

```
In [1]: import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import pandas as pd
```

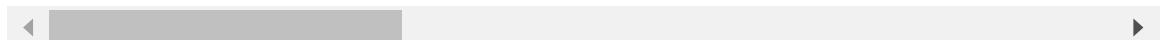
```
In [2]: tele_df=pd.read_csv("telecom_churn_data.csv")
```

```
In [3]: tele_df
```

Out[3]:

	year	customer_id	phone_no	gender	age	no_of_days_subscribed	multi_screen
0	2015	100198	409-8743	Female	36	62	no
1	2015	100643	340-5930	Female	39	149	no
2	2015	100756	372-3750	Female	65	126	no
3	2015	101595	331-4902	Female	24	131	no
4	2015	101653	351-8398	Female	40	191	no
...
1995	2015	997132	385-7387	Female	54	75	no
1996	2015	998086	383-9255	Male	45	127	no
1997	2015	998474	353-2080	NaN	53	94	no
1998	2015	998934	359-7788	Male	40	94	no
1999	2015	999961	414-1496	Male	37	73	no

2000 rows × 16 columns



Data Frame Size

```
In [4]: tele_df.size
```

```
Out[4]: 32000
```

```
In [ ]:
```

Data Frame Shape

```
In [5]: print(f" the total columns is {tele_df.shape[1]}")
print(f" the total rows {tele_df.shape[0]}")
```

the total columns is 16
the total rows 2000

```
In [6]: tele_df.shape
```

```
Out[6]: (2000, 16)
```

```
In [ ]:
```

Data Types

```
In [7]: tele_df.dtypes
```

```
Out[7]: year           int64
         customer_id    int64
         phone_no        object
         gender          object
         age             int64
         no_of_days_subscribed  int64
         multi_screen    object
         mail_subscribed object
         weekly_mins_watched float64
         minimum_daily_mins float64
         maximum_daily_mins float64
         weekly_max_night_mins int64
         videos_watched   int64
         maximum_days_inactive float64
         customer_support_calls int64
         churn            float64
         dtype: object
```

```
In [ ]:
```

Filter categorical columns and numerical columns

```
In [8]: cat_clm=tele_df.select_dtypes(include="object").columns
num_clm=tele_df.select_dtypes(exclude="object").columns
```

```
In [9]: cat_clm
```

```
Out[9]: Index(['phone_no', 'gender', 'multi_screen', 'mail_subscribed'], dtype='object')
```

```
In [10]: num_clm
```

```
Out[10]: Index(['year', 'customer_id', 'age', 'no_of_days_subscribed',
                 'weekly_mins_watched', 'minimum_daily_mins', 'maximum_daily_mins',
                 'weekly_max_night_mins', 'videos_watched', 'maximum_days_inactive',
                 'customer_support_calls', 'churn'],
                 dtype='object')
```

```
In [ ]:
```

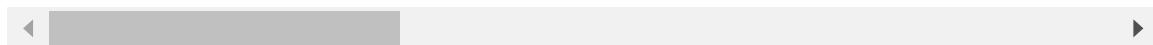
Check null values

```
In [11]: tele_df.isnull()
```

Out[11]:

	year	customer_id	phone_no	gender	age	no_of_days_subscribed	multi_screen
0	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False
...
1995	False	False	False	False	False	False	False
1996	False	False	False	False	False	False	False
1997	False	False	False	True	False	False	False
1998	False	False	False	False	False	False	False
1999	False	False	False	False	False	False	False

2000 rows × 16 columns

In [12]: `tele_df.isnull().sum()`

Out[12]:

year	0
customer_id	0
phone_no	0
gender	24
age	0
no_of_days_subscribed	0
multi_screen	0
mail_subscribed	0
weekly_mins_watched	0
minimum_daily_mins	0
maximum_daily_mins	0
weekly_max_night_mins	0
videos_watched	0
maximum_days_inactive	28
customer_support_calls	0
churn	35
dtype: int64	

In []:

FILL THE NULL VALUES WITH MODE FOR CATEGORICAL COLUMNS

In [14]:

```
mode_data=tele_df["gender"].mode()
md=mode_data.values[0]
tele_df["gender"]=tele_df["gender"].fillna(md)

tele_df.isnull().sum()
```

```
Out[14]: year          0  
customer_id      0  
phone_no         0  
gender           0  
age              0  
no_of_days_subscribed 0  
multi_screen     0  
mail_subscribed  0  
weekly_mins_watched 0  
minimum_daily_mins 0  
maximum_daily_mins 0  
weekly_max_night_mins 0  
videos_watched   0  
maximum_days_inactive 28  
customer_support_calls 0  
churn            35  
dtype: int64
```

```
In [15]: mode=tele_df["maximum_days_inactive"].mode()  
md=mode.values[0]  
tele_df["maximum_days_inactive"]=tele_df["maximum_days_inactive"].fillna(md)  
  
tele_df.isnull().sum()
```

```
Out[15]: year          0  
customer_id      0  
phone_no         0  
gender           0  
age              0  
no_of_days_subscribed 0  
multi_screen     0  
mail_subscribed  0  
weekly_mins_watched 0  
minimum_daily_mins 0  
maximum_daily_mins 0  
weekly_max_night_mins 0  
videos_watched   0  
maximum_days_inactive 0  
customer_support_calls 0  
churn            35  
dtype: int64
```

```
In [16]: mode=tele_df["churn"].mode()  
md=mode.values[0]  
tele_df["churn"]=tele_df["churn"].fillna(md)
```

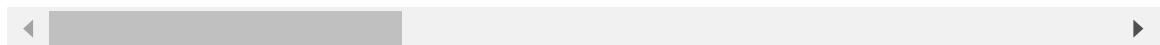
```
In [17]: tele_df.isnull().sum()
```

```
Out[17]: year          0
customer_id      0
phone_no         0
gender           0
age              0
no_of_days_subscribed 0
multi_screen     0
mail_subscribed 0
weekly_mins_watched 0
minimum_daily_mins 0
maximum_daily_mins 0
weekly_max_night_mins 0
videos_watched   0
maximum_days_inactive 0
customer_support_calls 0
churn            0
dtype: int64
```

In [18]: tele_df

	year	customer_id	phone_no	gender	age	no_of_days_subscribed	multi_screen	
0	2015	100198	409-8743	Female	36		62	no
1	2015	100643	340-5930	Female	39		149	no
2	2015	100756	372-3750	Female	65		126	no
3	2015	101595	331-4902	Female	24		131	no
4	2015	101653	351-8398	Female	40		191	no
...
1995	2015	997132	385-7387	Female	54		75	no
1996	2015	998086	383-9255	Male	45		127	no
1997	2015	998474	353-2080	Male	53		94	no
1998	2015	998934	359-7788	Male	40		94	no
1999	2015	999961	414-1496	Male	37		73	no

2000 rows × 16 columns



In []:

Find the Numerical values and categorycal columns

```
In [20]: cat_clm=tele_df.select_dtypes(include="object").columns
num_clm=tele_df.select_dtypes(exclude="object").columns
```

In [21]: cat_clm

```
Out[21]: Index(['phone_no', 'gender', 'multi_screen', 'mail_subscribed'], dtype='object')
```

In [22]: num_clm

```
Out[22]: Index(['year', 'customer_id', 'age', 'no_of_days_subscribed',
   'weekly_mins_watched', 'minimum_daily_mins', 'maximum_daily_mins',
   'weekly_max_night_mins', 'videos_watched', 'maximum_days_inactive',
   'customer_support_calls', 'churn'],
  dtype='object')
```

In [23]: data=tele_df["gender"].value_counts()
data

```
Out[23]: gender
Male      1077
Female     923
Name: count, dtype: int64
```

In []:

Find the keys and values

In [25]: keys=tele_df["gender"].value_counts().keys()
keys

```
Out[25]: Index(['Male', 'Female'], dtype='object', name='gender')
```

In [26]: values=tele_df["gender"].value_counts().values
values

```
Out[26]: array([1077, 923], dtype=int64)
```

In [27]: gender_df=pd.DataFrame(zip(keys,values),columns=["Label","counts"])
gender_df

Out[27]:

	Label	counts
0	Male	1077
1	Female	923

In []:

Frequency table of categorical columns

In [29]: cat_clm=tele_df.select_dtypes(include="object").columns
for i in cat_clm[1:]:
 print("=====>>>>>")
 keys=tele_df[i].value_counts().keys()
 values=tele_df[i].value_counts().values

 gender_df=pd.DataFrame(zip(keys,values),columns=["Label","counts"])
 print(gender_df)

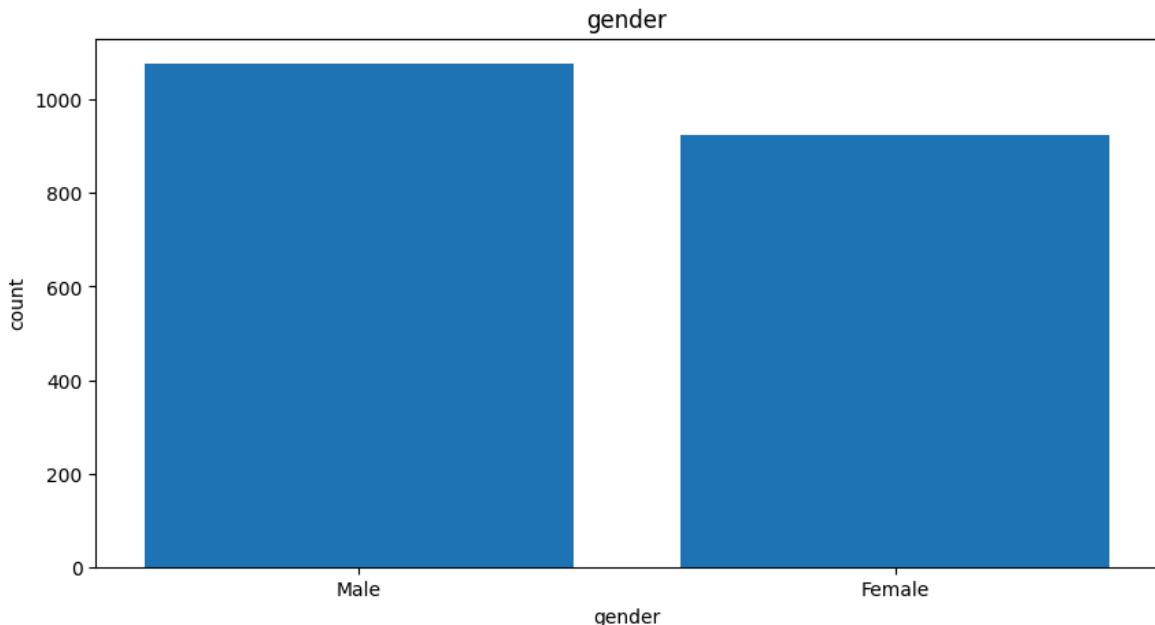
```
======>>>>>
Label  counts
0    Male    1077
1   Female    923
======>>>>>
Label  counts
0     no    1802
1   yes     198
======>>>>>
Label  counts
0     no    1430
1   yes     570
```

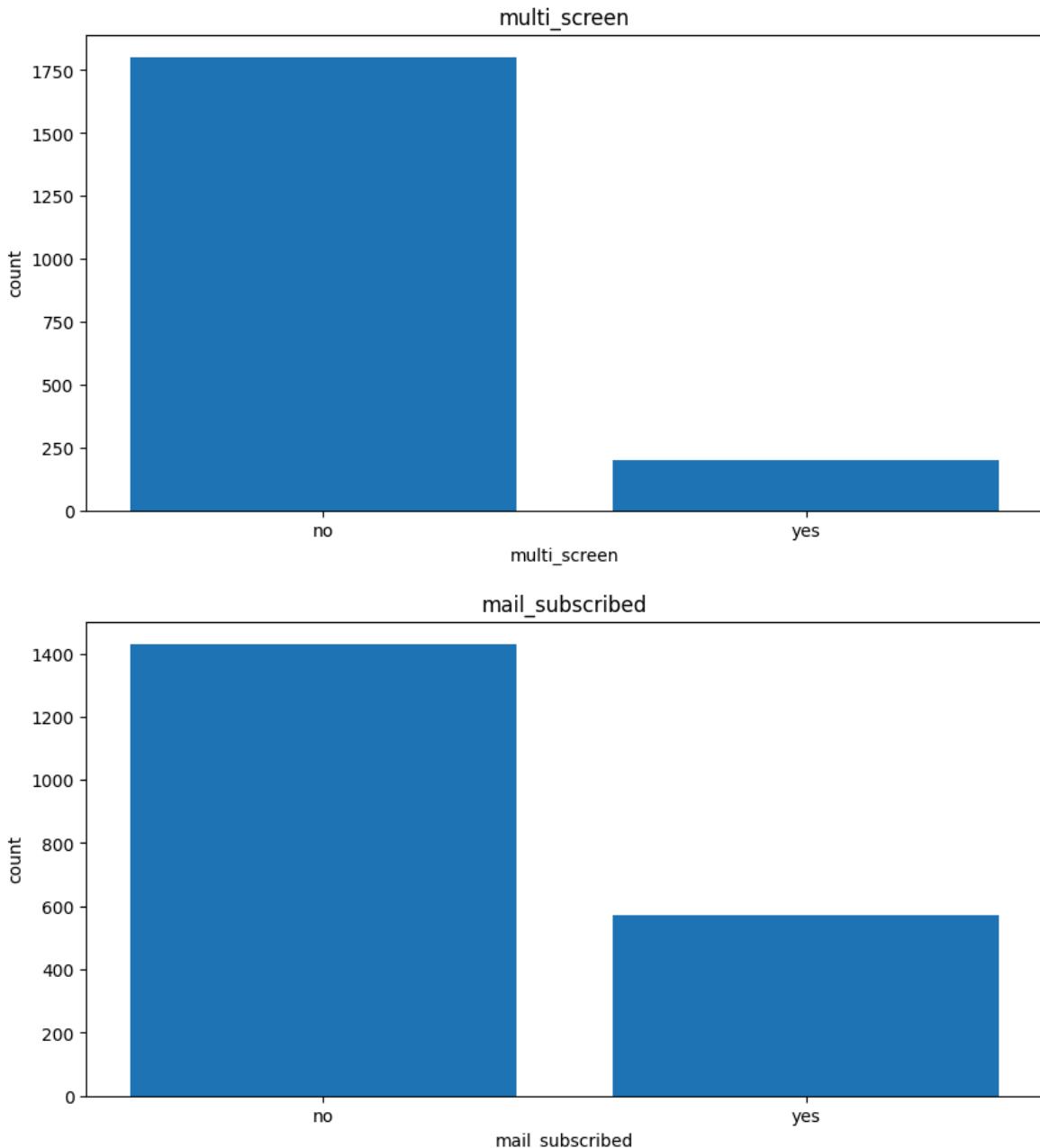
In []:

BAR CHART

```
In [31]: for i in cat_clm[1:]:
    counts=tele_df[i].value_counts()
    keys=counts.keys()
    values=counts.values

    plt.figure(figsize=(10,5))
    plt.bar(keys,values)
    plt.xlabel(i)
    plt.title(i)
    plt.ylabel("count")
```



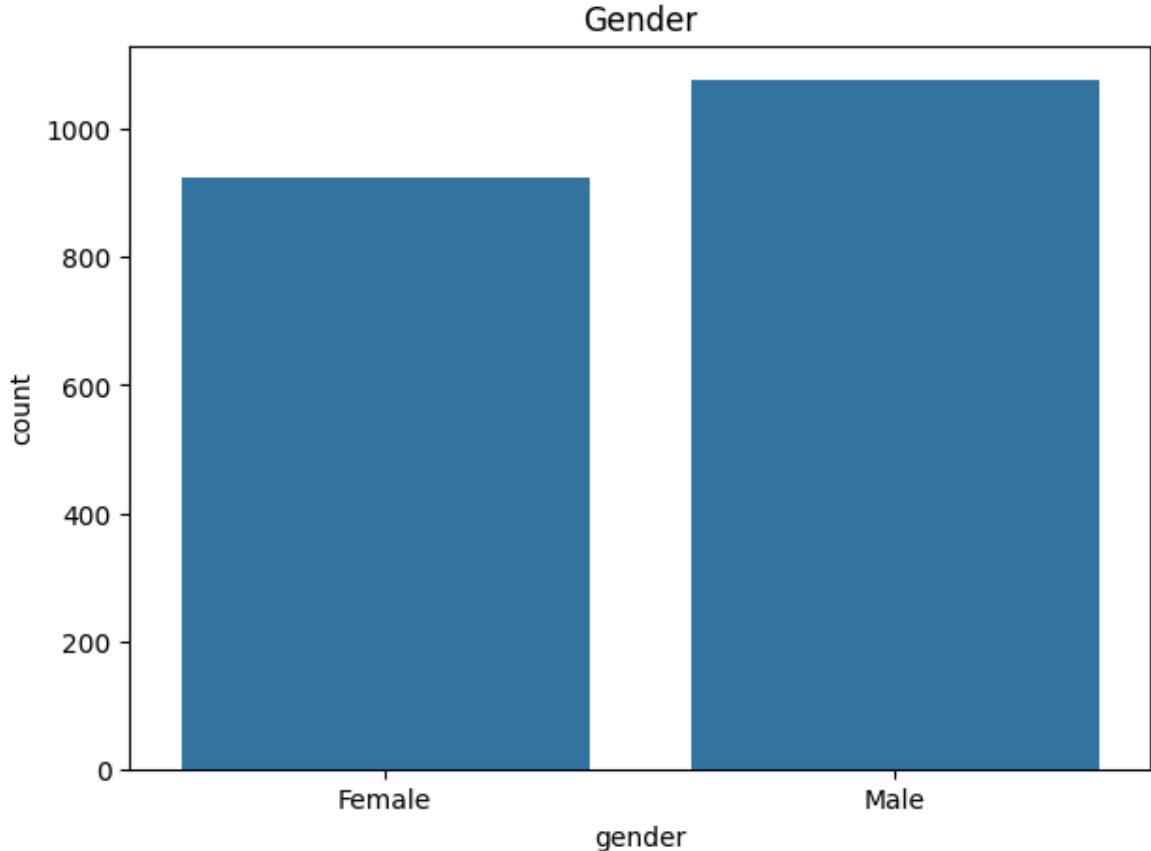


In []:

BAR CHART USING COUNTPLOT METHOD

```
In [33]: plt.figure(figsize=(7,5))
plt.title("Gender")
sns.countplot(data=tele_df,x="gender")
```

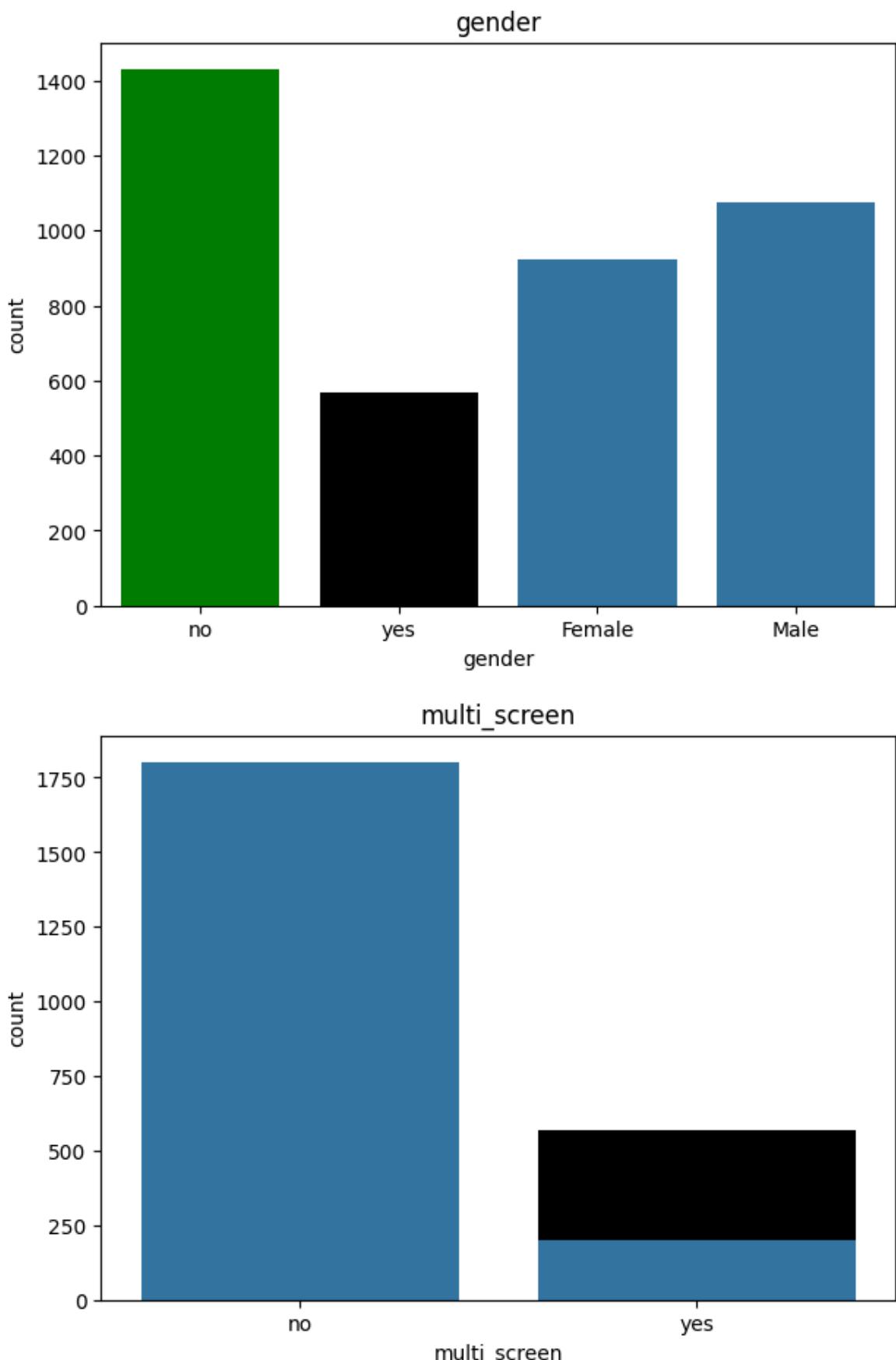
```
Out[33]: <Axes: title={'center': 'Gender'}, xlabel='gender', ylabel='count'>
```

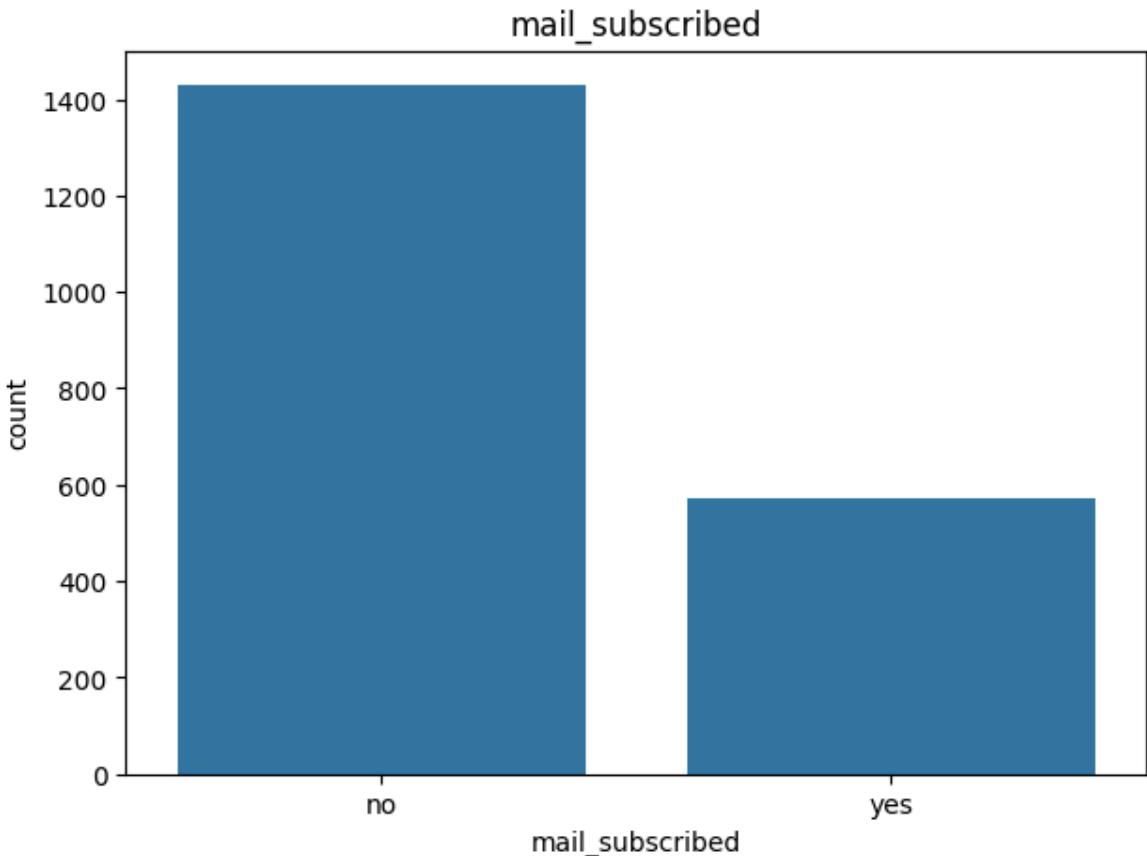


In []:

