8906007e+23	023.41	88.668460413e+076	16.194792
min	1.00	0010.0000000e+00	0.00	0010.0000000e+00	0.000000
25%	4.00	0010.0640000e+02	2500.00	10.000000000e+04	0.000000

	50%	4.300040.0723000e+039400.0050.000000000e+050.000000								
	75%	4.500070.0131325e+27000.0050.000000000e+064								0.000000
	max	5.000070.0815831e+07100000.010.0000000e+400.0000009								

```
In [[85]: # Reviews should not be more than installs as only those who
          installe # If there are any such records, drop them.
          df['review_check']=df['Reviews']>df['Installs']
```

```
In [92]: df.shape
```

```
Out[92]: (8886, 14)
```

```
In [93]: df[df['review_check']==True].head(2)
```

Out[93]:

	App	Category	Rating	Reviews	Size	Installs	Type	Price	Content Rating	Ge Rating
2454	KBA-EZ Health Guide	MEDICAL	5.0	425000.0	1.0M	Free	0.0	E	0veryone	Me
4663	Alarmy (Sleep If Can) - Pro	LIFESTYLE	4.8	10249	0.010000M	Paid	2.49	E	9veryone	Li

```
In [94]: df=df[df['review_check']== False]
         df.shape
```

```
Out[94]: (8879, 14)
```

```
In [95]: df['review_check'].unique()
```

```
Out[95]: array([False])
```

```
In [96]: df.drop('review_check',axis=1,inplace=True)
         df.head(1)
```

Out[96]:

	App	Category	Rating	Reviews	Size	Installs	Type	Price	Cont Rat
0	Photo Editor & Candy Camera & Grid & ScrapBook	ART_AND_DESI	4.1	159	19000	1.0	0.0	Free	0.0	Eve

```
In [9[85]: # For free apps (type = "Free"), the price should not be >0.
           Drop an df[(df['Type']=='Free')&(df['Price']>0)]
```

Out[98]:

App	Category	Rating	Reviews	Size	Installs	Type	Price	Content Rating	Genres	Las Updated

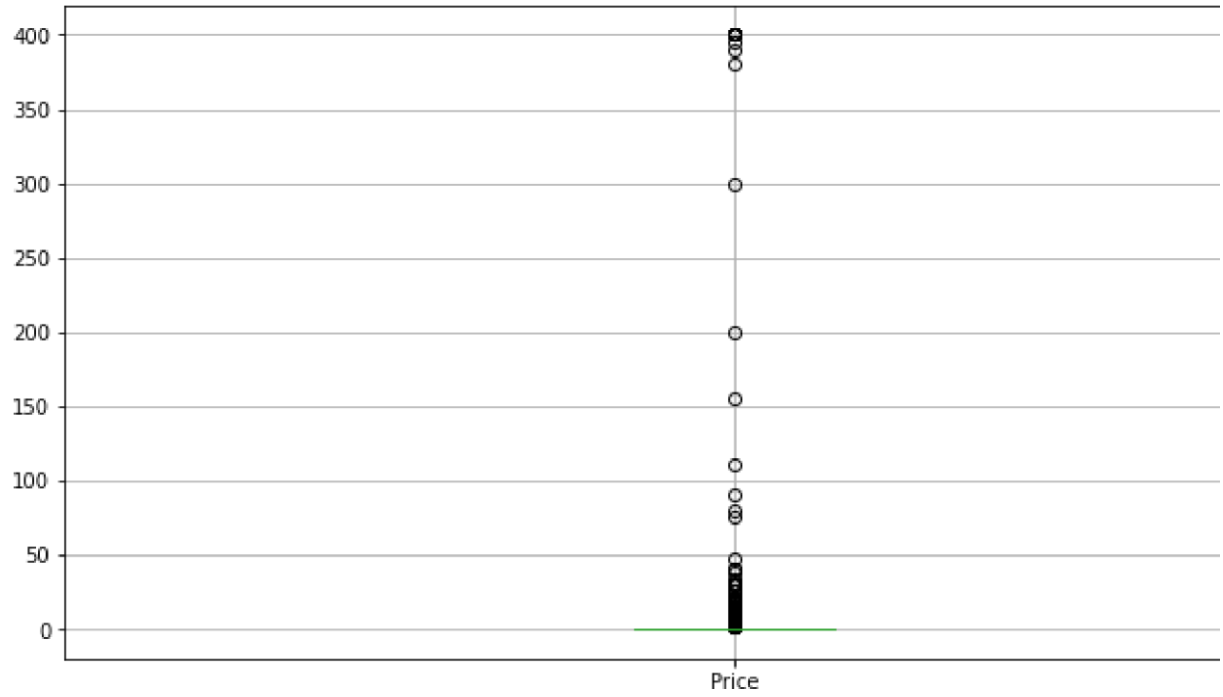
In [174]:
```
# ~ is used to Negate/reverse the df selected using condition

df=df[~((df['Type']=='Free')& (df['Price']>0))]
df.shape
```

Out[174]: (6981, 13)

In [34]:
```
plt.figure(figsize=(12,6))
df.boxplot(column ='Price')
```

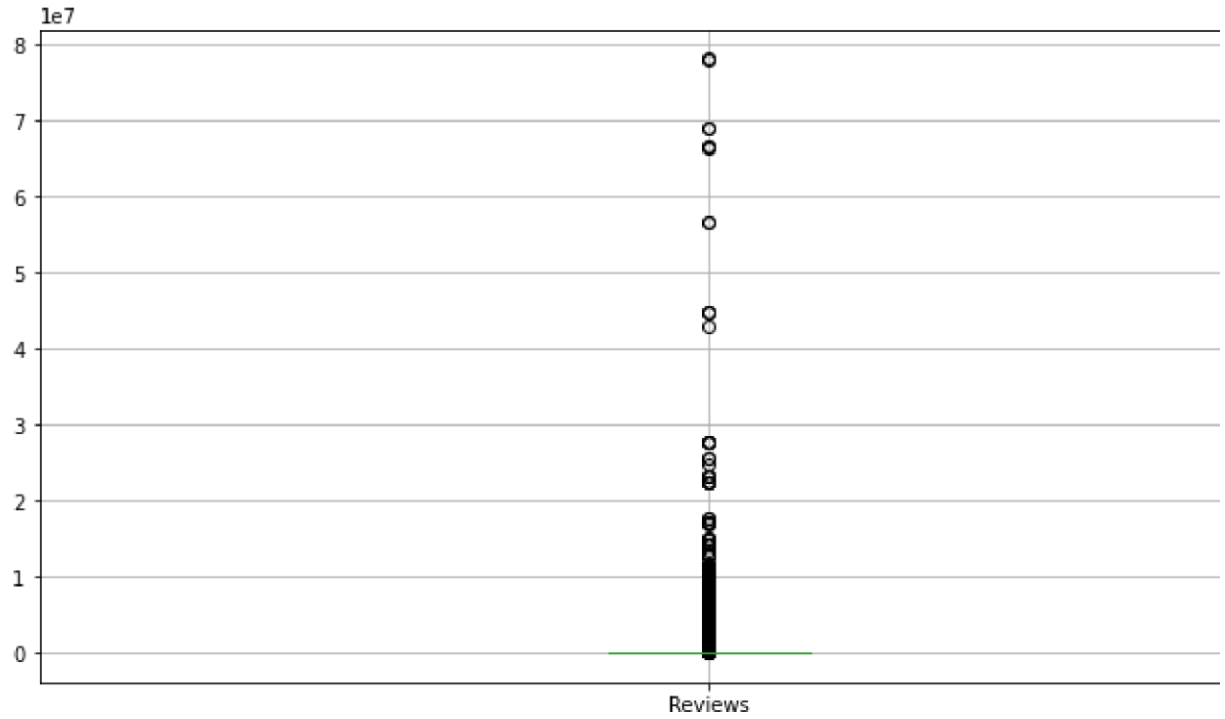
Out[34]: <AxesSubplot:>



In [[85]:
```
# outlier are present in dataset , anything above than 300 will
be co
```

In [37]:
```
plt.figure(figsize=(12,6))
df.boxplot('Reviews')
```

Out[37]: <AxesSubplot:>

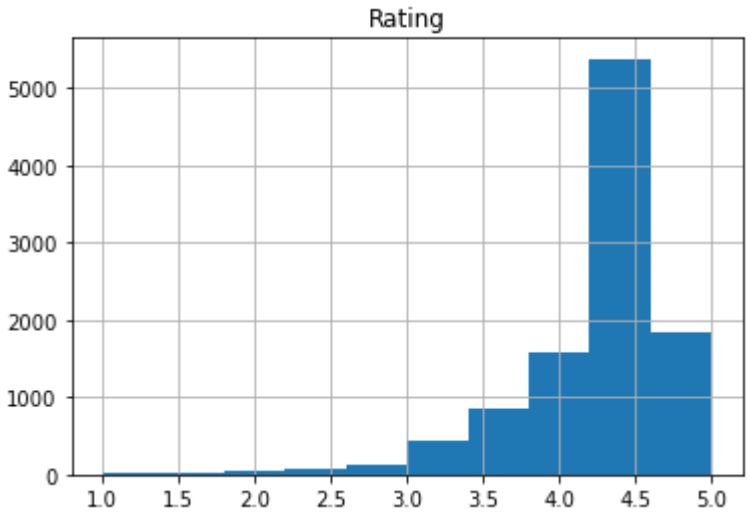


In [[85]: # values above than 3 to le7 are the outliers in box plot shown abov

In [43]:
```
plt.figure(figsize=(12,6))
df.hist('Rating')
```

Out[43]: array([[<AxesSubplot:title={'center':'Rating'}>]], dtype=object)

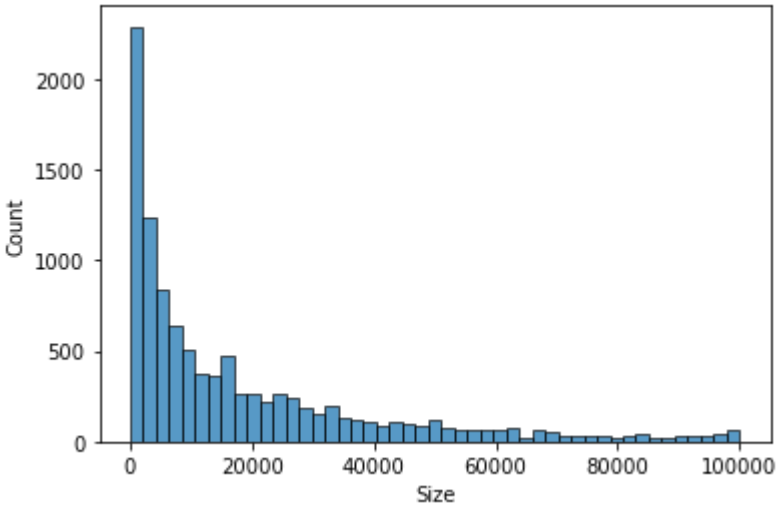
<Figure size 864x432 with 0 Axes>



In [49]: # Histogram for Size

```
sns.histplot(df['Size'])
```

Out[49]: <AxesSubplot:xlabel='Size', ylabel='Count'>



In [50]: # Most(50%) of the apps are below 20MB of size.

In [5[85]: # From the box plot, it seems like there are some apps with very high # A price of $200 for an application on the Play Store is very high a # Check out the records with very high price.
Is 200 indeed a high price?

```
df=df[df['Price']>200]
df
```

Out[51]:

	App	Category	Rating	Reviews	Size	Installs	Type	Price	Con Ra
4197	most expensiveFAMILY4.3 6 1500.0100.P0aid399.9E9ve app (H)								
4362	I'LmI FrEiScThYLE3.8 71826000.100000.0Paid399.9E9ve								
4367	I'm Rich - TrumpLIFESTYLE3.6 2757300.100000.0Paid400.0E0ve Edition								
5351	I am LrIiFcEhSTYLE3.8 35471800.100000.0Paid399.9E9ve								
5354	I am Rich FAMILY4.0 8568700.100000.0Paid399.9E9ve Plus								
5355	I am riLcIhF EVSITPYLE3.8 4112600.100000.0Paid299.9E9ve								
5356	I Am Rich FINANCE4.1 18674700.500000.0Paid399.9E9ve Premium								
5357	I am extremelyLIFESTYLE2.9 Rich				412900.01000.P0aid379.9E9ve				
5358	I am RiFcIhN!ANCE3.8				9322000.10000.0Paid399.9E9ve				
5359	I am FINANCE3.5 rich(premium)				472 965.05000.P0aid399.9E9ve				
5362	I Am RichF APMrIoLY4.4				2012700.05000.P0aid399.9E9ve				

I am rich (Most expensive app)
5364 FINANCE4.1 1292700.01000.P0aid399.99T

5366 I Am RicFhAMILY3.6 2174900.100000.0Paid389.9E9ve

5369 I am RiFcIhNANCE4.3 1803800.05000.P0aid399.9E9ve

I AM RICH PRO PLUS
5373 FINANCE4.0 3641000.10000.0Paid399.9E9ve

9917 Eu Sou RFiIcNoANCE4.4 0 1400.0 0.0Paid394.9E9ve

I'm Rich/Eu
9934 sou Rico/LIFESTYLE4.4
040000.00.0Paid399.9E9ve

In [100]:
```
# Yes $200 indeed  is a high price. #
Drop these as most seem to be junk apps

df=df[df['Price']<200]
df['Price'].unique()
```

Out[100]: array([0.   ,
4.99, 3.99, 6.99,
7.99, 5.99, 2.99,
3.49,
1.99,
 9.99, 7.49, 0.99, 9. , 5.49, 10. , 24.99, 11.99,
79.99, 16.99, 14.99, 29.99, 12.99, 2.49, 10.99, 1.5
, 19.99, 15.99, 33.99, 39.99, 3.95, 4.49, 1.7 ,
8.99, 1.49, 3.88,
17.99,
 3.02, 1.76, 4.84, 4.77, 1.61, 2.5 , 1.59, 6.49,
1.29,
 37.99, 18.99, 8.49, 1.75, 14. , 2. , 3.08, 2.59,
19.4 ,
 3.9 , 4.59, 15.46, 3.04, 13.99, 4.29, 3.28, 4.6 ,
1. ,
 2.95, 2.9 , 1.97, 2.56, 1.2])

In [85]:
```
# Reviews: Very few apps have very high number of reviews. These
are a # in fact, will skew it. Drop records having more than 2
million revie

df=df[df['Reviews']<2000000]
df.head(2)
```

Out[101]:

	App	Category	Rating	Reviews	Size	Installs	Type	Price	Co R
0	Photo Editor & Candy Camera & Grid & ScrapBook	ART_AND_DESI4G.N1 15919000.100000.0Free0.0Ev							
1	Coloring book moana	ART_AND_DESI3G.N9 967140005.000000.0Free0.0Ev							

```
In [1[85]: # Installs:  There seems to be some outliers in this field too.
       # Find out the different percentiles - 10, 25, 50, 70, 90, 95, 99
        # Decide a threshold as cutoff for outlier and drop records having
        v


       df.Installs.quantile([0.10, 0.25, 0.50, 0.70, 0.90, 0.95, 0.99])
```

```
Out[102]: 0.10         1000.0
          0.25        10000.0
          0.50       100000.0
          0.70      1000000.0
          0.90     10000000.0
          0.95     10000000.0
          0.99    100000000.0
          Name: Installs, dtype: float64
```

```
In [[85]: # Keeping 95% value as a threshold/cutoff for outlier and drop
record

       df=df[df['Installs']<10000000.0]
       df.head(2)
```

Out[103]:

	App	Category	Rating	Reviews	Size	Installs	Type	Price	Co R
0	Photo Editor & Candy	ART_AND_DESI4G.N1			15919000.100000.0Free0.0Ev				
1	Camera & Grid & ScrapBook Coloring bookART_AND_DESI3G.N9 moana				967140005.000000.0Free0.0Ev				

```
In [104]: df.shape
```

```
Out[104]: (6981, 13)
```

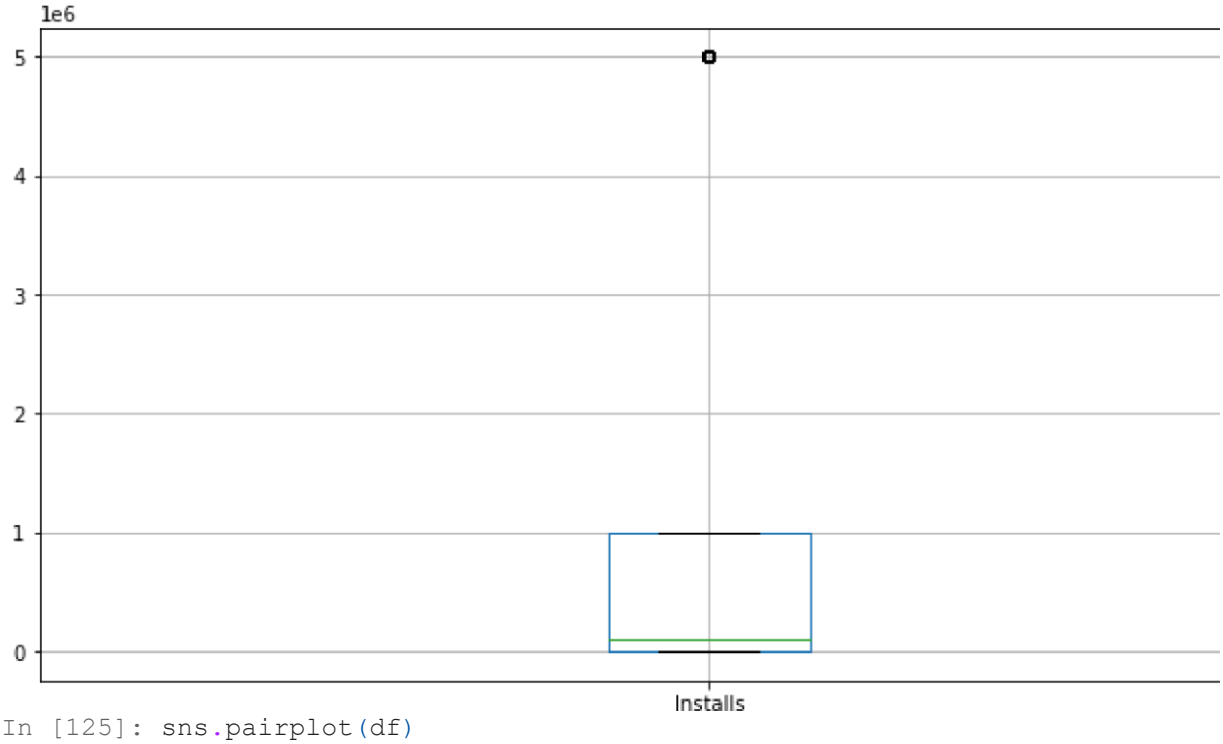
```
In [[85]: # Make scatter plot/joinplot for Rating vs. Price
      # What pattern do you observe? Does rating increase with price?
      # Yes, it is showing positive correlation as the price increasing
      Rati

      # Make scatter plot/joinplot for Rating vs.
      Size # Are heavier apps rated better?
      # No relation as we can see everyone is downloading any size of the
      ap

      # Make scatter plot/joinplot for Rating vs.
      Reviews # Does more review mean a better rating
      always?
      # Apps which are having higher ratings
      # The app which are having higher rating are getting somewhat of a
      mor # Most of the ratings are on the higher end side of the ratings.
```

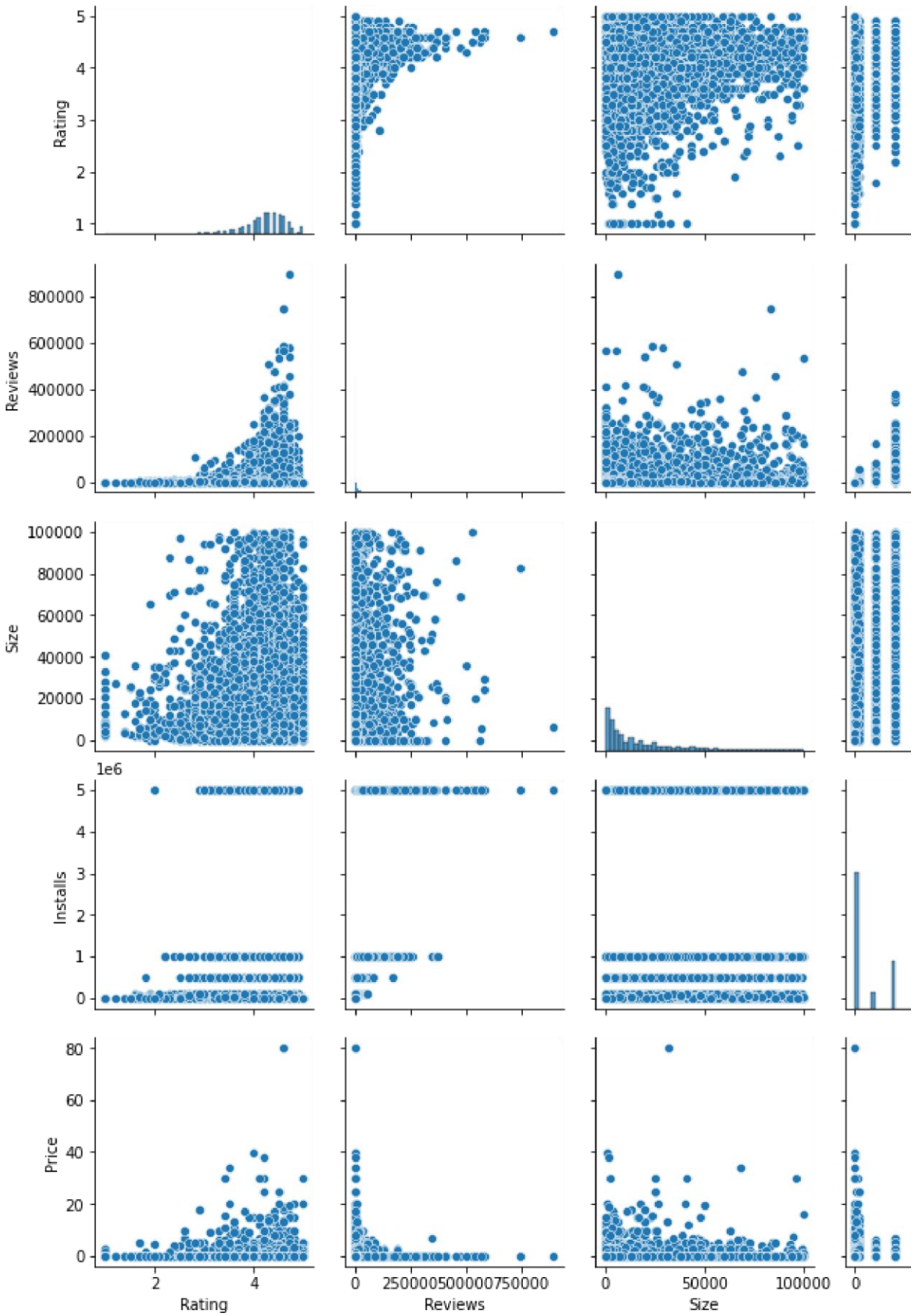
In [105]:
```
plt.figure(figsize=(12,6))
df.boxplot('Installs')
```

Out[105]: <AxesSubplot:>



In [125]:
```
sns.pairplot(df)
```

Out[125]: <seaborn.axisgrid.PairGrid at 0x1832d8b8310>



In [106]: ## value count of top most app on google

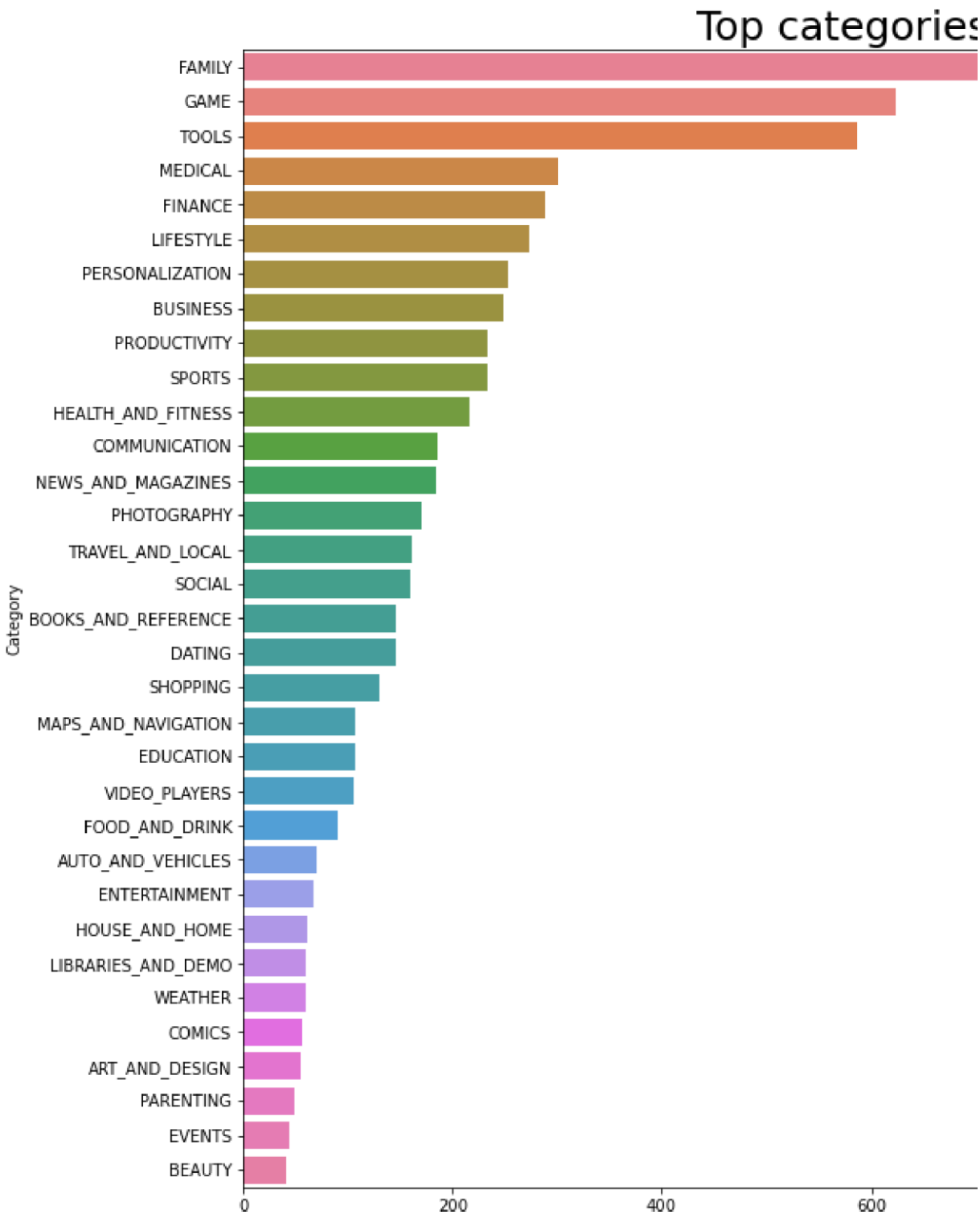
```
x = df['Category'].value_counts() y =
df['Category'].value_counts().index

x_axis =[] y_axis = []
for i in
range(len(x)):
x_axis.append(x[i])
y_axis.append(y[i])
```

In [107[85]:

```
plt.figure(figsize=(18,13))
plt.xlabel("Count")
plt.ylabel("Category")

graph = sns.barplot(x = x_axis, y = y_axis, palette= "husl")
graph.set_title("Top categories on Google Playstore", fontsize = 2
```


Top categories
FAMILY
GAME
TOOLS
MEDICAL
FINANCE
LIFESTYLE
PERSONALIZATION
BUSINESS
PRODUCTIVITY
SPORTS
HEALTH_AND_FITNESS
COMMUNICATION
NEWS_AND_MAGAZINES
PHOTOGRAPHY
TRAVEL_AND_LOCAL
SOCIAL
BOOKS_AND_REFERENCE
DATING
SHOPPING
MAPS_AND_NAVIGATION
EDUCATION
VIDEO_PLAYERS
FOOD_AND_DRINK
AUTO_AND_VEHICLES
ENTERTAINMENT
HOUSE_AND_HOME
LIBRARIES_AND_DEMO
WEATHER
COMICS
ART_AND_DESIGN
PARENTING
EVENTS
BEAUTY
Category
0
200
400
600

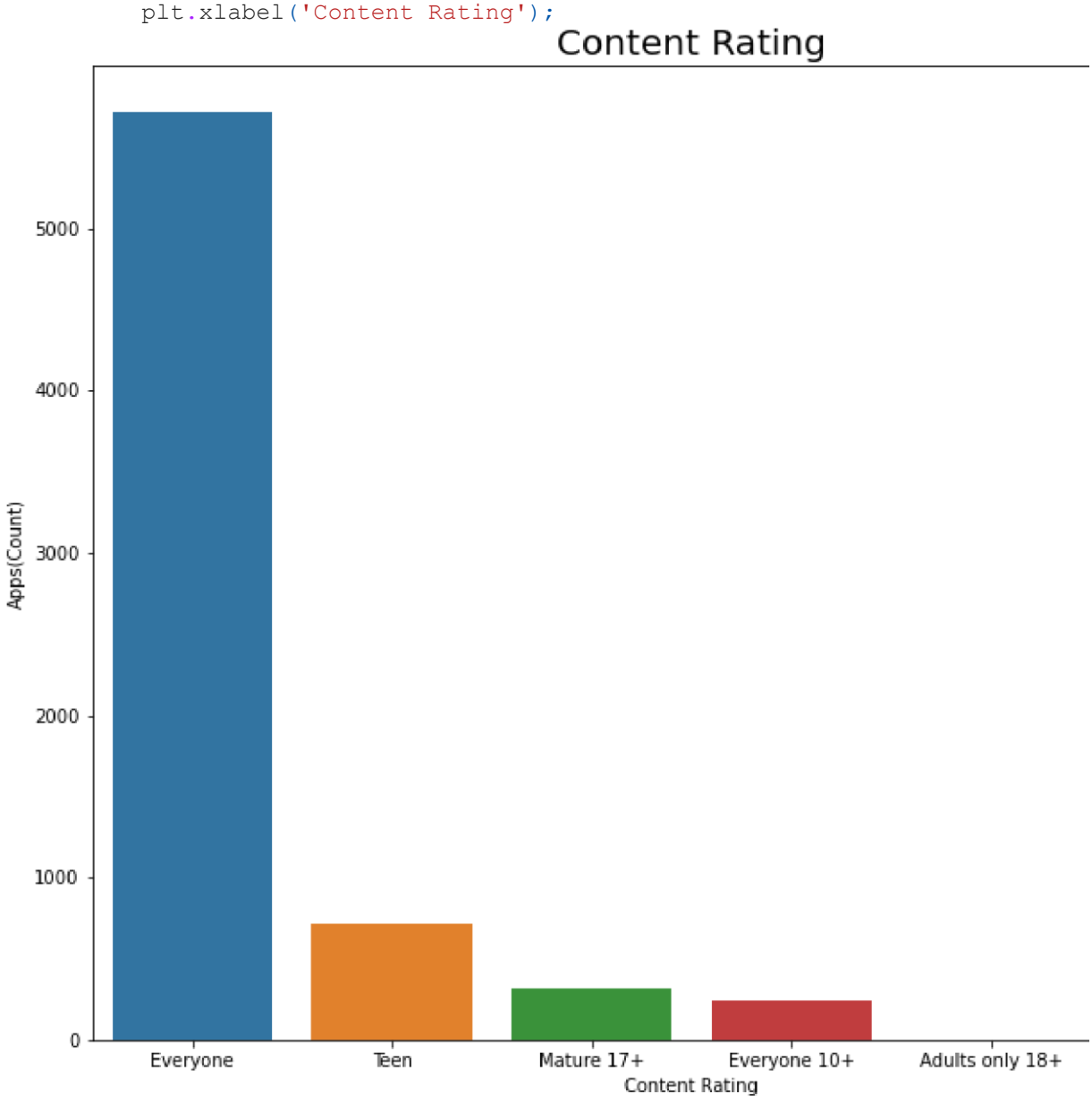
In [126]:
```
## Bar chart for Content Rating

x1 = df['Content Rating'].value_counts().index
y1 = df['Content Rating'].value_counts()

x1_axis = []
y1_axis = []

for i in range(len(x1)):
x1_axis.append(x1[i])
y1_axis.append(y1[i])
```

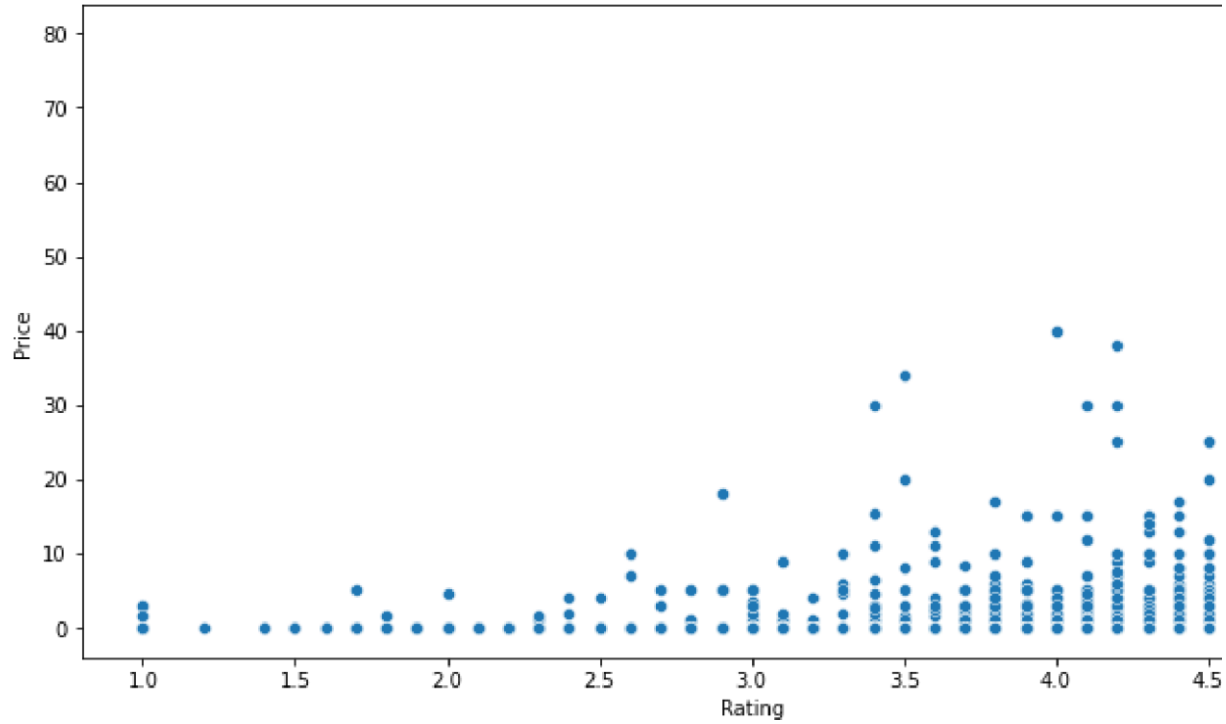
In [110]:
```
plt.figure(figsize=(12,10))
sns.barplot(x= x1_axis, y= y1_axis)
plt.title('Content Rating',size =
20); plt.ylabel('Apps(Count)');
plt.xlabel('Content Rating');
```



In [127]:
```
## Scatter plot for rating Vs Price
```

In [128]:
```
plt.figure(figsize=(12,6))
sns.scatterplot(x='Rating',y='Price',data=df)
```

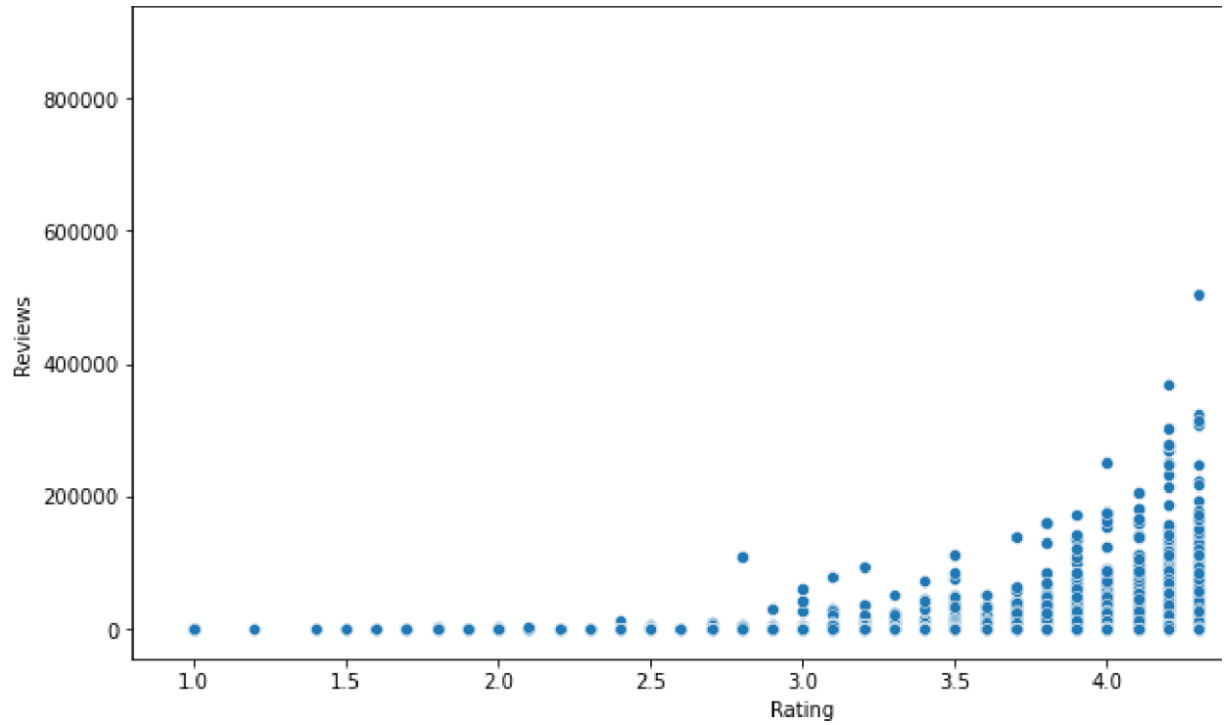
Out[128]: <AxesSubplot:xlabel='Rating', ylabel='Price'>



In [130]:
```
## Scatter plot for rating vs Reviews
```

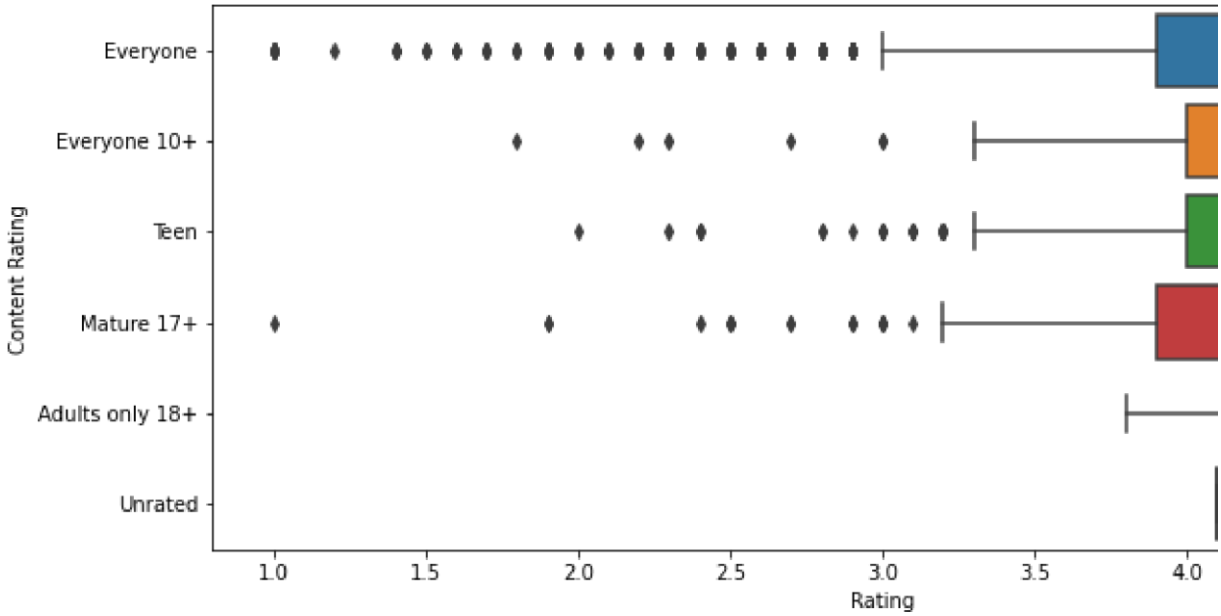
In [131]:
```
plt.figure(figsize=(12,6))
sns.scatterplot(x='Rating',y='Reviews',data=df)
```

Out[131]: <AxesSubplot:xlabel='Rating', ylabel='Reviews'>



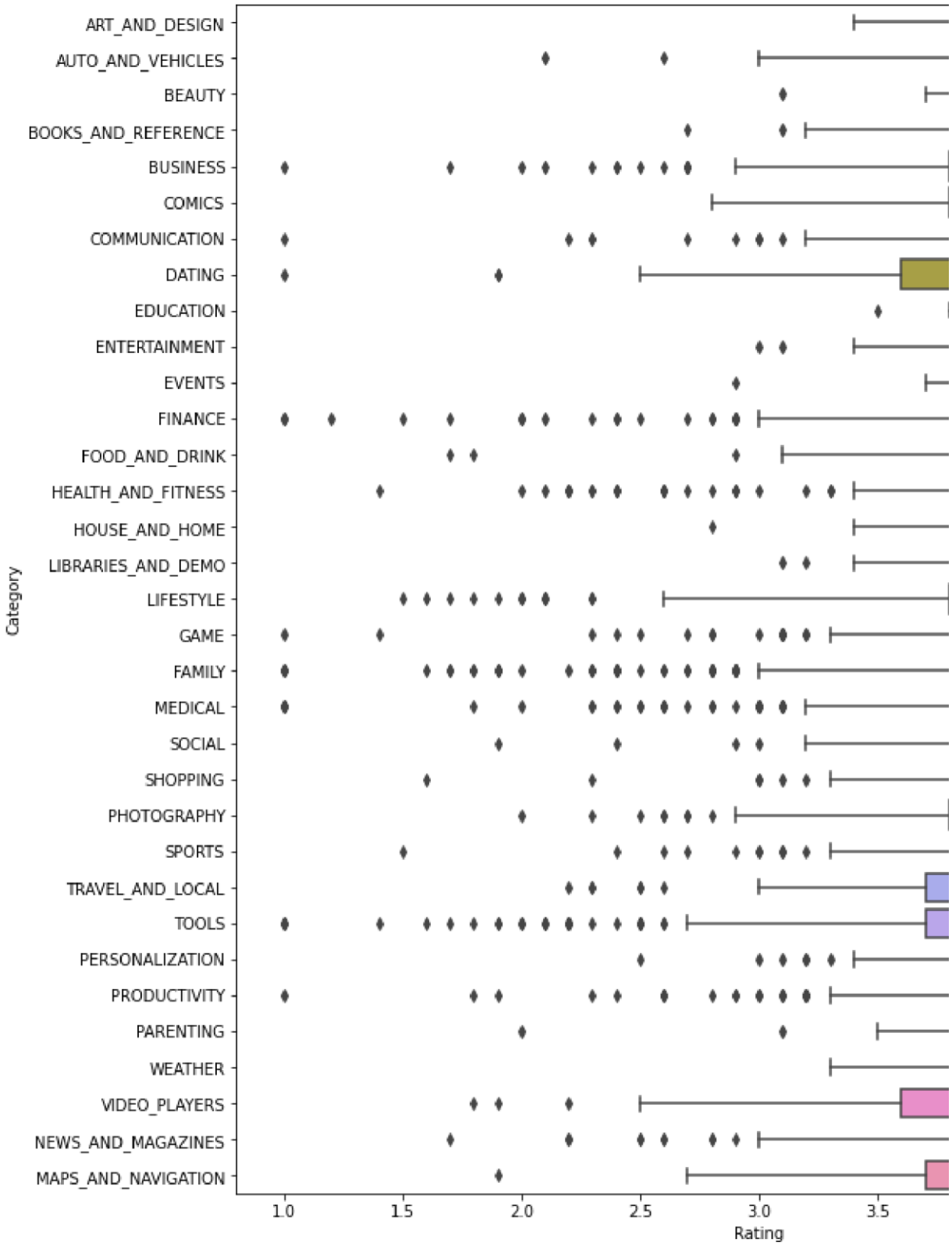
In [134]:
```
plt.figure(figsize=[12,5]) sns.boxplot("Rating",
"Content Rating", data=df )
```

Out[134]: <AxesSubplot:xlabel='Rating', ylabel='Content Rating'>



In [135]:
```
plt.figure(figsize=[12,14])
sns.boxplot("Rating", "Category", data=df )
```

Out[135]: <AxesSubplot:xlabel='Rating', ylabel='Category'>




```
In [85]: # Make boxplot for Ratings vs.
         Category # Which genre has the best
         ratings?
        # Here Q2 (Median) is higher in 'BOOKS_AND_REFERENCES' and 'EVENTS' an
```

```
In [[85]: # For the steps below, create a copy of the dataframe to make all
the
```

In [157]:
```
inp1 =df.copy()
```

In [158]:
```
inp1
```

Out[158]:

	App	Category	Rating	Reviews	Size	Installs	Typ
0	Photo Editor & Candy Camera & Grid & ScrapBook	ART_AND_DESI4G.N1	15919000.100000.Fr				
1	Coloring book	ART_AND_DESI3G.N9	96714000.5000000.Fr	moana			
2	U Launcher Lite – FREE Live Cool Themes, Hide ...	ART_AND_DESI4G.N7	875108700.5000000.Fr				
4	Pixel Draw - Number Art Coloring Book	ART_AND_DESI4G.N3	9672800.1000000.Fr				
5	Paper flowers instructions	ART_AND_DESI4G.N4			1675600.050000F.r		
...	...	...	...	...	...	...	
10833	Chemin (OfOrK)	S_AND_REFERENCE4.8		44	619.01000.F0r		
10834	FR Calculator	FAMILY4.0		7	2600.0500.F0r		
10836	Sya9a Maroc - FR	FAMILY4.5		3853000.05000.F0r			
10837	Fr. Mike Schmitz Audio Teachings	FAMILY5.0		4	3600.0100.F0r		

The SCP 114 0.0 1000.F0r
10839 FoundationBOOKS_AND_REFERE4N.C5E DB fr nn5n

6981 rows × 13 columns

```
In [159]: df['Reviews'].describe()

Out[159]: count      6981.000000
          mean      18564.907606
          std       47341.662556
          min           1.000000
          25%          78.000000
          50%        1213.000000
          75%       15192.000000
          max      896118.000000
          Name: Reviews, dtype: float64
In [160]: ## apply log transformation (np.Log1p) on Reviwes and Installs

In [161]: inp1['Reviews']=np.log1p(inp1['Reviews'])
          inp1['Installs']=np.log1p(inp1['Installs'])

In [162]: inp1.head(1)
```

Out[162]:

	App	Category	Rating	Reviews	Size	Installs	Type	Price	Con	R
0	Photo Editor & Candy Camera & Grid & ScrapBook	ART_AND_DESI	4G.N15.07511794000.09.210F4r4ee0.0Ev							

```
In [163]: ## droping the columns which are not usefull for further working

In [16[85]: inp1.drop(['App','Last Updated','Current Ver','Android Ver'],
        axis=1 inp1.head(2)
```

Out[164]:

	Category	Rating	Reviews	Size	Installs	Type	Price	Content	Rating
0	ART_AND_DESI	4G.N15.07511794000.09.2104F4r0ee0.0EveryonA							
1	ART_AND_DESI	3G.N96.87521342000.013.122F3r6e5e0.0EveryonDe							

```
In [16[85]: # Get dummy columns for Category, Genres, and Content Rating.

        inp2=pd.get_dummies(inp1,columns=['Content Rating','Genres','Categor
        inp2.head(2)

Out[165]:                                                    Content          Content
```

	Rating	Reviews	Size	Installs	Price	Rating_Adults		Rating
							Rating_Everyone only	
						18+		
0	4.1	5.075117	94000.0	9.210440	0.0	0	1	
1	3.9	6.875213	42000.0	13.122306	50	0	1	
2	rows × 156 columns							

```
In [166]: x=inp2.drop('Rating',axis=1)    # independent Variable
          y=inp2['Rating']                 # Dependent Variable

In [167]: from sklearn.model_selection import train_test_split
        In [1[85]: # Train test split  and apply 70-30 split. Name the new

 dataframes d # Separate the dataframes into X_train, y_train, X_test, and

          y_test. x_train, x_test, y_train, y_test= train_test_split(x, y,

                                                                test_size=0


        from sklearn.linear_model import LinearRegression as LR

In [169]: x_train.head(1)
```

Out[169]:

	Reviews	Size	Installs	Price	Content Rating_Adults		Content Rating_Eve
						Rating_Everyone only	
					18+		
9588	7.9	5543295	000.0	13.122306	50	0	1

1 rows × 155 columns

```
In [170]: # Use linear regression as the technique

          model=LR()
          model.fit(x_train, y_train)
```

Out[170]: LinearRegression()

```
In [171]: # Report the R2 on the train set

          model.score(x_train, y_train)
```

Out[171]: 0.15268919030909045

```
In [172]: # Make predictions on test set and report R2.

          model.score(x_test, y_test)
```

Out[172]: 0.11400450481740809

In [173]:
```
model.predict(x_test)
```

Out[173]:
```
array([3.74056292, 4.03368424, 4.14940299, ..., 4.21011289,
       4.29769173,
              4.27070364])
```

In [ ]:

In [ ]:

In [ ]: