

Importing Libraries

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

import warnings
warnings.filterwarnings("ignore")

data=pd.read_csv('telcom_data.csv')
```

data

	ms \	Bearer Id	Start	Start ms	End	End
0	662.0	1.311448e+19	4/4/2019 12:01	770.0	4/25/2019 14:35	
1	606.0	1.311448e+19	4/9/2019 13:04	235.0	4/25/2019 8:15	
2	652.0	1.311448e+19	4/9/2019 17:42	1.0	4/25/2019 11:58	
3	171.0	1.311448e+19	4/10/2019 0:31	486.0	4/25/2019 7:36	
4	954.0	1.311448e+19	4/12/2019 20:10	565.0	4/25/2019 10:40	
...	
149996	214.0	7.277826e+18	4/29/2019 7:28	451.0	4/30/2019 6:02	
149997	187.0	7.349883e+18	4/29/2019 7:28	483.0	4/30/2019 10:41	
149998	810.0	1.311448e+19	4/29/2019 7:28	283.0	4/30/2019 10:46	
149999	327.0	1.311448e+19	4/29/2019 7:28	696.0	4/30/2019 10:40	
150000	NaN	NaN	NaN	NaN	NaN	
NaN						

	Dur. (ms)	IMSI	MSISDN/Number	IMEI \
0	1823652.0	2.082014e+14	3.366496e+10	3.552121e+13
1	1365104.0	2.082019e+14	3.368185e+10	3.579401e+13
2	1361762.0	2.082003e+14	3.376063e+10	3.528151e+13
3	1321509.0	2.082014e+14	3.375034e+10	3.535661e+13
4	1089009.0	2.082014e+14	3.369980e+10	3.540701e+13
...
149996	81230.0	2.082022e+14	3.365069e+10	3.548311e+13
149997	97970.0	2.082019e+14	3.366345e+10	3.566051e+13
149998	98249.0	2.082017e+14	3.362189e+10	3.572121e+13

149999	97910.0	2.082021e+14	3.361962e+10	8.618620e+13
150000	NaN	NaN	NaN	NaN

	Last Location Name	...	Youtube DL (Bytes)	Youtube UL
(Bytes) \				
0	9.16456699548519E+015	...	1.585461e+07	
2.501332e+06				
1	L77566A	...	2.024740e+07	
1.911173e+07				
2	D42335A	...	1.972566e+07	
1.469958e+07				
3	T21824A	...	2.138812e+07	
1.514664e+07				
4	D88865A	...	1.525938e+07	
1.896287e+07				
...	
...				
149996	D20434A	...	1.619167e+07	
1.176343e+07				
149997	D10223C	...	1.387723e+07	
8.288284e+06				
149998	T51102A	...	2.266051e+07	
1.855903e+06				
149999	L88342B	...	8.817106e+06	
8.305402e+06				
150000	NaN	...	1.163407e+07	
1.100941e+07				

	Netflix DL (Bytes)	Netflix UL (Bytes)	Gaming DL (Bytes)	\
0	8.198936e+06	9.656251e+06	2.780823e+08	
1	1.833841e+07	1.722713e+07	6.087501e+08	
2	1.758779e+07	6.163408e+06	2.295846e+08	
3	1.399465e+07	1.097942e+06	7.995382e+08	
4	1.712458e+07	4.152180e+05	5.277072e+08	
...	
149996	1.788370e+07	1.967816e+07	5.266097e+08	
149997	1.935015e+07	2.129315e+07	6.268931e+08	
149998	9.963942e+06	5.065760e+06	5.535395e+08	
149999	3.322253e+06	1.317259e+07	3.525370e+08	
150000	1.162685e+07	1.100175e+07	4.220447e+08	

	Gaming UL (Bytes)	Other DL (Bytes)	Other UL (Bytes)	\
0	1.434415e+07	1.717444e+08	8.814393e+06	
1	1.170709e+06	5.269042e+08	1.505514e+07	
2	3.956300e+05	4.106926e+08	4.215763e+06	
3	1.084972e+07	7.490399e+08	1.279728e+07	
4	3.529801e+06	5.507095e+08	1.391032e+07	
...	
149996	9.197207e+06	3.264510e+06	1.348742e+07	
149997	4.735033e+06	7.121804e+08	2.457758e+06	

149998	1.339432e+07	1.211009e+08	1.131473e+07
149999	2.529475e+06	8.147131e+08	1.406930e+06
150000	8.288398e+06	4.211005e+08	8.264799e+06

	Total UL (Bytes)	Total DL (Bytes)
0	36749741.0	308879636.0
1	53800391.0	653384965.0
2	27883638.0	279807335.0
3	43324218.0	846028530.0
4	38542814.0	569138589.0
...
149996	57628851.0	574175259.0
149997	39135081.0	666648844.0
149998	34912224.0	592786405.0
149999	29626096.0	371895920.0
150000	NaN	NaN

[150001 rows x 55 columns]

data["Bearer Id"].unique

```
<bound method Series.unique of 0          1.311448e+19
1          1.311448e+19
2          1.311448e+19
3          1.311448e+19
4          1.311448e+19
...
149996     7.277826e+18
149997     7.349883e+18
149998     1.311448e+19
149999     1.311448e+19
150000          NaN
```

Name: Bearer Id, Length: 150001, dtype: float64>

data.head()

	Bearer Id	Start	Start ms	End	End ms	\
0	1.311448e+19	4/4/2019 12:01	770.0	4/25/2019 14:35	662.0	
1	1.311448e+19	4/9/2019 13:04	235.0	4/25/2019 8:15	606.0	
2	1.311448e+19	4/9/2019 17:42	1.0	4/25/2019 11:58	652.0	
3	1.311448e+19	4/10/2019 0:31	486.0	4/25/2019 7:36	171.0	
4	1.311448e+19	4/12/2019 20:10	565.0	4/25/2019 10:40	954.0	

	Dur. (ms)	IMSI	MSISDN/Number	IMEI	\
0	1823652.0	2.082014e+14	3.366496e+10	3.552121e+13	
1	1365104.0	2.082019e+14	3.368185e+10	3.579401e+13	
2	1361762.0	2.082003e+14	3.376063e+10	3.528151e+13	
3	1321509.0	2.082014e+14	3.375034e+10	3.535661e+13	
4	1089009.0	2.082014e+14	3.369980e+10	3.540701e+13	

Last Location Name ... Youtube DL (Bytes) Youtube UL (Bytes)

\	0	9.16456699548519E+015	...	15854611.0	2501332.0
	1	L77566A	...	20247395.0	19111729.0
	2	D42335A	...	19725661.0	14699576.0
	3	T21824A	...	21388122.0	15146643.0
	4	D88865A	...	15259380.0	18962873.0

	Netflix DL (Bytes)	Netflix UL (Bytes)	Gaming DL (Bytes)	\
0	8198936.0	9656251.0	278082303.0	
1	18338413.0	17227132.0	608750074.0	
2	17587794.0	6163408.0	229584621.0	
3	13994646.0	1097942.0	799538153.0	
4	17124581.0	415218.0	527707248.0	

	Gaming UL (Bytes)	Other DL (Bytes)	Other UL (Bytes)	Total UL
(Bytes) \				
0	14344150.0	171744450.0	8814393.0	
36749741.0				
1	1170709.0	526904238.0	15055145.0	
53800391.0				
2	395630.0	410692588.0	4215763.0	
27883638.0				
3	10849722.0	749039933.0	12797283.0	
43324218.0				
4	3529801.0	550709500.0	13910322.0	
38542814.0				

	Total DL (Bytes)
0	308879636.0
1	653384965.0
2	279807335.0
3	846028530.0
4	569138589.0

[5 rows x 55 columns]

data.tail()

	Bearer Id	Start	Start ms	End	End
ms \					
149996	7.277826e+18	4/29/2019 7:28	451.0	4/30/2019 6:02	
214.0					
149997	7.349883e+18	4/29/2019 7:28	483.0	4/30/2019 10:41	
187.0					
149998	1.311448e+19	4/29/2019 7:28	283.0	4/30/2019 10:46	

810.0						
149999	1.311448e+19	4/29/2019	7:28	696.0	4/30/2019	10:40
327.0						
150000	NaN		NaN	NaN		NaN
NaN						

	Dur. (ms)	IMSI	MSISDN/Number	IMEI	\
149996	81230.0	2.082022e+14	3.365069e+10	3.548311e+13	
149997	97970.0	2.082019e+14	3.366345e+10	3.566051e+13	
149998	98249.0	2.082017e+14	3.362189e+10	3.572121e+13	
149999	97910.0	2.082021e+14	3.361962e+10	8.618620e+13	
150000	NaN	NaN	NaN	NaN	

	Last Location Name	...	Youtube DL (Bytes)	Youtube UL (Bytes)
\				
149996	D20434A	...	1.619167e+07	1.176343e+07
149997	D10223C	...	1.387723e+07	8.288284e+06
149998	T51102A	...	2.266051e+07	1.855903e+06
149999	L88342B	...	8.817106e+06	8.305402e+06
150000	NaN	...	1.163407e+07	1.100941e+07

	Netflix DL (Bytes)	Netflix UL (Bytes)	Gaming DL (Bytes)	\
149996	1.788370e+07	1.967816e+07	5.266097e+08	
149997	1.935015e+07	2.129315e+07	6.268931e+08	
149998	9.963942e+06	5.065760e+06	5.535395e+08	
149999	3.322253e+06	1.317259e+07	3.525370e+08	
150000	1.162685e+07	1.100175e+07	4.220447e+08	

	Gaming UL (Bytes)	Other DL (Bytes)	Other UL (Bytes)	\
149996	9.197207e+06	3.264510e+06	1.348742e+07	
149997	4.735033e+06	7.121804e+08	2.457758e+06	
149998	1.339432e+07	1.211009e+08	1.131473e+07	
149999	2.529475e+06	8.147131e+08	1.406930e+06	
150000	8.288398e+06	4.211005e+08	8.264799e+06	

	Total UL (Bytes)	Total DL (Bytes)
149996	57628851.0	574175259.0
149997	39135081.0	666648844.0
149998	34912224.0	592786405.0
149999	29626096.0	371895920.0
150000	NaN	NaN

[5 rows x 55 columns]

data.shape

(150001, 55)

data.dtypes

Bearer Id	float64
Start	object
Start ms	float64
End	object
End ms	float64
Dur. (ms)	float64
IMSI	float64
MSISDN/Number	float64
IMEI	float64
Last Location Name	object
Avg RTT DL (ms)	float64
Avg RTT UL (ms)	float64
Avg Bearer TP DL (kbps)	float64
Avg Bearer TP UL (kbps)	float64
TCP DL Retrans. Vol (Bytes)	float64
TCP UL Retrans. Vol (Bytes)	float64
DL TP < 50 Kbps (%)	float64
50 Kbps < DL TP < 250 Kbps (%)	float64
250 Kbps < DL TP < 1 Mbps (%)	float64
DL TP > 1 Mbps (%)	float64
UL TP < 10 Kbps (%)	float64
10 Kbps < UL TP < 50 Kbps (%)	float64
50 Kbps < UL TP < 300 Kbps (%)	float64
UL TP > 300 Kbps (%)	float64
HTTP DL (Bytes)	float64
HTTP UL (Bytes)	float64
Activity Duration DL (ms)	float64
Activity Duration UL (ms)	float64
Dur. (ms).1	float64
Handset Manufacturer	object
Handset Type	object
Nb of sec with 125000B < Vol DL	float64
Nb of sec with 1250B < Vol UL < 6250B	float64
Nb of sec with 31250B < Vol DL < 125000B	float64
Nb of sec with 37500B < Vol UL	float64
Nb of sec with 6250B < Vol DL < 31250B	float64
Nb of sec with 6250B < Vol UL < 37500B	float64
Nb of sec with Vol DL < 6250B	float64
Nb of sec with Vol UL < 1250B	float64
Social Media DL (Bytes)	float64
Social Media UL (Bytes)	float64
Google DL (Bytes)	float64
Google UL (Bytes)	float64
Email DL (Bytes)	float64
Email UL (Bytes)	float64

```

Youtube DL (Bytes)          float64
Youtube UL (Bytes)          float64
Netflix DL (Bytes)          float64
Netflix UL (Bytes)          float64
Gaming DL (Bytes)           float64
Gaming UL (Bytes)           float64
Other DL (Bytes)            float64
Other UL (Bytes)            float64
Total UL (Bytes)            float64
Total DL (Bytes)            float64
dtype: object

```

```
data.describe()
```

	Bearer Id	Start ms	End ms	Dur. (ms)
IMSI \				
count	1.490100e+05	150000.000000	150000.000000	1.500000e+05
1.494310e+05				
mean	1.013887e+19	499.188200	498.800880	1.046086e+05
2.082016e+14				
std	2.893173e+18	288.611834	288.097653	8.103762e+04
2.148809e+10				
min	6.917538e+18	0.000000	0.000000	7.142000e+03
2.040471e+14				
25%	7.349883e+18	250.000000	251.000000	5.744050e+04
2.082014e+14				
50%	7.349883e+18	499.000000	500.000000	8.639900e+04
2.082015e+14				
75%	1.304243e+19	749.000000	750.000000	1.324302e+05
2.082018e+14				
max	1.318654e+19	999.000000	999.000000	1.859336e+06
2.140743e+14				

	MSISDN/Number	IMEI	Avg RTT DL (ms)	Avg RTT UL
(ms) \				
count	1.489350e+05	1.494290e+05	122172.000000	122189.000000
mean	4.188282e+10	4.847455e+13	109.795706	17.662883
std	2.447443e+12	2.241637e+13	619.782739	84.793524
min	3.360100e+10	4.400152e+11	0.000000	0.000000
25%	3.365130e+10	3.546071e+13	32.000000	2.000000
50%	3.366371e+10	3.572201e+13	45.000000	5.000000
75%	3.368349e+10	8.611970e+13	70.000000	15.000000
max	8.823971e+14	9.900120e+13	96923.000000	7120.000000

	Avg Bearer TP DL (kbps)	...	Youtube DL (Bytes)	Youtube UL (Bytes)
count	150000.000000	...	1.500010e+05	
mean	13300.045927	...	1.163407e+07	
std	23971.878541	...	6.710569e+06	
min	0.000000	...	5.300000e+01	
25%	43.000000	...	5.833501e+06	
50%	63.000000	...	1.161602e+07	
75%	19710.750000	...	1.744852e+07	
max	378160.000000	...	2.325910e+07	

	Netflix DL (Bytes)	Netflix UL (Bytes)	Gaming DL (Bytes)	\
count	1.500010e+05	1.500010e+05	1.500010e+05	
mean	1.162685e+07	1.100175e+07	4.220447e+08	
std	6.725218e+06	6.359490e+06	2.439675e+08	
min	4.200000e+01	3.500000e+01	2.516000e+03	
25%	5.777156e+06	5.475981e+06	2.104733e+08	
50%	1.164222e+07	1.099638e+07	4.234081e+08	
75%	1.747048e+07	1.650727e+07	6.331742e+08	
max	2.325919e+07	2.201196e+07	8.434419e+08	

	Gaming UL (Bytes)	Other DL (Bytes)	Other UL (Bytes)	\
count	1.500010e+05	1.500010e+05	1.500010e+05	
mean	8.288398e+06	4.211005e+08	8.264799e+06	
std	4.782700e+06	2.432050e+08	4.769004e+06	
min	5.900000e+01	3.290000e+03	1.480000e+02	
25%	4.128476e+06	2.101869e+08	4.145943e+06	
50%	8.291208e+06	4.218030e+08	8.267071e+06	
75%	1.243162e+07	6.316918e+08	1.238415e+07	
max	1.655879e+07	8.434425e+08	1.655882e+07	

	Total UL (Bytes)	Total DL (Bytes)
count	1.500000e+05	1.500000e+05
mean	4.112121e+07	4.546434e+08
std	1.127639e+07	2.441429e+08
min	2.866892e+06	7.114041e+06
25%	3.322201e+07	2.431068e+08
50%	4.114331e+07	4.558411e+08
75%	4.903424e+07	6.657055e+08
max	7.833131e+07	9.029696e+08

[8 rows x 50 columns]

data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150001 entries, 0 to 150000
Data columns (total 55 columns):

#	Column	Non-Null Count	Dtype
0	Bearer Id	149010 non-null	
float64			
1	Start	150000 non-null	object
2	Start ms	150000 non-null	
float64			
3	End	150000 non-null	object
4	End ms	150000 non-null	
float64			
5	Dur. (ms)	150000 non-null	
float64			
6	IMSI	149431 non-null	
float64			
7	MSISDN/Number	148935 non-null	
float64			
8	IMEI	149429 non-null	
float64			
9	Last Location Name	148848 non-null	object
10	Avg RTT DL (ms)	122172 non-null	
float64			
11	Avg RTT UL (ms)	122189 non-null	
float64			
12	Avg Bearer TP DL (kbps)	150000 non-null	
float64			
13	Avg Bearer TP UL (kbps)	150000 non-null	
float64			
14	TCP DL Retrans. Vol (Bytes)	61855 non-null	
float64			
15	TCP UL Retrans. Vol (Bytes)	53352 non-null	
float64			
16	DL TP < 50 Kbps (%)	149247 non-null	
float64			
17	50 Kbps < DL TP < 250 Kbps (%)	149247 non-null	
float64			
18	250 Kbps < DL TP < 1 Mbps (%)	149247 non-null	
float64			

19 DL TP > 1 Mbps (%)	149247 non-null	
float64		
20 UL TP < 10 Kbps (%)	149209 non-null	
float64		
21 10 Kbps < UL TP < 50 Kbps (%)	149209 non-null	
float64		
22 50 Kbps < UL TP < 300 Kbps (%)	149209 non-null	
float64		
23 UL TP > 300 Kbps (%)	149209 non-null	
float64		
24 HTTP DL (Bytes)	68527 non-null	
float64		
25 HTTP UL (Bytes)	68191 non-null	
float64		
26 Activity Duration DL (ms)	150000 non-null	
float64		
27 Activity Duration UL (ms)	150000 non-null	
float64		
28 Dur. (ms).1	150000 non-null	
float64		
29 Handset Manufacturer	149429 non-null	object
30 Handset Type	149429 non-null	object
31 Nb of sec with 125000B < Vol DL	52463 non-null	
float64		
32 Nb of sec with 1250B < Vol UL < 6250B	57107 non-null	
float64		
33 Nb of sec with 31250B < Vol DL < 125000B	56415 non-null	
float64		
34 Nb of sec with 37500B < Vol UL	19747 non-null	
float64		
35 Nb of sec with 6250B < Vol DL < 31250B	61684 non-null	
float64		
36 Nb of sec with 6250B < Vol UL < 37500B	38158 non-null	
float64		
37 Nb of sec with Vol DL < 6250B	149246 non-null	
float64		
38 Nb of sec with Vol UL < 1250B	149208 non-null	
float64		
39 Social Media DL (Bytes)	150001 non-null	
float64		
40 Social Media UL (Bytes)	150001 non-null	
float64		
41 Google DL (Bytes)	150001 non-null	
float64		
42 Google UL (Bytes)	150001 non-null	
float64		
43 Email DL (Bytes)	150001 non-null	
float64		

```

44 Email UL (Bytes) 150001 non-null
float64
45 Youtube DL (Bytes) 150001 non-null
float64
46 Youtube UL (Bytes) 150001 non-null
float64
47 Netflix DL (Bytes) 150001 non-null
float64
48 Netflix UL (Bytes) 150001 non-null
float64
49 Gaming DL (Bytes) 150001 non-null
float64
50 Gaming UL (Bytes) 150001 non-null
float64
51 Other DL (Bytes) 150001 non-null
float64
52 Other UL (Bytes) 150001 non-null
float64
53 Total UL (Bytes) 150000 non-null
float64
54 Total DL (Bytes) 150000 non-null
float64
dtypes: float64(50), object(5)
memory usage: 62.9+ MB

```

```
data.duplicated()
```

```

0      False
1      False
2      False
3      False
4      False

```

```

...
149996  False
149997  False
149998  False
149999  False
150000  False

```

```
Length: 150001, dtype: bool
```

```
null_pct= data.isnull().sum() / len(data) * 100
```

```
null_pct
```

```

Bearer Id      0.660662
Start          0.000667
Start ms       0.000667
End            0.000667
End ms         0.000667
Dur. (ms)      0.000667
IMSI           0.379997

```

MSISDN/Number	0.710662
IMEI	0.381331
Last Location Name	0.768662
Avg RTT DL (ms)	18.552543
Avg RTT UL (ms)	18.541210
Avg Bearer TP DL (kbps)	0.000667
Avg Bearer TP UL (kbps)	0.000667
TCP DL Retrans. Vol (Bytes)	58.763608
TCP UL Retrans. Vol (Bytes)	64.432237
DL TP < 50 Kbps (%)	0.502663
50 Kbps < DL TP < 250 Kbps (%)	0.502663
250 Kbps < DL TP < 1 Mbps (%)	0.502663
DL TP > 1 Mbps (%)	0.502663
UL TP < 10 Kbps (%)	0.527996
10 Kbps < UL TP < 50 Kbps (%)	0.527996
50 Kbps < UL TP < 300 Kbps (%)	0.527996
UL TP > 300 Kbps (%)	0.527996
HTTP DL (Bytes)	54.315638
HTTP UL (Bytes)	54.539636
Activity Duration DL (ms)	0.000667
Activity Duration UL (ms)	0.000667
Dur. (ms).1	0.000667
Handset Manufacturer	0.381331
Handset Type	0.381331
Nb of sec with 125000B < Vol DL	65.024900
Nb of sec with 1250B < Vol UL < 6250B	61.928920
Nb of sec with 31250B < Vol DL < 125000B	62.390251
Nb of sec with 37500B < Vol UL	86.835421
Nb of sec with 6250B < Vol DL < 31250B	58.877607
Nb of sec with 6250B < Vol UL < 37500B	74.561503
Nb of sec with Vol DL < 6250B	0.503330
Nb of sec with Vol UL < 1250B	0.528663
Social Media DL (Bytes)	0.000000
Social Media UL (Bytes)	0.000000
Google DL (Bytes)	0.000000
Google UL (Bytes)	0.000000
Email DL (Bytes)	0.000000
Email UL (Bytes)	0.000000
Youtube DL (Bytes)	0.000000
Youtube UL (Bytes)	0.000000
Netflix DL (Bytes)	0.000000
Netflix UL (Bytes)	0.000000
Gaming DL (Bytes)	0.000000
Gaming UL (Bytes)	0.000000
Other DL (Bytes)	0.000000
Other UL (Bytes)	0.000000
Total UL (Bytes)	0.000667
Total DL (Bytes)	0.000667
dtype: float64	

```
cols_to_drop = null_pct[null_pct > 50].index.tolist()
cols_to_drop
```

```
['TCP DL Retrans. Vol (Bytes)',
 'TCP UL Retrans. Vol (Bytes)',
 'HTTP DL (Bytes)',
 'HTTP UL (Bytes)',
 'Nb of sec with 125000B < Vol DL',
 'Nb of sec with 1250B < Vol UL < 6250B',
 'Nb of sec with 31250B < Vol DL < 125000B',
 'Nb of sec with 37500B < Vol UL',
 'Nb of sec with 6250B < Vol DL < 31250B',
 'Nb of sec with 6250B < Vol UL < 37500B']
```

```
data=data.drop(cols_to_drop,axis=1)
data
```

	ms \	Bearer Id	Start	Start ms	End	End
0	1.311448e+19	4/4/2019 12:01	770.0	4/25/2019 14:35		
662.0						
1	1.311448e+19	4/9/2019 13:04	235.0	4/25/2019 8:15		
606.0						
2	1.311448e+19	4/9/2019 17:42	1.0	4/25/2019 11:58		
652.0						
3	1.311448e+19	4/10/2019 0:31	486.0	4/25/2019 7:36		
171.0						
4	1.311448e+19	4/12/2019 20:10	565.0	4/25/2019 10:40		
954.0						
...
149996	7.277826e+18	4/29/2019 7:28	451.0	4/30/2019 6:02		
214.0						
149997	7.349883e+18	4/29/2019 7:28	483.0	4/30/2019 10:41		
187.0						
149998	1.311448e+19	4/29/2019 7:28	283.0	4/30/2019 10:46		
810.0						
149999	1.311448e+19	4/29/2019 7:28	696.0	4/30/2019 10:40		
327.0						
150000	NaN	NaN	NaN	NaN		
NaN						

	Dur. (ms)	IMSI	MSISDN/Number	IMEI \
0	1823652.0	2.082014e+14	3.366496e+10	3.552121e+13
1	1365104.0	2.082019e+14	3.368185e+10	3.579401e+13
2	1361762.0	2.082003e+14	3.376063e+10	3.528151e+13
3	1321509.0	2.082014e+14	3.375034e+10	3.535661e+13
4	1089009.0	2.082014e+14	3.369980e+10	3.540701e+13
...
149996	81230.0	2.082022e+14	3.365069e+10	3.548311e+13
149997	97970.0	2.082019e+14	3.366345e+10	3.566051e+13

149998	98249.0	2.082017e+14	3.362189e+10	3.572121e+13
149999	97910.0	2.082021e+14	3.361962e+10	8.618620e+13
150000	NaN	NaN	NaN	NaN

	Last Location Name	...	Youtube DL (Bytes)	Youtube UL
(Bytes) \				
0	9.16456699548519E+015	...	1.585461e+07	
2.501332e+06				
1	L77566A	...	2.024740e+07	
1.911173e+07				
2	D42335A	...	1.972566e+07	
1.469958e+07				
3	T21824A	...	2.138812e+07	
1.514664e+07				
4	D88865A	...	1.525938e+07	
1.896287e+07				
...	
...				
149996	D20434A	...	1.619167e+07	
1.176343e+07				
149997	D10223C	...	1.387723e+07	
8.288284e+06				
149998	T51102A	...	2.266051e+07	
1.855903e+06				
149999	L88342B	...	8.817106e+06	
8.305402e+06				
150000	NaN	...	1.163407e+07	
1.100941e+07				

	Netflix DL (Bytes)	Netflix UL (Bytes)	Gaming DL (Bytes)	\
0	8.198936e+06	9.656251e+06	2.780823e+08	
1	1.833841e+07	1.722713e+07	6.087501e+08	
2	1.758779e+07	6.163408e+06	2.295846e+08	
3	1.399465e+07	1.097942e+06	7.995382e+08	
4	1.712458e+07	4.152180e+05	5.277072e+08	
...	
149996	1.788370e+07	1.967816e+07	5.266097e+08	
149997	1.935015e+07	2.129315e+07	6.268931e+08	
149998	9.963942e+06	5.065760e+06	5.535395e+08	
149999	3.322253e+06	1.317259e+07	3.525370e+08	
150000	1.162685e+07	1.100175e+07	4.220447e+08	

	Gaming UL (Bytes)	Other DL (Bytes)	Other UL (Bytes)	\
0	1.434415e+07	1.717444e+08	8.814393e+06	
1	1.170709e+06	5.269042e+08	1.505514e+07	
2	3.956300e+05	4.106926e+08	4.215763e+06	
3	1.084972e+07	7.490399e+08	1.279728e+07	
4	3.529801e+06	5.507095e+08	1.391032e+07	
...	
149996	9.197207e+06	3.264510e+06	1.348742e+07	

149997	4.735033e+06	7.121804e+08	2.457758e+06
149998	1.339432e+07	1.211009e+08	1.131473e+07
149999	2.529475e+06	8.147131e+08	1.406930e+06
150000	8.288398e+06	4.211005e+08	8.264799e+06

	Total UL (Bytes)	Total DL (Bytes)
0	36749741.0	308879636.0
1	53800391.0	653384965.0
2	27883638.0	279807335.0
3	43324218.0	846028530.0
4	38542814.0	569138589.0
...
149996	57628851.0	574175259.0
149997	39135081.0	666648844.0
149998	34912224.0	592786405.0
149999	29626096.0	371895920.0
150000	NaN	NaN

[150001 rows x 45 columns]

data.isnull().sum()

Bearer Id	991
Start	1
Start ms	1
End	1
End ms	1
Dur. (ms)	1
IMSI	570
MSISDN/Number	1066
IMEI	572
Last Location Name	1153
Avg RTT DL (ms)	27829
Avg RTT UL (ms)	27812
Avg Bearer TP DL (kbps)	1
Avg Bearer TP UL (kbps)	1
DL TP < 50 Kbps (%)	754
50 Kbps < DL TP < 250 Kbps (%)	754
250 Kbps < DL TP < 1 Mbps (%)	754
DL TP > 1 Mbps (%)	754
UL TP < 10 Kbps (%)	792
10 Kbps < UL TP < 50 Kbps (%)	792
50 Kbps < UL TP < 300 Kbps (%)	792
UL TP > 300 Kbps (%)	792
Activity Duration DL (ms)	1
Activity Duration UL (ms)	1
Dur. (ms).1	1
Handset Manufacturer	572
Handset Type	572
Nb of sec with Vol DL < 6250B	755

```

Nb of sec with Vol UL < 1250B      793
Social Media DL (Bytes)            0
Social Media UL (Bytes)            0
Google DL (Bytes)                  0
Google UL (Bytes)                  0
Email DL (Bytes)                   0
Email UL (Bytes)                   0
Youtube DL (Bytes)                 0
Youtube UL (Bytes)                 0
Netflix DL (Bytes)                 0
Netflix UL (Bytes)                 0
Gaming DL (Bytes)                  0
Gaming UL (Bytes)                  0
Other DL (Bytes)                   0
Other UL (Bytes)                   0
Total UL (Bytes)                   1
Total DL (Bytes)                   1
dtype: int64

```

```

data=data.dropna(how='any',axis=0)
data

```

	ms \	Bearer Id	Start	Start ms	End	End
0	1.311448e+19	4/4/2019 12:01	770.0	4/25/2019 14:35		
662.0						
1	1.311448e+19	4/9/2019 13:04	235.0	4/25/2019 8:15		
606.0						
6	1.311448e+19	4/13/2019 8:41	612.0	4/25/2019 8:16		
168.0						
7	1.304243e+19	4/14/2019 2:11	592.0	4/25/2019 2:26		
512.0						
9	1.304243e+19	4/15/2019 0:32	0.0	4/25/2019 0:40		
284.0						
...
.						
149995	1.304243e+19	4/29/2019 7:28	615.0	4/30/2019 0:01		
407.0						
149996	7.277826e+18	4/29/2019 7:28	451.0	4/30/2019 6:02		
214.0						
149997	7.349883e+18	4/29/2019 7:28	483.0	4/30/2019 10:41		
187.0						
149998	1.311448e+19	4/29/2019 7:28	283.0	4/30/2019 10:46		
810.0						
149999	1.311448e+19	4/29/2019 7:28	696.0	4/30/2019 10:40		
327.0						
	Dur. (ms)	IMSI	MSISDN/Number	IMEI \		
0	1823652.0	2.082014e+14	3.366496e+10	3.552121e+13		
1	1365104.0	2.082019e+14	3.368185e+10	3.579401e+13		

6	1035261.0	2.082014e+14	3.366537e+10	8.676270e+13
7	951292.0	2.082010e+14	3.376349e+10	8.654640e+13
9	864482.0	2.082003e+14	3.365922e+10	3.557311e+13
...
149995	59587.0	2.082014e+14	3.366865e+10	3.533251e+13
149996	81230.0	2.082022e+14	3.365069e+10	3.548311e+13
149997	97970.0	2.082019e+14	3.366345e+10	3.566051e+13
149998	98249.0	2.082017e+14	3.362189e+10	3.572121e+13
149999	97910.0	2.082021e+14	3.361962e+10	8.618620e+13

	Last Location Name	...	Youtube DL (Bytes)	Youtube UL
(Bytes) \				
0	9.16456699548519E+015	...	15854611.0	
2501332.0				
1	L77566A	...	20247395.0	
19111729.0				
6	9.16456701058919E+015	...	7234465.0	
1886295.0				
7	CELL_208_20_520025_1	...	12189786.0	
19735183.0				
9	L20434C	...	16338345.0	
6573194.0				
...	
...				
149995	T85721A	...	6550499.0	
18003146.0				
149996	D20434A	...	16191667.0	
11763428.0				
149997	D10223C	...	13877234.0	
8288284.0				
149998	T51102A	...	22660510.0	
1855903.0				
149999	L88342B	...	8817106.0	
8305402.0				

	Netflix DL (Bytes)	Netflix UL (Bytes)	Gaming DL (Bytes)	\
0	8198936.0	9656251.0	278082303.0	
1	18338413.0	17227132.0	608750074.0	
6	13231006.0	17981760.0	44796223.0	
7	6858496.0	12075206.0	748906189.0	
9	12211149.0	974201.0	811355188.0	
...	
149995	22468983.0	7149728.0	784435351.0	
149996	17883703.0	19678161.0	526609673.0	
149997	19350146.0	21293148.0	626893062.0	
149998	9963942.0	5065760.0	553539484.0	
149999	3322253.0	13172589.0	352536971.0	

	Gaming UL (Bytes)	Other DL (Bytes)	Other UL (Bytes)	\
0	14344150.0	171744450.0	8814393.0	

1	1170709.0	526904238.0	15055145.0
6	14084934.0	803653557.0	10649768.0
7	16149622.0	537601820.0	7714657.0
9	1591237.0	64856042.0	12815208.0
...
149995	12724751.0	321383162.0	14890486.0
149996	9197207.0	3264510.0	13487416.0
149997	4735033.0	712180387.0	2457758.0
149998	13394316.0	121100856.0	11314729.0
149999	2529475.0	814713113.0	1406930.0

	Total UL (Bytes)	Total DL (Bytes)
0	36749741.0	308879636.0
1	53800391.0	653384965.0
6	47925246.0	70562047.0
7	58813016.0	775350343.0
9	22417975.0	850570347.0
...
149995	53745392.0	819016468.0
149996	57628851.0	574175259.0
149997	39135081.0	666648844.0
149998	34912224.0	592786405.0
149999	29626096.0	371895920.0

[120739 rows x 45 columns]

Task 1 - User Overview Analysis

- Start by identifying the top 10 handsets used by the customers.
- Then, identify the top 3 handset manufacturers
- Next, identify the top 5 handsets per top 3 handset manufacturer
- Make a short interpretation and recommendation to marketing teams

```
top_10_handsets = data["Handset
Type"].value_counts(ascending=False).head(10)
print(top_10_handsets)
```

Huawei B528S-23A	19310
Apple iPhone 6S (A1688)	9244
Apple iPhone 6 (A1586)	8786
Apple iPhone 7 (A1778)	6134
Apple iPhone Se (A1723)	5062
Apple iPhone 8 (A1905)	4900
undefined	4816
Apple iPhone Xr (A2105)	4490
Apple iPhone X (A1901)	3746

```
Apple iPhone 8 Plus (A1897)      2968
Name: Handset Type, dtype: int64
```

```
top_3_Manufacturer = data["Handset
Manufacturer"].value_counts(ascending=False).head(3)
print(top_3_Manufacturer )
```

```
Apple      58190
Huawei      33306
Samsung    19167
Name: Handset Manufacturer, dtype: int64
```

```
top_5_handsets_per_Manufacturer = {}
for Manufacturer in top_3_Manufacturer .index:
    Manufacturer_data = data[data['Handset Manufacturer'] ==
Manufacturer]
    top_5_handsets = Manufacturer_data['Handset
Type'].value_counts().head(5)
    top_5_handsets_per_Manufacturer[ Manufacturer] = top_5_handsets
print("top_5_handsets_per_Manufacturer:")
for Manufacturer, top_5_handsets in
top_5_handsets_per_Manufacturer.items():
    print("Manufacturer: {Manufacturer}")
    print(top_5_handsets)
```

```
top_5_handsets_per_Manufacturer:
Manufacturer: {Manufacturer}
Apple iPhone 6S (A1688)      9244
Apple iPhone 6 (A1586)      8786
Apple iPhone 7 (A1778)      6134
Apple iPhone Se (A1723)     5062
Apple iPhone 8 (A1905)      4900
Name: Handset Type, dtype: int64
Manufacturer: {Manufacturer}
Huawei B528S-23A              19310
Huawei E5180                  2054
Huawei P20 Lite Huawei Nova 3E 1980
Huawei P20                    1394
Huawei Y6 2018                975
Name: Handset Type, dtype: int64
Manufacturer: {Manufacturer}
Samsung Galaxy S8 (Sm-G950F)  1820
Samsung Galaxy J5 (Sm-J530)   1813
Samsung Galaxy J3 (Sm-J330)   1547
Samsung Galaxy A5 Sm-A520F    1488
Samsung Galaxy S7 (Sm-G930X)  1328
Name: Handset Type, dtype: int64
```

Task 1.1 - Your employer wants to have an overview of the users' behaviour on those applications.

- Aggregate per user the following information in the column
- number of xDR sessions
- Session duration
- the total download (DL) and upload (UL) data
- the total data volume (in Bytes) during this session for each application

```
categorical = [var for var in data.columns if data[var].dtypes=='0']  
numerical = [var for var in data.columns if data[var].dtypes!='0']
```

```
categorical
```

```
['Start', 'End', 'Last Location Name', 'Handset Manufacturer',  
'Handset Type']
```

```
numerical
```

```
['Bearer Id',  
'Start ms',  
'End ms',  
'Dur. (ms)',  
'IMSI',  
'MSISDN/Number',  
'IMEI',  
'Avg RTT DL (ms)',  
'Avg RTT UL (ms)',  
'Avg Bearer TP DL (kbps)',  
'Avg Bearer TP UL (kbps)',  
'DL TP < 50 Kbps (%)',  
'50 Kbps < DL TP < 250 Kbps (%)',  
'250 Kbps < DL TP < 1 Mbps (%)',  
'DL TP > 1 Mbps (%)',  
'UL TP < 10 Kbps (%)',  
'10 Kbps < UL TP < 50 Kbps (%)',  
'50 Kbps < UL TP < 300 Kbps (%)',  
'UL TP > 300 Kbps (%)',  
'Activity Duration DL (ms)',  
'Activity Duration UL (ms)',  
'Dur. (ms).1',  
'Nb of sec with Vol DL < 6250B',  
'Nb of sec with Vol UL < 1250B',  
'Social Media DL (Bytes)',  
'Social Media UL (Bytes)',  
'Google DL (Bytes)',  
'Google UL (Bytes)']
```

```

'Email DL (Bytes)',
'Email UL (Bytes)',
'Youtube DL (Bytes)',
'Youtube UL (Bytes)',
'Netflix DL (Bytes)',
'Netflix UL (Bytes)',
'Gaming DL (Bytes)',
'Gaming UL (Bytes)',
'Other DL (Bytes)',
'Other UL (Bytes)',
'Total UL (Bytes)',
'Total DL (Bytes)']

```

data.columns

```

Index(['Bearer Id', 'Start', 'Start ms', 'End', 'End ms', 'Dur. (ms)',
      'IMSI',
      'MSISDN/Number', 'IMEI', 'Last Location Name', 'Avg RTT DL
(ms)',
      'Avg RTT UL (ms)', 'Avg Bearer TP DL (kbps)', 'Avg Bearer TP UL
(kbps)',
      'DL TP < 50 Kbps (%)', '50 Kbps < DL TP < 250 Kbps (%)',
      '250 Kbps < DL TP < 1 Mbps (%)', 'DL TP > 1 Mbps (%)',
      'UL TP < 10 Kbps (%)', '10 Kbps < UL TP < 50 Kbps (%)',
      '50 Kbps < UL TP < 300 Kbps (%)', 'UL TP > 300 Kbps (%)',
      'Activity Duration DL (ms)', 'Activity Duration UL (ms)', 'Dur.
(ms).1',
      'Handset Manufacturer', 'Handset Type', 'Nb of sec with Vol DL
< 6250B',
      'Nb of sec with Vol UL < 1250B', 'Social Media DL (Bytes)',
      'Social Media UL (Bytes)', 'Google DL (Bytes)', 'Google UL
(Bytes)',
      'Email DL (Bytes)', 'Email UL (Bytes)', 'Youtube DL (Bytes)',
      'Youtube UL (Bytes)', 'Netflix DL (Bytes)', 'Netflix UL
(Bytes)',
      'Gaming DL (Bytes)', 'Gaming UL (Bytes)', 'Other DL (Bytes)',
      'Other UL (Bytes)', 'Total UL (Bytes)', 'Total DL (Bytes)'],
      dtype='object')

```

```

for i in data.columns:
    print(i)

```

```

Bearer Id
Start
Start ms
End
End ms
Dur. (ms)
IMSI
MSISDN/Number
IMEI

```

Last Location Name
 Avg RTT DL (ms)
 Avg RTT UL (ms)
 Avg Bearer TP DL (kbps)
 Avg Bearer TP UL (kbps)
 DL TP < 50 Kbps (%)
 50 Kbps < DL TP < 250 Kbps (%)
 250 Kbps < DL TP < 1 Mbps (%)
 DL TP > 1 Mbps (%)
 UL TP < 10 Kbps (%)
 10 Kbps < UL TP < 50 Kbps (%)
 50 Kbps < UL TP < 300 Kbps (%)
 UL TP > 300 Kbps (%)
 Activity Duration DL (ms)
 Activity Duration UL (ms)
 Dur. (ms).1
 Handset Manufacturer
 Handset Type
 Nb of sec with Vol DL < 6250B
 Nb of sec with Vol UL < 1250B
 Social Media DL (Bytes)
 Social Media UL (Bytes)
 Google DL (Bytes)
 Google UL (Bytes)
 Email DL (Bytes)
 Email UL (Bytes)
 Youtube DL (Bytes)
 Youtube UL (Bytes)
 Netflix DL (Bytes)
 Netflix UL (Bytes)
 Gaming DL (Bytes)
 Gaming UL (Bytes)
 Other DL (Bytes)
 Other UL (Bytes)
 Total UL (Bytes)
 Total DL (Bytes)

```

user_behaviour=data.groupby('Bearer Id').agg({
    'Dur. (ms)': 'sum',
    'Activity Duration DL (ms)': 'sum',
    'Activity Duration UL (ms)': 'sum',
    'Social Media DL (Bytes)': 'sum',
    'Social Media UL (Bytes)': 'sum',
    'Google DL (Bytes)': 'sum',
    'Google DL (Bytes)' : 'sum',

    'Google UL (Bytes)' : 'sum',

    'Email DL (Bytes)' : 'sum',

```

```

        'Email UL (Bytes)' : 'sum',
        'Youtube DL (Bytes)' : 'sum',

        'Youtube UL (Bytes)' : 'sum',

        'Netflix DL (Bytes)' : 'sum',

        'Netflix UL (Bytes)' : 'sum',

        'Gaming DL (Bytes)' : 'sum',

        'Gaming UL (Bytes)' : 'sum',

        'Other DL (Bytes)' : 'sum',

        'Other UL (Bytes)' : 'sum',
    }).reset_index()
    print(user_behaviour)

```

	Bearer Id	Dur. (ms)	Activity Duration DL (ms)	\
0	6.917538e+18	24534.0	131798.0	
1	6.917538e+18	27786.0	401941.0	
2	6.917538e+18	15635.0	73347.0	
3	6.917538e+18	24264.0	117340.0	
4	6.917538e+18	15548.0	76969.0	
...	
110427	1.318654e+19	80024.0	2512362.0	
110428	1.318654e+19	145291.0	2067.0	
110429	1.318654e+19	86399.0	3968131.0	
110430	1.318654e+19	86399.0	1689999.0	
110431	1.318654e+19	103113.0	0.0	

	Activity Duration UL (ms)	Social Media DL (Bytes)	\
0	101470.0	2404741.0	
1	399092.0	944612.0	
2	81378.0	1817239.0	
3	347852.0	1867318.0	
4	80241.0	1502346.0	
...	
110427	2437668.0	3240226.0	
110428	45217.0	3062671.0	
110429	3537154.0	720996.0	
110430	1513764.0	2492460.0	
110431	30367.0	1314234.0	

	Social Media UL (Bytes)	Google DL (Bytes)	Google UL (Bytes)
\			
0	2410.0	5791591.0	2871336.0

1	2827.0	10373157.0	56392.0
2	19827.0	269988.0	3696393.0
3	18928.0	1689296.0	195216.0
4	37348.0	8917833.0	701876.0
...
110427	38284.0	2036152.0	2271168.0
110428	48953.0	9363661.0	4001970.0
110429	42836.0	1541915.0	2100839.0
110430	39905.0	11318188.0	466218.0
110431	27938.0	6969652.0	3756009.0

	Email DL (Bytes)	Email UL (Bytes)	Youtube DL (Bytes) \
0	782388.0	806920.0	6139644.0
1	128003.0	34038.0	5385159.0
2	3191192.0	896670.0	12347020.0
3	740633.0	590043.0	15231815.0
4	332813.0	537902.0	14602694.0
...
110427	2410615.0	387548.0	12404964.0
110428	2192057.0	866373.0	22147919.0
110429	2315638.0	839789.0	11879062.0
110430	2612190.0	618629.0	22163800.0
110431	3317462.0	408257.0	12099319.0

	Youtube UL (Bytes)	Netflix DL (Bytes)	Netflix UL (Bytes) \
0	2071526.0	19494278.0	14668354.0
1	4295851.0	15755839.0	1300571.0
2	11089528.0	2859358.0	1738176.0
3	8401567.0	21563985.0	2817981.0
4	1237356.0	4506463.0	17006591.0
...
110427	4343114.0	11108134.0	21649273.0
110428	2152449.0	21468525.0	8603105.0
110429	1290963.0	22596930.0	11943452.0
110430	16763435.0	9522397.0	8346624.0
110431	1636122.0	9992219.0	17624886.0

	Gaming DL (Bytes)	Gaming UL (Bytes)	Other DL (Bytes) \
0	466109357.0	5333340.0	670751043.0

1	821879090.0	8521398.0	472846860.0
2	805301713.0	16257481.0	24303797.0
3	583864716.0	6992868.0	685122214.0
4	542337372.0	15820659.0	463698729.0
...
110427	114093049.0	2834548.0	695881178.0
110428	328766801.0	7569327.0	371261255.0
110429	833634251.0	10607174.0	697260277.0
110430	338246033.0	1845068.0	17385489.0
110431	293519955.0	16295588.0	440290470.0

	Other UL (Bytes)
0	15950724.0
1	1337849.0
2	15907613.0
3	2966860.0
4	11939642.0
...	...
110427	3888729.0
110428	12947410.0
110429	9094407.0
110430	12797797.0
110431	6398758.0

[110432 rows x 18 columns]

```
data1=pd.DataFrame(user_behaviour)
```

```
data1
```

	Bearer Id	Dur. (ms)	Activity Duration DL (ms)	\
0	6.917538e+18	24534.0	131798.0	
1	6.917538e+18	27786.0	401941.0	
2	6.917538e+18	15635.0	73347.0	
3	6.917538e+18	24264.0	117340.0	
4	6.917538e+18	15548.0	76969.0	
...	
110427	1.318654e+19	80024.0	2512362.0	
110428	1.318654e+19	145291.0	2067.0	
110429	1.318654e+19	86399.0	3968131.0	
110430	1.318654e+19	86399.0	1689999.0	
110431	1.318654e+19	103113.0	0.0	

	Activity Duration UL (ms)	Social Media DL (Bytes)	\
0	101470.0	2404741.0	
1	399092.0	944612.0	
2	81378.0	1817239.0	
3	347852.0	1867318.0	
4	80241.0	1502346.0	
...	

110427	2437668.0	3240226.0
110428	45217.0	3062671.0
110429	3537154.0	720996.0
110430	1513764.0	2492460.0
110431	30367.0	1314234.0

	Social Media UL (Bytes)	Google DL (Bytes)	Google UL (Bytes)
\			
0	2410.0	5791591.0	2871336.0
1	2827.0	10373157.0	56392.0
2	19827.0	269988.0	3696393.0
3	18928.0	1689296.0	195216.0
4	37348.0	8917833.0	701876.0
...

110427	38284.0	2036152.0	2271168.0
110428	48953.0	9363661.0	4001970.0
110429	42836.0	1541915.0	2100839.0
110430	39905.0	11318188.0	466218.0
110431	27938.0	6969652.0	3756009.0

	Email DL (Bytes)	Email UL (Bytes)	Youtube DL (Bytes)	\
0	782388.0	806920.0	6139644.0	
1	128003.0	34038.0	5385159.0	
2	3191192.0	896670.0	12347020.0	
3	740633.0	590043.0	15231815.0	
4	332813.0	537902.0	14602694.0	
...	
110427	2410615.0	387548.0	12404964.0	
110428	2192057.0	866373.0	22147919.0	
110429	2315638.0	839789.0	11879062.0	
110430	2612190.0	618629.0	22163800.0	
110431	3317462.0	408257.0	12099319.0	

	Youtube UL (Bytes)	Netflix DL (Bytes)	Netflix UL (Bytes)	\
0	2071526.0	19494278.0	14668354.0	
1	4295851.0	15755839.0	1300571.0	
2	11089528.0	2859358.0	1738176.0	
3	8401567.0	21563985.0	2817981.0	

4	1237356.0	4506463.0	17006591.0
...
110427	4343114.0	11108134.0	21649273.0
110428	2152449.0	21468525.0	8603105.0
110429	1290963.0	22596930.0	11943452.0
110430	16763435.0	9522397.0	8346624.0
110431	1636122.0	9992219.0	17624886.0

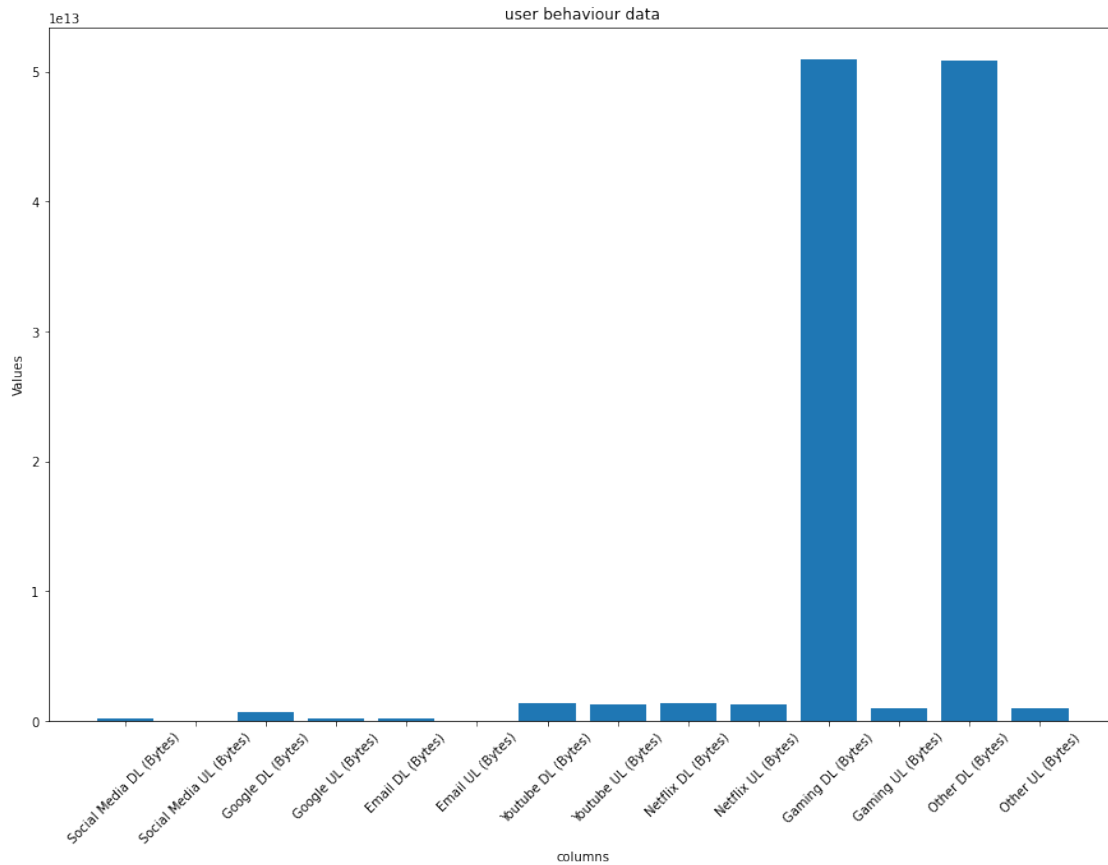
	Gaming DL (Bytes)	Gaming UL (Bytes)	Other DL (Bytes) \
0	466109357.0	5333340.0	670751043.0
1	821879090.0	8521398.0	472846860.0
2	805301713.0	16257481.0	24303797.0
3	583864716.0	6992868.0	685122214.0
4	542337372.0	15820659.0	463698729.0
...
110427	114093049.0	2834548.0	695881178.0
110428	328766801.0	7569327.0	371261255.0
110429	833634251.0	10607174.0	697260277.0
110430	338246033.0	1845068.0	17385489.0
110431	293519955.0	16295588.0	440290470.0

	Other UL (Bytes)
0	15950724.0
1	1337849.0
2	15907613.0
3	2966860.0
4	11939642.0
...	...
110427	3888729.0
110428	12947410.0
110429	9094407.0
110430	12797797.0
110431	6398758.0

[110432 rows x 18 columns]

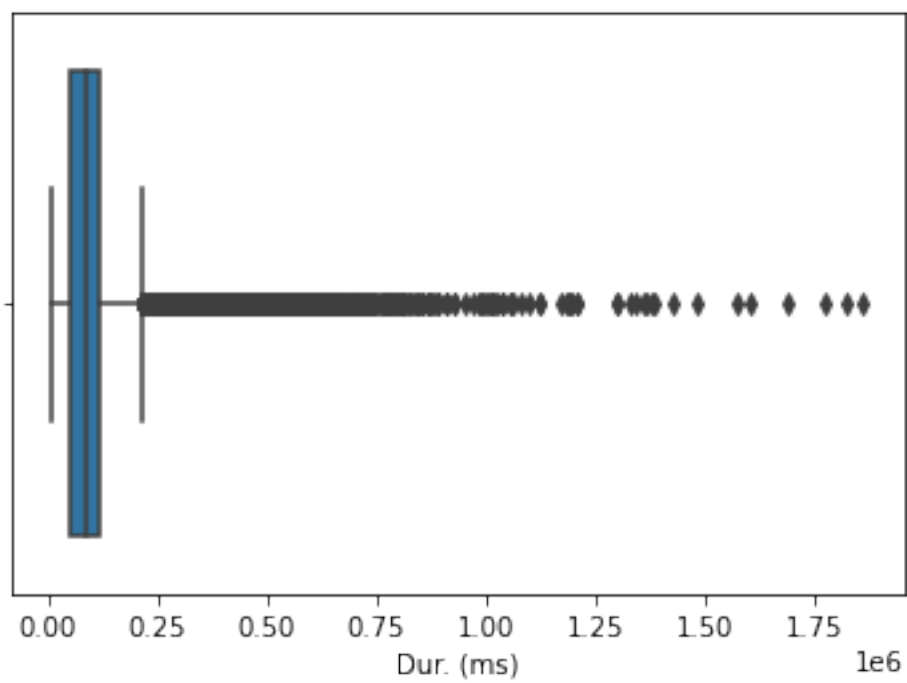
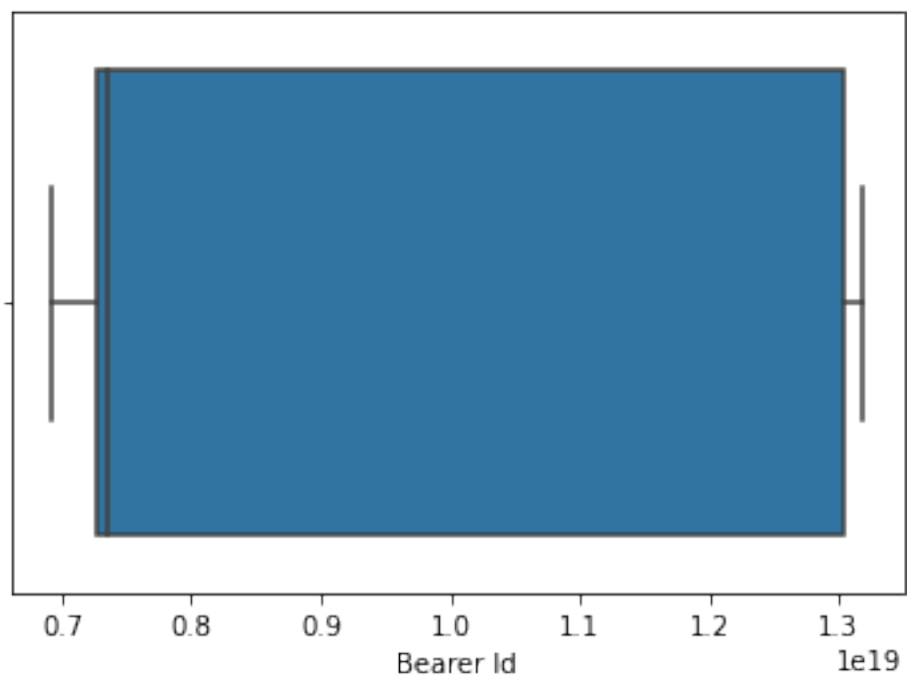
```
# Plot the bar chart using seaborn
plt.figure(figsize=(15,10))
Bar_columns = [ 'Social Media DL (Bytes)', 'Social Media UL
(Bytes)', 'Google DL (Bytes)', 'Google DL (Bytes)', 'Google UL (Bytes)',
                'Email DL (Bytes)', 'Email UL (Bytes)', 'Youtube DL
(Bytes)', 'Youtube UL (Bytes)', 'Netflix DL (Bytes)',
                'Netflix UL (Bytes)', 'Gaming DL (Bytes)', 'Gaming UL
(Bytes)', 'Other DL (Bytes)', 'Other UL (Bytes)']
values = user_behaviour[Bar_columns].sum()
plt.bar(Bar_columns, values)
plt.title('user behaviour data')
plt.xlabel('columns')
plt.ylabel('Values')
```

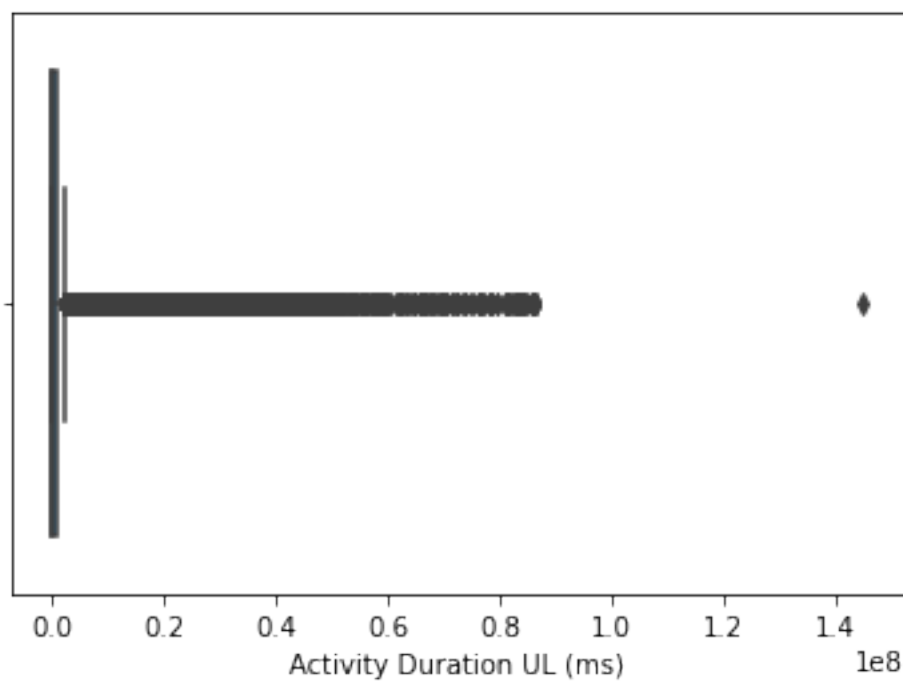
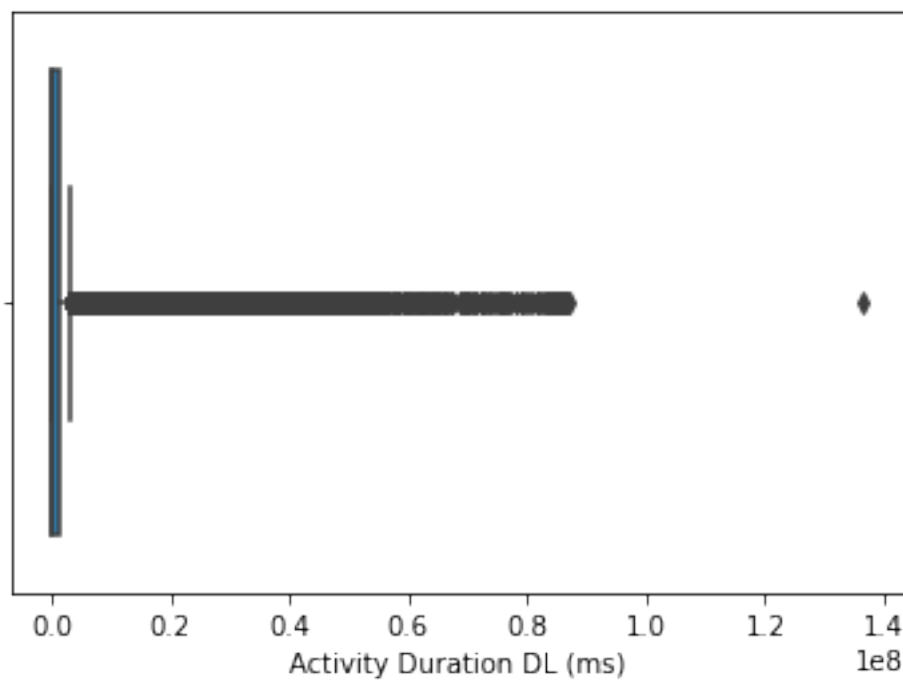
```
plt.xticks(rotation=45)
plt.show()
```

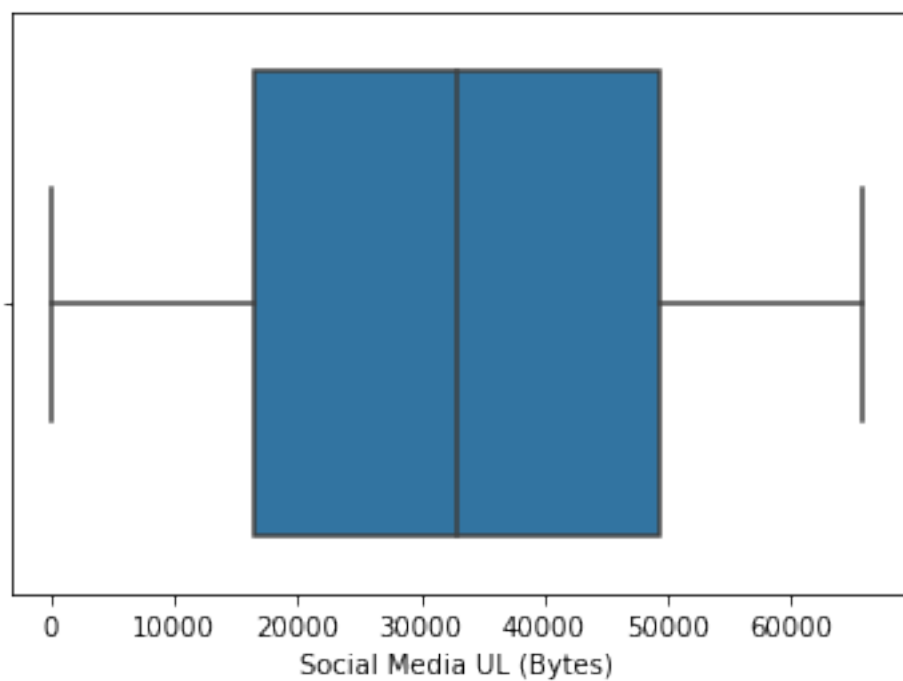
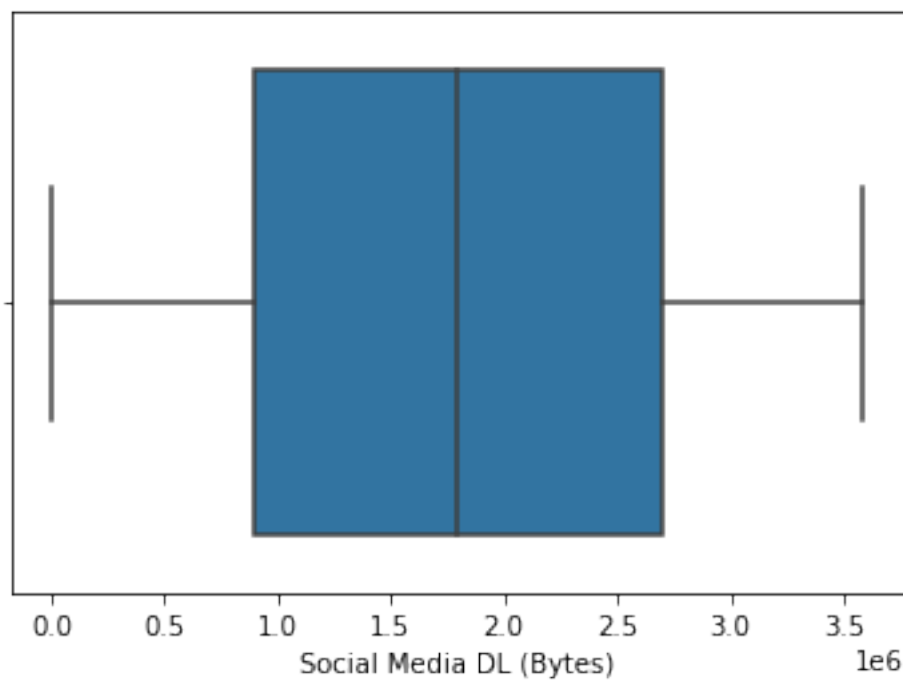


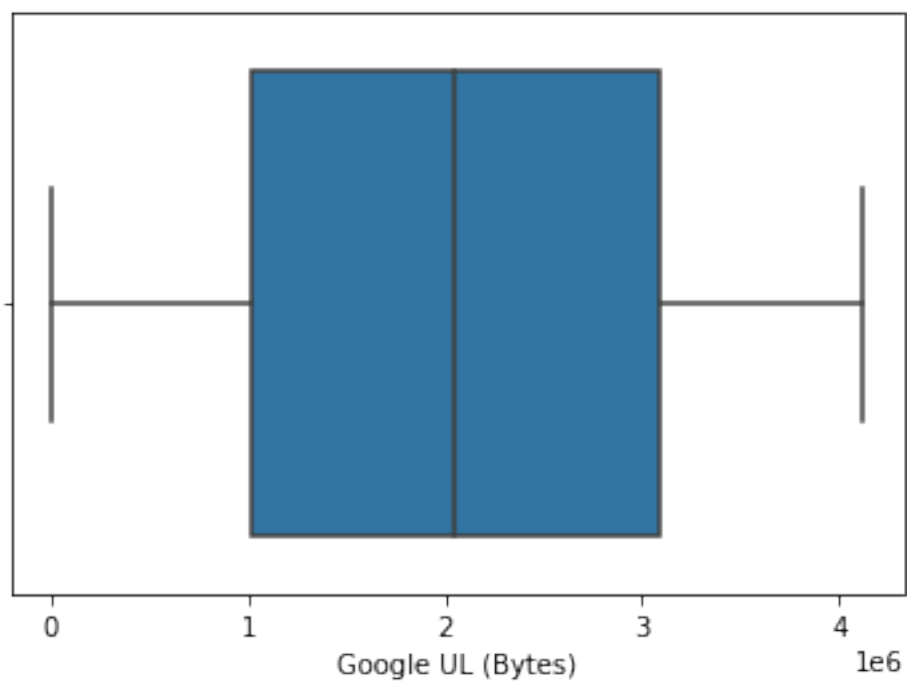
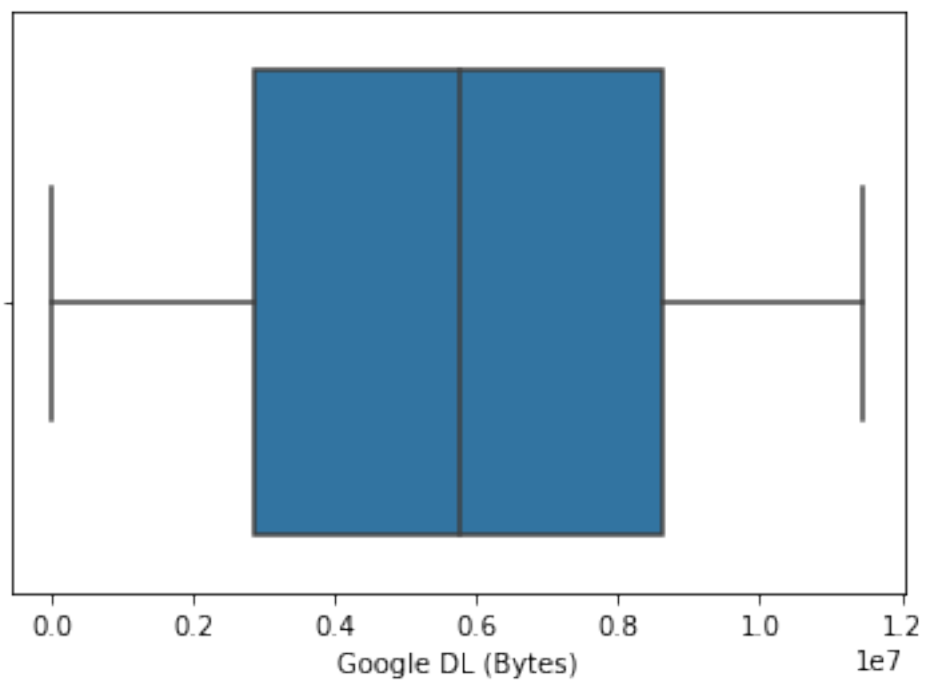
```
for i in data.columns:
    if data[i].dtype == 'object':
        mode_value = data[i].mode()[0]
        data[i].fillna(mode_value, inplace=True)
    else:
        mean_value = data[i].mean()
        data[i].fillna(mean_value, inplace=True)

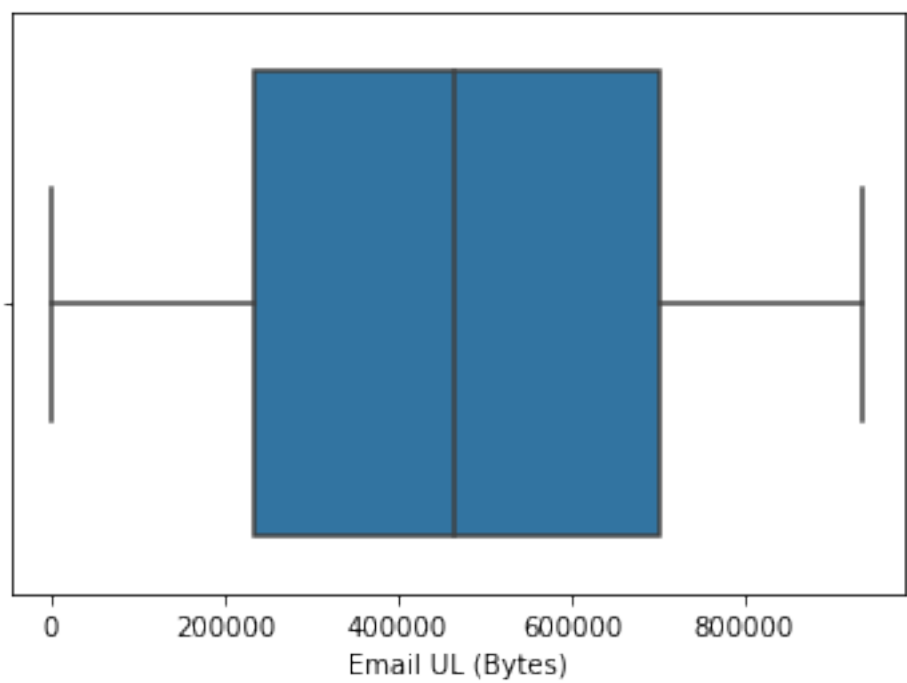
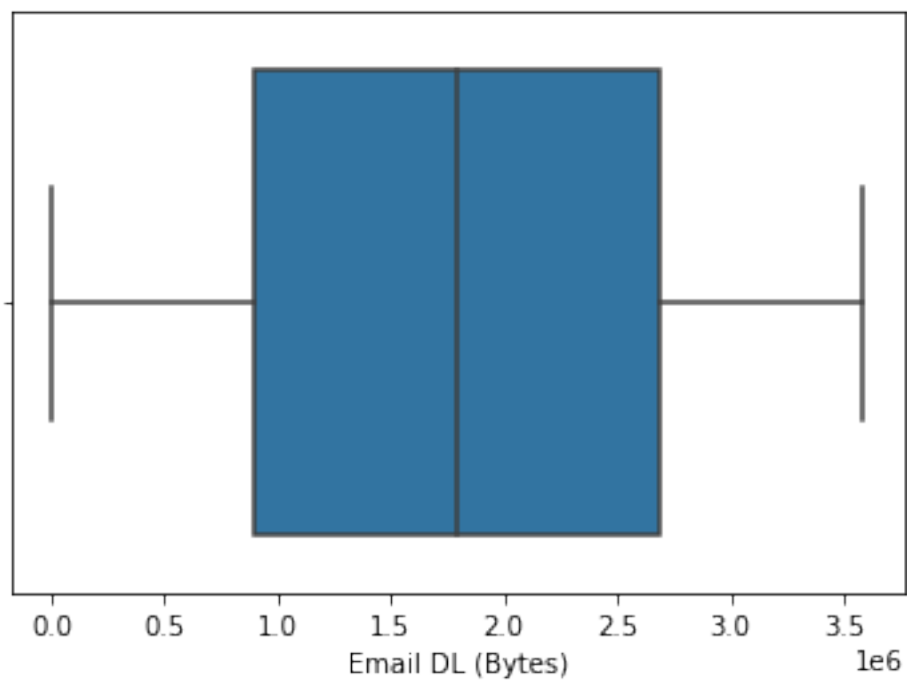
for i in data1.columns:
    sns.boxplot(x=i, data=data)
plt.show()
```

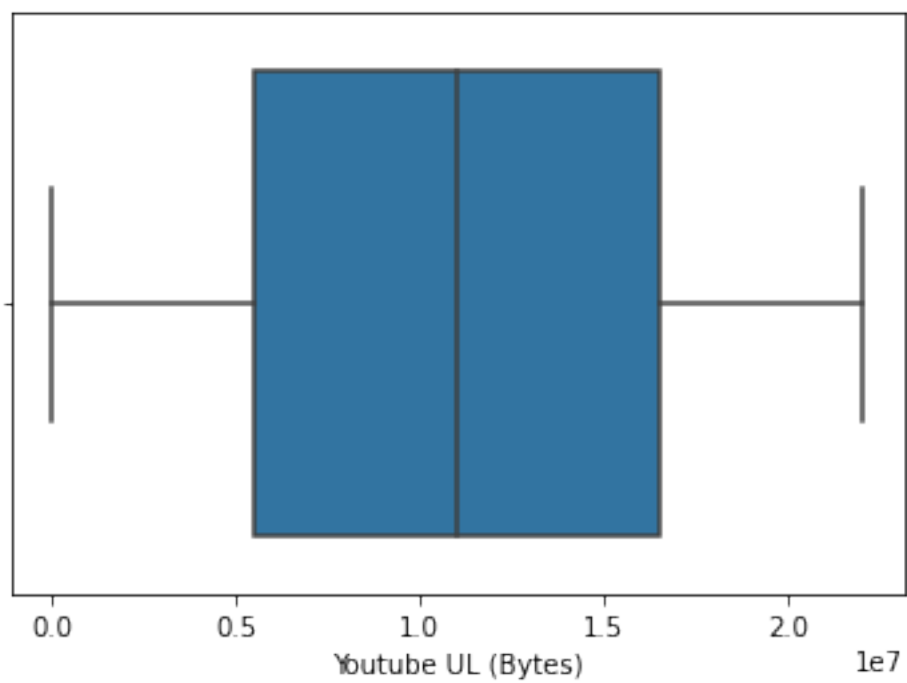
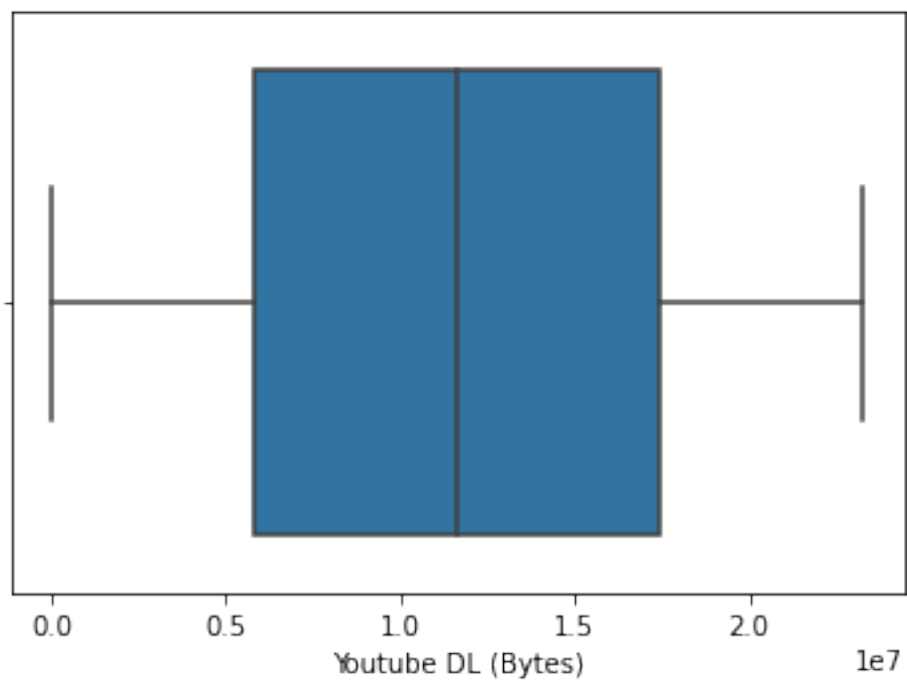


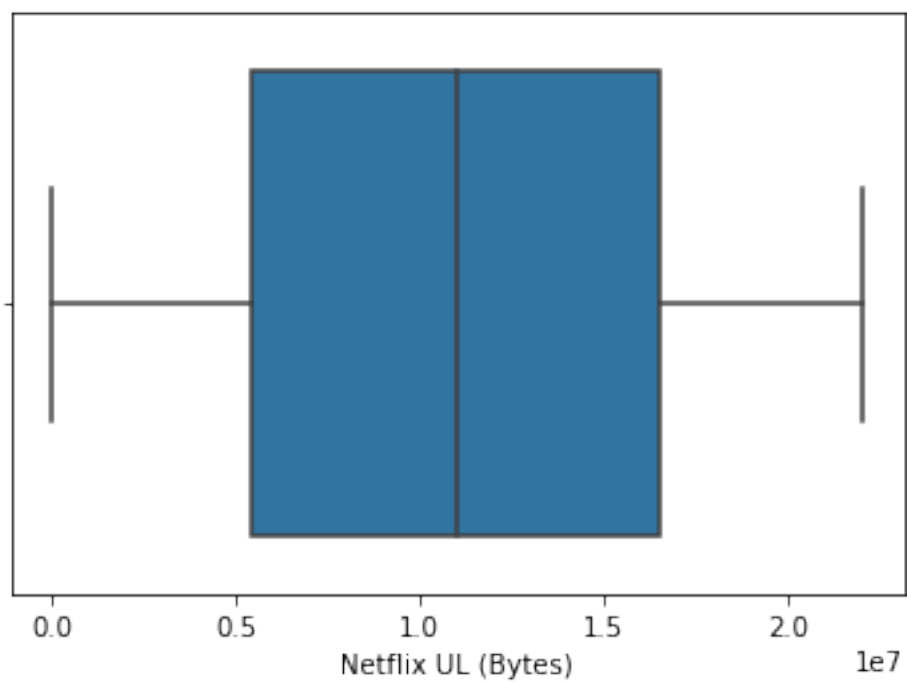
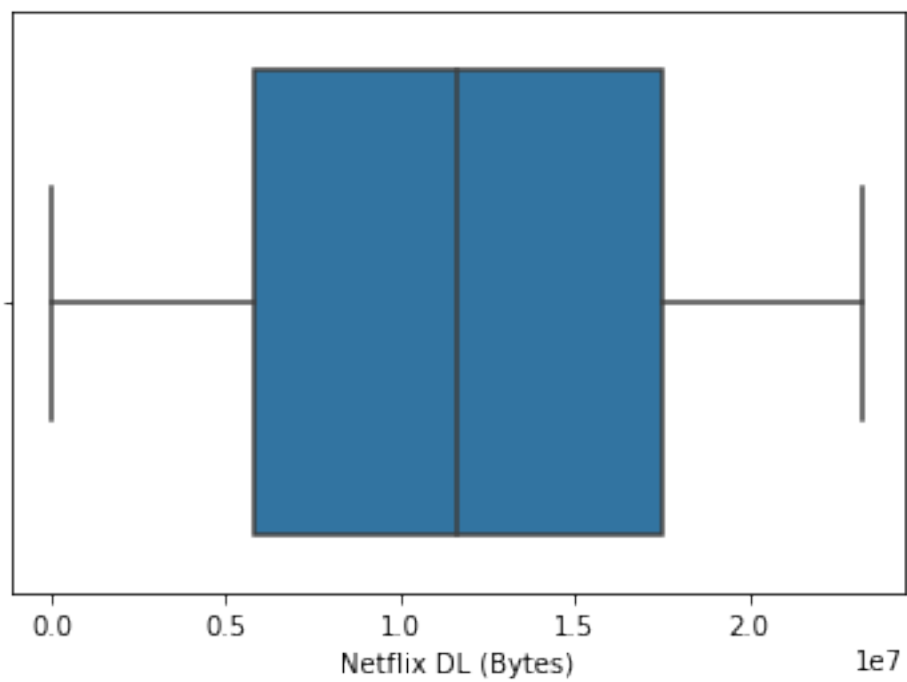


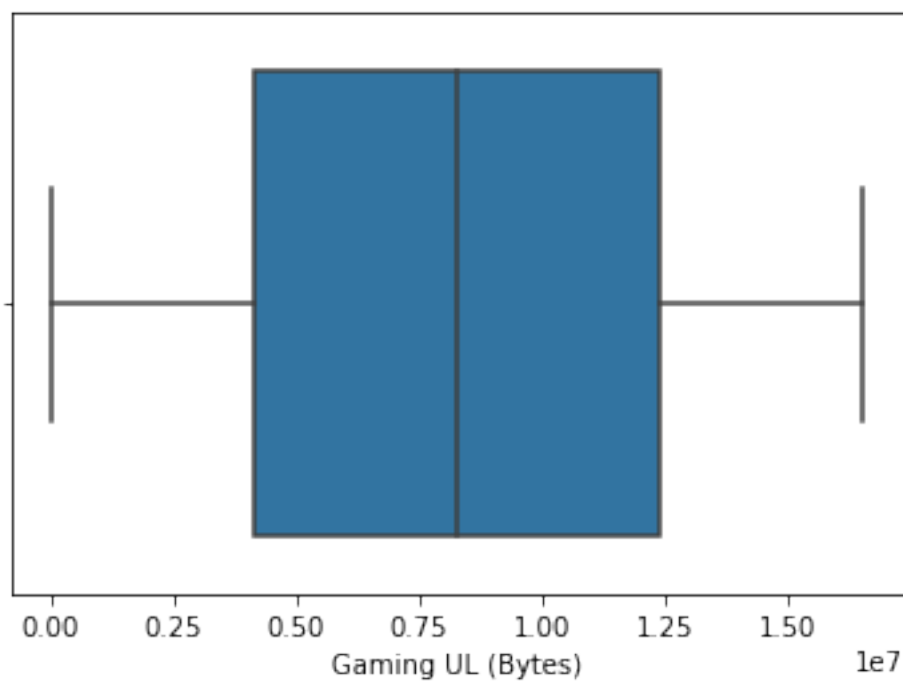
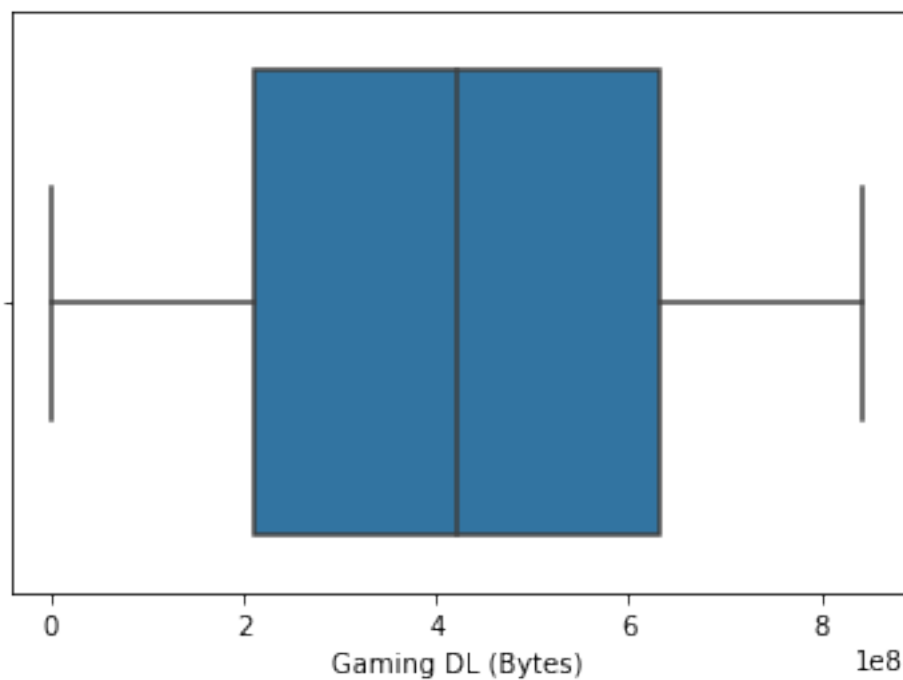


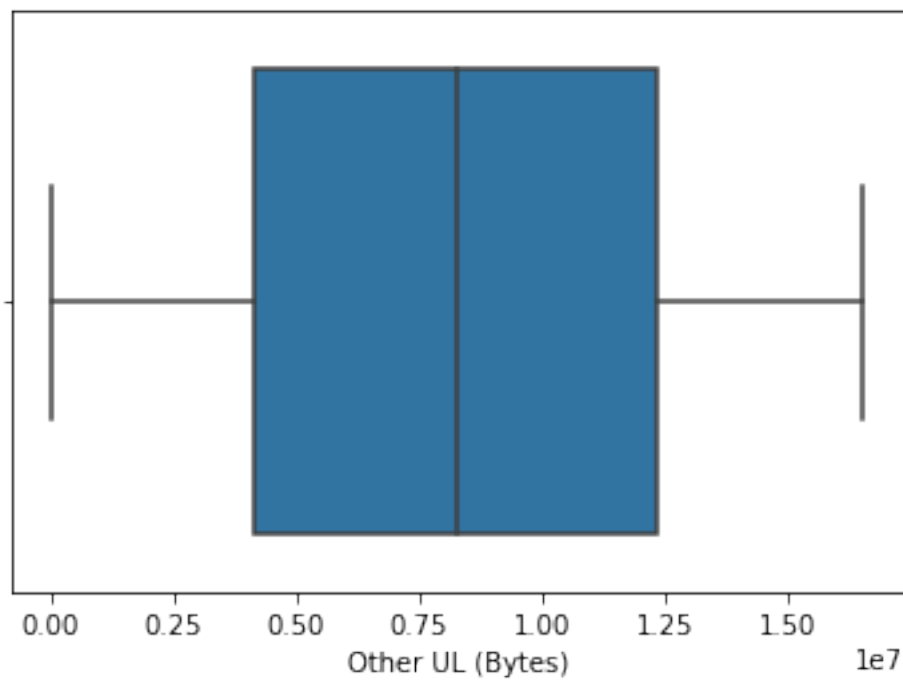
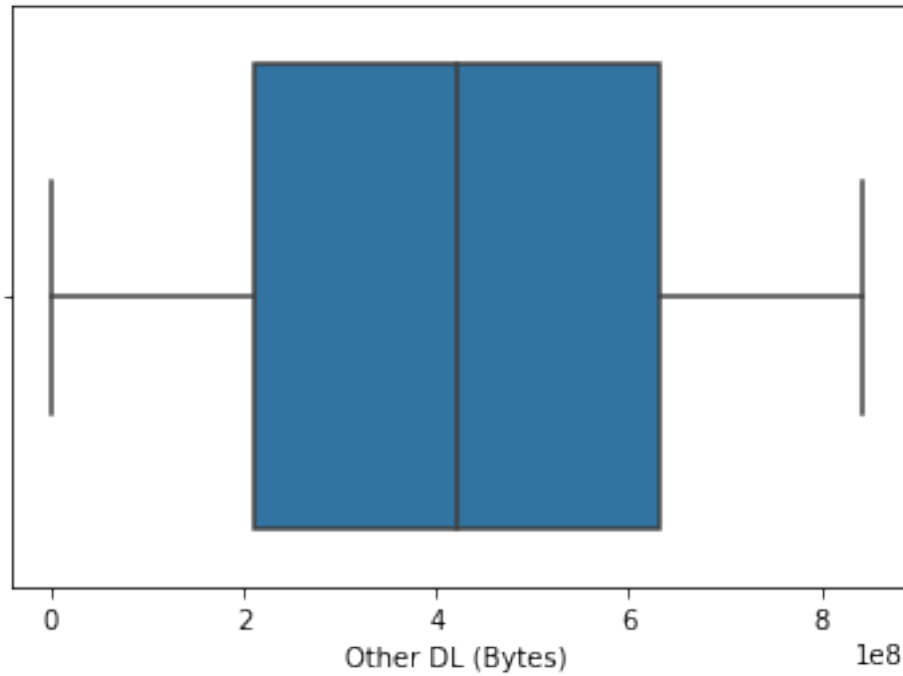












```
user_behaviour.skew()
```

Bearer Id	0.181440
Dur. (ms)	8.478570
Activity Duration DL (ms)	5.583385
Activity Duration UL (ms)	6.859796
Social Media DL (Bytes)	1.199202
Social Media UL (Bytes)	1.118963

Google DL (Bytes)	1.190602
Google UL (Bytes)	1.263205
Email DL (Bytes)	1.207259
Email UL (Bytes)	1.203918
Youtube DL (Bytes)	1.184562
Youtube UL (Bytes)	1.203035
Netflix DL (Bytes)	1.096394
Netflix UL (Bytes)	1.238333
Gaming DL (Bytes)	1.119912
Gaming UL (Bytes)	1.118836
Other DL (Bytes)	1.243104
Other UL (Bytes)	1.174505
dtype: float64	

Task 1.2 - Conduct exploratory data analysis on those data & communicate useful insights. Ensure that you identify and treat all missing values and outliers in the dataset by replacing them with the mean of the corresponding column.

You're expected to report about the following using Python script and slide :

- Describe all relevant variables and associated data types (slide).
- Analyze the basic metrics (mean, median, etc) in the Dataset (explain) & their importance for the global objective.
- Conduct a Non-Graphical Univariate Analysis by computing dispersion parameters for each quantitative variable and provide useful interpretation.
- Conduct a Graphical Univariate Analysis by identifying the most suitable plotting options for each variable and interpreting your findings.
- Bivariate Analysis – explore the relationship between each application & the total DL+UL data using appropriate methods and interpret your findings.
- Variable transformations – segment the users into the top five decile classes based on the total duration for all sessions and compute the total data (DL+UL) per decile class.
- Correlation Analysis – compute a correlation matrix for the following variables and interpret your findings: Social Media data, Google data, Email data, Youtube data, Netflix data, Gaming data, and Other data
- Dimensionality Reduction – perform a principal component analysis to reduce the dimensions of your data and provide a useful interpretation of the results (Provide your interpretation in four (4) bullet points maximum).

data1

	Bearer Id	Dur. (ms)	Activity Duration DL (ms)	\
0	6.917538e+18	24534.0	131798.0	
1	6.917538e+18	27786.0	401941.0	
2	6.917538e+18	15635.0	73347.0	
3	6.917538e+18	24264.0	117340.0	
4	6.917538e+18	15548.0	76969.0	
...	
110427	1.318654e+19	80024.0	2512362.0	
110428	1.318654e+19	145291.0	2067.0	
110429	1.318654e+19	86399.0	3968131.0	
110430	1.318654e+19	86399.0	1689999.0	
110431	1.318654e+19	103113.0	0.0	

	Activity Duration UL (ms)	Social Media DL (Bytes)	\
0	101470.0	2404741.0	
1	399092.0	944612.0	
2	81378.0	1817239.0	
3	347852.0	1867318.0	
4	80241.0	1502346.0	
...	
110427	2437668.0	3240226.0	
110428	45217.0	3062671.0	
110429	3537154.0	720996.0	
110430	1513764.0	2492460.0	
110431	30367.0	1314234.0	

	Social Media UL (Bytes)	Google DL (Bytes)	Google UL (Bytes)
\			
0	2410.0	5791591.0	2871336.0
1	2827.0	10373157.0	56392.0
2	19827.0	269988.0	3696393.0
3	18928.0	1689296.0	195216.0
4	37348.0	8917833.0	701876.0
...
110427	38284.0	2036152.0	2271168.0
110428	48953.0	9363661.0	4001970.0
110429	42836.0	1541915.0	2100839.0
110430	39905.0	11318188.0	466218.0
110431	27938.0	6969652.0	3756009.0

	Email DL (Bytes)	Email UL (Bytes)	Youtube DL (Bytes)	\
0	782388.0	806920.0	6139644.0	
1	128003.0	34038.0	5385159.0	
2	3191192.0	896670.0	12347020.0	
3	740633.0	590043.0	15231815.0	
4	332813.0	537902.0	14602694.0	
...	
110427	2410615.0	387548.0	12404964.0	
110428	2192057.0	866373.0	22147919.0	
110429	2315638.0	839789.0	11879062.0	
110430	2612190.0	618629.0	22163800.0	
110431	3317462.0	408257.0	12099319.0	

	Youtube UL (Bytes)	Netflix DL (Bytes)	Netflix UL (Bytes)	\
0	2071526.0	19494278.0	14668354.0	
1	4295851.0	15755839.0	1300571.0	
2	11089528.0	2859358.0	1738176.0	
3	8401567.0	21563985.0	2817981.0	
4	1237356.0	4506463.0	17006591.0	
...	
110427	4343114.0	11108134.0	21649273.0	
110428	2152449.0	21468525.0	8603105.0	
110429	1290963.0	22596930.0	11943452.0	
110430	16763435.0	9522397.0	8346624.0	
110431	1636122.0	9992219.0	17624886.0	

	Gaming DL (Bytes)	Gaming UL (Bytes)	Other DL (Bytes)	\
0	466109357.0	5333340.0	670751043.0	
1	821879090.0	8521398.0	472846860.0	
2	805301713.0	16257481.0	24303797.0	
3	583864716.0	6992868.0	685122214.0	
4	542337372.0	15820659.0	463698729.0	
...	
110427	114093049.0	2834548.0	695881178.0	
110428	328766801.0	7569327.0	371261255.0	
110429	833634251.0	10607174.0	697260277.0	
110430	338246033.0	1845068.0	17385489.0	
110431	293519955.0	16295588.0	440290470.0	

	Other UL (Bytes)
0	15950724.0
1	1337849.0
2	15907613.0
3	2966860.0
4	11939642.0
...	...
110427	3888729.0
110428	12947410.0


```
110429      9094407.0
110430     12797797.0
110431     6398758.0
```

```
[110432 rows x 18 columns]
```

```
data1.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 110432 entries, 0 to 110431
Data columns (total 18 columns):
```

#	Column	Non-Null Count	Dtype
0	Bearer Id	110432 non-null	float64
1	Dur. (ms)	110432 non-null	float64
2	Activity Duration DL (ms)	110432 non-null	float64
3	Activity Duration UL (ms)	110432 non-null	float64
4	Social Media DL (Bytes)	110432 non-null	float64
5	Social Media UL (Bytes)	110432 non-null	float64
6	Google DL (Bytes)	110432 non-null	float64
7	Google UL (Bytes)	110432 non-null	float64
8	Email DL (Bytes)	110432 non-null	float64
9	Email UL (Bytes)	110432 non-null	float64
10	Youtube DL (Bytes)	110432 non-null	float64
11	Youtube UL (Bytes)	110432 non-null	float64
12	Netflix DL (Bytes)	110432 non-null	float64
13	Netflix UL (Bytes)	110432 non-null	float64
14	Gaming DL (Bytes)	110432 non-null	float64
15	Gaming UL (Bytes)	110432 non-null	float64
16	Other DL (Bytes)	110432 non-null	float64
17	Other UL (Bytes)	110432 non-null	float64

```
dtypes: float64(18)
```

```
memory usage: 15.2 MB
```

```
data1.head()
```

	Bearer Id	Dur. (ms)	Activity Duration DL (ms)	\
0	6.917538e+18	24534.0	131798.0	
1	6.917538e+18	27786.0	401941.0	
2	6.917538e+18	15635.0	73347.0	
3	6.917538e+18	24264.0	117340.0	
4	6.917538e+18	15548.0	76969.0	

	Activity Duration UL (ms)	Social Media DL (Bytes)	\
0	101470.0	2404741.0	
1	399092.0	944612.0	
2	81378.0	1817239.0	
3	347852.0	1867318.0	
4	80241.0	1502346.0	

```
Social Media UL (Bytes)  Google DL (Bytes)  Google UL (Bytes)  \
```

0	2410.0	5791591.0	2871336.0
1	2827.0	10373157.0	56392.0
2	19827.0	269988.0	3696393.0
3	18928.0	1689296.0	195216.0
4	37348.0	8917833.0	701876.0

	Email DL (Bytes)	Email UL (Bytes)	Youtube DL (Bytes)	Youtube UL (Bytes)
0	782388.0	806920.0	6139644.0	2071526.0
1	128003.0	34038.0	5385159.0	4295851.0
2	3191192.0	896670.0	12347020.0	11089528.0
3	740633.0	590043.0	15231815.0	8401567.0
4	332813.0	537902.0	14602694.0	1237356.0

	Netflix DL (Bytes)	Netflix UL (Bytes)	Gaming DL (Bytes)
0	19494278.0	14668354.0	466109357.0
1	15755839.0	1300571.0	821879090.0
2	2859358.0	1738176.0	805301713.0
3	21563985.0	2817981.0	583864716.0
4	4506463.0	17006591.0	542337372.0

	Gaming UL (Bytes)	Other DL (Bytes)	Other UL (Bytes)
0	5333340.0	670751043.0	15950724.0
1	8521398.0	472846860.0	1337849.0
2	16257481.0	24303797.0	15907613.0
3	6992868.0	685122214.0	2966860.0
4	15820659.0	463698729.0	11939642.0

data1.describe()

	Bearer Id	Dur. (ms)	Activity Duration DL (ms)
count	1.104320e+05	1.104320e+05	1.104320e+05
mean	9.907303e+18	1.051041e+05	2.464303e+06
std	2.883127e+18	9.217577e+04	7.058841e+06
min	6.917538e+18	7.142000e+03	0.000000e+00
25%	7.277826e+18	5.291850e+04	2.131075e+04
50%	7.349883e+18	8.639900e+04	6.175800e+04
75%	1.304243e+19	1.357822e+05	1.219347e+06
max	1.318654e+19	4.838428e+06	2.035294e+08

	Activity Duration UL (ms)	Social Media DL (Bytes)
count	1.104320e+05	1.104320e+05
mean	1.893064e+06	1.962976e+06
std	5.711516e+06	1.264421e+06
min	0.000000e+00	1.200000e+01

25%	2.755000e+04	9.625455e+05
50%	6.907150e+04	1.910490e+06
75%	1.040554e+06	2.836990e+06
max	1.666611e+08	2.618898e+07

	Social Media UL (Bytes)	Google DL (Bytes)	Google UL
(Bytes) \			
count	110432.000000	1.104320e+05	1.104320e+05
mean	35980.638031	6.288150e+06	2.247206e+06
std	23141.331972	4.048132e+06	1.456667e+06
min	0.000000	2.970000e+02	3.000000e+00
25%	17606.750000	3.090234e+06	1.097197e+06
50%	35040.000000	6.122278e+06	2.181378e+06
75%	51920.000000	9.070152e+06	3.253726e+06
max	478166.000000	9.556064e+07	3.861748e+07

	Email DL (Bytes)	Email UL (Bytes)	Youtube DL (Bytes)	\
count	1.104320e+05	1.104320e+05	1.104320e+05	
mean	1.958303e+06	5.107500e+05	1.272504e+07	
std	1.263999e+06	3.300887e+05	8.183277e+06	
min	1.400000e+01	2.000000e+00	5.300000e+01	
25%	9.572890e+05	2.499518e+05	6.278096e+06	
50%	1.904660e+06	4.949325e+05	1.235056e+07	
75%	2.830856e+06	7.386615e+05	1.839123e+07	
max	3.379861e+07	6.744930e+06	1.796289e+08	

	Youtube UL (Bytes)	Netflix DL (Bytes)	Netflix UL (Bytes)	\
count	1.104320e+05	1.104320e+05	1.104320e+05	
mean	1.203117e+07	1.271781e+07	1.202267e+07	
std	7.756424e+06	8.171737e+06	7.759512e+06	
min	1.050000e+02	4.200000e+01	3.500000e+01	
25%	5.896152e+06	6.219047e+06	5.876785e+06	
50%	1.169876e+07	1.235935e+07	1.170558e+07	
75%	1.737368e+07	1.840726e+07	1.738361e+07	
max	1.879087e+08	1.452646e+08	2.277824e+08	

	Gaming DL (Bytes)	Gaming UL (Bytes)	Other DL (Bytes)	\
count	1.104320e+05	1.104320e+05	1.104320e+05	
mean	4.608041e+08	9.066019e+06	4.607876e+08	
std	2.972070e+08	5.831056e+06	2.966063e+08	
min	1.251100e+04	5.900000e+01	4.626900e+04	

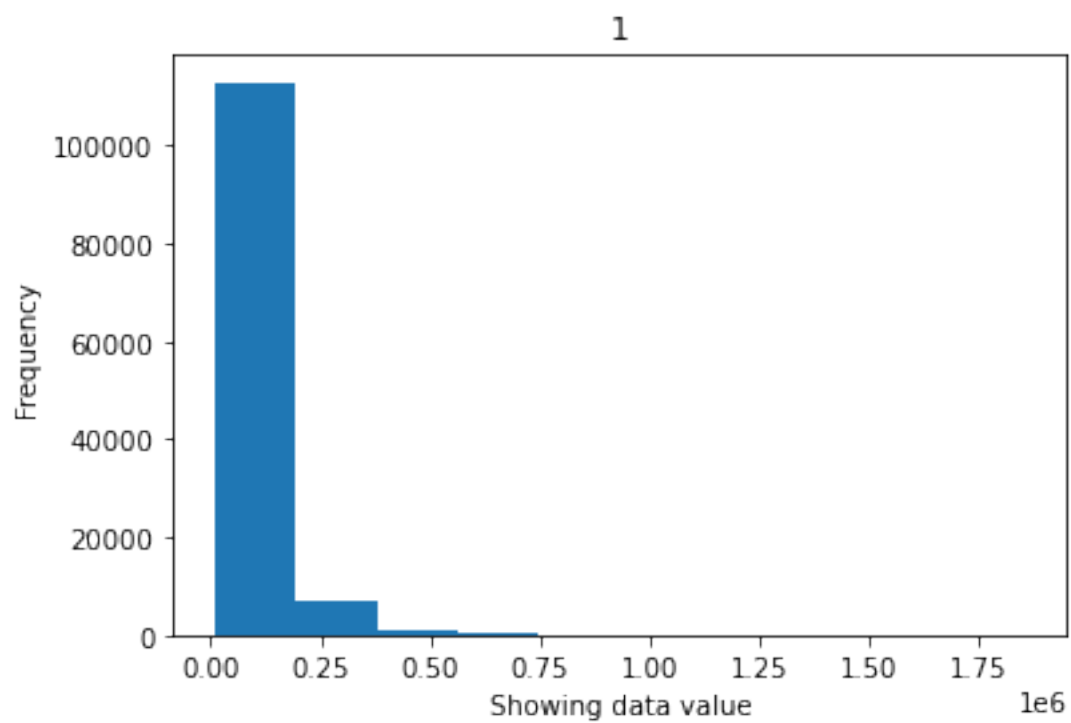
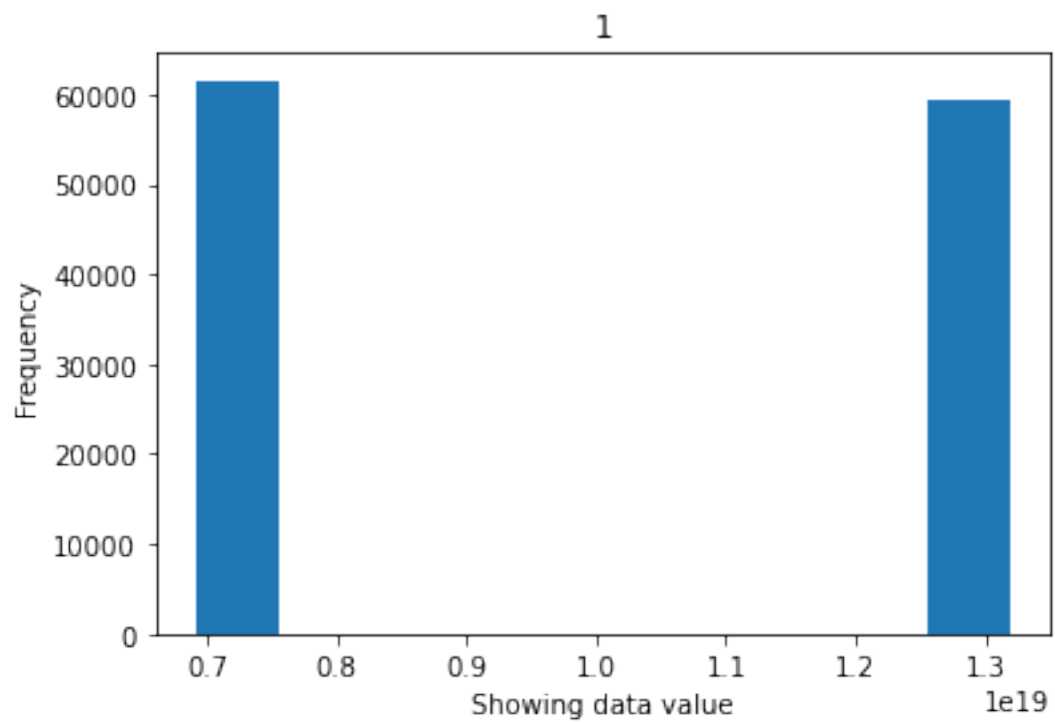
25%	2.254630e+08	4.442423e+06	2.262967e+08
50%	4.489663e+08	8.820961e+06	4.482374e+08
75%	6.660761e+08	1.305845e+07	6.657577e+08
max	5.808028e+09	1.118985e+08	8.051625e+09

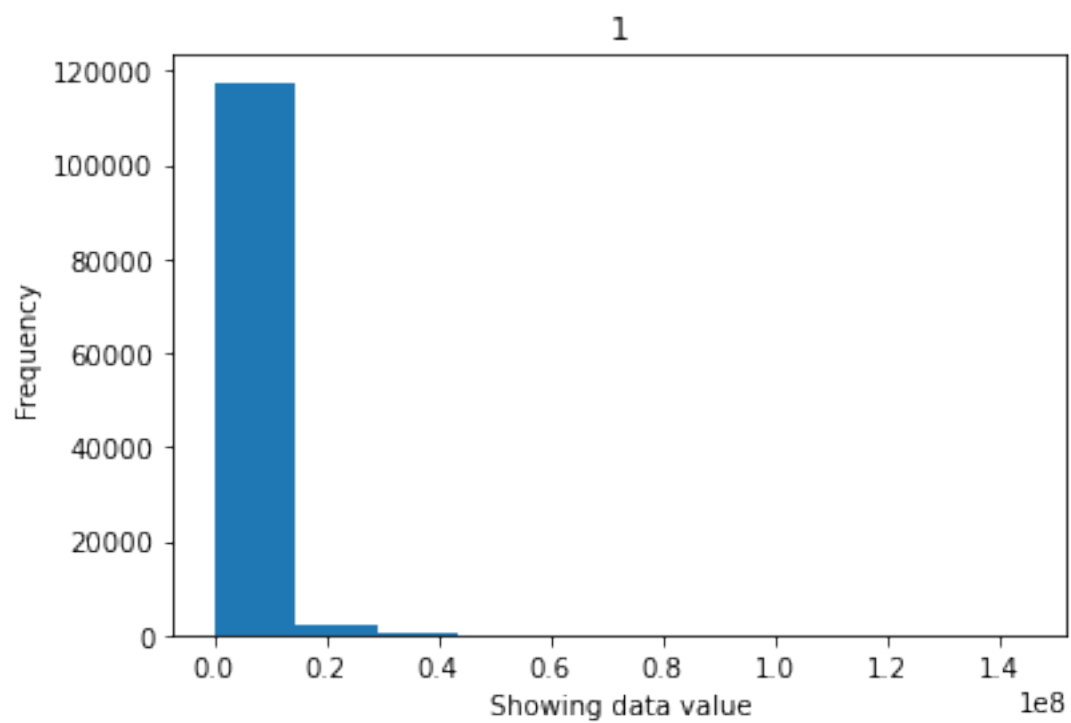
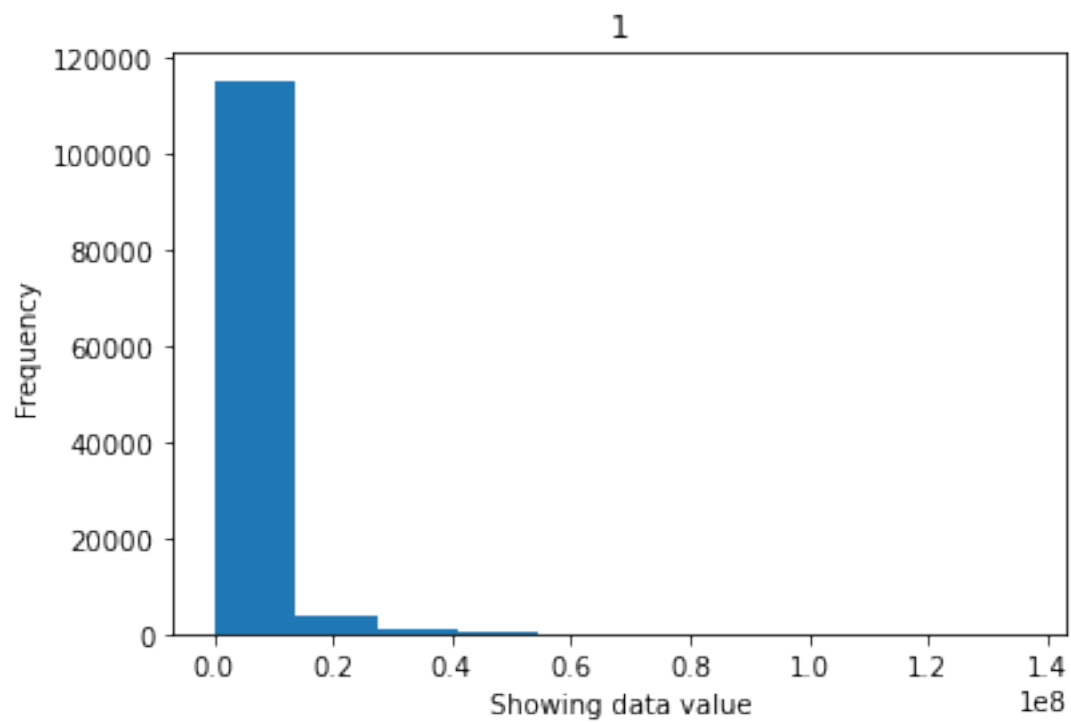
	Other UL (Bytes)
count	1.104320e+05
mean	9.024526e+06
std	5.824731e+06
min	1.480000e+02
25%	4.445894e+06
50%	8.754301e+06
75%	1.302956e+07
max	1.287153e+08

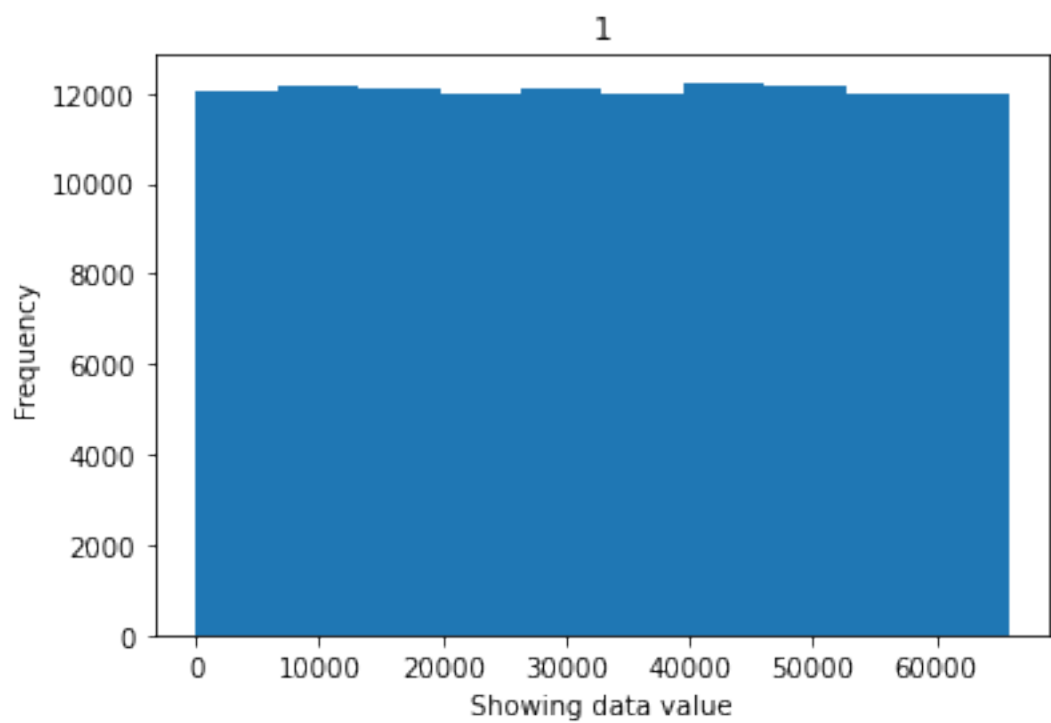
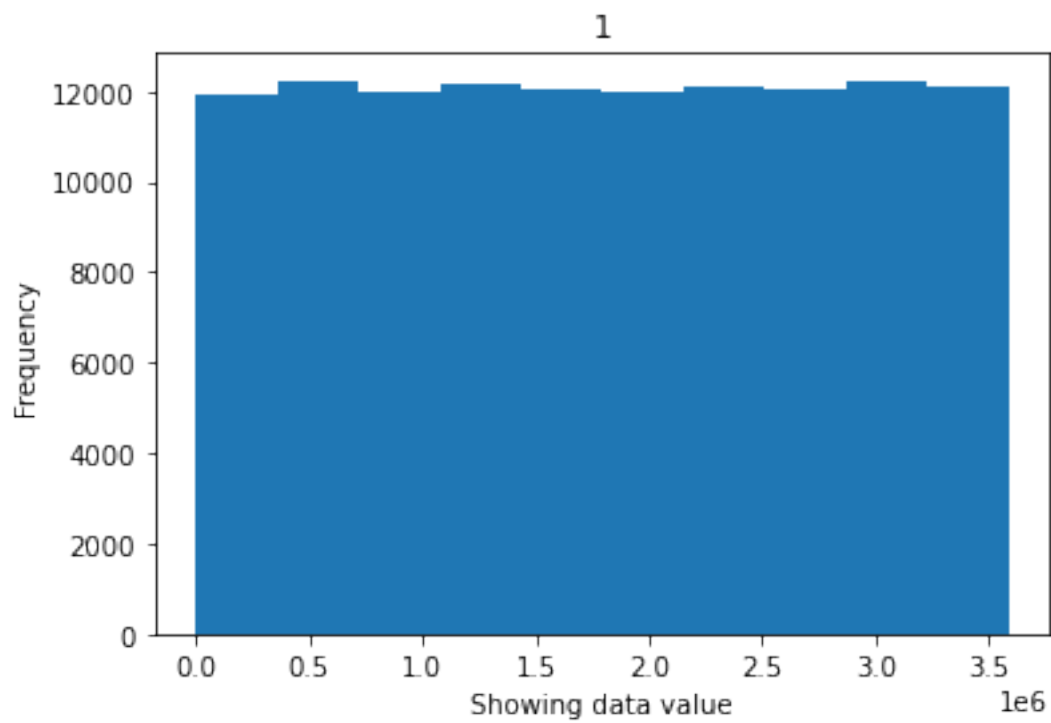
data1.columns

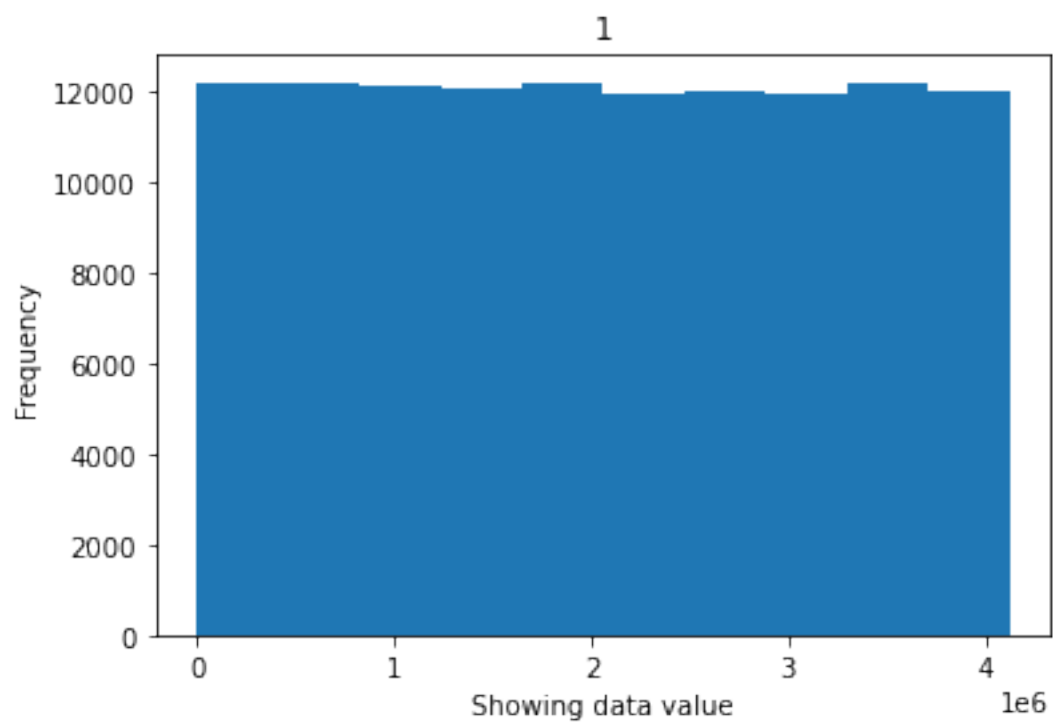
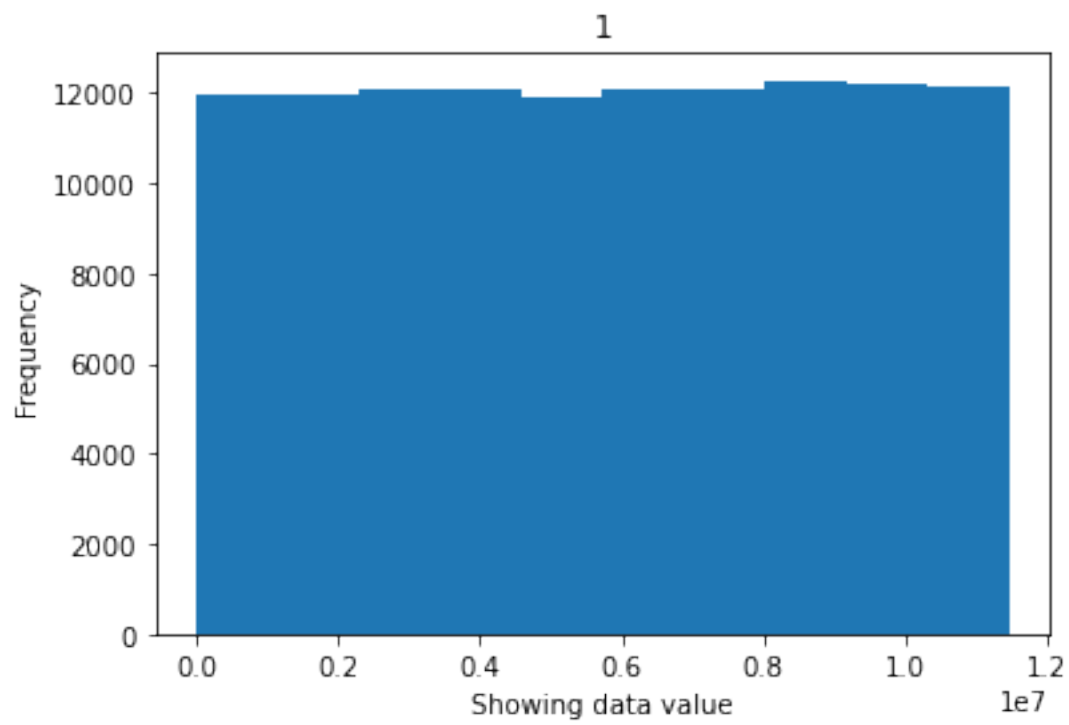
```
Index(['Bearer Id', 'Dur. (ms)', 'Activity Duration DL (ms)',
      'Activity Duration UL (ms)', 'Social Media DL (Bytes)',
      'Social Media UL (Bytes)', 'Google DL (Bytes)', 'Google UL
(Bytes)',
      'Email DL (Bytes)', 'Email UL (Bytes)', 'Youtube DL (Bytes)',
      'Youtube UL (Bytes)', 'Netflix DL (Bytes)', 'Netflix UL
(Bytes)',
      'Gaming DL (Bytes)', 'Gaming UL (Bytes)', 'Other DL (Bytes)',
      'Other UL (Bytes)'],
      dtype='object')
```

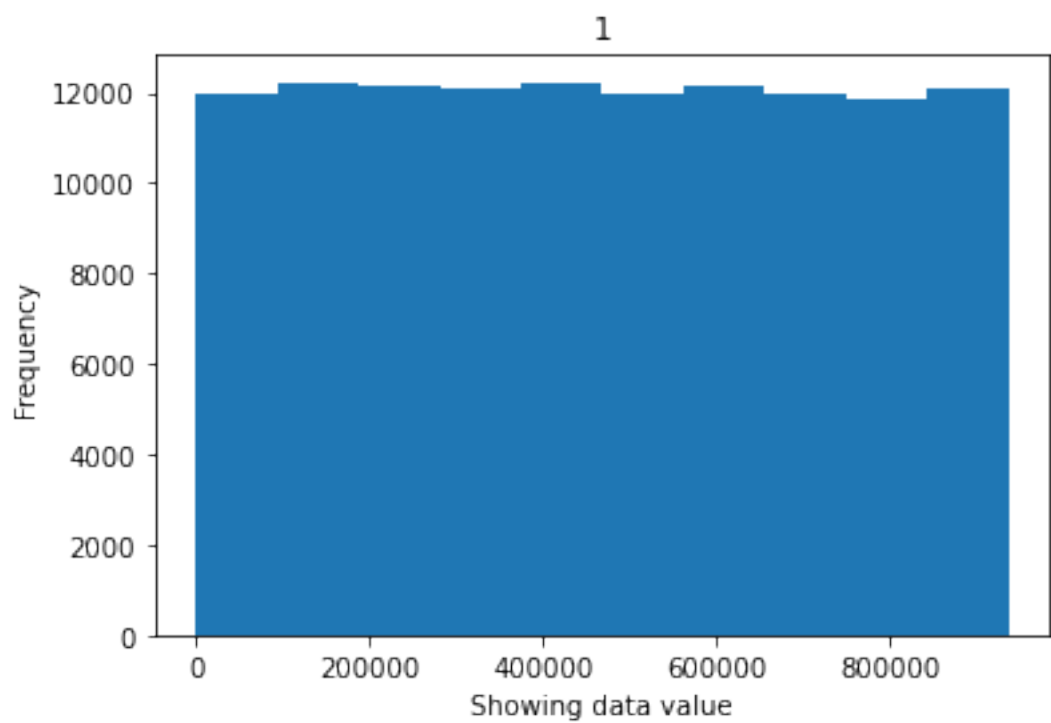
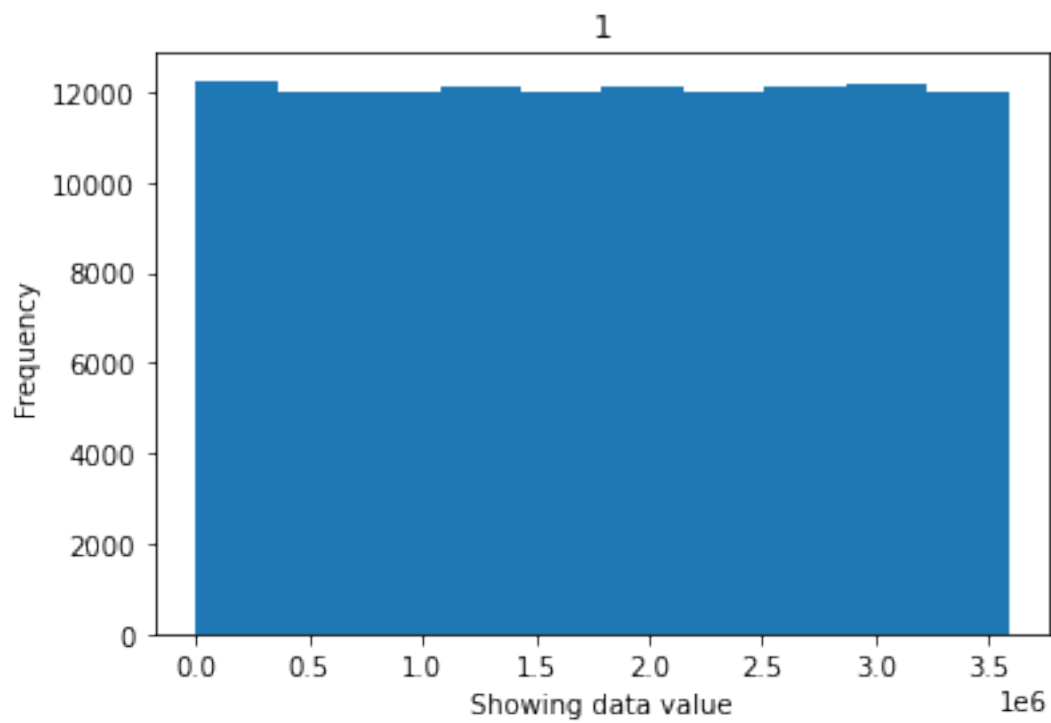
```
for i in data1.columns:
    plt.hist(data[i],bins=10)
    plt.title(1)
    plt.xlabel("Showing data value")
    plt.ylabel("Frequency")
    plt.savefig(f"{i}_histogram.png")
    plt.show()
```

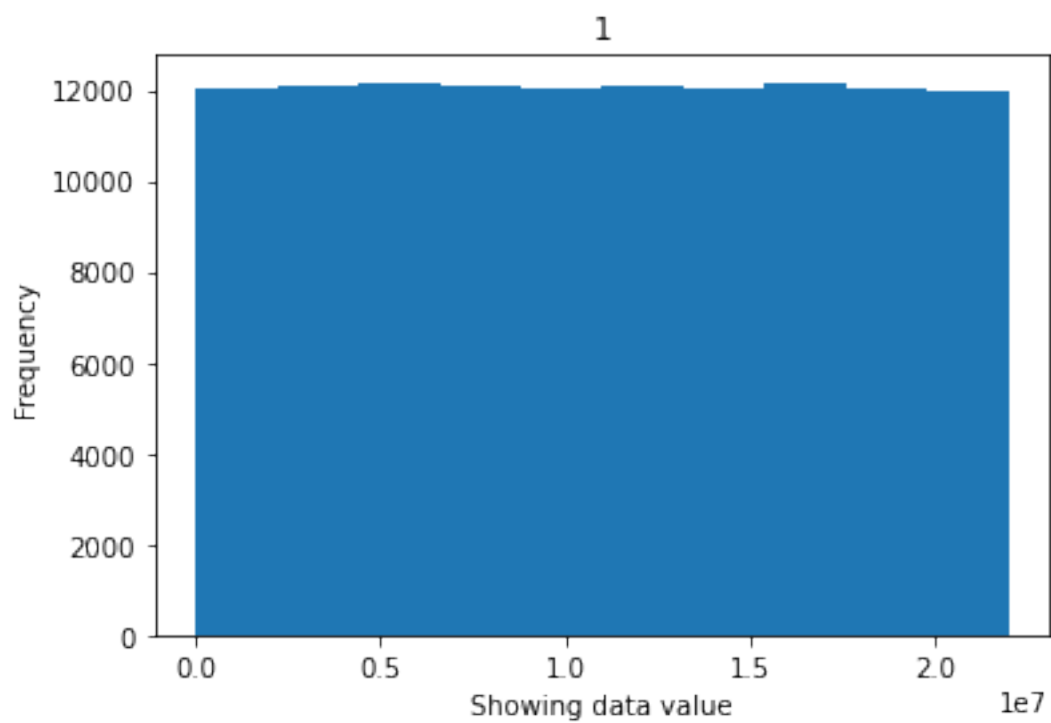
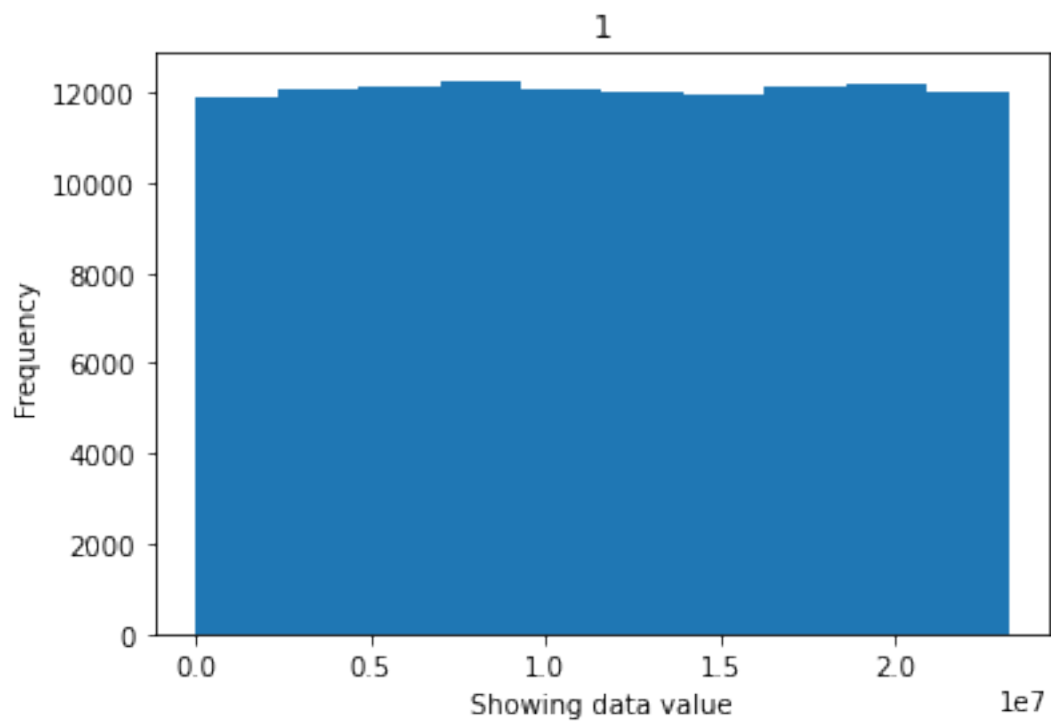


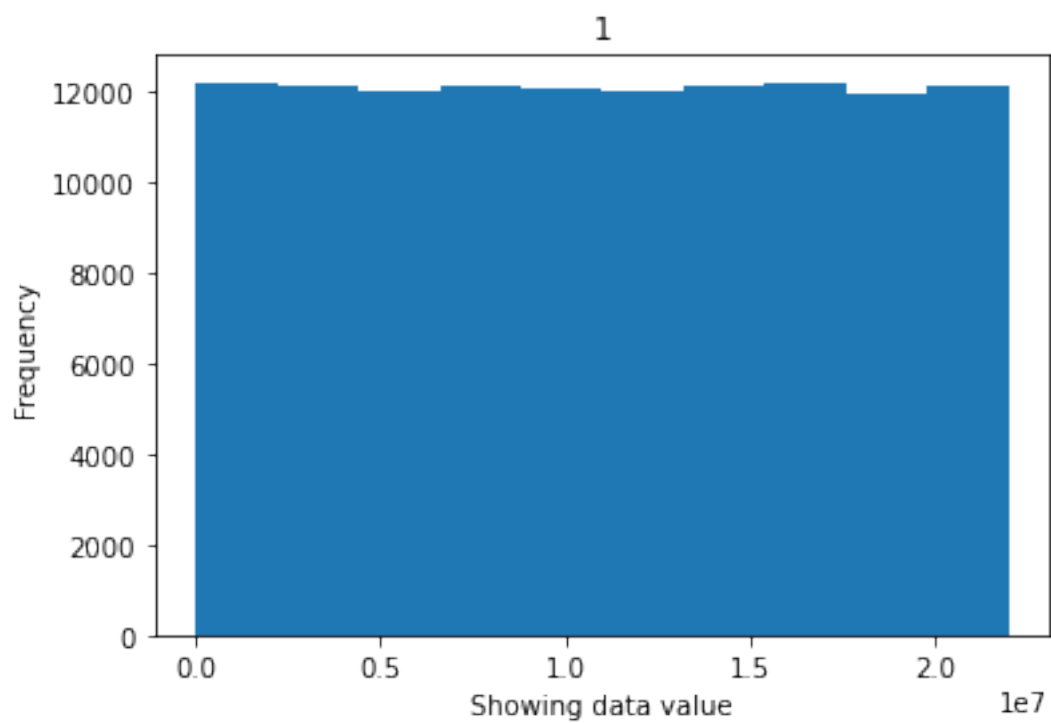
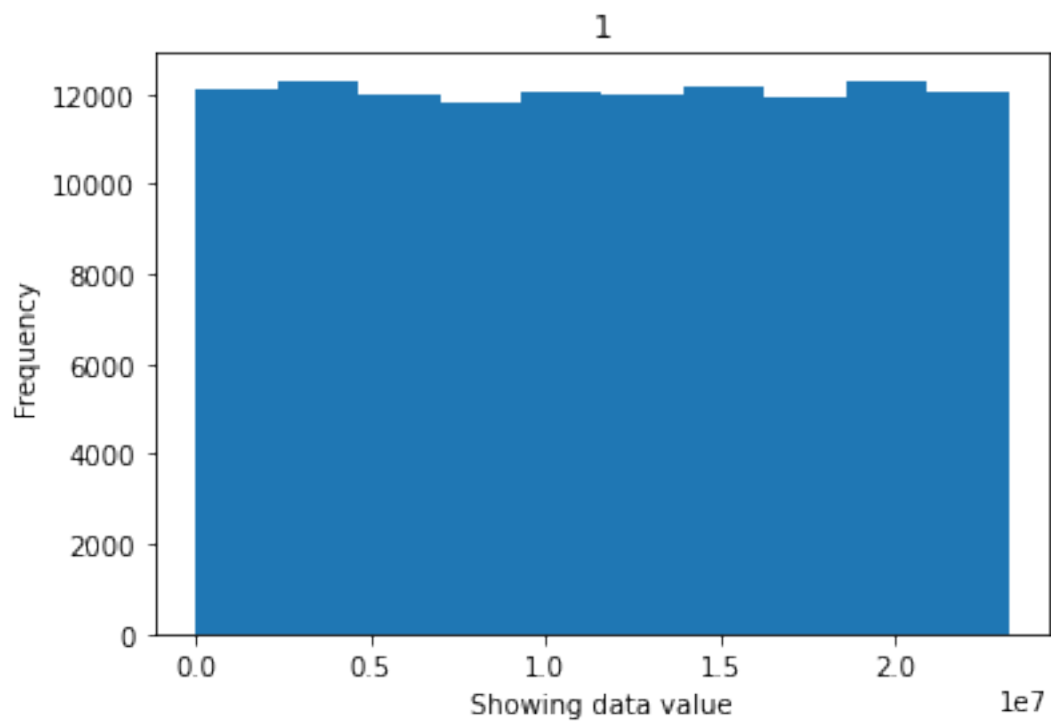


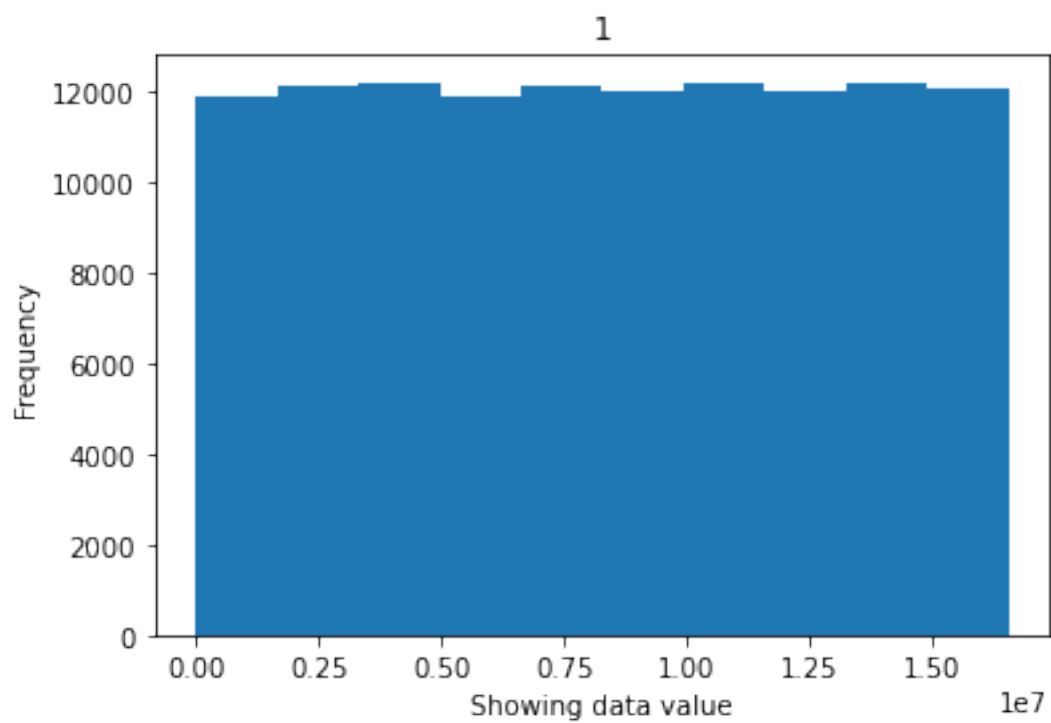
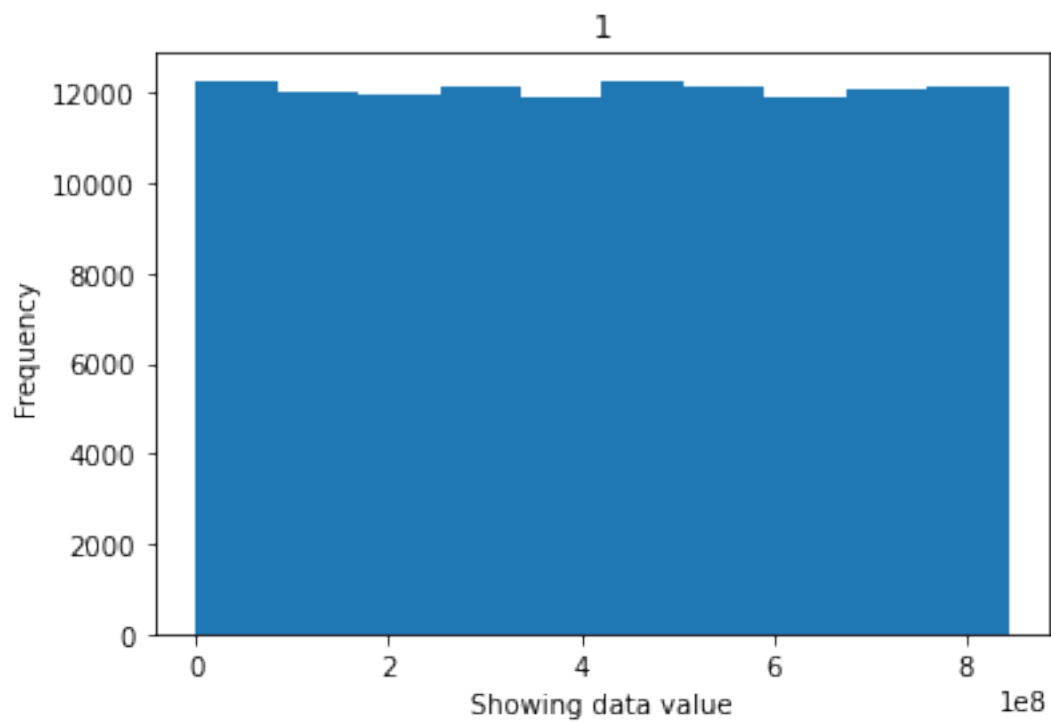


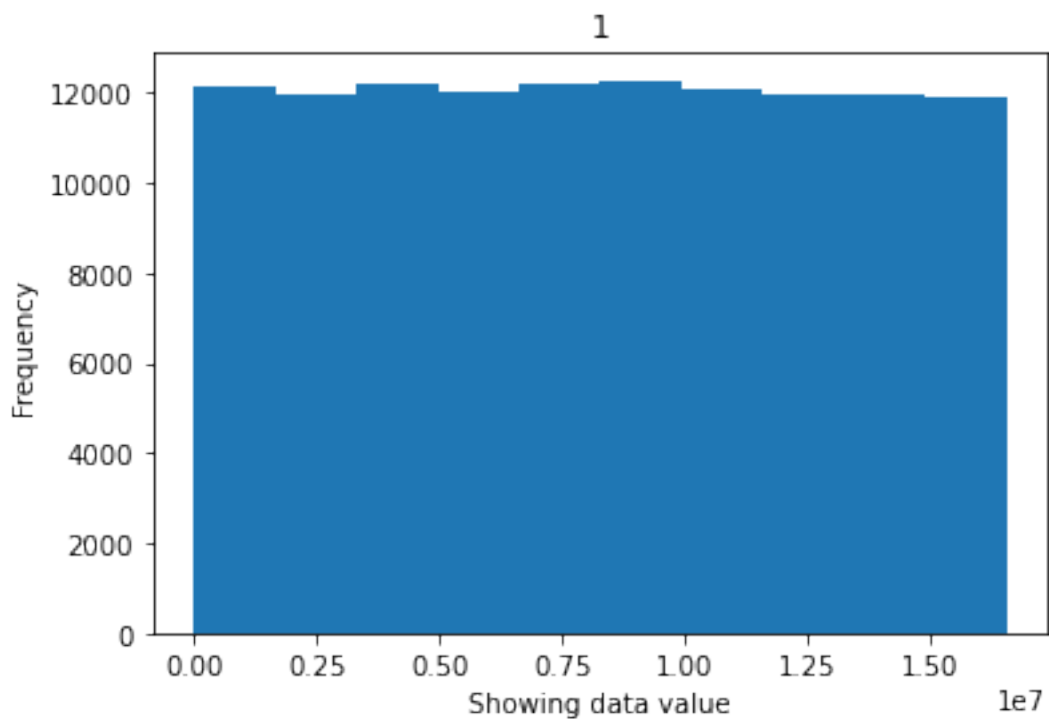
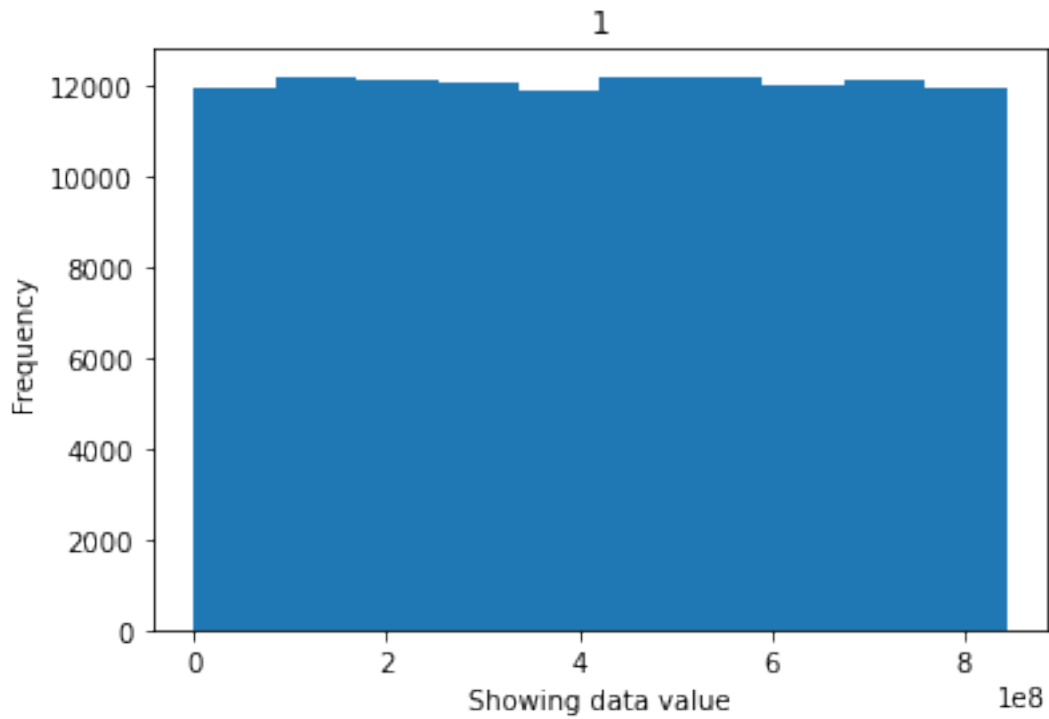












```
top_10_handsets = data["Handset
Type"].value_counts(ascending=False).head(10)
print(top_10_handsets)
```

Huawei B528S-23A	19310
Apple iPhone 6S (A1688)	9244

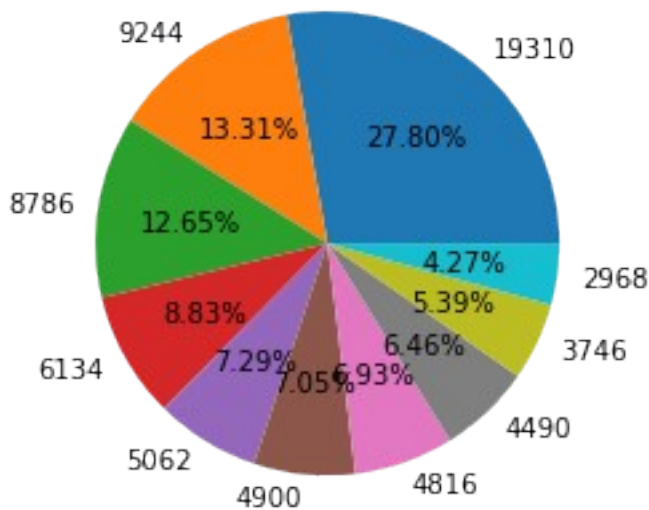
```

Apple iPhone 6 (A1586)      8786
Apple iPhone 7 (A1778)      6134
Apple iPhone Se (A1723)     5062
Apple iPhone 8 (A1905)      4900
undefined                   4816
Apple iPhone Xr (A2105)      4490
Apple iPhone X (A1901)       3746
Apple iPhone 8 Plus (A1897)  2968
Name: Handset Type, dtype: int64

```

```
data['top_10_handsets']=top_10_handsets
```

```
plt.pie(x=top_10_handsets,labels=top_10_handsets,autopct = '%1.2f%%')
plt.show()
```



```

top_3_Manufacturer = data["Handset
Manufacturer"].value_counts(ascending=False).head(3)
print(top_3_Manufacturer )

```

```

Apple      58190
Huawei      33306
Samsung    19167
Name: Handset Manufacturer, dtype: int64

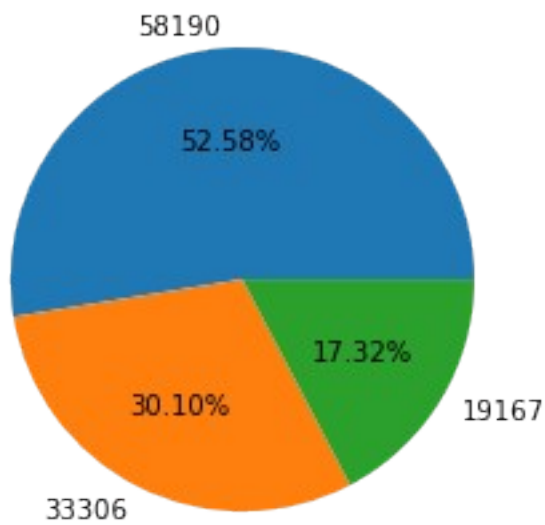
```

```
data['top_10_handsets']=top_10_handsets
```

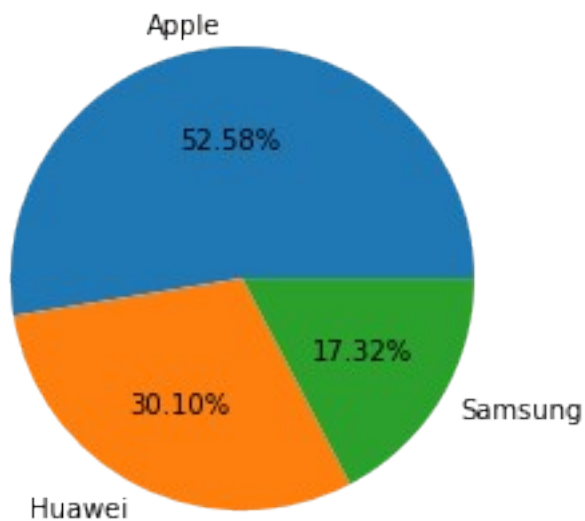
```

plt.pie(x=top_3_Manufacturer,labels=top_3_Manufacturer,autopct = '%1.2f
%%')
plt.show()

```



```
plt.pie(labels=top_3_Manufacturer.index,x=top_3_Manufacturer,autopct =
"%1.2f%%")
plt.show();
```



```
data1.columns
```

```
Index(['Bearer Id', 'Dur. (ms)', 'Activity Duration DL (ms)',
      'Activity Duration UL (ms)', 'Social Media DL (Bytes)',
      'Social Media UL (Bytes)', 'Google DL (Bytes)', 'Google UL
(Bytes)',
      'Email DL (Bytes)', 'Email UL (Bytes)', 'Youtube DL (Bytes)',
      'Youtube UL (Bytes)', 'Netflix DL (Bytes)', 'Netflix UL
(Bytes)'],
      dtype='object', name='Index')
```

```

        'Gaming DL (Bytes)', 'Gaming UL (Bytes)', 'Other DL (Bytes)',
        'Other UL (Bytes)'],
dtype='object')

```

data1.skew

<bound method NDFrame._add_numeric_operations.<locals>.skew of

Bearer Id	Dur. (ms)	Activity Duration DL (ms)	\
0	6.917538e+18	24534.0	131798.0
1	6.917538e+18	27786.0	401941.0
2	6.917538e+18	15635.0	73347.0
3	6.917538e+18	24264.0	117340.0
4	6.917538e+18	15548.0	76969.0
...
110427	1.318654e+19	80024.0	2512362.0
110428	1.318654e+19	145291.0	2067.0
110429	1.318654e+19	86399.0	3968131.0
110430	1.318654e+19	86399.0	1689999.0
110431	1.318654e+19	103113.0	0.0

	Activity Duration UL (ms)	Social Media DL (Bytes)	\
0	101470.0	2404741.0	
1	399092.0	944612.0	
2	81378.0	1817239.0	
3	347852.0	1867318.0	
4	80241.0	1502346.0	
...
110427	2437668.0	3240226.0	
110428	45217.0	3062671.0	
110429	3537154.0	720996.0	
110430	1513764.0	2492460.0	
110431	30367.0	1314234.0	

	Social Media UL (Bytes)	Google DL (Bytes)	Google UL (Bytes)
\			
0	2410.0	5791591.0	2871336.0
1	2827.0	10373157.0	56392.0
2	19827.0	269988.0	3696393.0
3	18928.0	1689296.0	195216.0
4	37348.0	8917833.0	701876.0
...
110427	38284.0	2036152.0	2271168.0
110428	48953.0	9363661.0	4001970.0

110429	42836.0	1541915.0	2100839.0
110430	39905.0	11318188.0	466218.0
110431	27938.0	6969652.0	3756009.0

	Email DL (Bytes)	Email UL (Bytes)	Youtube DL (Bytes)	\
0	782388.0	806920.0	6139644.0	
1	128003.0	34038.0	5385159.0	
2	3191192.0	896670.0	12347020.0	
3	740633.0	590043.0	15231815.0	
4	332813.0	537902.0	14602694.0	
...	
110427	2410615.0	387548.0	12404964.0	
110428	2192057.0	866373.0	22147919.0	
110429	2315638.0	839789.0	11879062.0	
110430	2612190.0	618629.0	22163800.0	
110431	3317462.0	408257.0	12099319.0	

	Youtube UL (Bytes)	Netflix DL (Bytes)	Netflix UL (Bytes)	\
0	2071526.0	19494278.0	14668354.0	
1	4295851.0	15755839.0	1300571.0	
2	11089528.0	2859358.0	1738176.0	
3	8401567.0	21563985.0	2817981.0	
4	1237356.0	4506463.0	17006591.0	
...	
110427	4343114.0	11108134.0	21649273.0	
110428	2152449.0	21468525.0	8603105.0	
110429	1290963.0	22596930.0	11943452.0	
110430	16763435.0	9522397.0	8346624.0	
110431	1636122.0	9992219.0	17624886.0	

	Gaming DL (Bytes)	Gaming UL (Bytes)	Other DL (Bytes)	\
0	466109357.0	5333340.0	670751043.0	
1	821879090.0	8521398.0	472846860.0	
2	805301713.0	16257481.0	24303797.0	
3	583864716.0	6992868.0	685122214.0	
4	542337372.0	15820659.0	463698729.0	
...	
110427	114093049.0	2834548.0	695881178.0	
110428	328766801.0	7569327.0	371261255.0	
110429	833634251.0	10607174.0	697260277.0	
110430	338246033.0	1845068.0	17385489.0	
110431	293519955.0	16295588.0	440290470.0	

	Other UL (Bytes)
0	15950724.0

```

1          1337849.0
2          15907613.0
3           2966860.0
4          11939642.0
...
110427      3888729.0
110428      12947410.0
110429       9094407.0
110430      12797797.0
110431       6398758.0

```

```
[110432 rows x 18 columns]>
```

```

for i in data.columns:
    if data[i].dtypes == "object":
        mode_value = data[i].mode()[0]
        data[i].fillna(mode_value, inplace = True)
    elif data[i].dtypes != "object":
        mean_value = data[i].mean()
        data[i].fillna(mean_value, inplace = True)

```

```
data.isnull().sum()
```

```

Bearer Id          0
Start              0
Start ms           0
End                0
End ms             0
Dur. (ms)          0
IMSI               0
MSISDN/Number      0
IMEI               0
Last Location Name 0
Avg RTT DL (ms)    0
Avg RTT UL (ms)    0
Avg Bearer TP DL (kbps) 0
Avg Bearer TP UL (kbps) 0
DL TP < 50 Kbps (%) 0
50 Kbps < DL TP < 250 Kbps (%) 0
250 Kbps < DL TP < 1 Mbps (%) 0
DL TP > 1 Mbps (%) 0
UL TP < 10 Kbps (%) 0
10 Kbps < UL TP < 50 Kbps (%) 0
50 Kbps < UL TP < 300 Kbps (%) 0
UL TP > 300 Kbps (%) 0
Activity Duration DL (ms) 0
Activity Duration UL (ms) 0
Dur. (ms).1        0
Handset Manufacturer 0
Handset Type        0

```

```

Nb of sec with Vol DL < 6250B      0
Nb of sec with Vol UL < 1250B     0
Social Media DL (Bytes)           0
Social Media UL (Bytes)           0
Google DL (Bytes)                 0
Google UL (Bytes)                 0
Email DL (Bytes)                  0
Email UL (Bytes)                  0
Youtube DL (Bytes)                0
Youtube UL (Bytes)                0
Netflix DL (Bytes)                0
Netflix UL (Bytes)                0
Gaming DL (Bytes)                 0
Gaming UL (Bytes)                 0
Other DL (Bytes)                  0
Other UL (Bytes)                  0
Total UL (Bytes)                  0
Total DL (Bytes)                  0
top_10_handsets                   120739
dtype: int64

```

```
IQR = 'Quartile3 - Quartile1'
```

```

Q1 = np.percentile(data1,25)
Q2 = np.percentile(data1,50)
Q3 = np.percentile(data1,75)
print(Q1,Q2,Q3)

```

```
541061.0 3936649.5 16100163.25
```

```

IQR=Q3-Q1
print(IQR)

```

```
15559102.25
```

```
upper = 'Q3 +1.5*IQR'
```

```
lower = 'Q1 - 1.5*IQR'
```

```

upper=Q3+1.5*IQR
upper_array=np.array(user_behaviour['Dur. (ms)']>=upper)
print("Upper Bound:",upper)
print(upper_array.sum())

```

```
#Below Lower bound
```

```

lower=Q1-1.5*IQR
lower_array=np.array(user_behaviour['Dur. (ms)']<=lower)
print("Lower Bound:",lower)
print(lower_array.sum())

```

```
Upper Bound: 39438816.625
```

```
0
```

Lower Bound: -22797592.375
0

```
upperrange=(Q3+IQR)*1.5  
print('upperrange=',(Q3+IQR)*1.5)  
lowerrange=(Q1-IQR)*1.5  
print('lowerrange=',(Q1-IQR)*1.5)
```

```
upperrange= 47488898.25  
lowerrange= -22527061.875
```

```
def find_outlier(data1,threshold=1.5):  
    outliers=pd.DataFrame()  
    for i in data1.columns:  
        if data1[i].dtype !='object':  
            i_outlier=data1[(data1[i]<lowerrange)|  
(data1[i]>upperrange)]  
            outlier=pd.DataFrame()  
            outlier=pd.concat([outlier,i_outlier])  
            data1[i]=np.where((data1[i]<lowerrange)|  
(data1[i]>upperrange),data1[i].mean(),data1[i])  
    return outlier
```

```
find_outlier(data1)
```

	Bearer Id	Dur. (ms)	Activity	Duration	DL (ms)	\
7880	9.907303e+18	1063952.0			6.601000e+05	
29394	9.907303e+18	1247317.0			8.702510e+05	
67253	9.907303e+18	106584.0			1.315224e+06	
68806	9.907303e+18	67198.0			8.002700e+04	
68924	9.907303e+18	1507998.0			5.890606e+06	
69801	9.907303e+18	518394.0			2.464303e+06	
69835	9.907303e+18	340788.0			1.571891e+07	
69905	9.907303e+18	431996.0			2.464303e+06	
70298	9.907303e+18	431997.0			2.021032e+07	
70336	9.907303e+18	1468804.0			2.464303e+06	
70424	9.907303e+18	1050221.0			8.532424e+06	
70447	9.907303e+18	431996.0			1.974265e+07	
70602	9.907303e+18	775239.0			2.581390e+07	
70818	9.907303e+18	864001.0			4.044378e+07	
71112	9.907303e+18	325535.0			4.041298e+07	
71133	9.907303e+18	431996.0			3.730937e+07	
71675	9.907303e+18	417975.0			2.464303e+06	
72634	9.907303e+18	1960914.0			1.596210e+07	
72645	9.907303e+18	1475653.0			2.331734e+07	
72903	9.907303e+18	864002.0			7.510138e+06	
73175	9.907303e+18	1117511.0			1.185512e+07	
73256	9.907303e+18	1319645.0			2.694907e+07	
73414	9.907303e+18	431135.0			2.464303e+06	
73484	9.907303e+18	2418727.0			2.091366e+07	

73730	9.907303e+18	1296005.0	4.287512e+06
74419	9.907303e+18	1734912.0	2.464303e+06
74670	9.907303e+18	1131647.0	4.433025e+06
74695	9.907303e+18	928854.0	1.037289e+07
74717	9.907303e+18	909952.0	2.187496e+07
74772	9.907303e+18	1556434.0	2.066912e+07
75671	9.907303e+18	880384.0	7.693302e+06
76321	9.907303e+18	1501420.0	7.032942e+06
76756	9.907303e+18	784483.0	2.464303e+06
77151	9.907303e+18	2377940.0	1.762254e+06
77259	9.907303e+18	1262072.0	2.626484e+07
77502	9.907303e+18	622221.0	2.385087e+07
80052	9.907303e+18	1534218.0	7.112424e+06
81674	9.907303e+18	830722.0	9.724392e+06
84259	9.907303e+18	2170150.0	9.523994e+06
84739	9.907303e+18	431996.0	2.464303e+06
84963	9.907303e+18	343221.0	1.361760e+07
85249	9.907303e+18	372584.0	1.900233e+07
85461	9.907303e+18	2484980.0	3.943613e+07
85795	9.907303e+18	2052703.0	4.238755e+06
87354	9.907303e+18	465565.0	1.790148e+07
87681	9.907303e+18	1317282.0	2.060609e+06
88288	9.907303e+18	2423726.0	8.160787e+06
88555	9.907303e+18	3990246.0	2.022999e+07
89327	9.907303e+18	518394.0	3.029330e+07
89341	9.907303e+18	1329280.0	2.464303e+06
89441	9.907303e+18	1042956.0	2.758022e+06
89680	9.907303e+18	2604267.0	2.585884e+06
89933	9.907303e+18	1390780.0	3.873719e+07
97244	9.907303e+18	1572096.0	1.132443e+06
100587	9.907303e+18	4120953.0	1.277930e+07
100662	9.907303e+18	2524675.0	2.023511e+06

	Activity Duration UL (ms)	Social Media DL (Bytes)	\
7880	6.271550e+05	12065608.0	
29394	8.813150e+05	5318154.0	
67253	1.169613e+06	5376189.0	
68806	1.039640e+05	8414716.0	
68924	5.281474e+06	9889718.0	
69801	1.893064e+06	13054280.0	
69835	1.556476e+07	7516441.0	
69905	1.893064e+06	6649767.0	
70298	4.395977e+07	10828963.0	
70336	2.302037e+07	10858194.0	
70424	7.510434e+06	9554325.0	
70447	1.300915e+07	11152831.0	
70602	2.286255e+07	10807105.0	
70818	1.358318e+07	3422143.0	
71112	2.560577e+07	11123354.0	
71133	2.443059e+07	5137157.0	

71675	1.893064e+06	7537420.0
72634	8.794107e+06	11238224.0
72645	2.194194e+07	8886660.0
72903	6.393964e+06	3867243.0
73175	1.051952e+07	5496294.0
73256	1.801094e+07	7400944.0
73414	1.893064e+06	8244196.0
73484	1.265915e+07	13445179.0
73730	3.297345e+06	9579071.0
74419	2.960062e+07	15691444.0
74670	1.479171e+06	9779800.0
74695	9.395395e+06	10950957.0
74717	1.531425e+07	11091536.0
74772	1.742055e+07	11486692.0
75671	6.323335e+06	12154285.0
76321	6.072272e+06	8724177.0
76756	3.086231e+07	9069813.0
77151	1.404338e+06	12565025.0
77259	2.507784e+07	10358833.0
77502	1.273152e+07	7105536.0
80052	5.217562e+06	9727711.0
81674	5.112691e+06	5942916.0
84259	8.136190e+06	12985618.0
84739	3.224348e+07	5035665.0
84963	1.278085e+07	8762084.0
85249	1.949376e+07	9338505.0
85461	2.910492e+07	17685208.0
85795	3.823250e+06	12387634.0
87354	1.282601e+07	10209251.0
87681	2.089353e+06	8333925.0
88288	7.424225e+06	17392073.0
88555	2.255607e+07	13127153.0
89327	2.026302e+07	11545789.0
89341	1.893064e+06	26188977.0
89441	1.848506e+06	12118788.0
89680	6.576600e+05	18205049.0
89933	3.077506e+07	11500407.0
97244	1.032171e+06	6485564.0
100587	1.117550e+07	10697509.0
100662	1.319980e+06	17325759.0

	Social Media UL (Bytes)	Google DL (Bytes)	Google UL (Bytes)
7880	192836.0	1.997650e+07	15961483.0
29394	154867.0	4.114084e+07	11369457.0
67253	102915.0	1.798342e+07	3839007.0
68806	116568.0	1.593428e+07	13860606.0

68924	138392.0	3.130054e+07	8320695.0
69801	191178.0	2.240049e+07	14137153.0
69835	73022.0	1.238805e+07	11990054.0
69905	224580.0	2.305591e+07	8820964.0
70298	146130.0	2.235237e+07	15170823.0
70336	251973.0	3.702023e+07	11627562.0
70424	170575.0	3.147998e+07	10324851.0
70447	70188.0	2.441095e+07	14271845.0
70602	148757.0	3.432890e+07	13367484.0
70818	61927.0	2.175851e+07	6683615.0
71112	162275.0	1.011205e+07	6098489.0
71133	209370.0	1.509313e+07	14685789.0
71675	161070.0	3.290942e+07	10970558.0
72634	158257.0	4.309459e+07	9152556.0
72645	195111.0	2.520216e+07	8523422.0
72903	163226.0	2.287799e+07	6568526.0
73175	109965.0	2.105290e+07	11014607.0
73256	170013.0	2.795737e+07	8393857.0
73414	174485.0	1.994129e+07	7234134.0
73484	247287.0	4.601201e+07	12333099.0
73730	226963.0	3.300066e+07	11295256.0
74419	314047.0	6.288150e+06	24423584.0
74670	238435.0	3.901363e+07	14067047.0
74695	63483.0	3.233953e+07	4269581.0

74717	86373.0	4.662386e+07	12251325.0
74772	180416.0	4.166681e+07	10748639.0
75671	171529.0	3.471458e+07	12284439.0
76321	195794.0	2.392670e+07	11751436.0
76756	108306.0	2.154879e+07	9557138.0
77151	335767.0	6.288150e+06	15865041.0
77259	151199.0	3.118401e+07	7666551.0
77502	138161.0	3.103030e+07	7151987.0
80052	234882.0	3.917658e+07	9308525.0
81674	158739.0	3.017979e+07	16186767.0
84259	114504.0	4.721672e+07	11206798.0
84739	211437.0	2.336369e+07	11918900.0
84963	180096.0	1.662024e+07	10393266.0
85249	136809.0	3.157844e+07	7690469.0
85461	206719.0	6.288150e+06	17813083.0
85795	131985.0	3.624851e+07	11220485.0
87354	123753.0	2.146870e+07	8217477.0
87681	228510.0	2.818442e+07	8181989.0
88288	261469.0	3.390403e+07	13650023.0
88555	165881.0	4.727747e+07	11731439.0
89327	132766.0	4.209783e+07	13554999.0
89341	478166.0	6.288150e+06	38617483.0
89441	249117.0	2.540394e+07	14660911.0
89680	289172.0	4.669842e+07	17138263.0

89933	258236.0	3.413013e+07	11168318.0
97244	186380.0	3.076997e+07	13146865.0
100587	260729.0	3.782076e+07	7857451.0
100662	257241.0	4.354167e+07	18335688.0

	Email DL (Bytes)	Email UL (Bytes)	Youtube DL (Bytes)	\
7880	11231575.0	1972235.0	2.769808e+07	
29394	5183039.0	3010617.0	1.272504e+07	
67253	10881218.0	2066085.0	1.272504e+07	
68806	8290179.0	3827418.0	1.272504e+07	
68924	7655312.0	1697073.0	1.272504e+07	
69801	5145847.0	2480733.0	1.272504e+07	
69835	3269195.0	1532864.0	4.209157e+07	
69905	11689441.0	2173748.0	1.272504e+07	
70298	10241272.0	1365449.0	4.520454e+07	
70336	16704195.0	4254396.0	4.366018e+07	
70424	10319386.0	3747944.0	1.272504e+07	
70447	6872814.0	2075624.0	4.505911e+07	
70602	12771512.0	2403445.0	1.272504e+07	
70818	8087758.0	1060245.0	3.987437e+07	
71112	6644610.0	1499829.0	1.272504e+07	
71133	6360247.0	3240041.0	1.272504e+07	
71675	9144172.0	2738173.0	2.774418e+07	
72634	15734320.0	2826234.0	1.272504e+07	
72645	9882214.0	1882167.0	1.272504e+07	
72903	9289515.0	1644461.0	1.272504e+07	
73175	7893657.0	1431100.0	1.272504e+07	
73256	8925766.0	1056722.0	3.097584e+07	
73414	9359438.0	2099315.0	4.488028e+07	
73484	15539104.0	3771618.0	1.272504e+07	
73730	10919494.0	2114602.0	1.272504e+07	
74419	12657446.0	4737415.0	1.272504e+07	
74670	16654353.0	4274289.0	1.272504e+07	
74695	6917289.0	2531472.0	4.444422e+07	
74717	11105577.0	2421490.0	1.272504e+07	
74772	9962993.0	3068303.0	1.272504e+07	
75671	9988060.0	1112004.0	3.947088e+07	
76321	11104268.0	2219763.0	1.272504e+07	
76756	12798077.0	2604676.0	1.272504e+07	
77151	10517911.0	3109426.0	1.272504e+07	
77259	8284082.0	2765985.0	1.272504e+07	
77502	7791103.0	1936662.0	1.272504e+07	
80052	12415544.0	920393.0	1.272504e+07	
81674	10084278.0	3547208.0	1.272504e+07	
84259	9066036.0	3757376.0	1.272504e+07	

84739	7725244.0	2639373.0	1.272504e+07
84963	7028221.0	2041557.0	3.606245e+07
85249	8472045.0	844196.0	1.272504e+07
85461	18155310.0	3064384.0	1.272504e+07
85795	9127323.0	2497270.0	1.272504e+07
87354	4561436.0	2176805.0	1.272504e+07
87681	8986004.0	1885918.0	1.272504e+07
88288	13329755.0	4640566.0	1.272504e+07
88555	11434548.0	2305977.0	1.272504e+07
89327	12297738.0	2251547.0	1.272504e+07
89341	33798609.0	6744930.0	1.272504e+07
89441	10620792.0	3279302.0	1.272504e+07
89680	12511511.0	3711363.0	1.272504e+07
89933	9822222.0	3082898.0	1.272504e+07
97244	7654421.0	1663475.0	1.272504e+07
100587	9225038.0	3264818.0	1.272504e+07
100662	10220519.0	2880932.0	1.272504e+07

	Youtube UL (Bytes)	Netflix DL (Bytes)	Netflix UL (Bytes)	\
7880	1.203117e+07	1.271781e+07	1.202267e+07	
29394	1.203117e+07	1.271781e+07	1.202267e+07	
67253	3.646618e+07	1.271781e+07	4.547011e+07	
68806	1.203117e+07	1.271781e+07	1.202267e+07	
68924	1.203117e+07	1.271781e+07	2.708597e+07	
69801	3.953070e+07	1.271781e+07	1.202267e+07	
69835	1.203117e+07	4.643770e+07	1.202267e+07	
69905	3.916596e+07	4.160956e+07	1.202267e+07	
70298	1.203117e+07	1.271781e+07	1.202267e+07	
70336	1.203117e+07	1.271781e+07	1.202267e+07	
70424	1.203117e+07	1.271781e+07	1.202267e+07	
70447	1.203117e+07	1.271781e+07	4.187393e+07	
70602	1.203117e+07	1.271781e+07	4.696836e+07	
70818	3.667662e+07	1.271781e+07	1.202267e+07	
71112	2.932434e+07	1.271781e+07	2.393736e+07	
71133	3.791663e+07	1.271781e+07	1.202267e+07	
71675	1.203117e+07	1.271781e+07	1.202267e+07	
72634	4.251627e+07	1.271781e+07	1.202267e+07	
72645	4.355817e+07	1.271781e+07	1.202267e+07	
72903	4.025200e+07	3.572694e+07	1.202267e+07	
73175	1.203117e+07	1.271781e+07	1.202267e+07	
73256	3.530932e+07	1.271781e+07	1.202267e+07	
73414	4.326230e+07	1.271781e+07	1.202267e+07	
73484	1.203117e+07	1.271781e+07	1.202267e+07	
73730	1.203117e+07	4.677622e+07	4.353004e+07	
74419	1.203117e+07	1.271781e+07	1.202267e+07	
74670	1.203117e+07	1.271781e+07	3.938948e+07	
74695	3.848046e+07	4.517117e+07	1.202267e+07	
74717	4.321166e+07	1.271781e+07	3.418807e+07	
74772	1.203117e+07	1.271781e+07	1.202267e+07	
75671	3.593694e+07	1.271781e+07	1.202267e+07	

76321	4.135406e+07	3.994501e+07	3.679697e+07
76756	1.203117e+07	3.197276e+07	1.202267e+07
77151	1.203117e+07	1.271781e+07	1.202267e+07
77259	1.203117e+07	1.271781e+07	2.453050e+07
77502	1.203117e+07	3.074802e+07	3.633603e+07
80052	1.203117e+07	1.271781e+07	1.202267e+07
81674	1.203117e+07	1.271781e+07	1.202267e+07
84259	1.203117e+07	1.271781e+07	1.202267e+07
84739	4.305971e+07	1.271781e+07	4.290793e+07
84963	2.284456e+07	3.765715e+07	2.565069e+07
85249	4.453732e+07	2.801131e+07	1.202267e+07
85461	1.203117e+07	1.271781e+07	1.202267e+07
85795	1.203117e+07	1.271781e+07	1.202267e+07
87354	1.203117e+07	1.271781e+07	1.202267e+07
87681	1.203117e+07	3.789540e+07	1.202267e+07
88288	1.203117e+07	1.271781e+07	1.202267e+07
88555	1.203117e+07	1.271781e+07	1.202267e+07
89327	1.203117e+07	1.271781e+07	1.202267e+07
89341	1.203117e+07	1.271781e+07	1.202267e+07
89441	1.788242e+07	1.271781e+07	1.202267e+07
89680	1.203117e+07	1.271781e+07	1.202267e+07
89933	1.203117e+07	1.271781e+07	1.202267e+07
97244	1.203117e+07	3.534024e+07	1.202267e+07
100587	1.203117e+07	1.271781e+07	1.202267e+07
100662	1.203117e+07	1.271781e+07	1.202267e+07

	Gaming DL (Bytes)	Gaming UL (Bytes)	Other DL (Bytes) \
7880	4.608041e+08	3.926920e+07	4.607876e+08
29394	4.608041e+08	4.199390e+07	4.607876e+08
67253	4.608041e+08	2.129808e+07	4.607876e+08
68806	4.608041e+08	3.072244e+07	4.607876e+08
68924	4.608041e+08	3.453025e+07	4.607876e+08
69801	4.608041e+08	9.066019e+06	4.607876e+08
69835	4.608041e+08	3.276557e+07	4.607876e+08
69905	4.608041e+08	9.066019e+06	4.607876e+08
70298	4.608041e+08	9.066019e+06	4.607876e+08
70336	4.608041e+08	9.066019e+06	4.607876e+08
70424	4.608041e+08	3.215753e+07	4.607876e+08
70447	4.608041e+08	2.912187e+07	4.607876e+08
70602	4.608041e+08	9.066019e+06	4.607876e+08
70818	4.608041e+08	2.254635e+07	4.607876e+08
71112	4.608041e+08	4.603988e+07	4.607876e+08
71133	4.608041e+08	9.066019e+06	4.607876e+08
71675	4.608041e+08	9.066019e+06	4.607876e+08
72634	4.608041e+08	4.367436e+07	4.607876e+08
72645	4.608041e+08	3.311396e+07	4.607876e+08
72903	4.608041e+08	3.598716e+07	4.607876e+08
73175	4.608041e+08	9.066019e+06	4.607876e+08
73256	4.608041e+08	9.066019e+06	4.607876e+08
73414	4.608041e+08	9.066019e+06	4.607876e+08

73484	4.608041e+08	9.066019e+06	4.607876e+08
73730	4.608041e+08	3.547314e+07	4.607876e+08
74419	4.608041e+08	9.066019e+06	4.607876e+08
74670	4.608041e+08	9.066019e+06	4.607876e+08
74695	4.608041e+08	3.187746e+07	4.607876e+08
74717	4.608041e+08	4.712999e+07	4.607876e+08
74772	4.608041e+08	4.504584e+07	4.607876e+08
75671	4.608041e+08	4.505545e+07	4.607876e+08
76321	4.608041e+08	4.424680e+07	4.607876e+08
76756	4.608041e+08	4.295631e+07	4.607876e+08
77151	4.608041e+08	9.066019e+06	4.607876e+08
77259	4.608041e+08	4.703218e+07	4.607876e+08
77502	4.608041e+08	3.756312e+07	4.607876e+08
80052	4.608041e+08	4.673146e+07	4.607876e+08
81674	4.608041e+08	9.066019e+06	4.607876e+08
84259	4.608041e+08	9.066019e+06	4.607876e+08
84739	4.608041e+08	9.066019e+06	4.607876e+08
84963	4.608041e+08	3.011592e+07	4.607876e+08
85249	4.608041e+08	2.644042e+07	4.607876e+08
85461	4.608041e+08	9.066019e+06	4.607876e+08
85795	4.608041e+08	2.898376e+07	4.607876e+08
87354	4.608041e+08	4.299542e+07	4.607876e+08
87681	4.608041e+08	3.743657e+07	4.607876e+08
88288	4.608041e+08	9.066019e+06	4.607876e+08
88555	4.608041e+08	9.066019e+06	4.607876e+08
89327	4.608041e+08	4.017621e+07	4.607876e+08
89341	4.608041e+08	9.066019e+06	4.607876e+08
89441	4.608041e+08	4.278891e+07	4.607876e+08
89680	4.608041e+08	4.252727e+07	4.607876e+08
89933	4.608041e+08	9.066019e+06	4.607876e+08
97244	4.608041e+08	2.269613e+07	4.607876e+08
100587	4.608041e+08	3.516079e+07	4.607876e+08
100662	4.608041e+08	2.893398e+07	4.607876e+08

	Other UL (Bytes)
7880	48706279.0
29394	60651254.0
67253	54043072.0
68806	51538897.0
68924	49706447.0
69801	66599994.0
69835	50753734.0
69905	60561728.0
70298	51209440.0
70336	70192623.0
70424	64474883.0
70447	54217928.0
70602	59246099.0
70818	48154317.0
71112	50913448.0

71133	47863045.0
71675	54478686.0
72634	61754055.0
72645	48141734.0
72903	49965988.0
73175	48023370.0
73256	48930687.0
73414	52685390.0
73484	53006405.0
73730	50671233.0
74419	79388455.0
74670	48177714.0
74695	48571116.0
74717	61767145.0
74772	52886243.0
75671	49063721.0
76321	58654734.0
76756	47534564.0
77151	54218645.0
77259	56062208.0
77502	47622908.0
80052	76549883.0
81674	49592823.0
84259	68100816.0
84739	52141513.0
84963	49591592.0
85249	48175673.0
85461	76617173.0
85795	56716734.0
87354	57891300.0
87681	50197623.0
88288	73736223.0
88555	68115759.0
89327	48669809.0
89341	128715313.0
89441	59836041.0
89680	75443538.0
89933	54963856.0
97244	56139536.0
100587	64857892.0
100662	50327587.0

data1.columns

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Index(['Bearer Id', 'Dur. (ms)', 'Activity Duration DL (ms)',
      'Activity Duration UL (ms)', 'Social Media DL (Bytes)',
      'Social Media UL (Bytes)', 'Google DL (Bytes)', 'Google UL
(Bytes)',
      'Email DL (Bytes)', 'Email UL (Bytes)', 'Youtube DL (Bytes)',
      'Youtube UL (Bytes)', 'Netflix DL (Bytes)', 'Netflix UL
(Bytes)'])
```

```
    'Gaming DL (Bytes)', 'Gaming UL (Bytes)', 'Other DL (Bytes)',  
    'Other UL (Bytes)'],  
    dtype='object')
```

```
# total DL data for each category
```

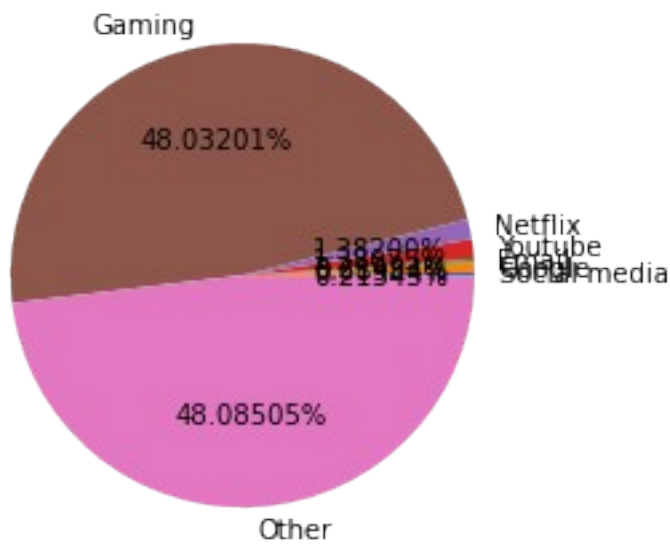
```
dl_data=[  
    data1['Social Media DL (Bytes)'].sum(),  
    data1['Google DL (Bytes)'].sum(),  
    data1['Email DL (Bytes)'].sum(),  
    data1['Youtube DL (Bytes)'].sum(),  
    data1['Netflix DL (Bytes)'].sum(),  
    data1['Gaming DL (Bytes)'].sum(),  
    data1['Other DL (Bytes)'].sum()  
]
```

```
# total UL data for each category
```

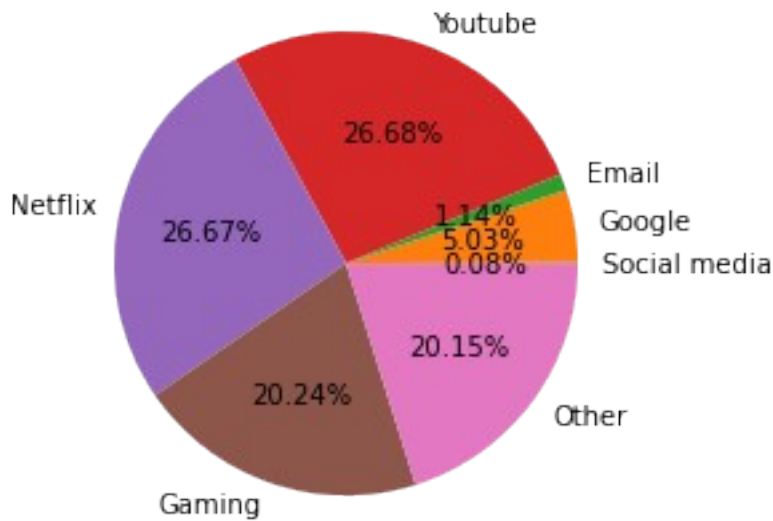
```
ul_data=[  
    data1['Social Media UL (Bytes)'].sum(),  
    data1['Google UL (Bytes)'].sum(),  
    data1['Email UL (Bytes)'].sum(),  
    data1['Youtube UL (Bytes)'].sum(),  
    data1['Netflix UL (Bytes)'].sum(),  
    data1['Gaming UL (Bytes)'].sum(),  
    data1['Other UL (Bytes)'].sum()  
]
```

```
labels=["Social  
media", "Google", "Email", "Youtube", "Netflix", "Gaming", "Other"]
```

```
plt.pie(labels=labels,  
        x=dl_data, autopct = '%1.5f%%')  
plt.show()
```



```
plt.pie(labels=labels,
        x=ul_data, autopct = '%1.2f%%')
plt.show()
```



```
data1.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 110432 entries, 0 to 110431
Data columns (total 18 columns):
 #   Column                Non-Null Count  Dtype
---  -
 0   Bearer Id             110432 non-null float64
 1   Dur. (ms)             110432 non-null float64
```

```

2   Activity Duration DL (ms)    110432 non-null float64
3   Activity Duration UL (ms)    110432 non-null float64
4   Social Media DL (Bytes)      110432 non-null float64
5   Social Media UL (Bytes)      110432 non-null float64
6   Google DL (Bytes)            110432 non-null float64
7   Google UL (Bytes)            110432 non-null float64
8   Email DL (Bytes)             110432 non-null float64
9   Email UL (Bytes)             110432 non-null float64
10  Youtube DL (Bytes)           110432 non-null float64
11  Youtube UL (Bytes)           110432 non-null float64
12  Netflix DL (Bytes)           110432 non-null float64
13  Netflix UL (Bytes)           110432 non-null float64
14  Gaming DL (Bytes)            110432 non-null float64
15  Gaming UL (Bytes)            110432 non-null float64
16  Other DL (Bytes)             110432 non-null float64
17  Other UL (Bytes)             110432 non-null float64
dtypes: float64(18)
memory usage: 15.2 MB

```

#Basic col Analysis

```

columns=['Dur. (ms)', 'Start ms', 'End ms', 'Total DL (Bytes)', 'Total UL
(Bytes)']
stat=data[columns].describe()
print(columns)

```

```

['Dur. (ms)', 'Start ms', 'End ms', 'Total DL (Bytes)', 'Total UL
(Bytes)']

```

Non Graphical Univariate Analysis

```

dispersion = data[columns].std()
print(dispersion)

```

```

Dur. (ms)          7.335495e+04
Start ms           2.887598e+02
End ms             2.879436e+02
Total DL (Bytes)   2.440239e+08
Total UL (Bytes)   1.128297e+07
dtype: float64

```

#Graphical Univariate Analysis

```

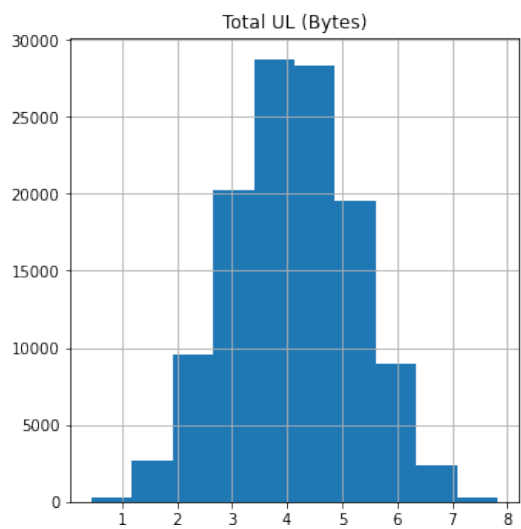
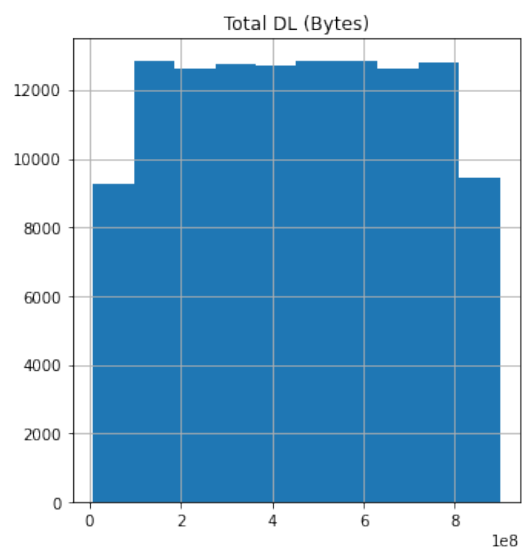
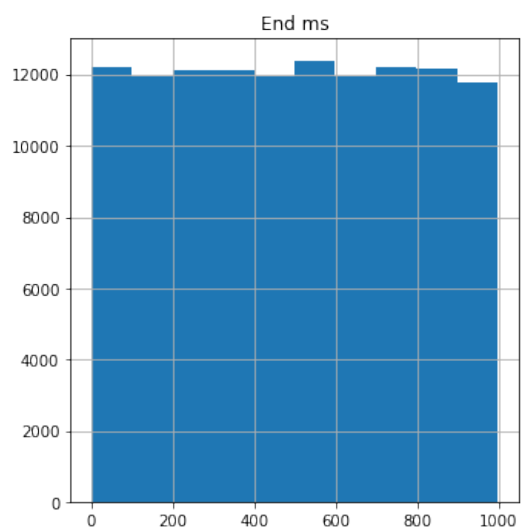
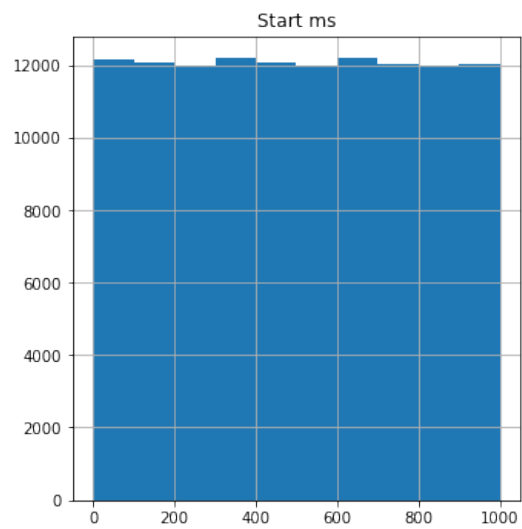
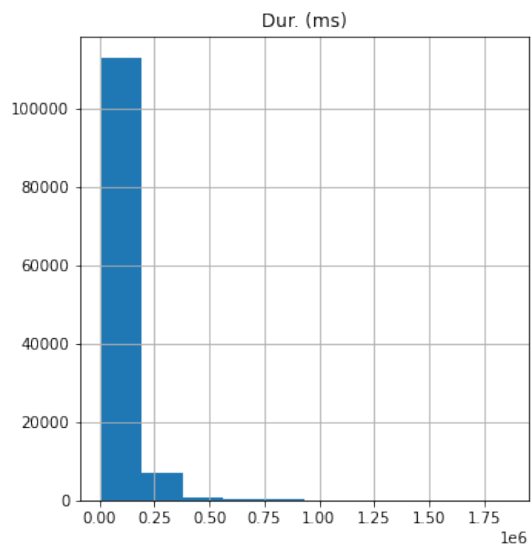
data.hist(column=columns,bins=10,figsize=(12,20))

```

```

array([[<AxesSubplot:title={'center':'Dur. (ms)'}>,
        <AxesSubplot:title={'center':'Start ms'}>],
       [<AxesSubplot:title={'center':'End ms'}>,
        <AxesSubplot:title={'center':'Total DL (Bytes)'}>],
       [<AxesSubplot:title={'center':'Total UL (Bytes)'}>,
        <AxesSubplot:>]], dtype=object)

```

```
corr_matrix = data[['Social Media DL (Bytes)', 'Social Media UL
(Bytes)',
'Google DL (Bytes)', 'Google UL (Bytes)', 'Email DL (Bytes)',
'Email UL (Bytes)', 'Youtube DL (Bytes)', 'Youtube UL (Bytes)',
'Netflix DL (Bytes)', 'Netflix UL (Bytes)', 'Gaming DL
(Bytes)',
'Gaming UL (Bytes)', 'Other DL (Bytes)', 'Other UL (Bytes)',
'Total UL (Bytes)', 'Total DL (Bytes)']]

print(corr_matrix)
```

	Social Media DL (Bytes)	Social Media UL (Bytes)	Google DL
(Bytes) \			
0	1545765.0	24420.0	
1634479.0			
1	1926113.0	7165.0	
3493924.0			
6	1623509.0	1482.0	
1837897.0			
7	3103940.0	57463.0	
1304524.0			
9	1541259.0	44154.0	
8601949.0			
...	
...			
149995	962097.0	31078.0	
1139573.0			
149996	3464974.0	52091.0	
9967603.0			
149997	2344568.0	7613.0	
2229420.0			
149998	1245845.0	14394.0	
3850890.0			
149999	801547.0	21562.0	
4189773.0			

	Google UL (Bytes)	Email DL (Bytes)	Email UL (Bytes)	\
0	1271433.0	3563542.0	137762.0	
1	920172.0	629046.0	308339.0	
6	3068975.0	1838947.0	252032.0	
7	2568189.0	2987408.0	512696.0	
9	176392.0	522457.0	243589.0	
...	
149995	126061.0	3459965.0	820142.0	
149996	2817311.0	57639.0	633237.0	
149997	2185941.0	1954414.0	167304.0	
149998	2734579.0	1525734.0	532543.0	
149999	3567494.0	2228270.0	622644.0	

	Youtube DL (Bytes)	Youtube UL (Bytes)	Netflix DL (Bytes)	\
--	--------------------	--------------------	--------------------	---

0	15854611.0	2501332.0	8198936.0
1	20247395.0	19111729.0	18338413.0
6	7234465.0	1886295.0	13231006.0
7	12189786.0	19735183.0	6858496.0
9	16338345.0	6573194.0	12211149.0
...
149995	6550499.0	18003146.0	22468983.0
149996	16191667.0	11763428.0	17883703.0
149997	13877234.0	8288284.0	19350146.0
149998	22660510.0	1855903.0	9963942.0
149999	8817106.0	8305402.0	3322253.0

	Netflix UL (Bytes)	Gaming DL (Bytes)	Gaming UL (Bytes) \
0	9656251.0	278082303.0	14344150.0
1	17227132.0	608750074.0	1170709.0
6	17981760.0	44796223.0	14084934.0
7	12075206.0	748906189.0	16149622.0
9	974201.0	811355188.0	1591237.0
...
149995	7149728.0	784435351.0	12724751.0
149996	19678161.0	526609673.0	9197207.0
149997	21293148.0	626893062.0	4735033.0
149998	5065760.0	553539484.0	13394316.0
149999	13172589.0	352536971.0	2529475.0

	Other DL (Bytes)	Other UL (Bytes)	Total UL (Bytes)	Total DL
(Bytes)				
0	171744450.0	8814393.0	36749741.0	
308879636.0				
1	526904238.0	15055145.0	53800391.0	
653384965.0				
6	803653557.0	10649768.0	47925246.0	
70562047.0				
7	537601820.0	7714657.0	58813016.0	
775350343.0				
9	64856042.0	12815208.0	22417975.0	
850570347.0				
...	
...				
149995	321383162.0	14890486.0	53745392.0	
819016468.0				
149996	3264510.0	13487416.0	57628851.0	
574175259.0				
149997	712180387.0	2457758.0	39135081.0	
666648844.0				
149998	121100856.0	11314729.0	34912224.0	
592786405.0				
149999	814713113.0	1406930.0	29626096.0	
371895920.0				

[120739 rows x 16 columns]

```
corr=data1.corr() # correlation in numerical feature
fig=plt.figure(figsize=(15,7))
sns.heatmap(corr,annot=True)
```

<AxesSubplot:>