BAR CHART USING COUNTPLOT METHOD IN FOR-LOOP FUNCTION

```
In [35]: colors = ['green', 'black']
for i in cat_clm[1:]:
    plt.figure(figsize=(7,5))
    plt.bar(keys, values, color=colors[:len(keys)])
    plt.title(f"{i}")
    sns.countplot(data=tele_df,
                  x=i)
```





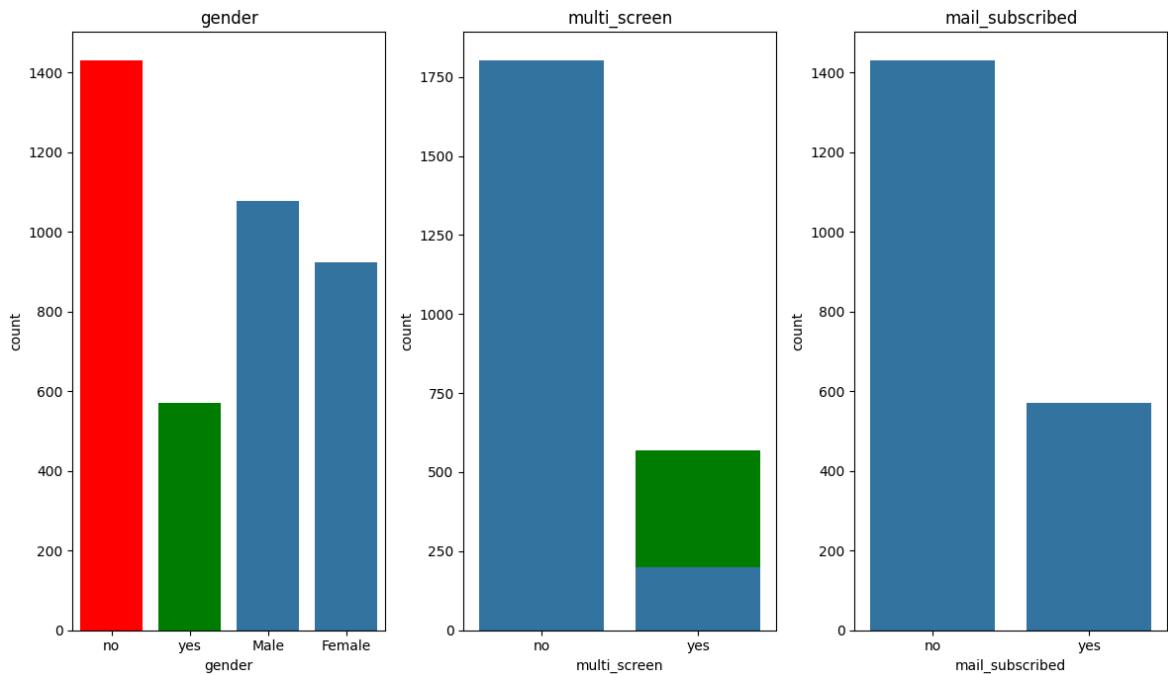
In []:

SUBPLOT

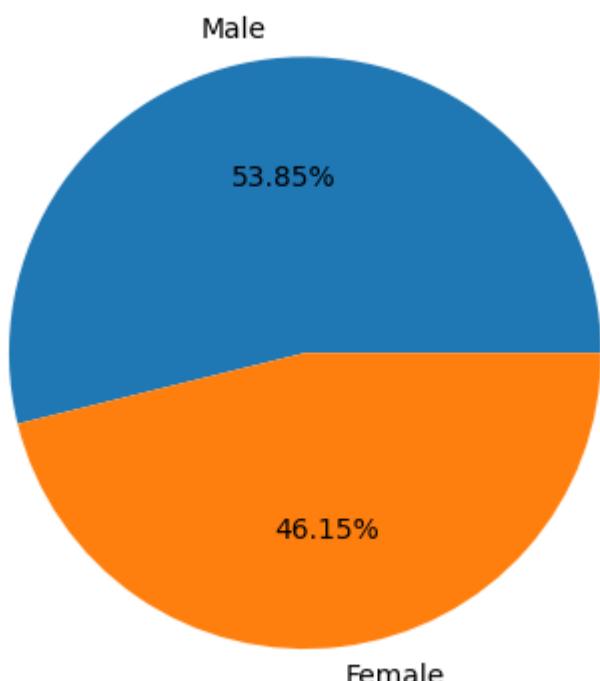
```
In [37]: import matplotlib.pyplot as plt
import seaborn as sns
colors = ['red', 'green', 'blue']

plt.figure(figsize=(12, 7))
for i in range(1, len(cat_clm)):
    keys1 = tele_df[cat_clm[i]].value_counts().keys()
    plt.subplot(1, 3, i)
    plt.bar(keys, values, color=colors[:len(keys)])
    plt.title(f'{cat_clm[i]}')
    sns.countplot(data=tele_df, x=cat_clm[i], order=keys1)

plt.tight_layout()
plt.show()
```



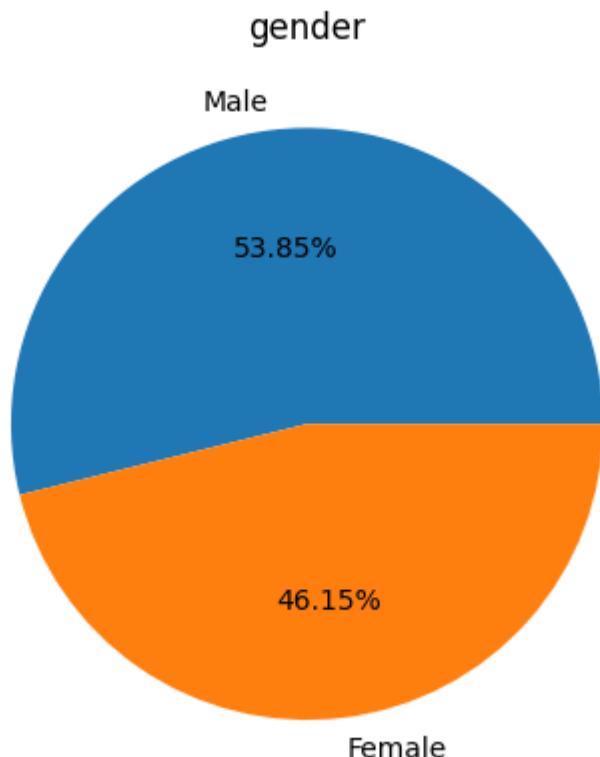
In []:

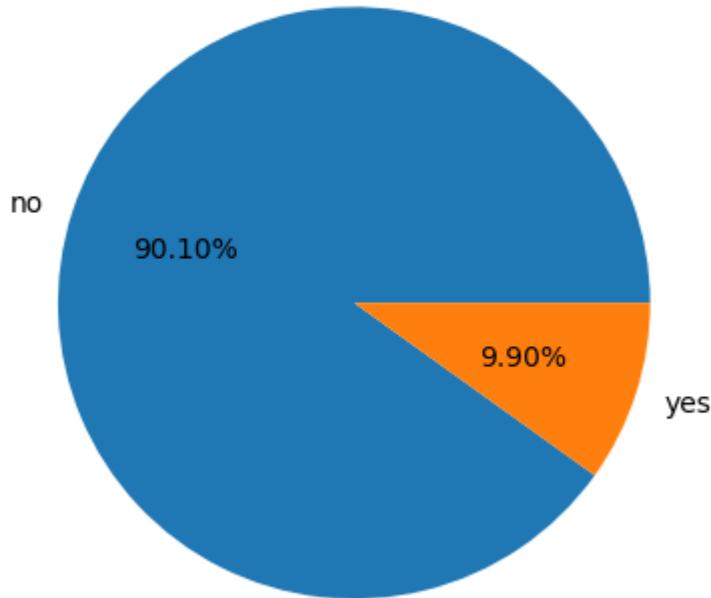
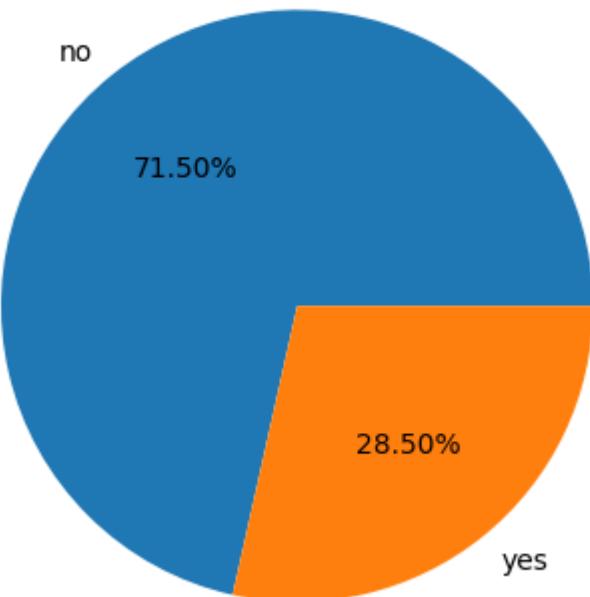
PIE CHARTIn [39]: `cat_clm`Out[39]: `Index(['phone_no', 'gender', 'multi_screen', 'mail_subscribed'], dtype='object')`In [40]: `keys=tele_df["gender"].value_counts().keys()
values=tele_df["gender"].value_counts().values
plt.pie(values, labels=keys, autopct="%0.2f%%")
plt.show()`

In []:

USING FOR-LOOP

```
In [43]: for i in cat_clm[1:]:
    keys=tele_df[i].value_counts().keys()
    values=tele_df[i].value_counts().values
    plt.title(i)
    plt.pie(values,labels=keys,autopct="%0.2f%%")
    plt.show()
```



multi_screen**mail_subscribed**

In []:

In []:

USING FOR-LOOP IN SUBPLOT

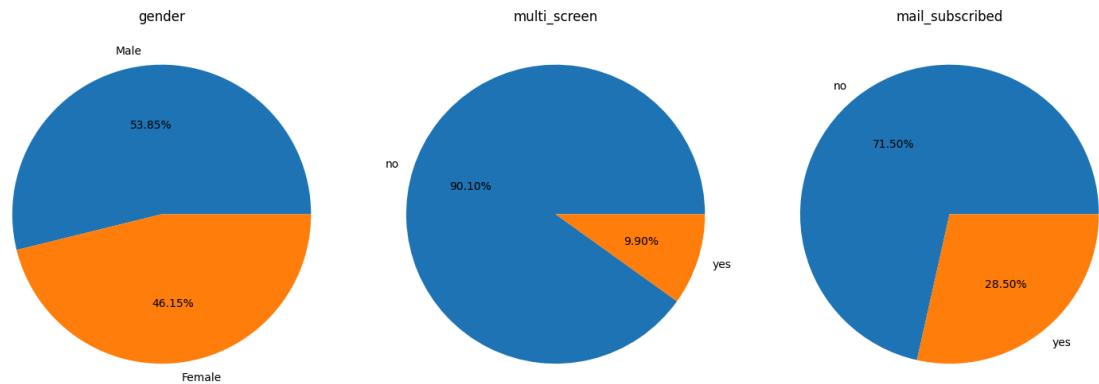
```
In [45]: plt.figure(figsize=(15, 5))
for idx, i in enumerate(cat_clm[1:], start=1):
    plt.subplot(1, 3, idx)
```

```

keys=tele_df[i].value_counts().keys()
values=tele_df[i].value_counts().values
plt.title(i)
plt.pie(values, labels=keys, autopct="%0.2f%%")

plt.tight_layout()
plt.show()

```



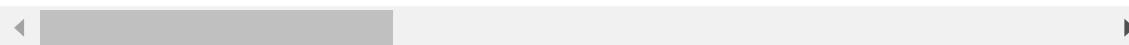
In []:

NUMERICAL COLUMNS ANALYSIS

In [47]: tele_df

	year	customer_id	phone_no	gender	age	no_of_days_subscribed	multi_screen
0	2015	100198	409-8743	Female	36	62	no
1	2015	100643	340-5930	Female	39	149	no
2	2015	100756	372-3750	Female	65	126	no
3	2015	101595	331-4902	Female	24	131	no
4	2015	101653	351-8398	Female	40	191	no
...
1995	2015	997132	385-7387	Female	54	75	no
1996	2015	998086	383-9255	Male	45	127	no
1997	2015	998474	353-2080	Male	53	94	no
1998	2015	998934	359-7788	Male	40	94	no
1999	2015	999961	414-1496	Male	37	73	no

2000 rows × 16 columns



In [48]: num_clm

```
Out[48]: Index(['year', 'customer_id', 'age', 'no_of_days_subscribed',
       'weekly_mins_watched', 'minimum_daily_mins', 'maximum_daily_mins',
       'weekly_max_night_mins', 'videos_watched', 'maximum_days_inactive',
       'customer_support_calls', 'churn'],
      dtype='object')
```

```
In [49]: # First we have to drop the churn columns
```

```
num_clm_2=num_clm.drop("churn")
num_clm_2
```

```
Out[49]: Index(['year', 'customer_id', 'age', 'no_of_days_subscribed',
       'weekly_mins_watched', 'minimum_daily_mins', 'maximum_daily_mins',
       'weekly_max_night_mins', 'videos_watched', 'maximum_days_inactive',
       'customer_support_calls'],
      dtype='object')
```

```
In [50]: # frequency table for numeral column
# We will use scribe method here
```

```
In [51]: tele_df.describe()
```

	year	customer_id	age	no_of_days_subscribed	weekly_mins_watched
count	2000.0	2000.000000	2000.000000	2000.000000	2000.000000
mean	2015.0	554887.157500	38.69050	99.750000	270.178425
std	0.0	261033.690318	10.20641	39.755386	80.551627
min	2015.0	100198.000000	18.00000	1.000000	0.000000
25%	2015.0	328634.750000	32.00000	73.000000	218.212500
50%	2015.0	567957.500000	37.00000	99.000000	269.925000
75%	2015.0	773280.250000	44.00000	127.000000	324.675000
max	2015.0	999961.000000	82.00000	243.000000	526.200000

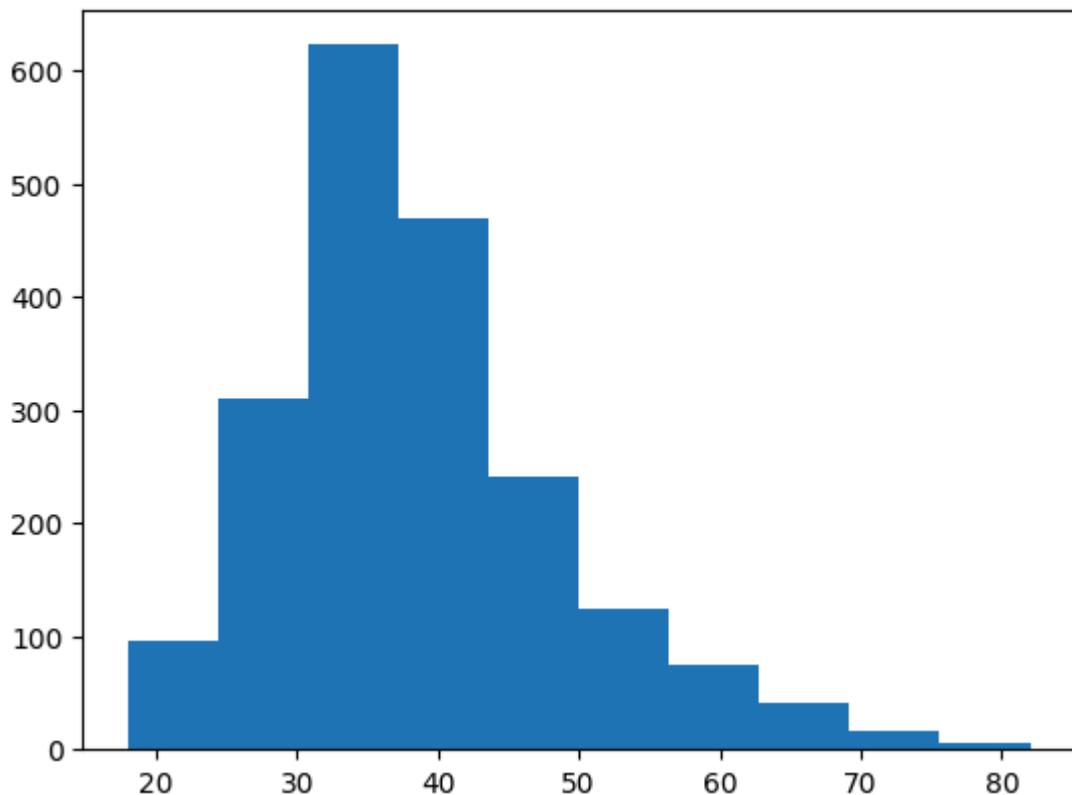
◀ ▶

```
In [ ]:
```

HOSTOGRAM

```
In [53]: data=tele_df["age"]
plt.hist(data)
plt.title("AGE HISTOGRAM")
plt.show()
```

AGE HISTOGRAM

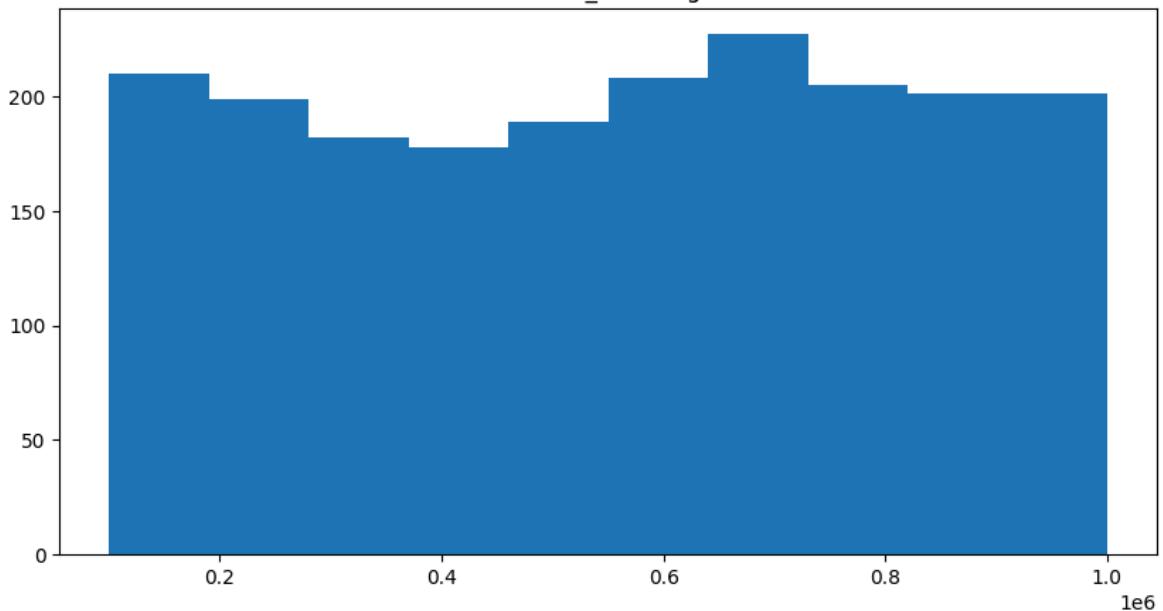


USING FOR-LOOP

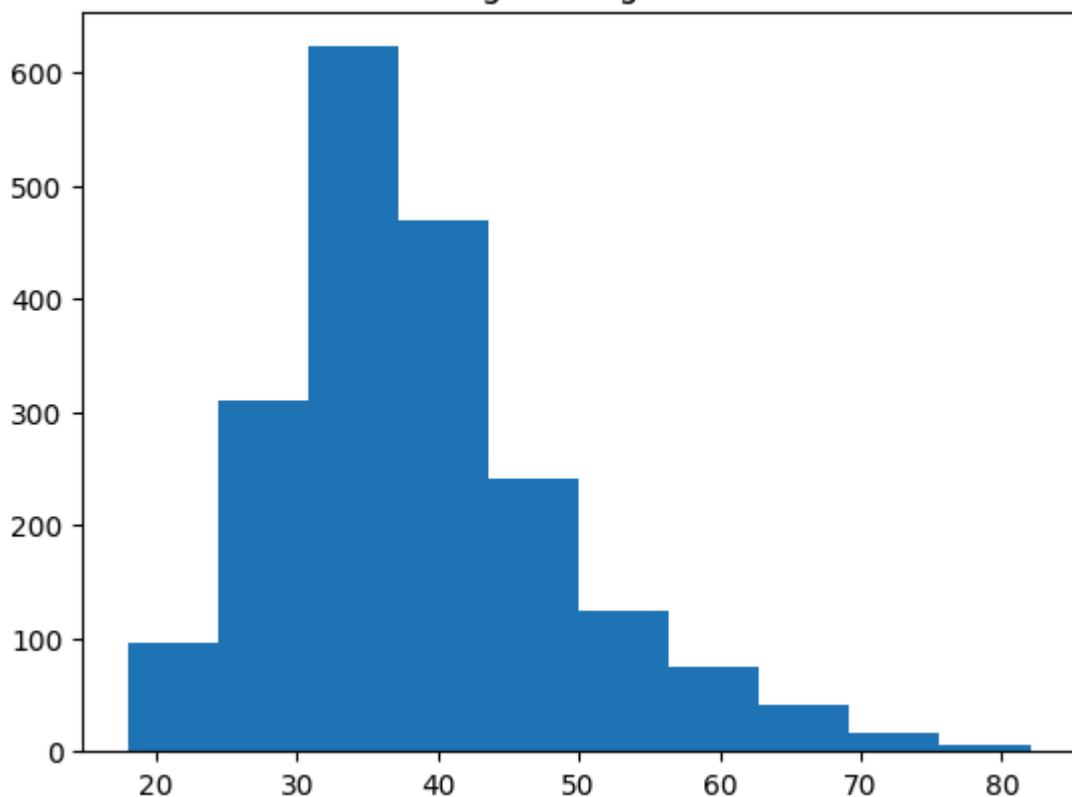
In [54]:

```
plt.figure(figsize=(10,5))
for i in num_clm_2[1:]:
    plt.hist(tele_df[i])
    plt.title(f" {i} Histogram")
    plt.show()
```

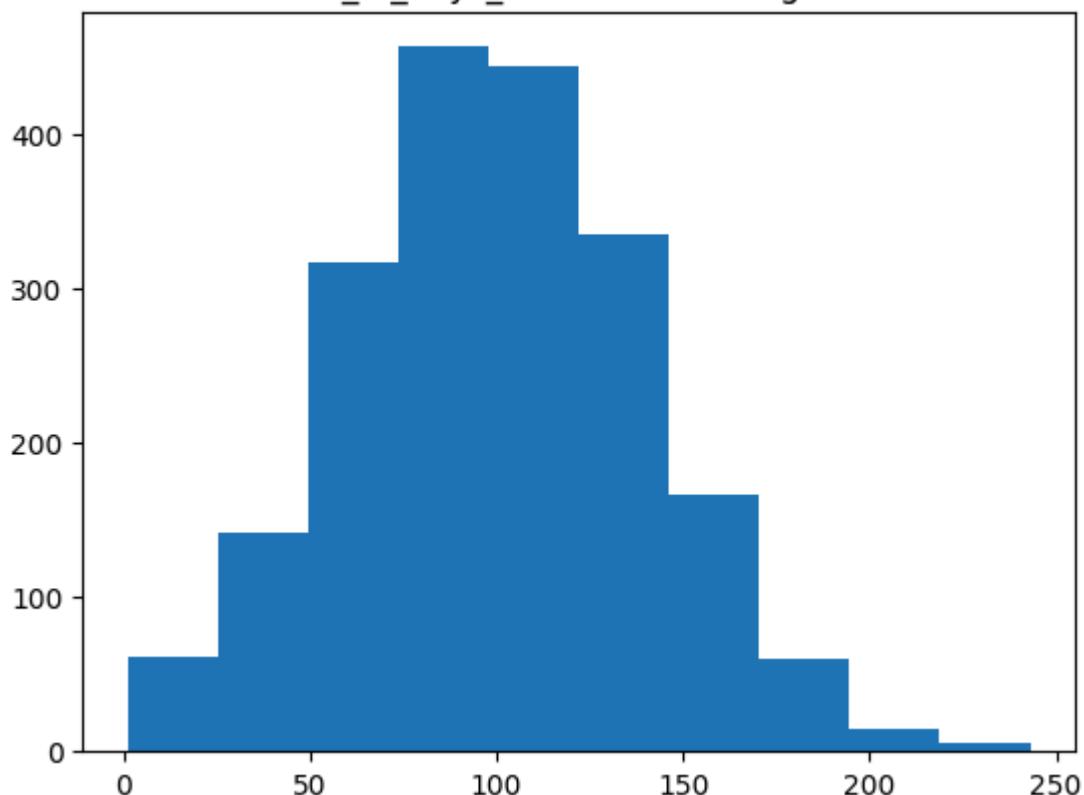
customer_id Histogram

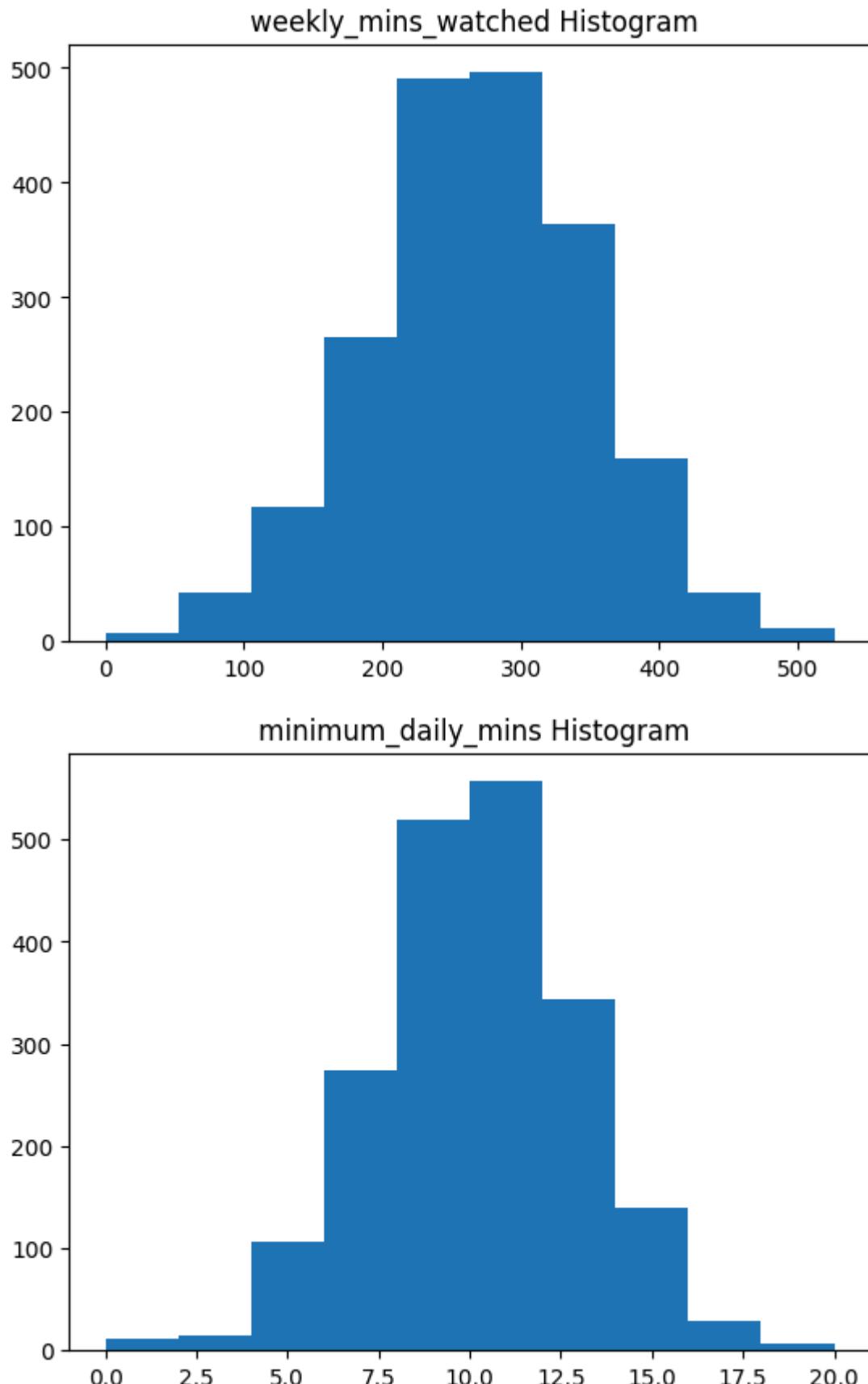


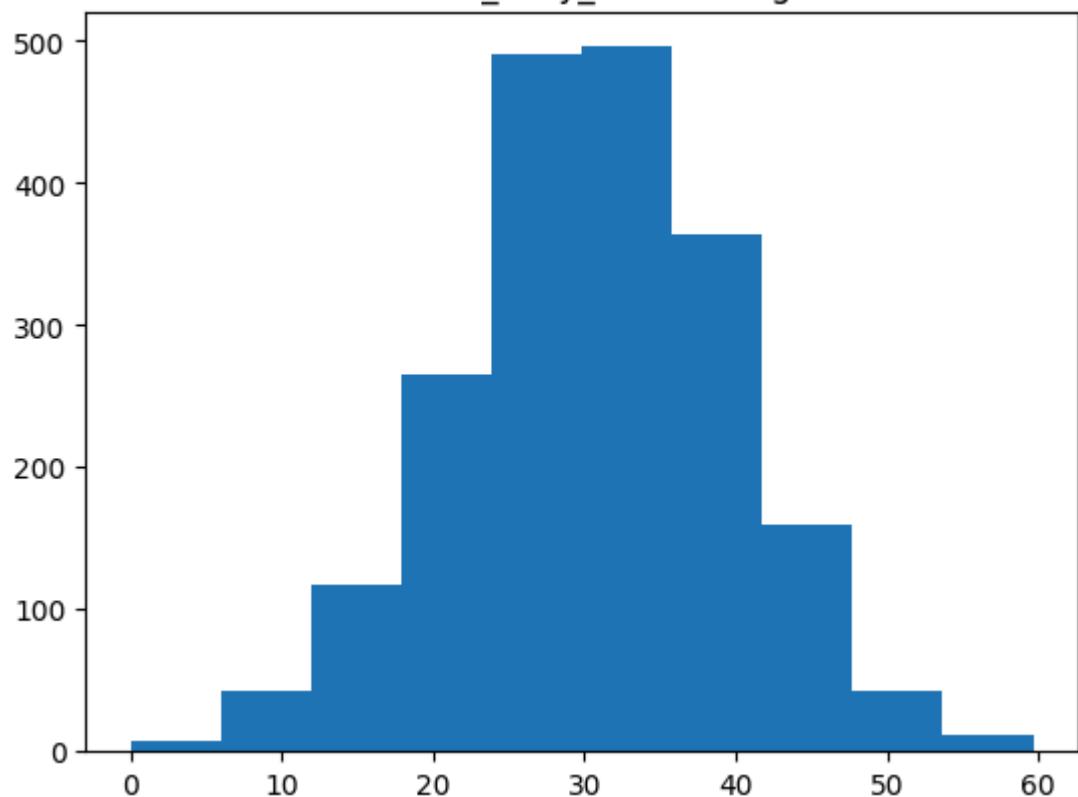
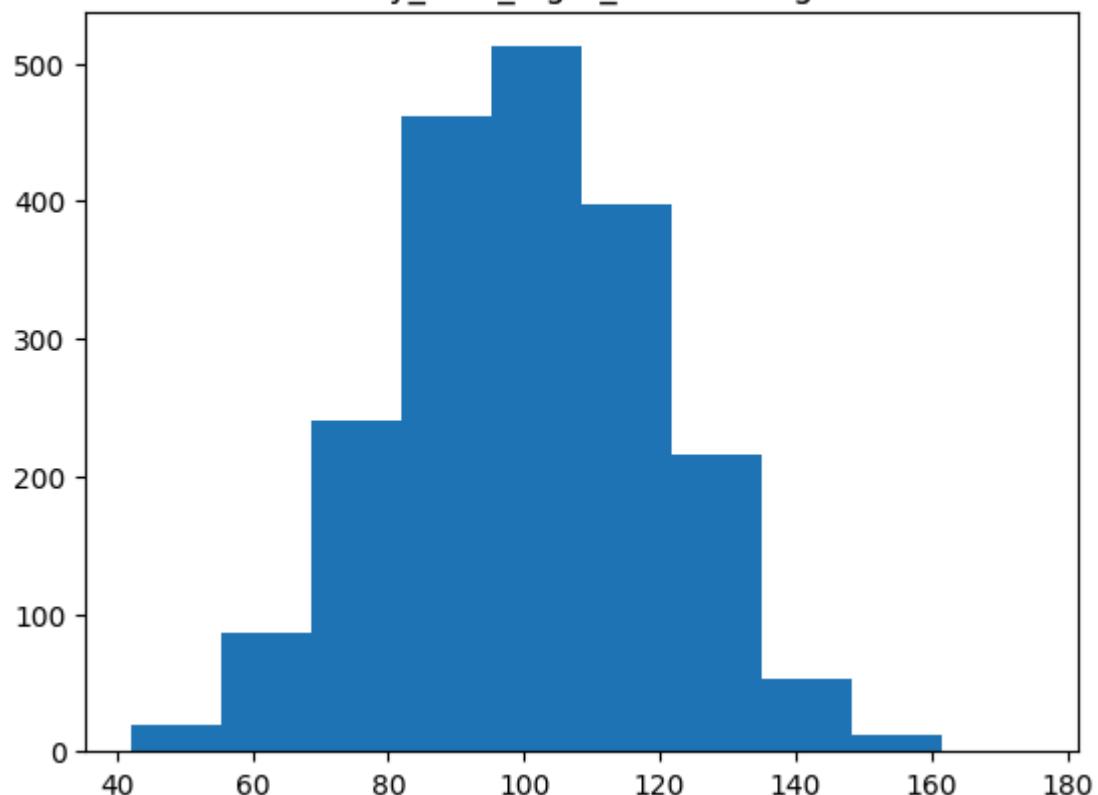
age Histogram



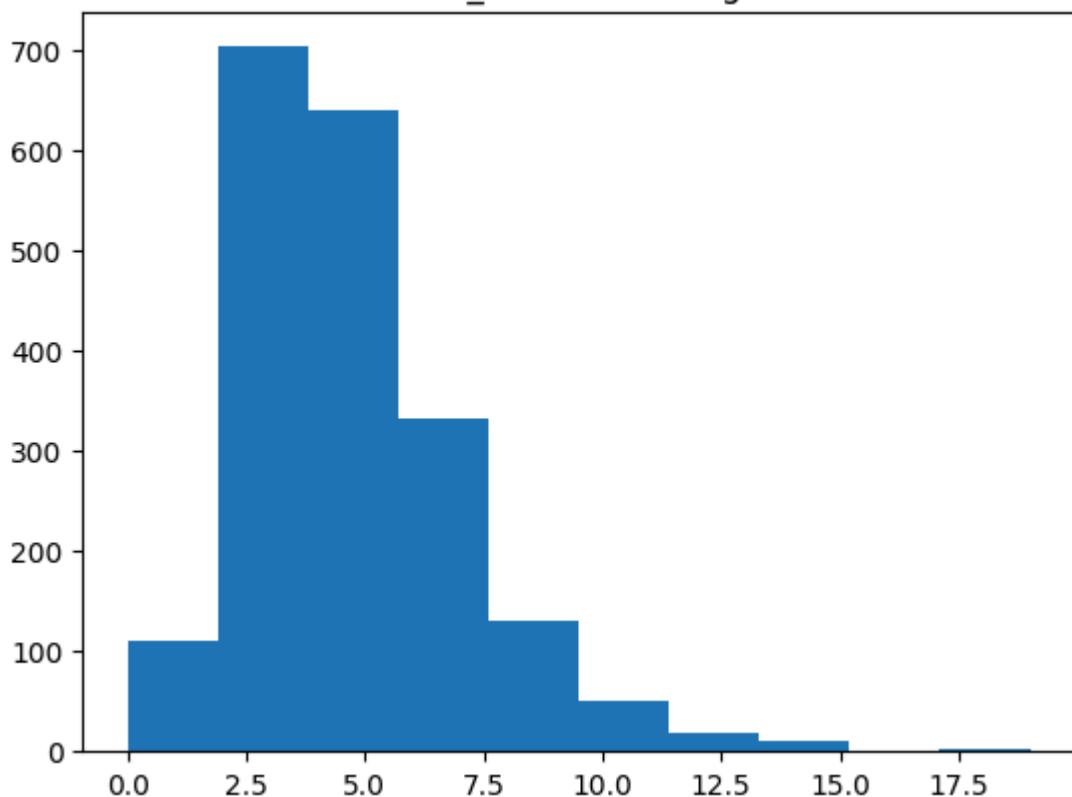
no_of_days_subscribed Histogram



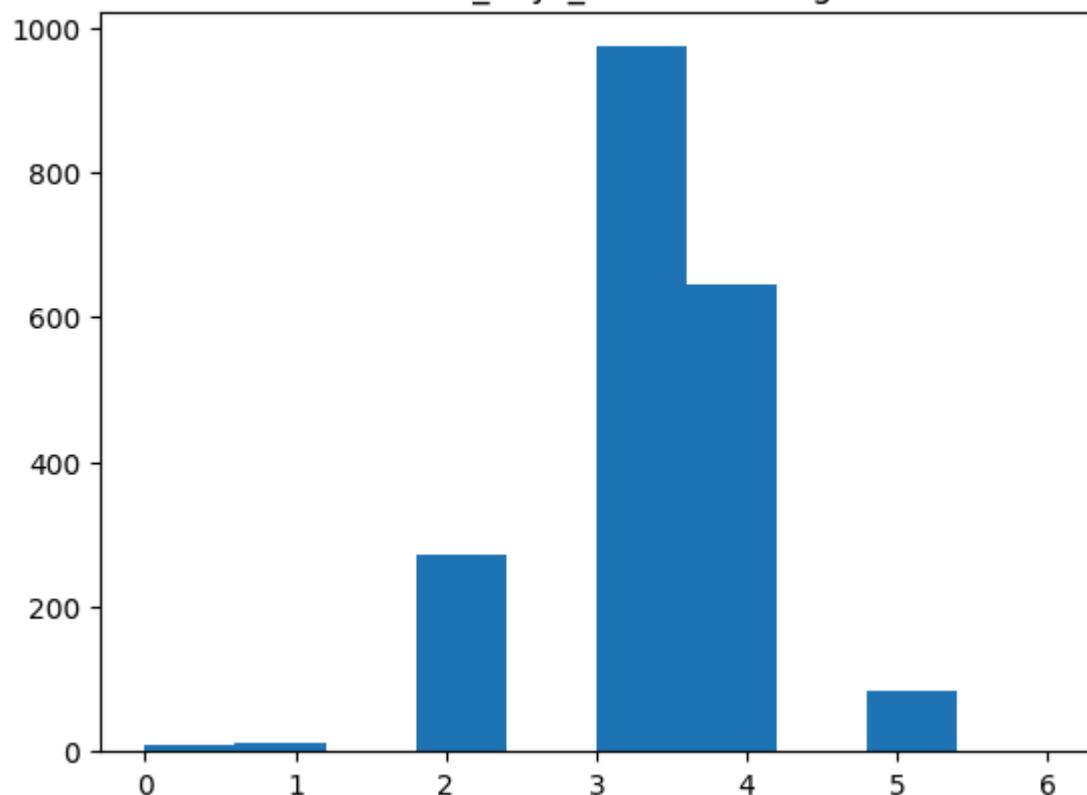


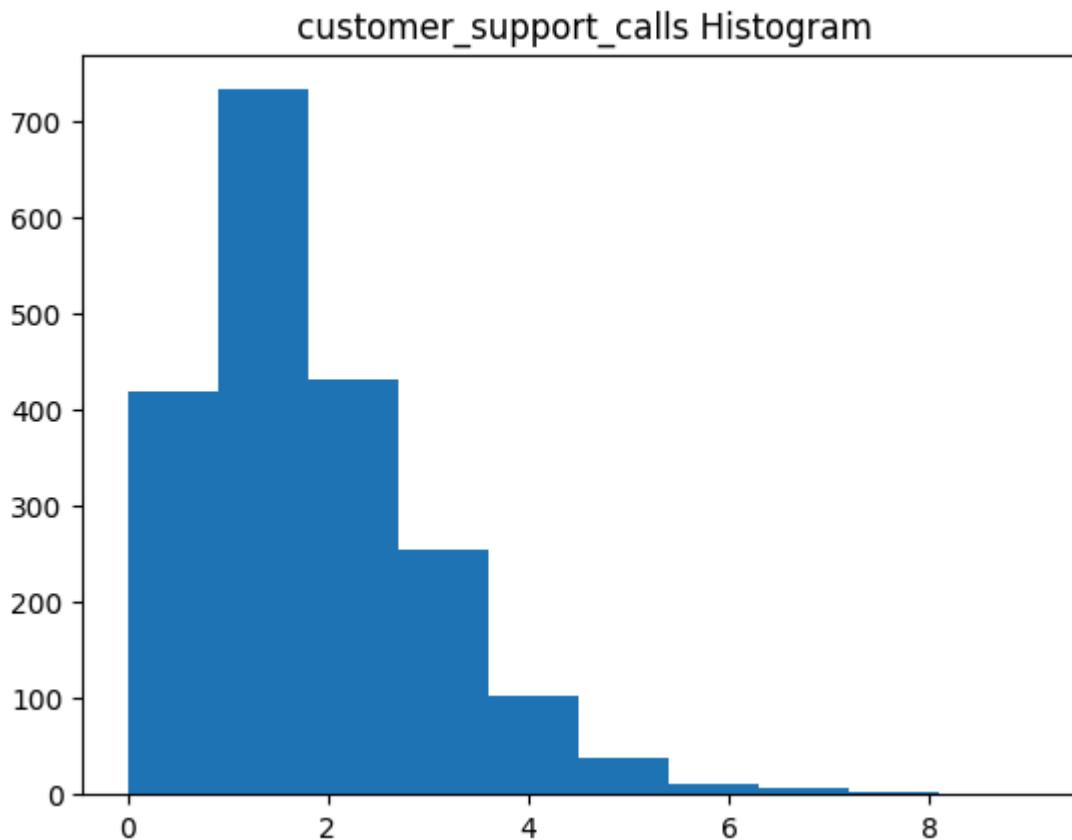
maximum_daily_mins Histogram**weekly_max_night_mins Histogram**

videos_watched Histogram



maximum_days_inactive Histogram





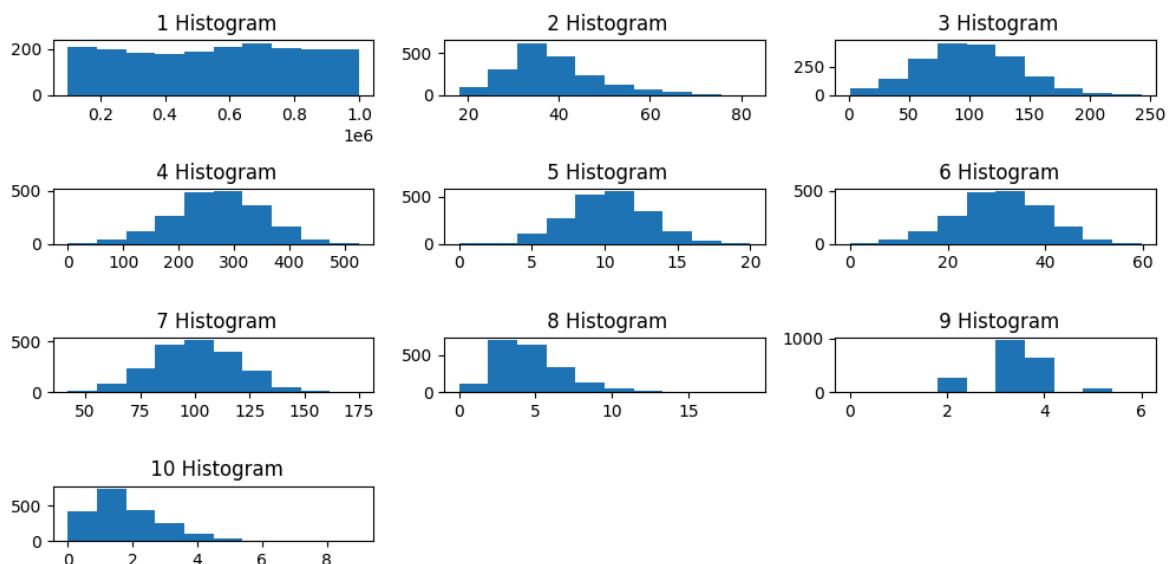
In []:

SUBPLOT

```
In [56]: plt.figure(figsize=(10,5))
for i in range(1,len(num_clm_2)):

    plt.subplot(4,3,i)
    plt.hist(tele_df[num_clm_2[i]])
    plt.title(f" {i} Histogram")

plt.tight_layout()
plt.show()
```



In []:

DATA DISTRIBUTION

```
In [58]: tele_df=pd.read_csv("telecom_churn_data.csv")
tele_df
```

Out[58]:

	year	customer_id	phone_no	gender	age	no_of_days_subscribed	multi_screen
0	2015	100198	409-8743	Female	36	62	no
1	2015	100643	340-5930	Female	39	149	no
2	2015	100756	372-3750	Female	65	126	no
3	2015	101595	331-4902	Female	24	131	no
4	2015	101653	351-8398	Female	40	191	no
...
1995	2015	997132	385-7387	Female	54	75	no
1996	2015	998086	383-9255	Male	45	127	no
1997	2015	998474	353-2080	NaN	53	94	no
1998	2015	998934	359-7788	Male	40	94	no
1999	2015	999961	414-1496	Male	37	73	no

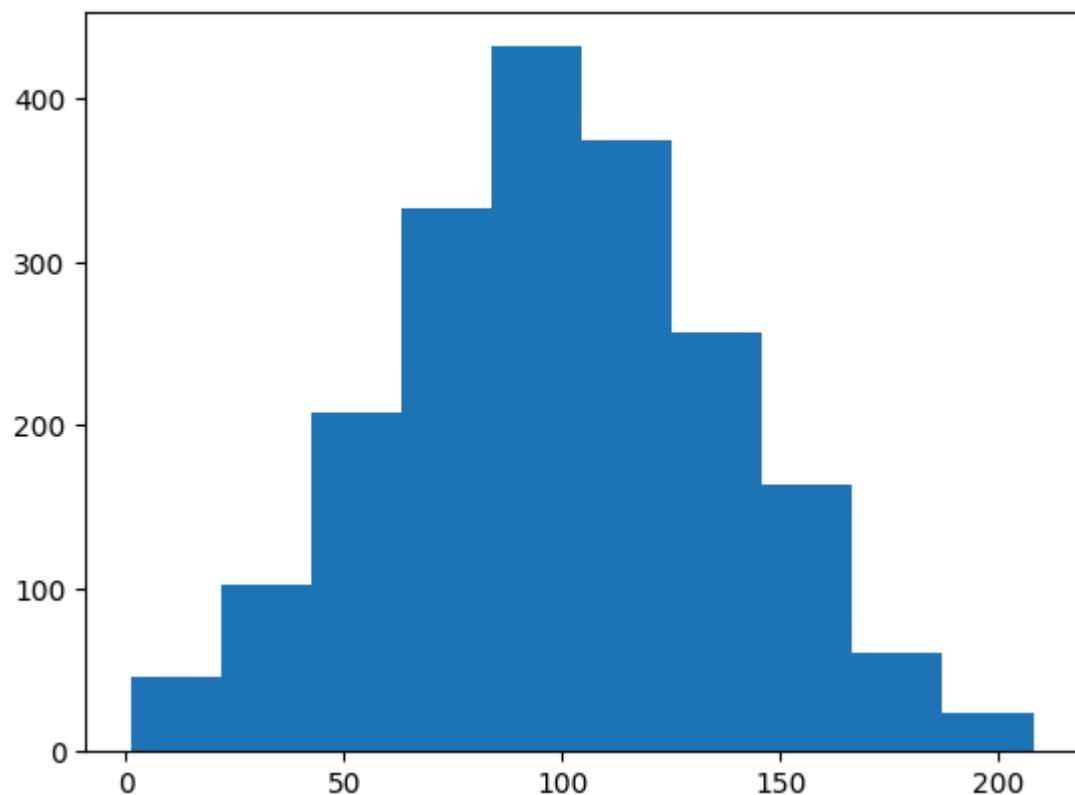
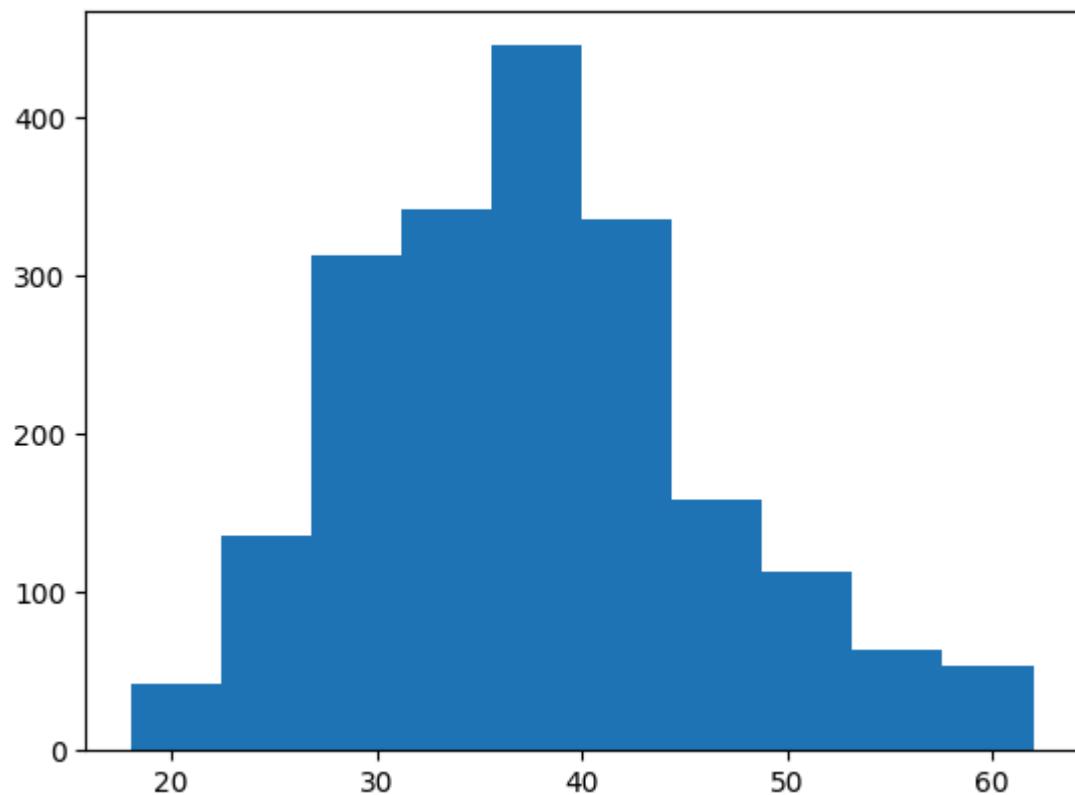
2000 rows × 16 columns

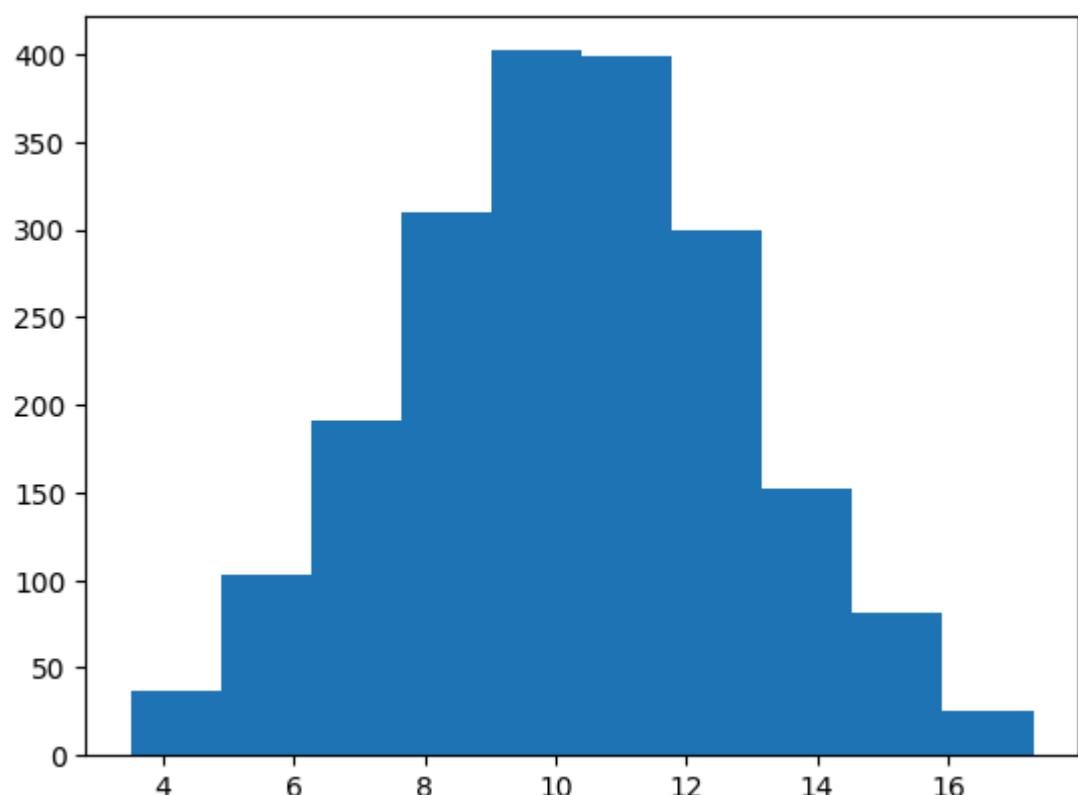
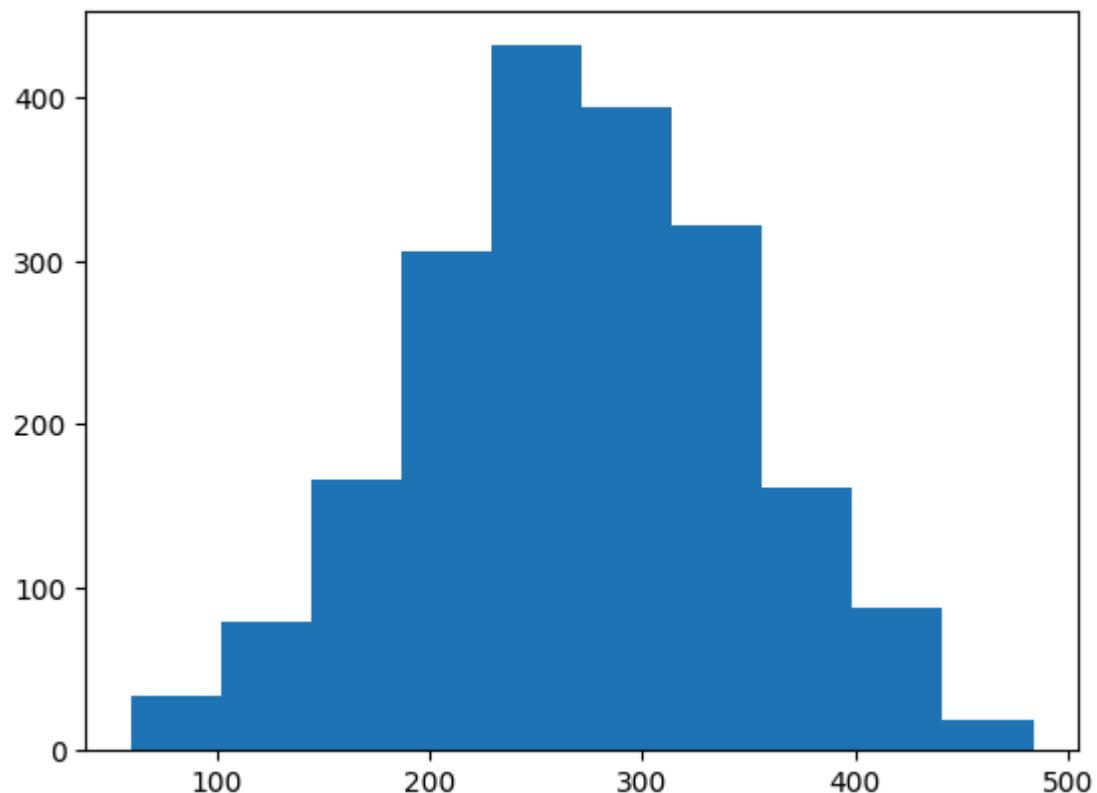
```
In [59]: for i in num_clm_2[2:]:
    q1=np.quantile(tele_df[i],q=0.25)
    q2=np.quantile(tele_df[i],q=0.50)
    q3=np.quantile(tele_df[i],q=0.75)
```

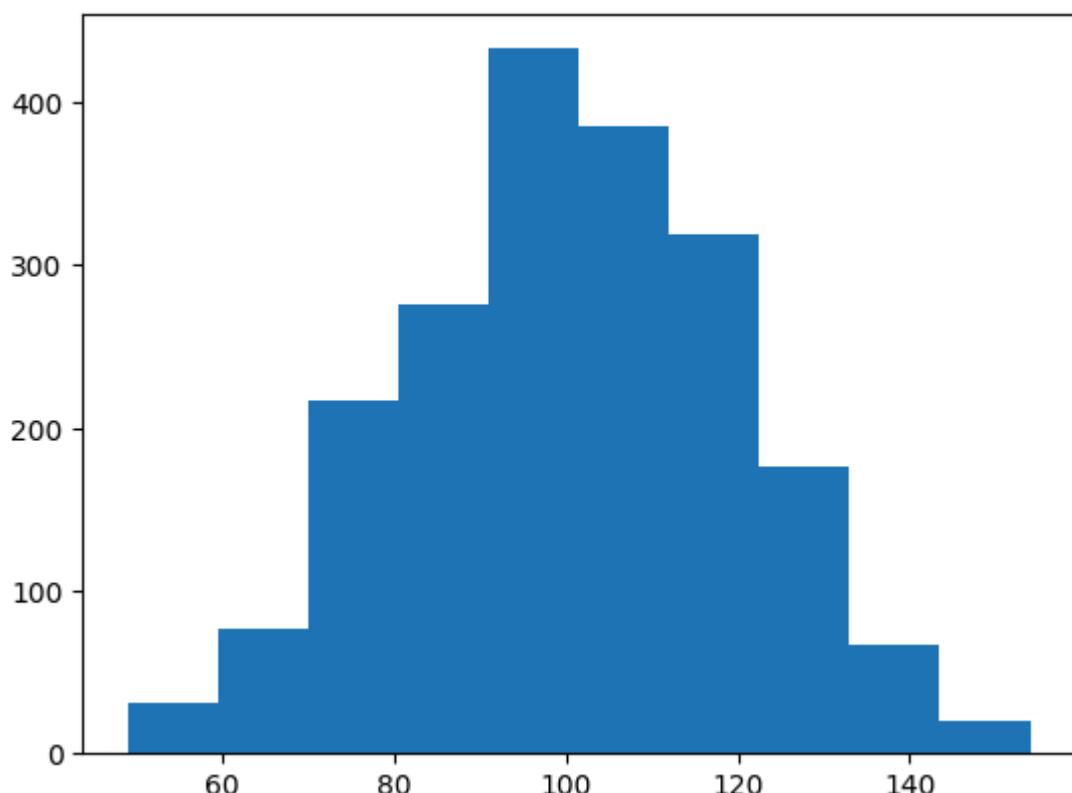
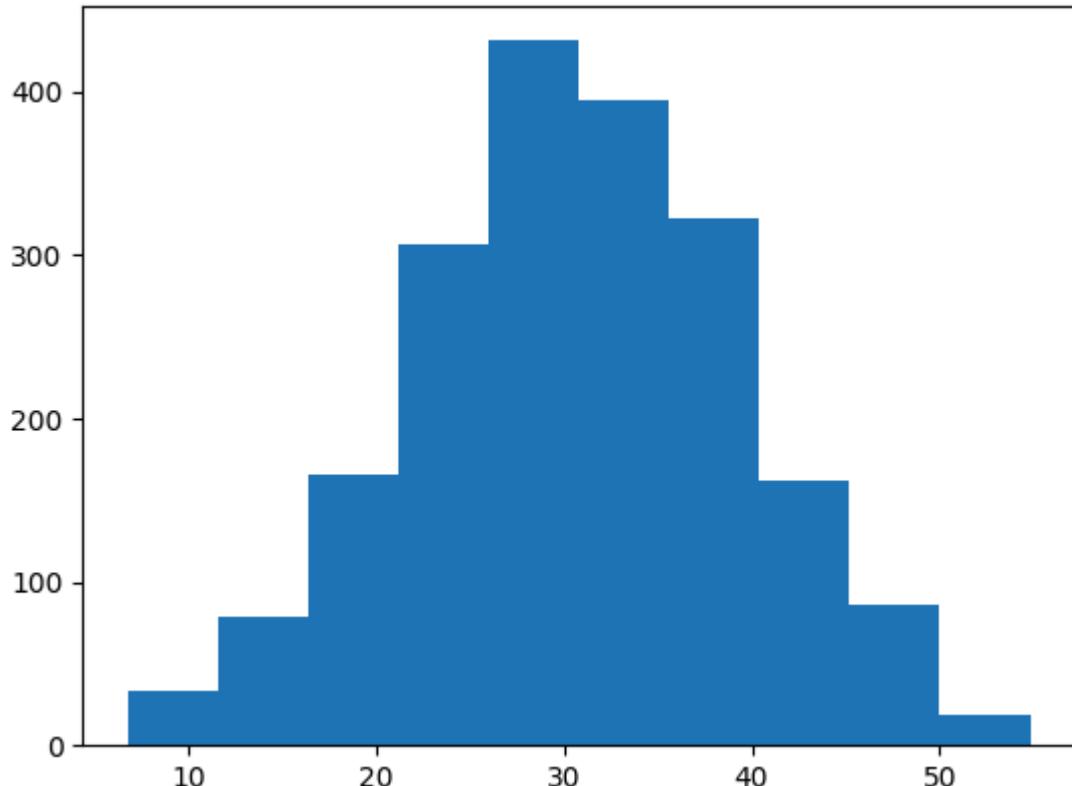
```
iqr=q3-q1
lb=q1-1.5*iqr
ub=q3+1.5*iqr
cont=(tele_df[i]>ub) | (tele_df[i]<lb)
true=np.median(tele_df[i])
false=tele_df[i]
```

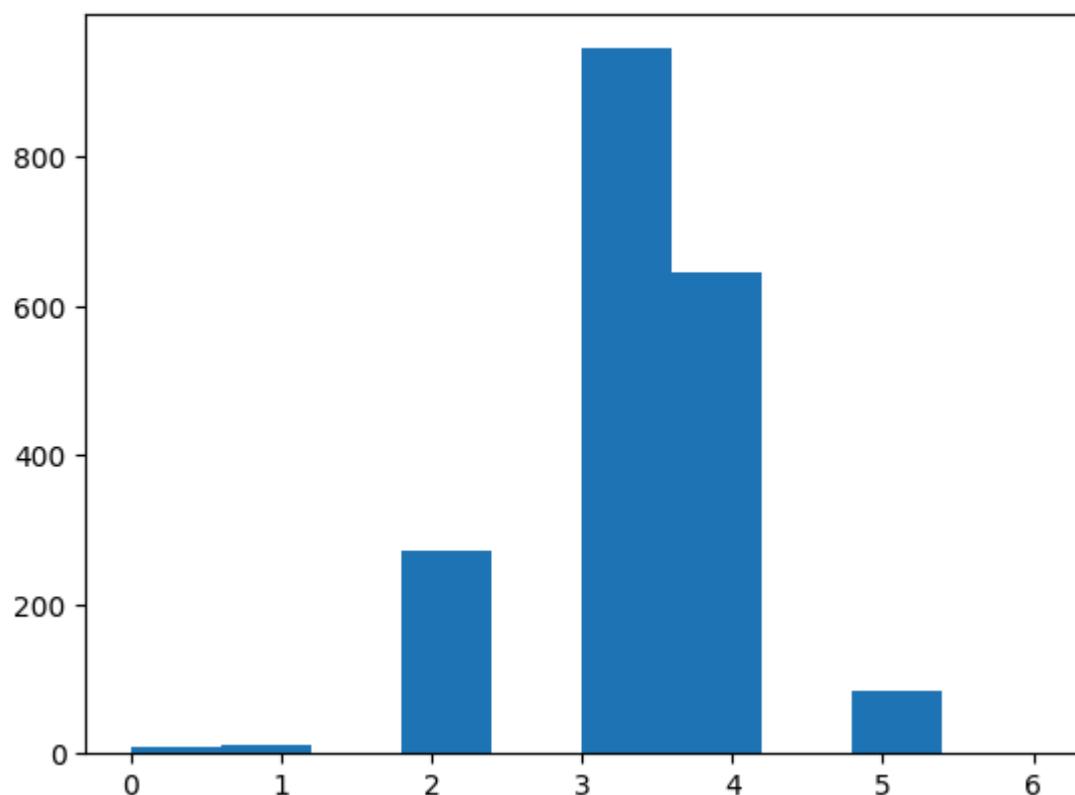
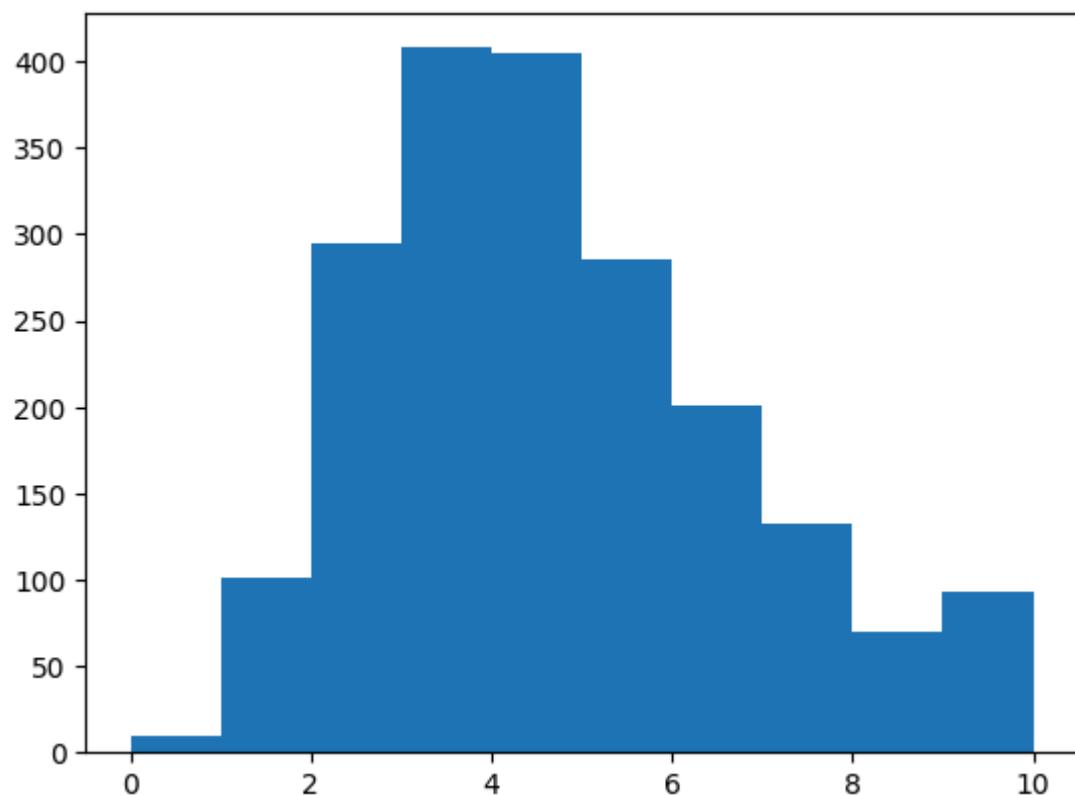
```
v=np.where(cont,true,false)
tele_df[i]=v
```

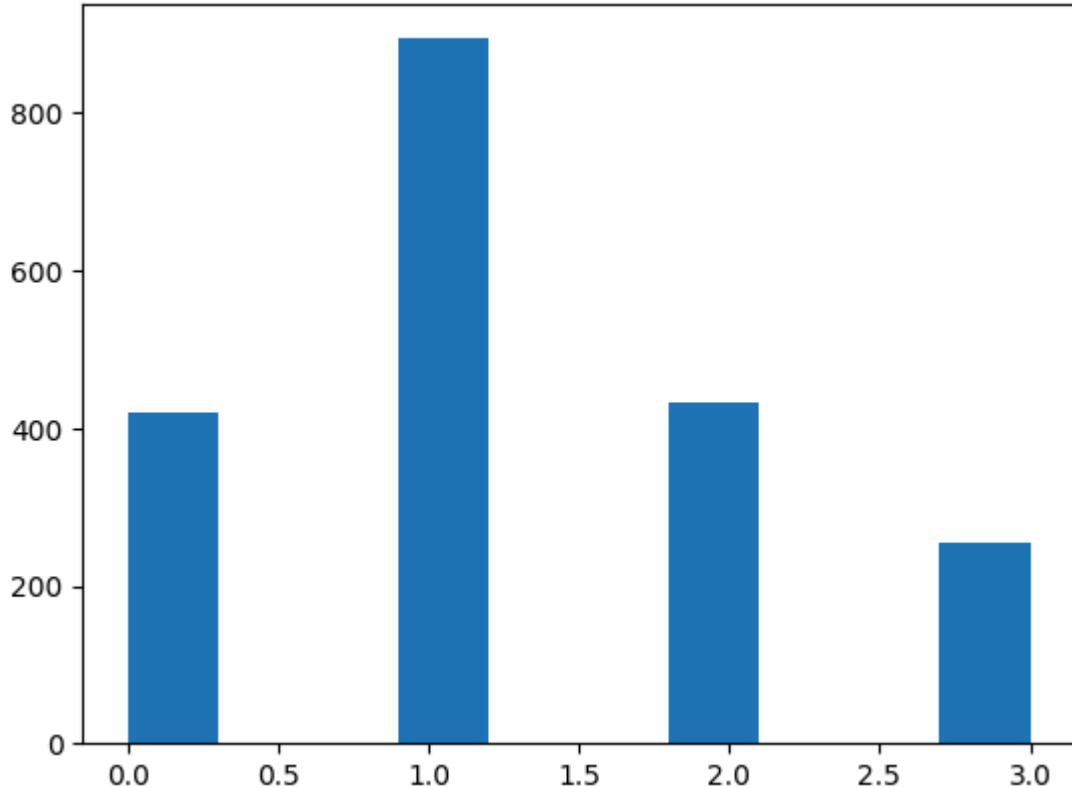
```
plt.hist(v)
plt.show()
```











In []:

DATA DISTRIBUTION WITH SUBPLOT

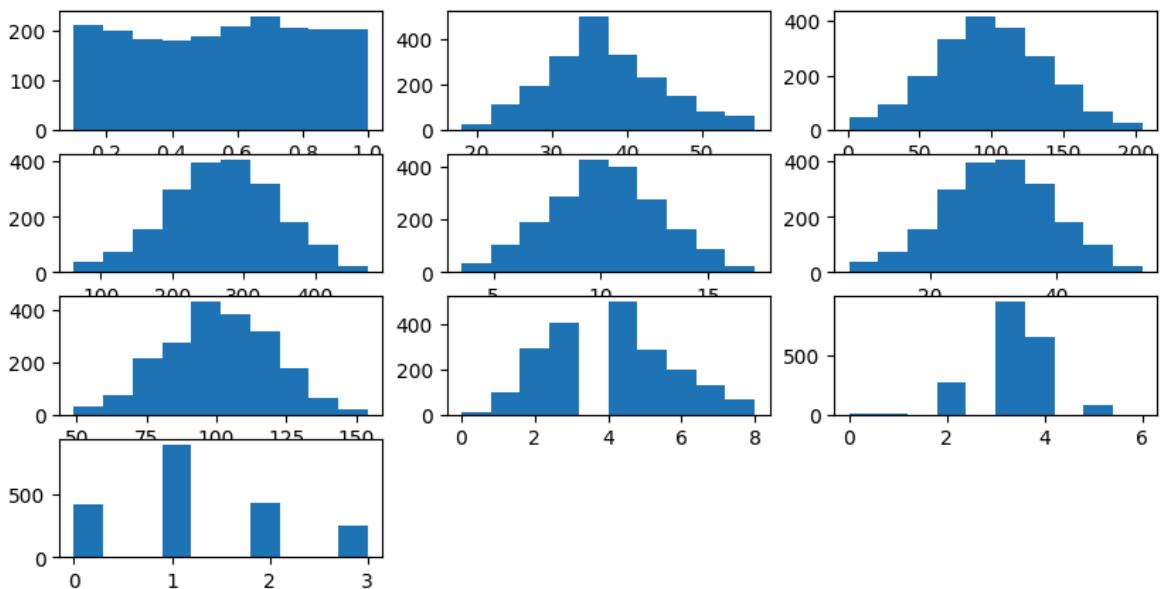
```
In [61]: plt.figure(figsize=(10,5))
for i in range(1,len(num_clm_2)):
    plt.subplot(4,3,i)

    q1=np.quantile(tele_df[num_clm[i]],q=0.25)
    q2=np.quantile(tele_df[num_clm[i]],q=0.50)
    q3=np.quantile(tele_df[num_clm[i]],q=0.75)

    iqr=q3-q1
    lb=q1-1.5*iqr
    ub=q3+1.5*iqr
    cont=(tele_df[num_clm[i]]>ub) | (tele_df[num_clm[i]]<lb)
    true=np.median(tele_df[num_clm[i]])
    false=tele_df[num_clm[i]]

    v=np.where(cont,true,false)
    tele_df[i]=v

    plt.hist(v)
plt.show()
```



In []:

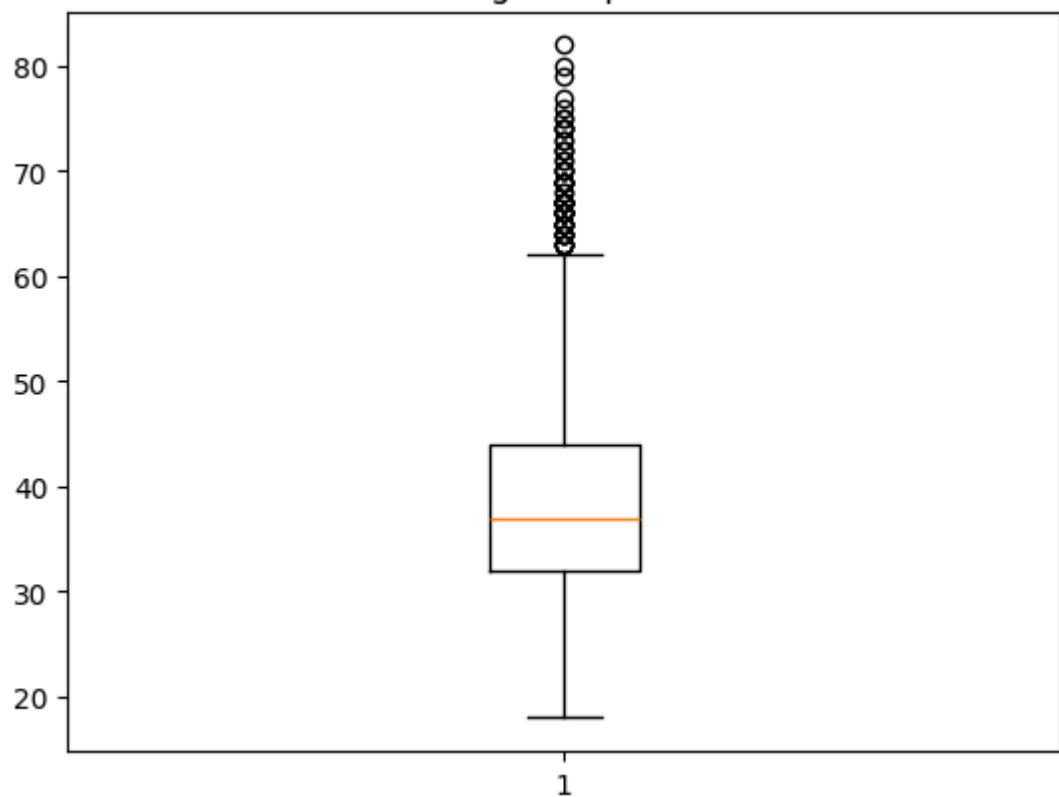
BOXPLOT

- box plot is used to identify the outliers
- Outliers is an observation having huge positive value or huge negative value

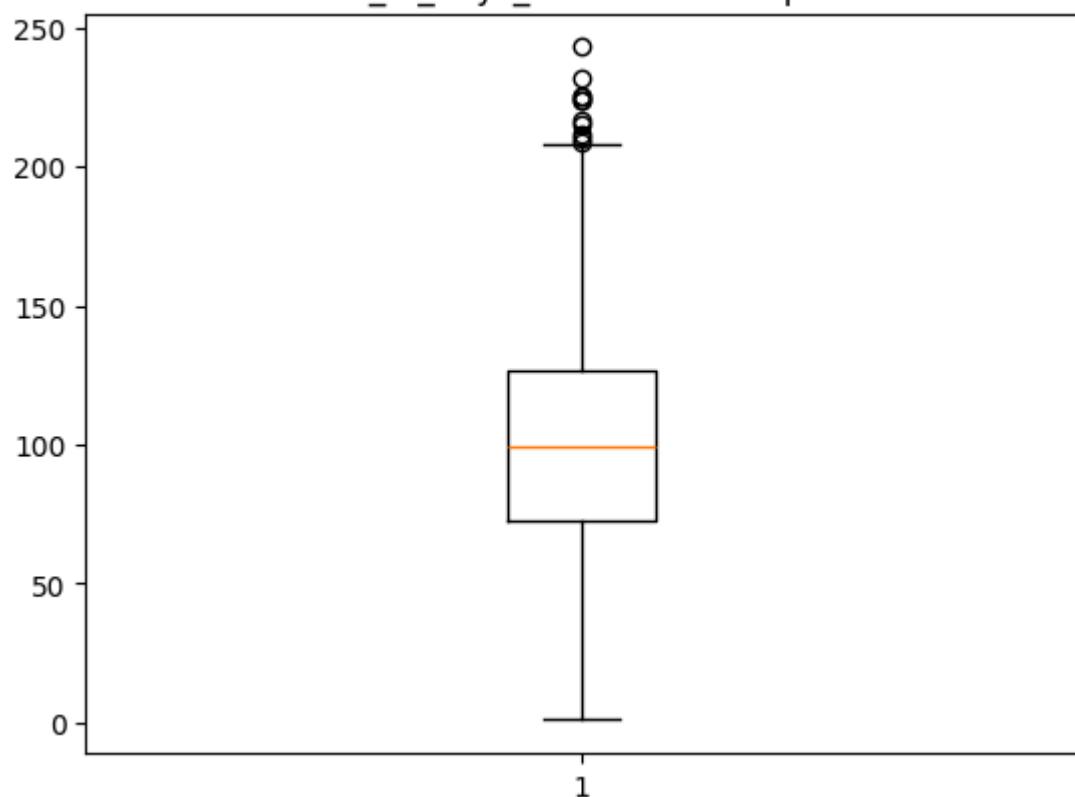
```
In [64]: def boxplot():
    tele_df=pd.read_csv("telecom_churn_data.csv")

    for i in num_clm_2[2:]:
        box_data=tele_df[i]
        plt.boxplot(box_data)
        plt.title(f" {i} Boxplot")
        plt.show()
boxplot()
```

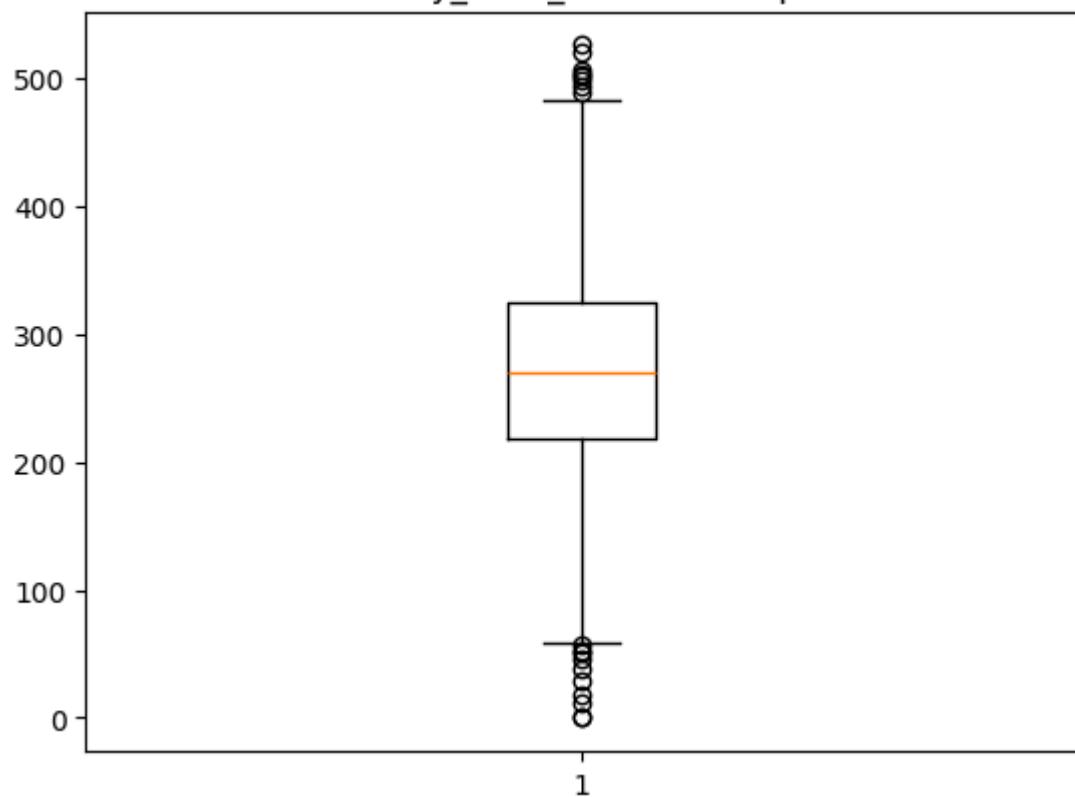
age Boxplot



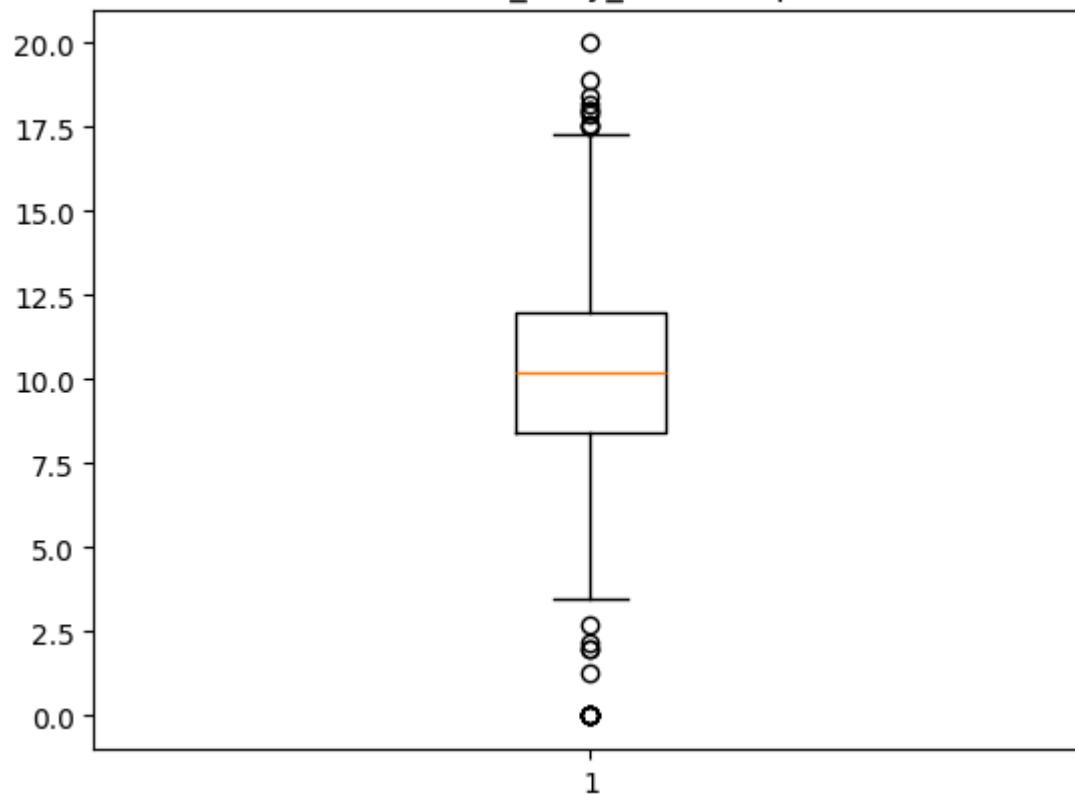
no_of_days_subscribed Boxplot



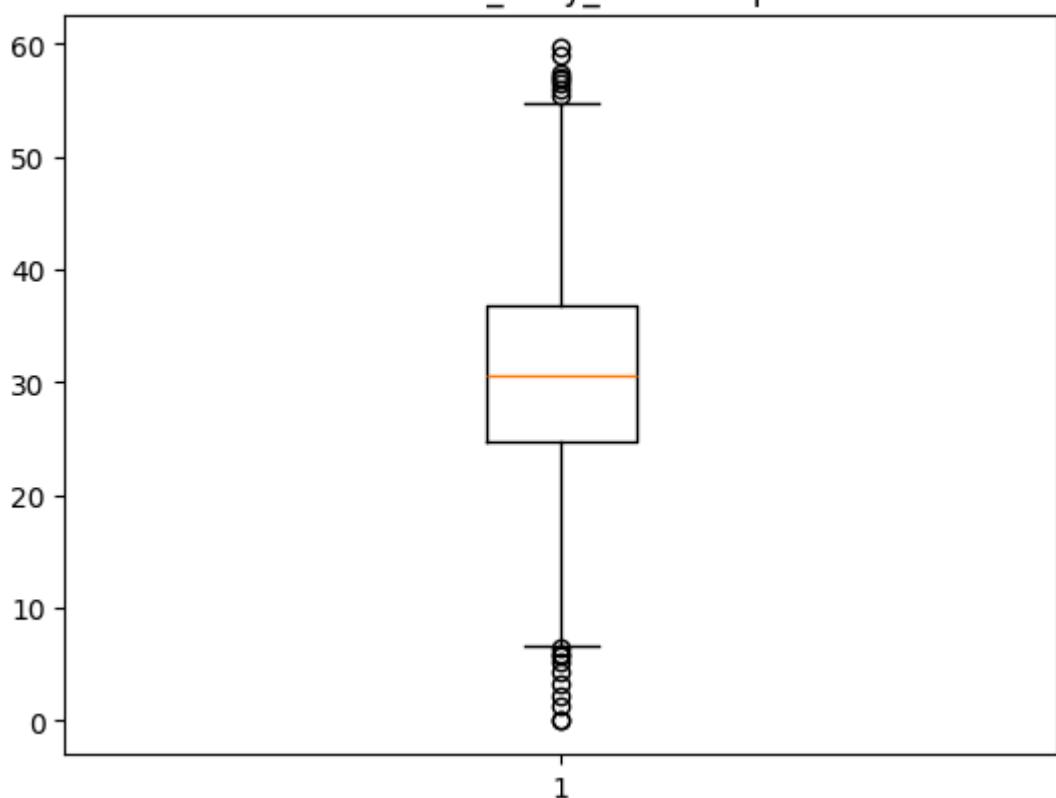
weekly_mins_watched Boxplot



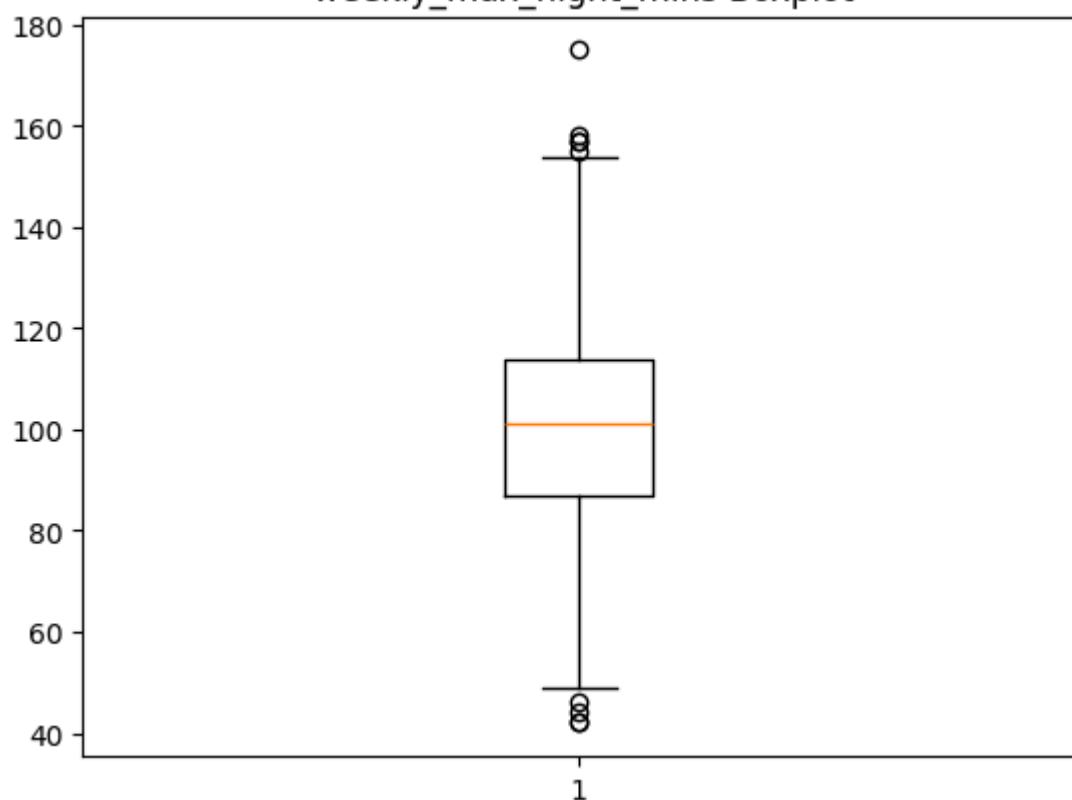
minimum_daily_mins Boxplot



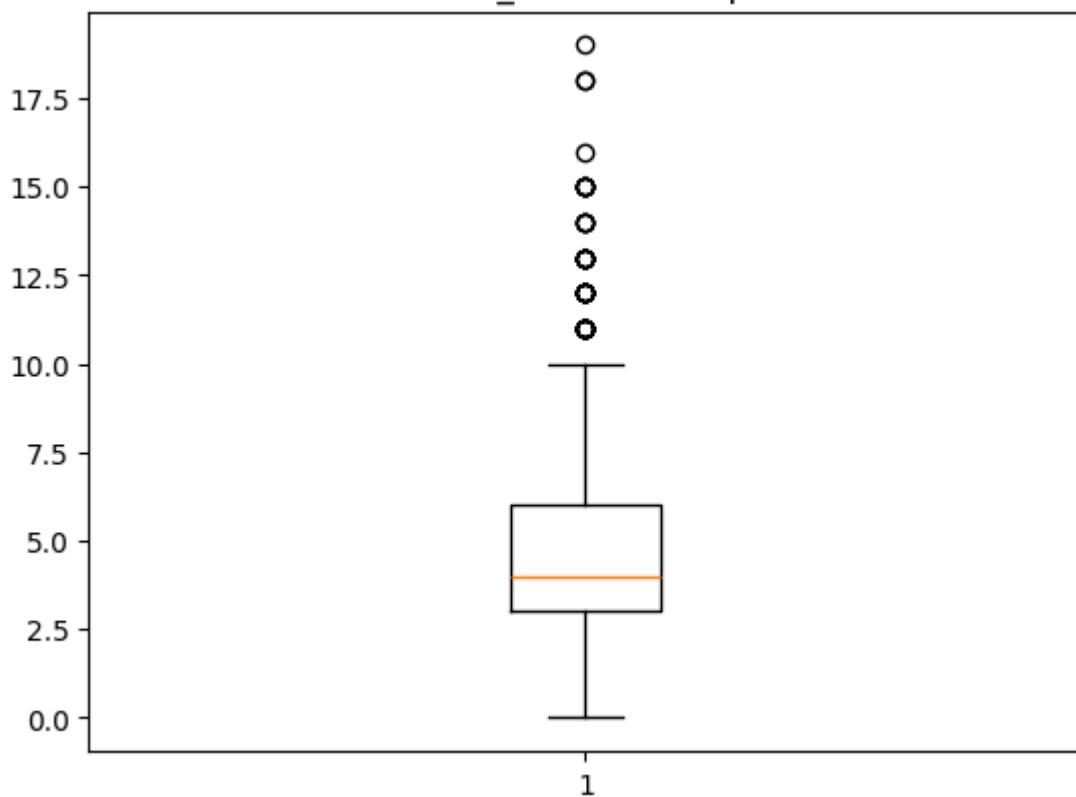
maximum_daily_mins Boxplot



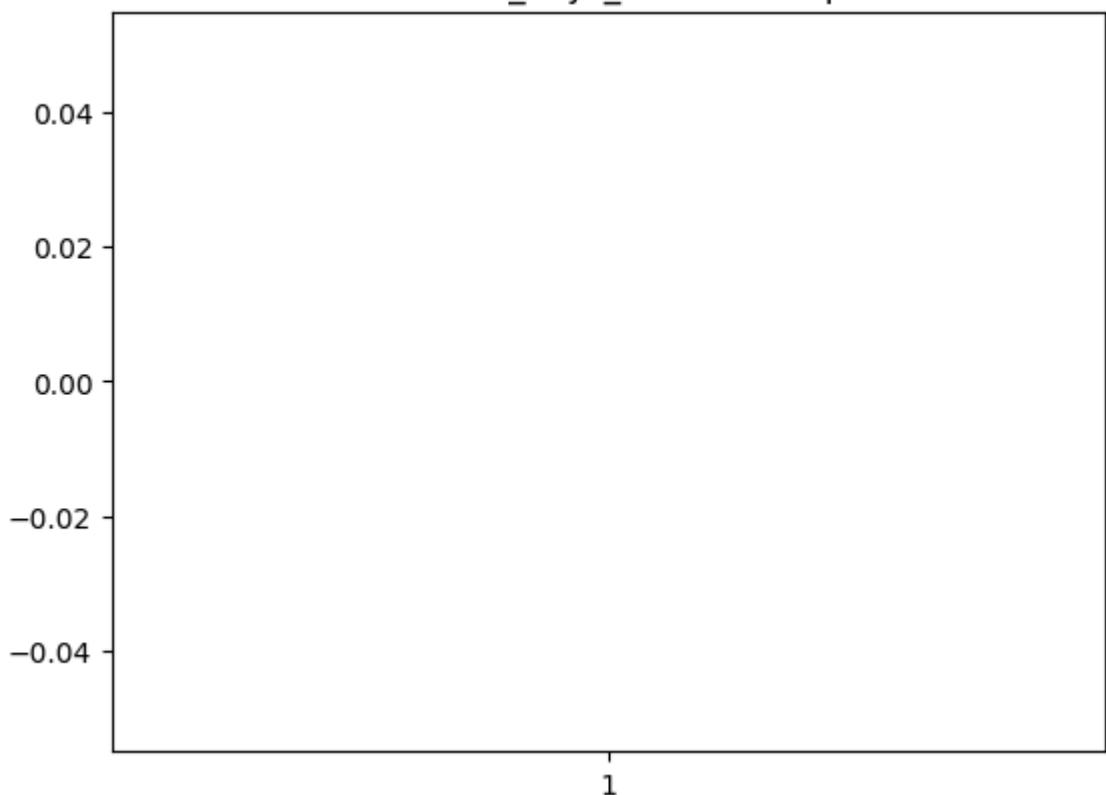
weekly_max_night_mins Boxplot

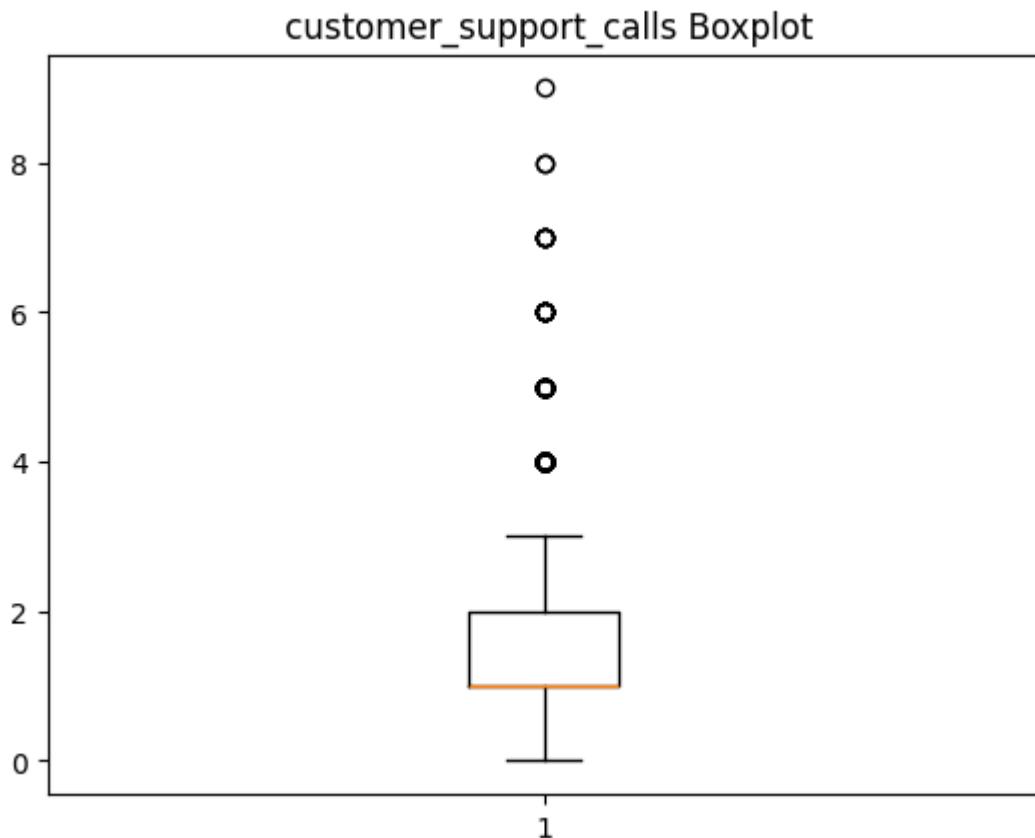


videos_watched Boxplot



maximum_days_inactive Boxplot





In []:

OUTLIERS ANALYSIS

- We know that Outliers available less than $Q1 - 1.5IQR$ and $Q3 + 1.5IQR$
- step-1:
 - Calculate $Q1 = 25p$
 - Calculate $Q2 = 50p$
 - Calculate $Q3 = 75p$
- step-2:
 - Calculate $IQR = Q3 - Q1$
- step-3 :
 - Calculate $lb = Q1 - 1.5IQR$
 - calculate $ub = Q1 + 1.5IQR$
- step-4 :
 - $con1 = \text{data} < lb$
 - $con2 = \text{data} > ub$
 - $con3 = con1 \text{ or-and } con2$ (bitwise operator)
- step-5
 - $\text{data}[con3]$

In []:

OUTLIERS DATA

```
In [68]: tele_df = pd.read_csv("telecom_churn_data.csv")
outliers_data = pd.DataFrame()

for i in num_clm_2:
    q1 = np.quantile(tele_df[i], q=0.25)
    q2 = np.quantile(tele_df[i], q=0.50)
    q3 = np.quantile(tele_df[i], q=0.75)

    iqr = q3 - q1

    lb = q1 - 1.5 * iqr
    ub = q3 + 1.5 * iqr

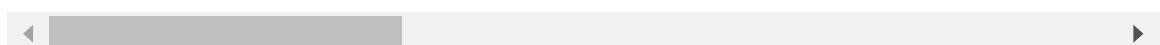
    con1 = tele_df[i] > ub
    con2 = tele_df[i] < lb
    con3 = con1 | con2

# Append outliers for this column
outliers_data = pd.concat([outliers_data, tele_df[con3]])
```

In [69]: outliers_data

	year	customer_id	phone_no	gender	age	no_of_days_subscribed	multi_screen
2	2015	100756	372-3750	Female	65	126	no
30	2015	111252	410-9633	Female	63	106	no
71	2015	124837	413-2241	Male	67	163	no
87	2015	131070	334-9182	Male	64	21	no
154	2015	165319	403-9733	Female	66	68	no
...
1884	2015	950907	412-2520	Male	69	73	no
1887	2015	952646	332-5521	Male	37	122	no
1926	2015	968500	345-1524	Male	36	101	no
1931	2015	969732	354-3237	Female	34	98	no
1933	2015	970012	368-3078	Male	47	102	no

356 rows × 16 columns



In []:

NONOUTLIERS DATA

```
In [71]: tele_df=pd.read_csv(r"C:\Users\HP\Desktop\Naresh IT\Data Set Analysis\Telecom_Ch
Nonoutliers_data=pd.DataFrame()
for i in num_clm_2:
```

```

q1=np.percentile(tele_df[i],q=0.25)
q2=np.percentile(tele_df[i],q=0.50)
q3=np.percentile(tele_df[i],q=0.75)

iqr=q3-q1

lb=q1-1.5*iqr
ub=q3+1.5*iqr

con1=tele_df[i]>lb
con2=tele_df[i]<ub
con3=con1 & con2
#Append outliers for this column
Nonoutliers_data=pd.concat([Nonoutliers_data,tele_df[con3]])

```

In [72]: Nonoutliers_data

Out[72]:

	year	customer_id	phone_no	gender	age	no_of_days_subscribed	multi_screen
0	2015	100198	409-8743	Female	36	62	no
1	2015	100643	340-5930	Female	39	149	no
2	2015	100756	372-3750	Female	65	126	no
3	2015	101595	331-4902	Female	24	131	no
4	2015	101653	351-8398	Female	40	191	no
...
1956	2015	979406	366-7360	Male	47	129	no
1963	2015	981937	380-5873	Male	37	46	no
1973	2015	986715	353-7461	Male	37	128	no
1974	2015	987122	364-8981	Female	26	54	no
1983	2015	989511	346-4216	Female	43	154	no

1291 rows × 16 columns

In []:

COMPARISON BOTH OUTLIERS DATA AND AFTER FILLING OUTLIERS DATA USING BOXPLOT

Outliers

In [73]:

```

tele_df=pd.read_csv("telecom_churn_data.csv")
num_clm=tele_df.select_dtypes(exclude="object").columns
num_clm

```

```
Out[73]: Index(['year', 'customer_id', 'age', 'no_of_days_subscribed',
       'weekly_mins_watched', 'minimum_daily_mins', 'maximum_daily_mins',
       'weekly_max_night_mins', 'videos_watched', 'maximum_days_inactive',
       'customer_support_calls', 'churn'],
      dtype='object')
```

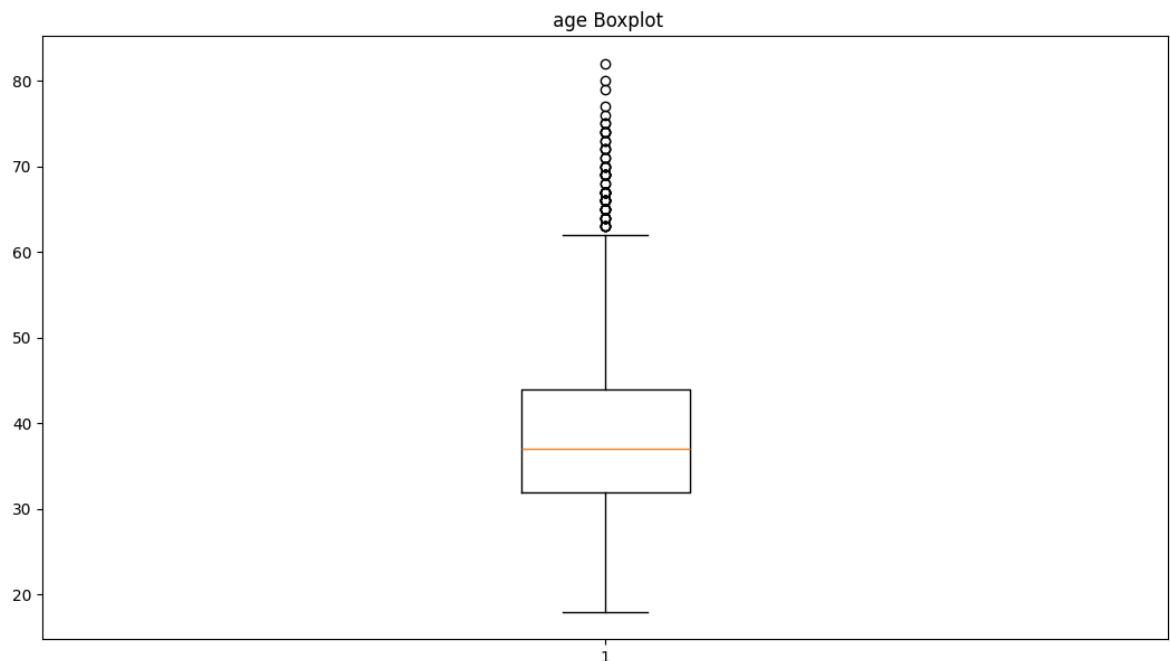
```
In [74]: num_clm_2=num_clm.drop("churn")
```

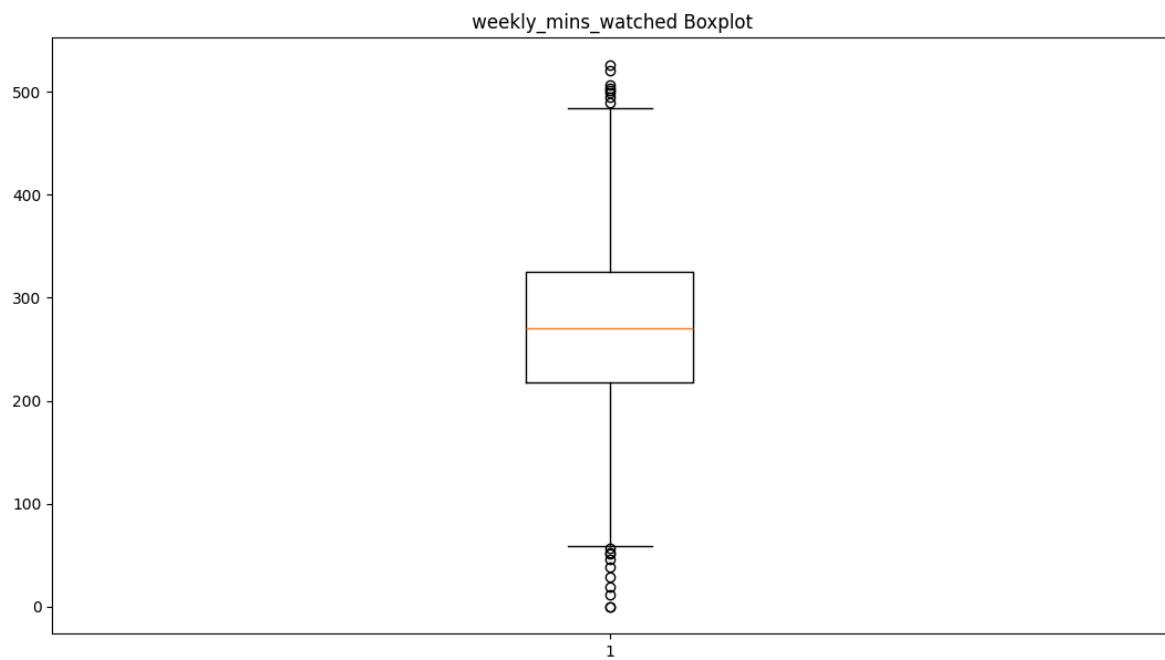
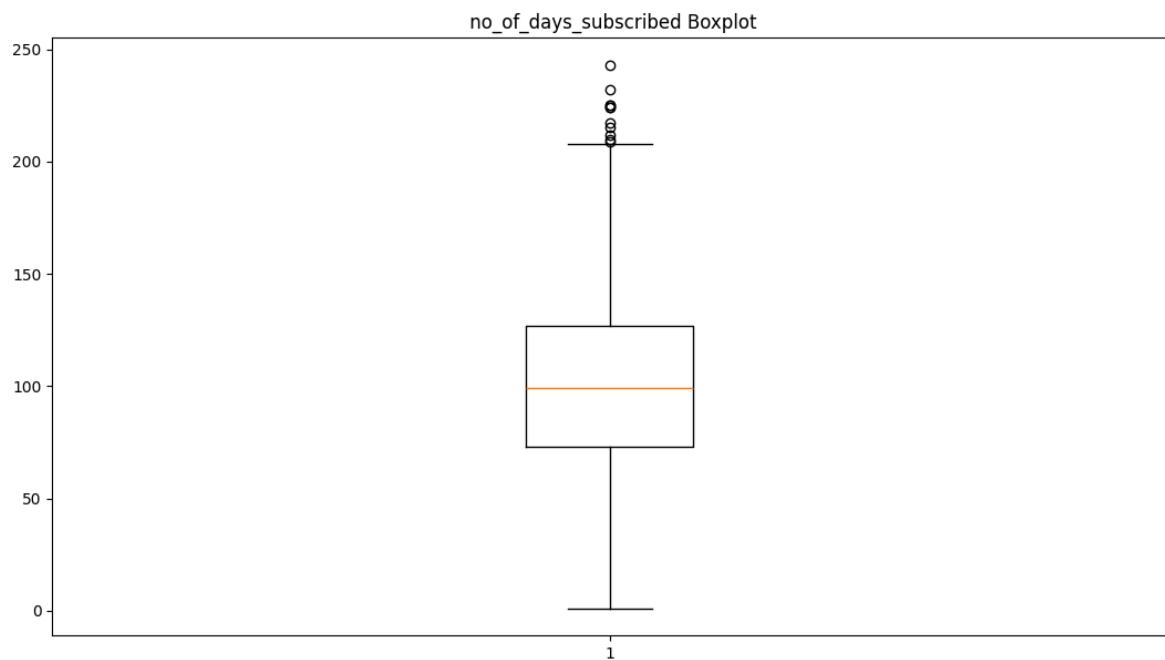
```
In [75]: num_clm_2
```

```
Out[75]: Index(['year', 'customer_id', 'age', 'no_of_days_subscribed',
       'weekly_mins_watched', 'minimum_daily_mins', 'maximum_daily_mins',
       'weekly_max_night_mins', 'videos_watched', 'maximum_days_inactive',
       'customer_support_calls'],
      dtype='object')
```

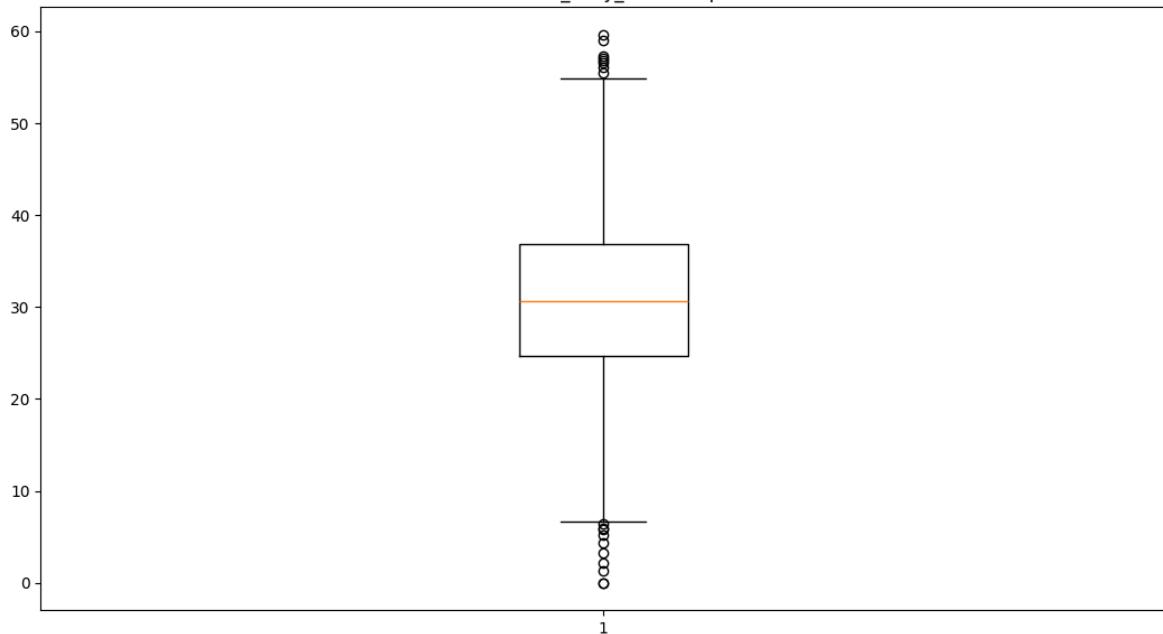
```
In [76]: tele_df=pd.read_csv("telecom_churn_data.csv")
for i in num_clm_2[2:]:
    plt.figure(figsize=(13,7))

    plt.boxplot(tele_df[i])
    plt.title(f" {i} Boxplot")
```

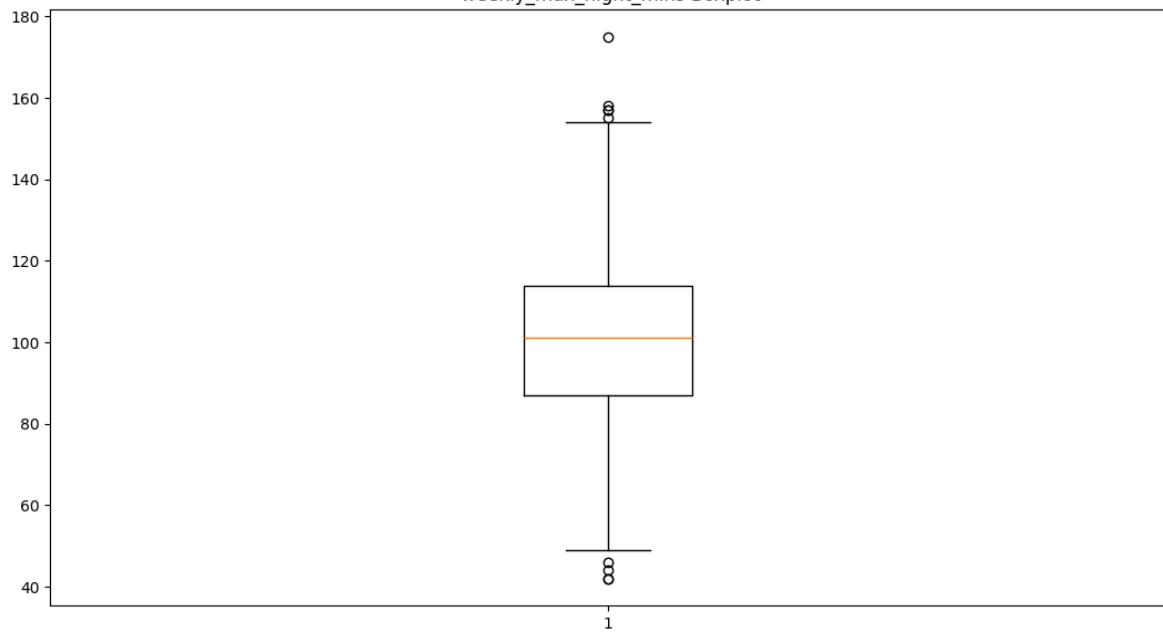




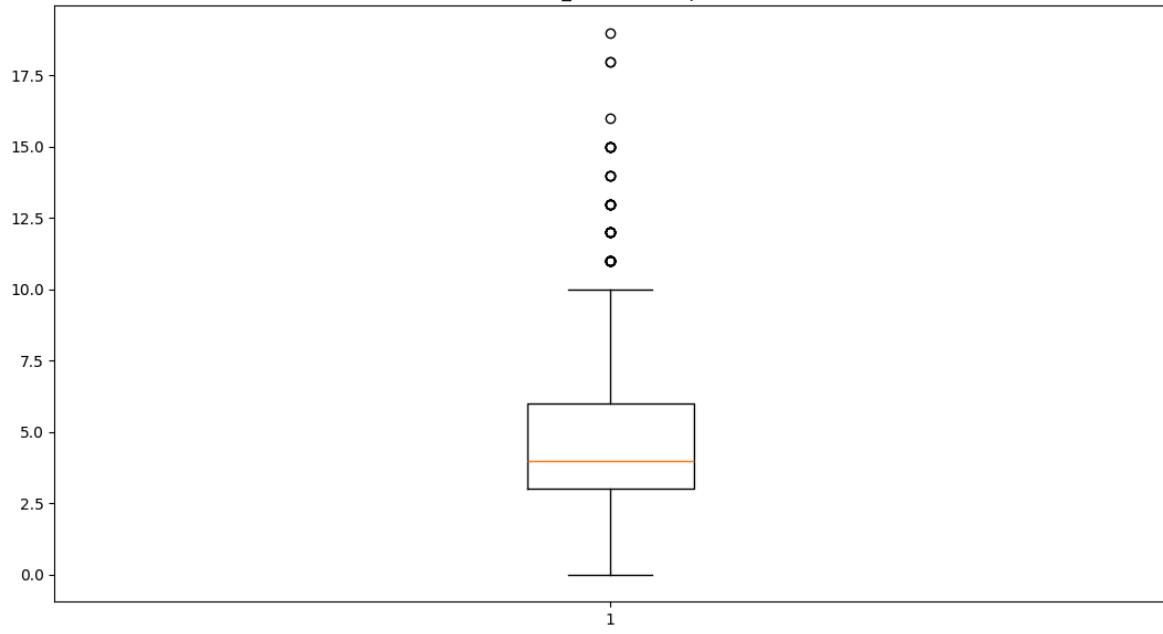
maximum_daily_mins Boxplot

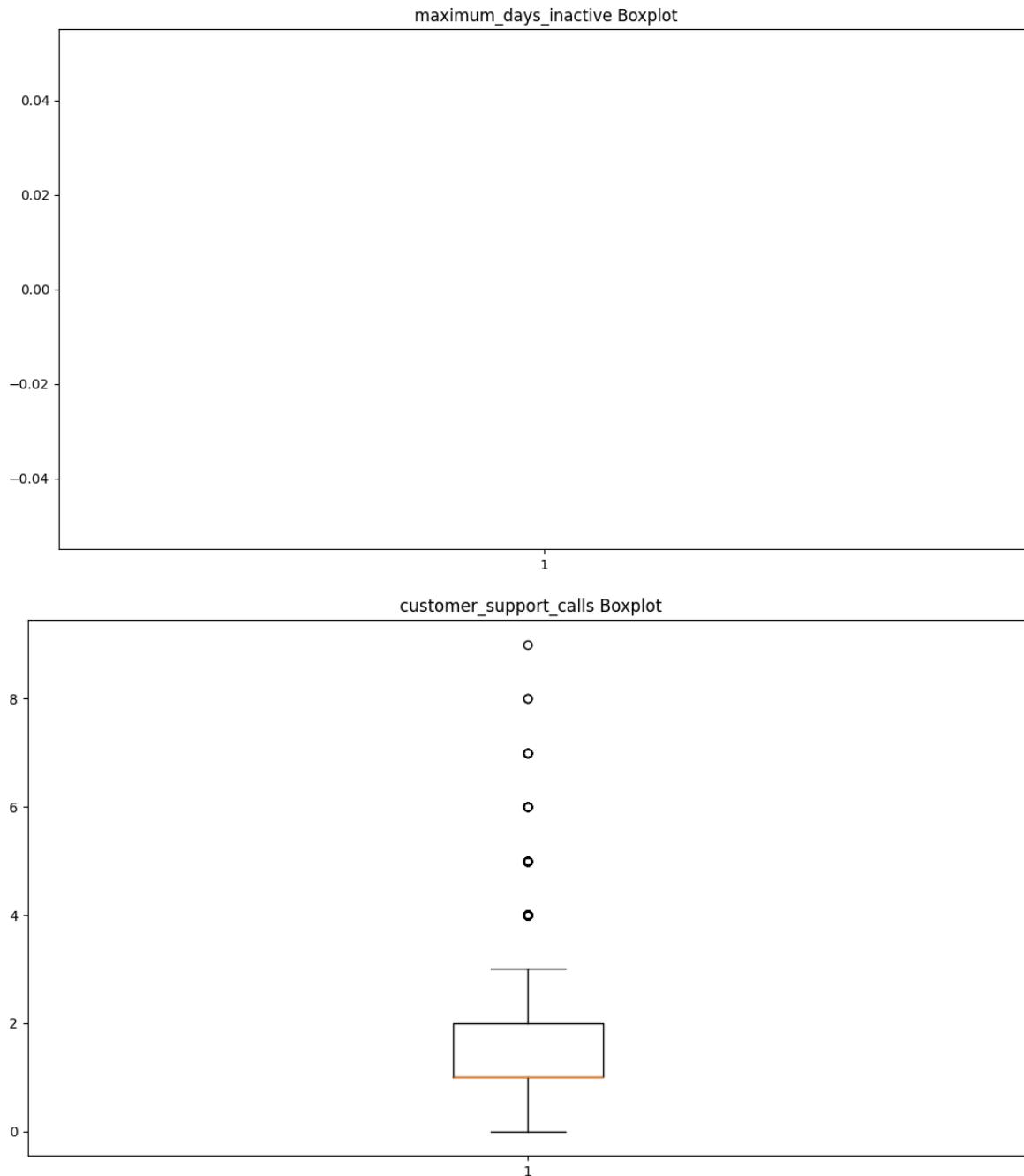


weekly_max_night_mins Boxplot



videos_watched Boxplot





In []:

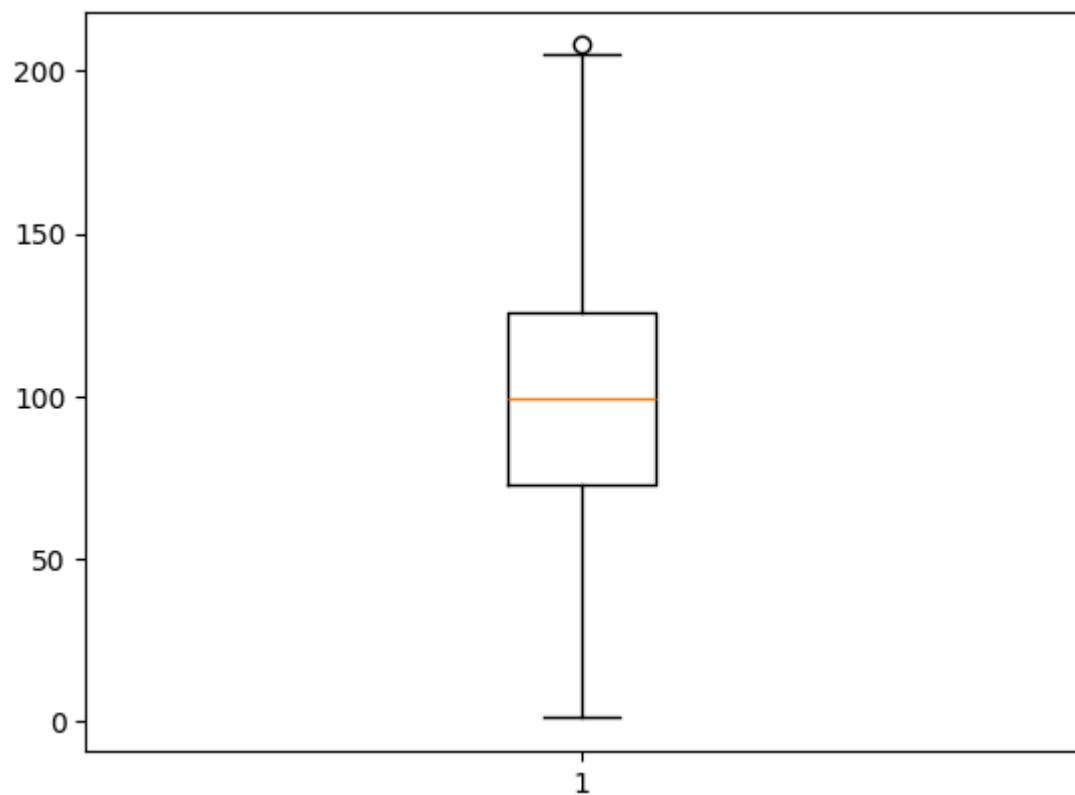
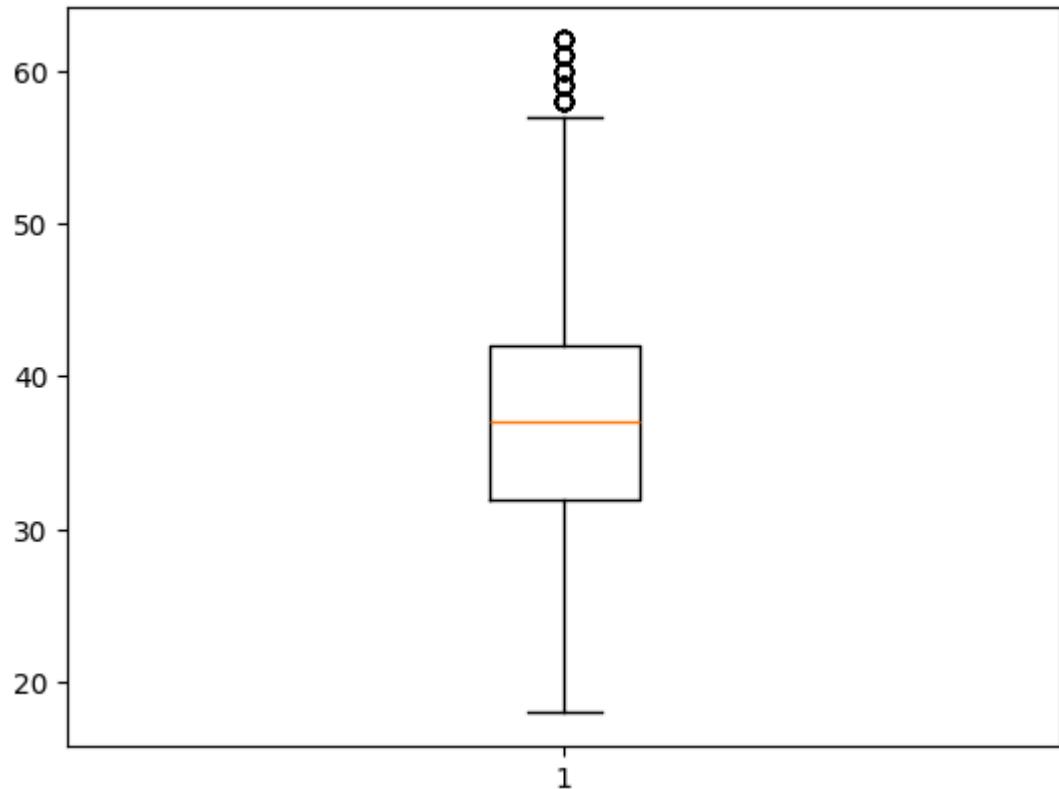
FILLING OUTLIERS DATA

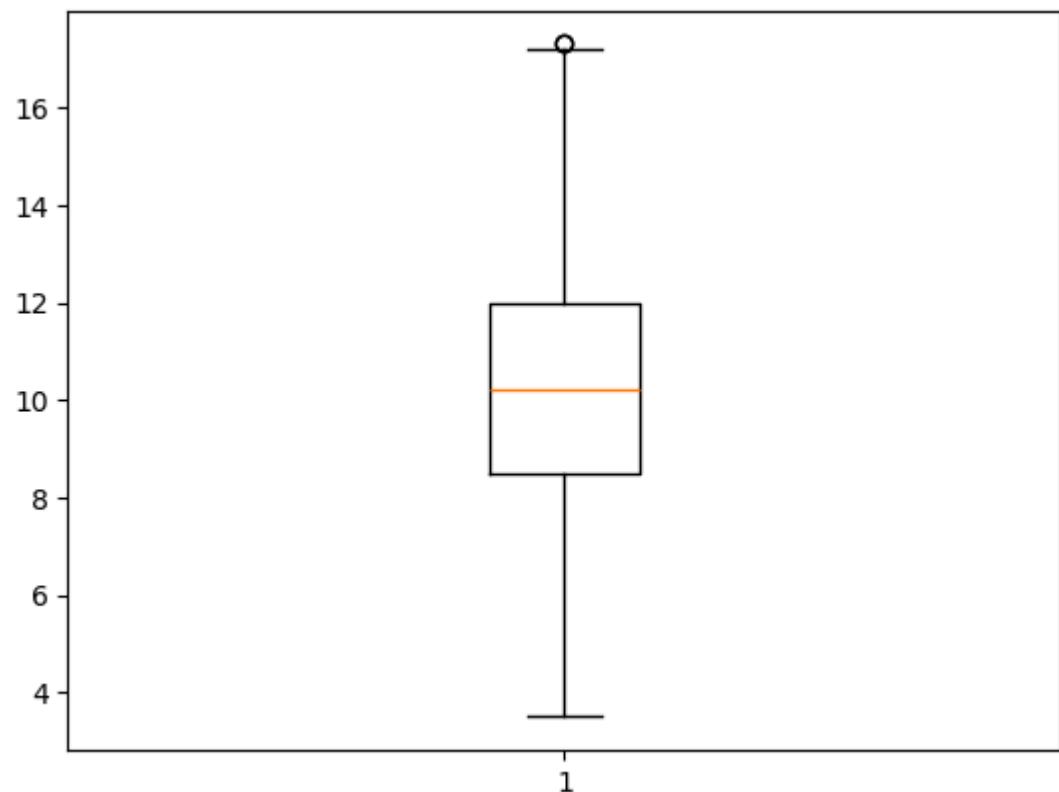
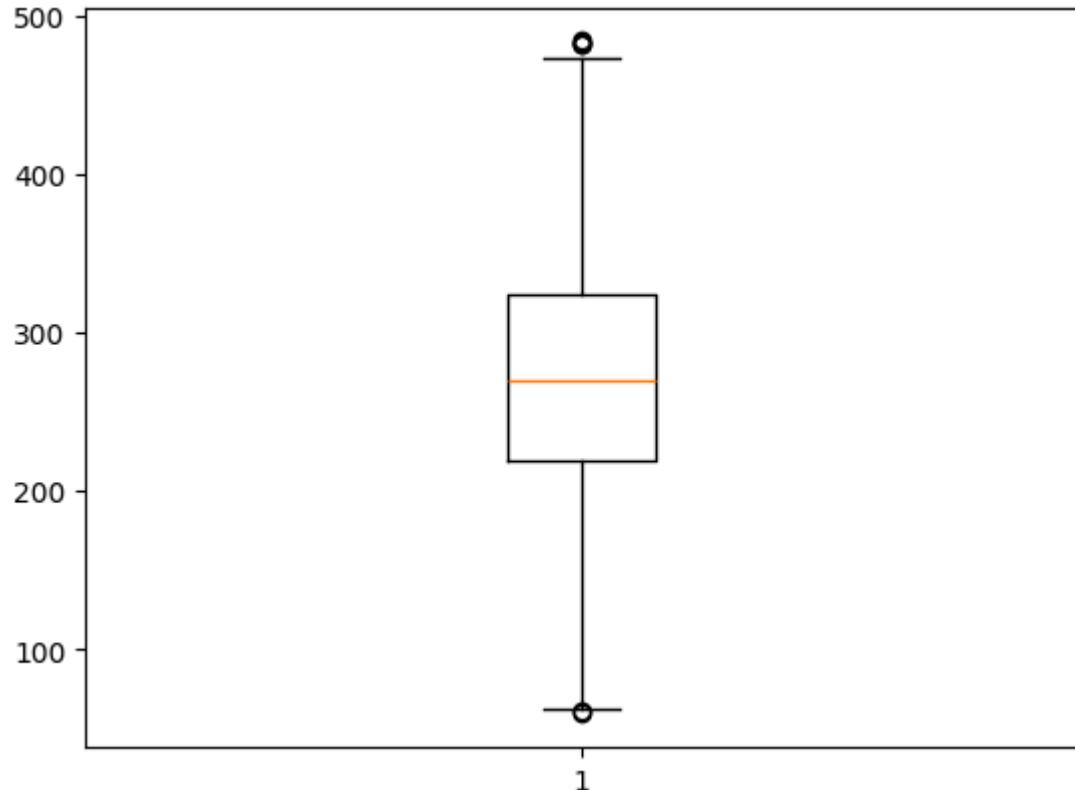
```
In [77]: for i in num_clm_2[2:]:  
    q1=np.quantile(tele_df[i],q=0.25)  
    q2=np.quantile(tele_df[i],q=0.50)  
    q3=np.quantile(tele_df[i],q=0.75)  
  
    iqr=q3-q1  
  
    lb=q1-1.5*iqr  
    ub=q3+1.5*iqr  
  
    con1=tele_df[i]<lb  
    con2=tele_df[i]>ub  
    con3=con1 | con2
```

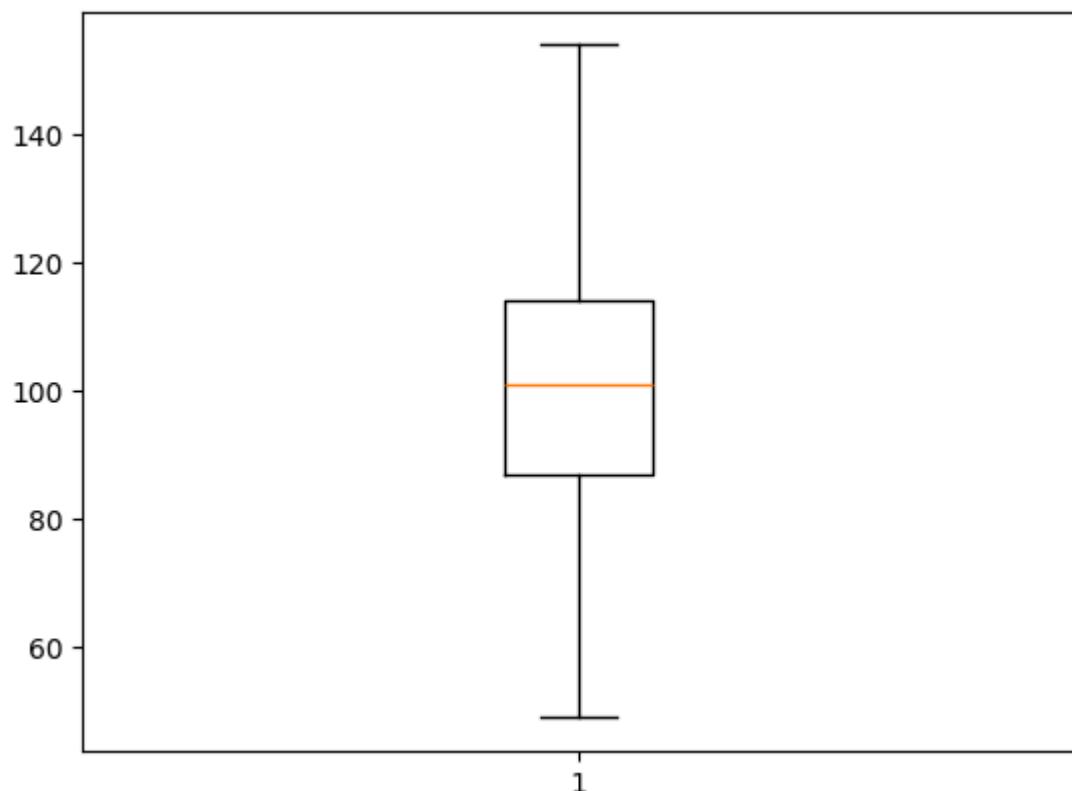
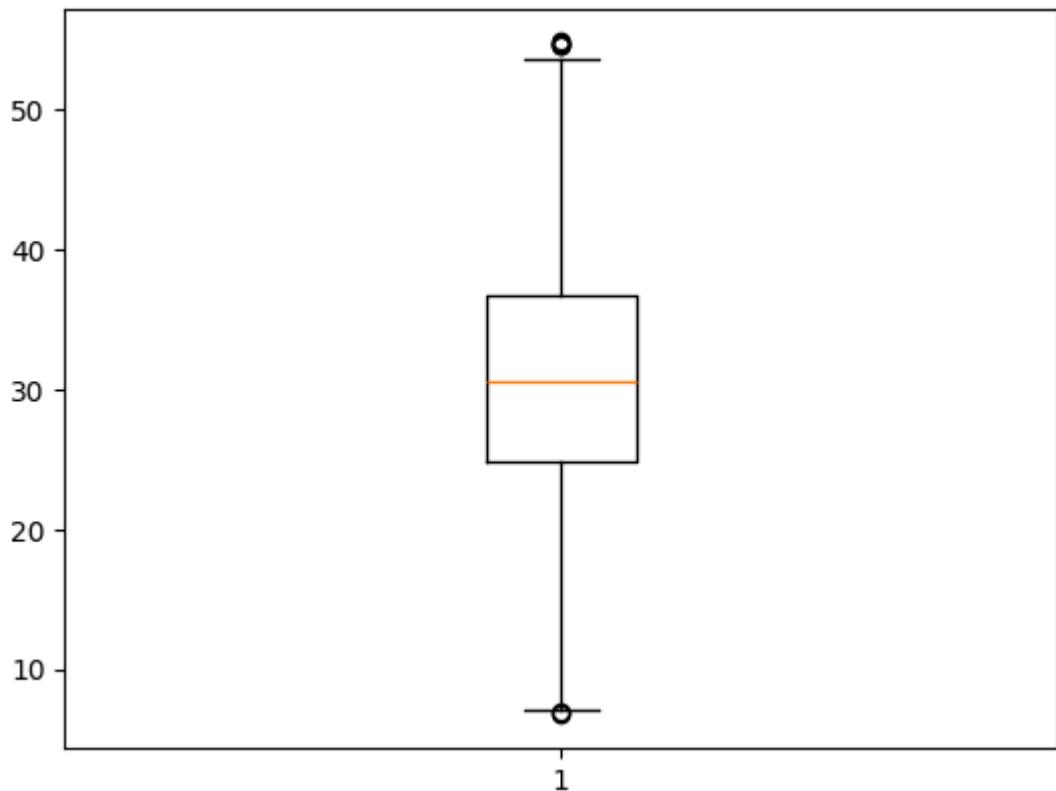
```
true=np.median(tele_df[i])
false=tele_df[i]

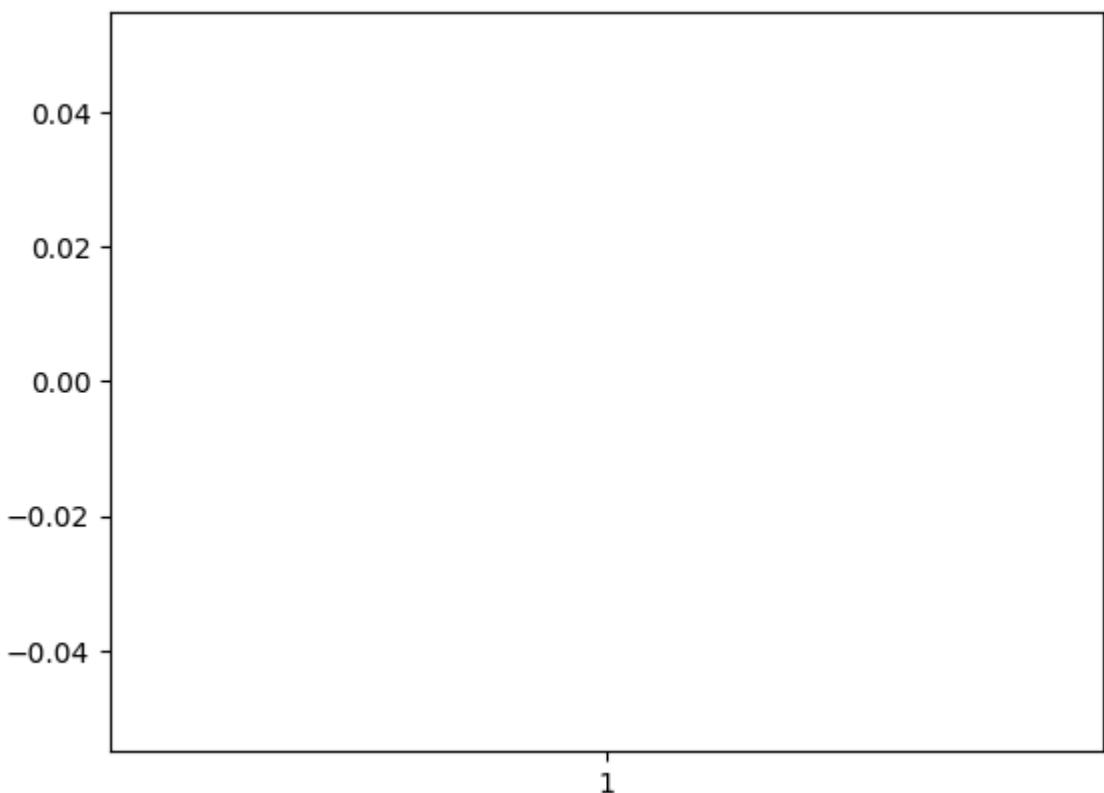
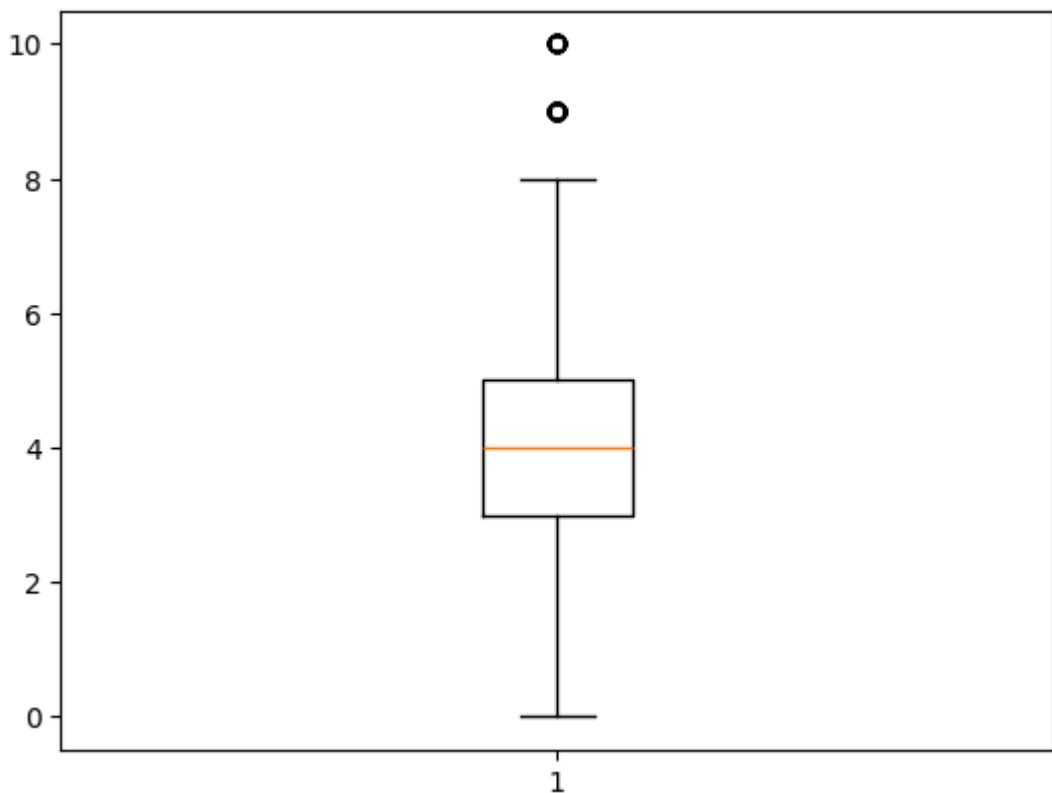
v=np.where(con3,true,false)
tele_df[i]=v

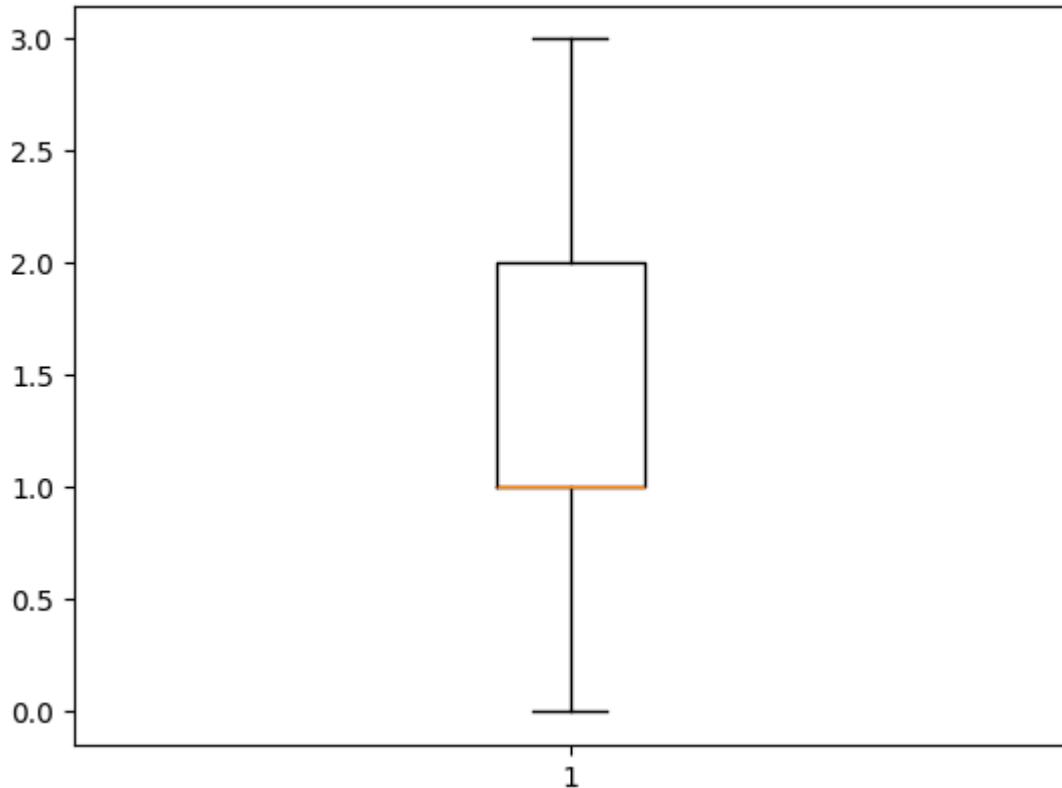
plt.boxplot(v)
plt.show()
```











In []:

COMPARISON BOTH

```
In [152]: tele_df=pd.read_csv("telecom_churn_data.csv")
for i in num_clm_2[2:]:
    # with outliers
    plt.figure(figsize=(13,7))
    plt.subplot(1,2,1)
    plt.boxplot(tele_df[i])
    plt.title(f" {i} Boxplot")

    # fill the outliers
    plt.subplot(1,2,2)
    q1=np.quantile(tele_df[i],q=0.25)
    q2=np.quantile(tele_df[i],q=0.50)
    q3=np.quantile(tele_df[i],q=0.75)

    iqr=q3-q1
    lb=q1-1.5*iqr
    ub=q3+1.5*iqr

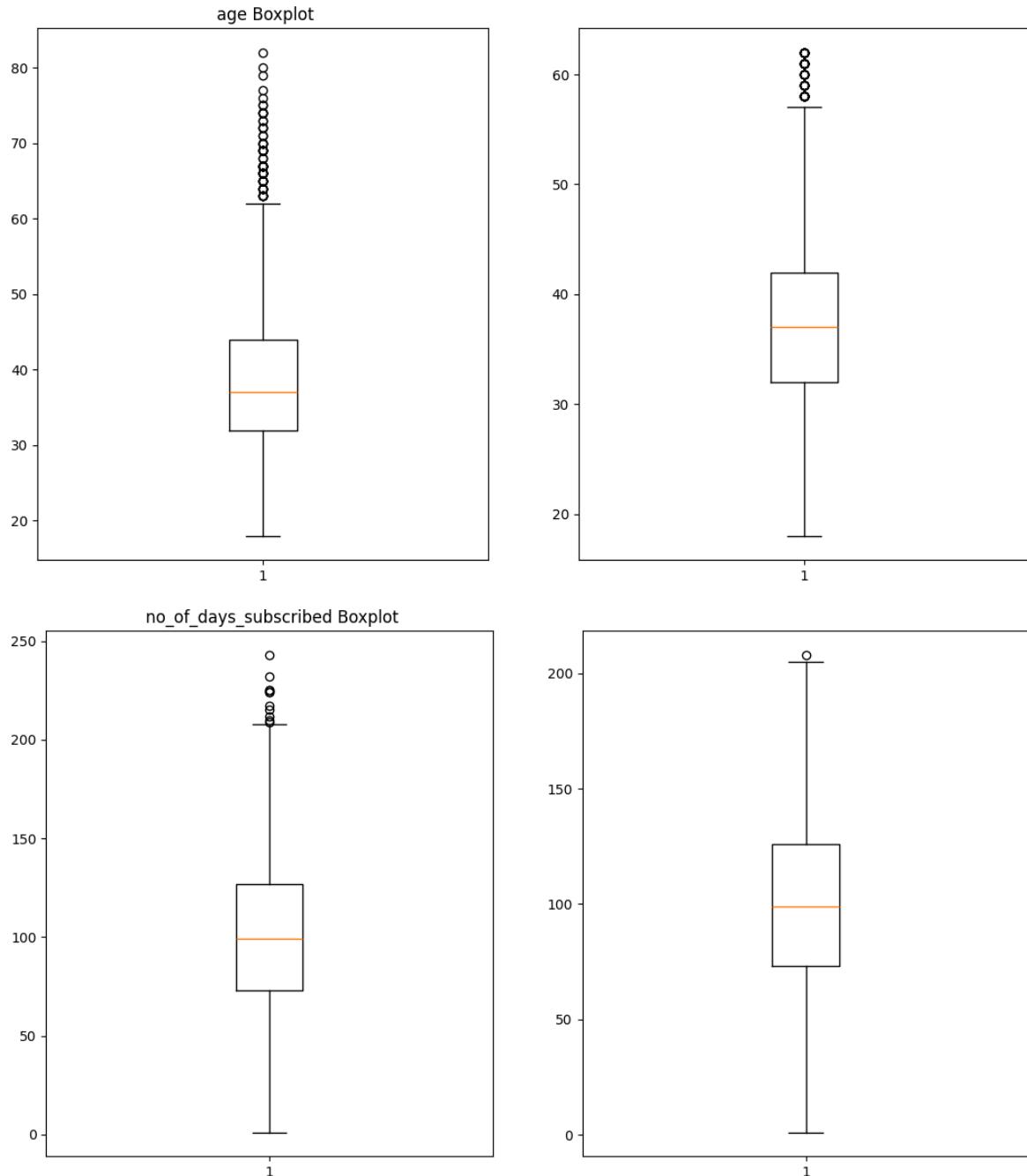
    con1=tele_df[i]<lb
    con2=tele_df[i]>ub
    con3=con1 | con2

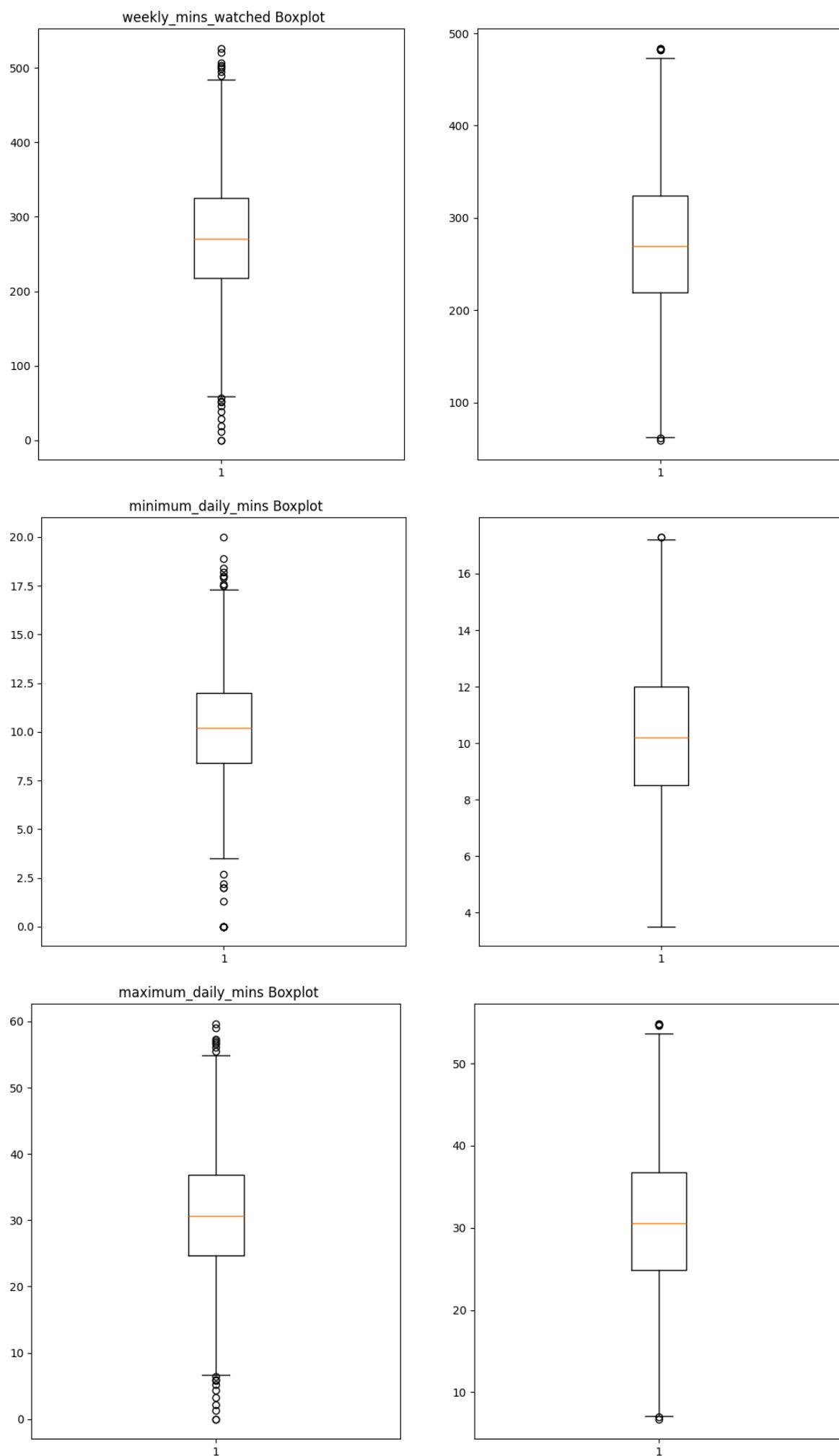
    true=np.median(tele_df[i])
    false=tele_df[i]

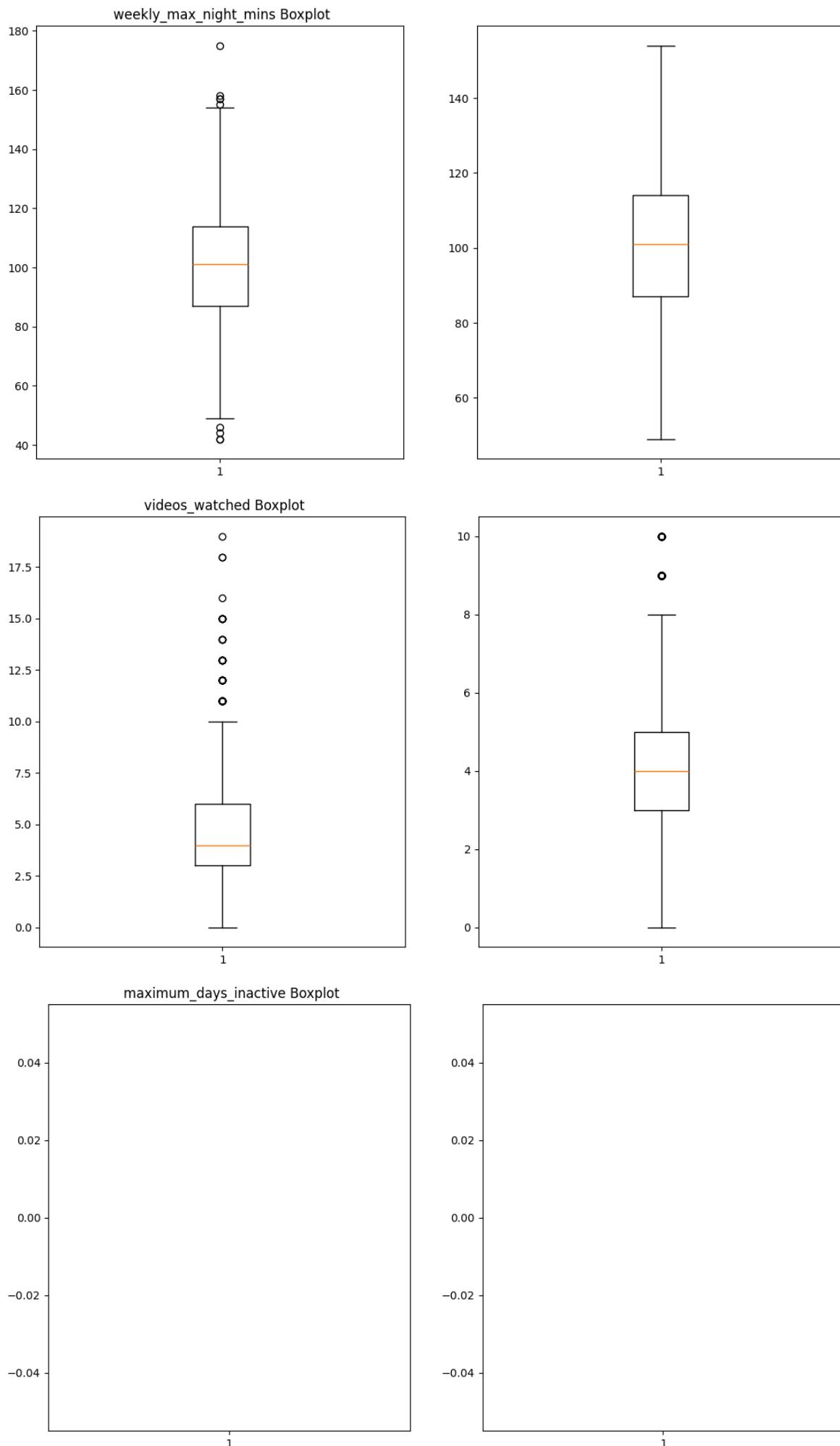
    v=np.where(con3,true,false)
```

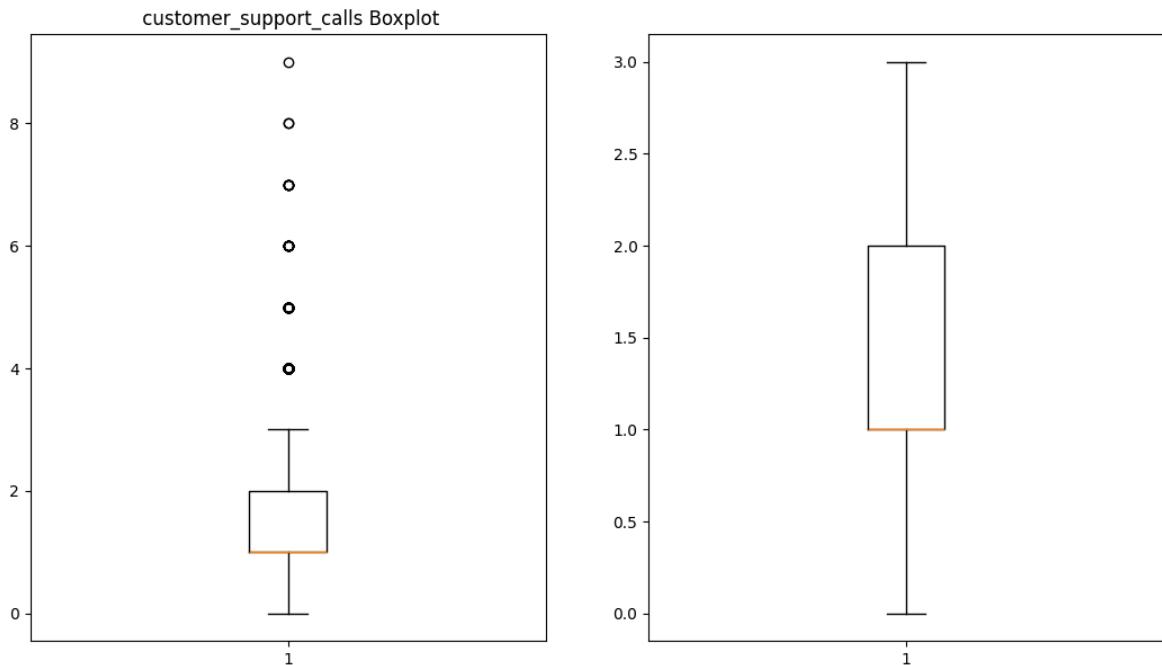
```
tele_df[i]=v
```

```
plt.boxplot(v)
plt.show()
```









In []:

Bivariate and Multi Variate analysis

In [159...]: cat_clm

Out[159...]: Index(['phone_no', 'gender', 'multi_screen', 'mail_subscribed'], dtype='object')

In [171...]:

```
index=tele_df['mail_subscribed']
con_1=tele_df['gender']
con_2=tele_df['multi_screen']
con_3=[con_1,con_2]
data=pd.crosstab(index,con_3)
data
```

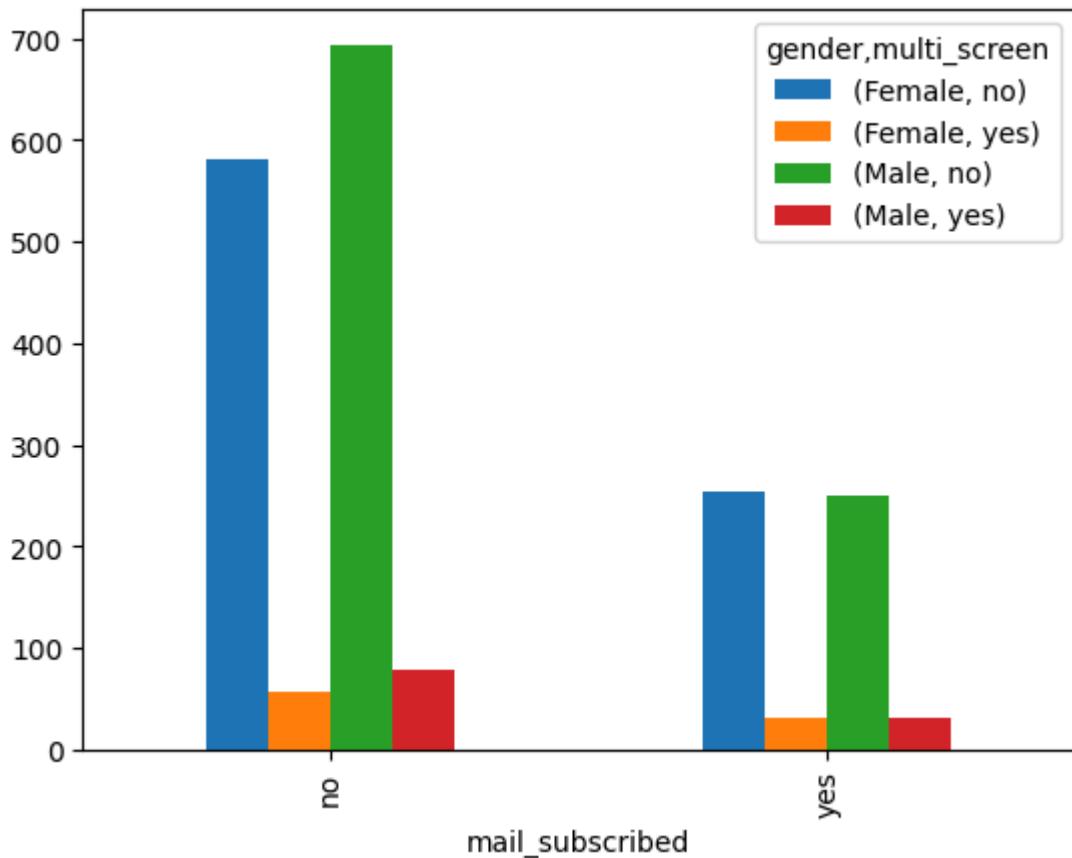
Out[171...]:

	gender		Male	
	Female	Male		
multi_screen	no	yes	no	yes
mail_subscribed				
no	582	57	694	79
yes	253	31	249	31

In [175...]:

```
plt.figure(figsize=(10,5))
data.plot(kind="bar")
plt.show()
```

<Figure size 1000x500 with 0 Axes>



In []:

Co relation between numerical column

In [183...]

```
co_dataframe=tele_df.corr(numeric_only=True)
co_dataframe
```

Out[183...]

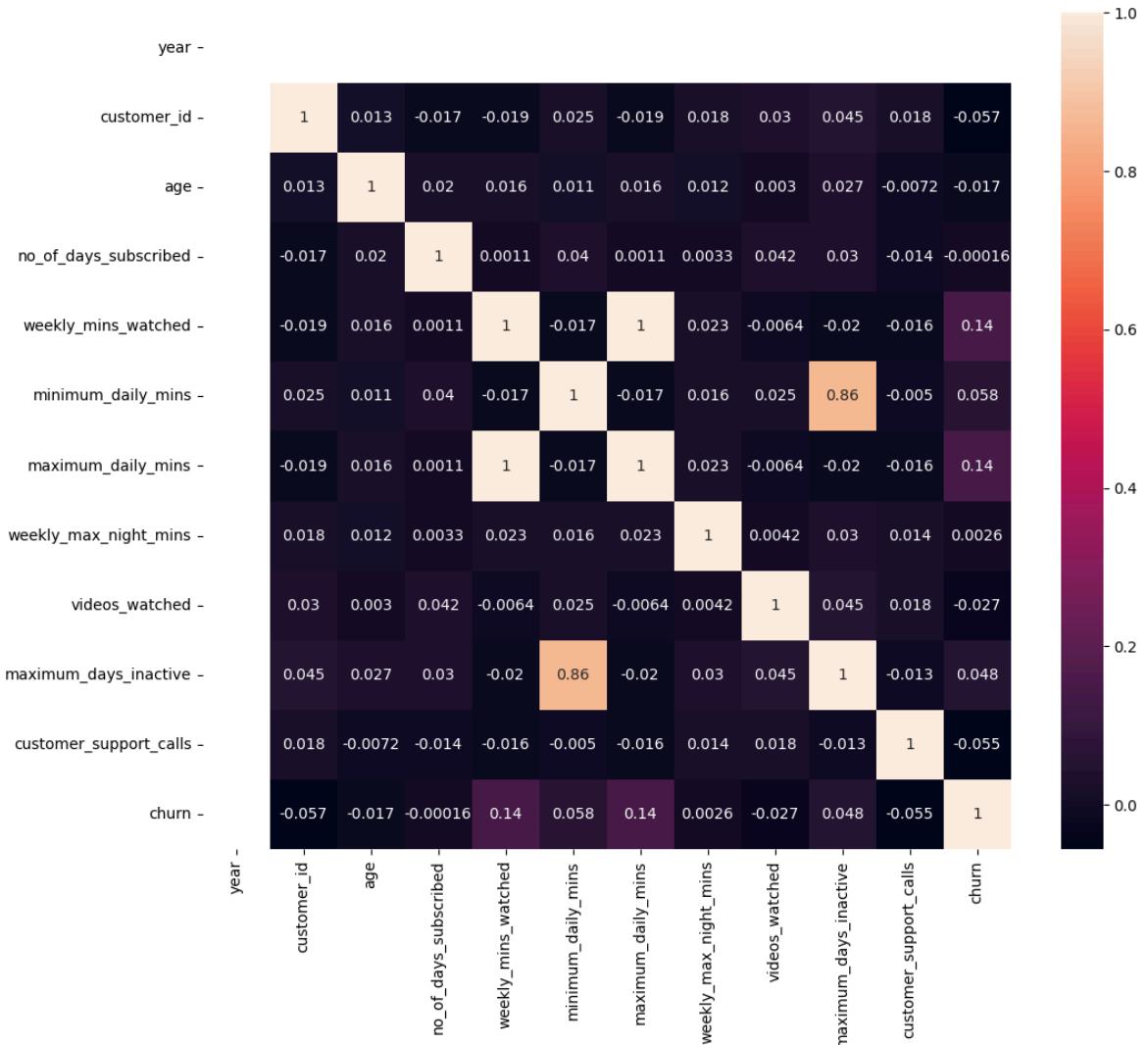
	year	customer_id	age	no_of_days_subscribed	weekly_mi
year	NaN	NaN	NaN	NaN	NaN
customer_id	NaN	1.000000	0.012506	-0.017422	
age	NaN	0.012506	1.000000	0.020416	
no_of_days_subscribed	NaN	-0.017422	0.020416	1.000000	
weekly_mins_watched	NaN	-0.018533	0.016250	0.001131	
minimum_daily_mins	NaN	0.024945	0.011062	0.039527	
maximum_daily_mins	NaN	-0.018538	0.016265	0.001123	
weekly_max_night_mins	NaN	0.017802	0.012450	0.003276	
videos_watched	NaN	0.029513	0.002979	0.041836	
maximum_days_inactive	NaN	0.045145	0.026899	0.029951	
customer_support_calls	NaN	0.018197	-0.007193	-0.014171	
churn	NaN	-0.056777	-0.017213	-0.000157	

In []:

HeatMaps

In [197...]

```
plt.figure(figsize=(12,10))
sns.heatmap(co_dataframe, annot=True)
plt.show()
```



In []:

Convert categorical to numerical data (Encoding)

LabelEncoder

In [209...]

```
tele_df=pd.read_csv("telecom_churn_data.csv")
for i in cat_clm:
    from sklearn.preprocessing import LabelEncoder
    label=LabelEncoder()
    tele_df[i]=label.fit_transform(tele_df[i])
tele_df
```

Out[209...]

	year	customer_id	phone_no	gender	age	no_of_days_subscribed	multi_screen
0	2015	100198	1754	0	36	62	0
1	2015	100643	299	0	39	149	0
2	2015	100756	957	0	65	126	0
3	2015	101595	75	0	24	131	0
4	2015	101653	517	0	40	191	0
...
1995	2015	997132	1224	0	54	75	0
1996	2015	998086	1188	1	45	127	0
1997	2015	998474	553	2	53	94	0
1998	2015	998934	706	1	40	94	0
1999	2015	999961	1834	1	37	73	0

2000 rows × 16 columns

--	--

In []:

Onehot Encoder

In [212...]

`pd.get_dummies(tele_df,dtype='int')`

Out[212...]

	year	customer_id	phone_no	gender	age	no_of_days_subscribed	multi_screen
0	2015	100198	1754	0	36	62	0
1	2015	100643	299	0	39	149	0
2	2015	100756	957	0	65	126	0
3	2015	101595	75	0	24	131	0
4	2015	101653	517	0	40	191	0
...
1995	2015	997132	1224	0	54	75	0
1996	2015	998086	1188	1	45	127	0
1997	2015	998474	553	2	53	94	0
1998	2015	998934	706	1	40	94	0
1999	2015	999961	1834	1	37	73	0

2000 rows × 16 columns

--	--

In []: