```
In [1]: #import all packages
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.decomposition import PCA
import seaborn as sns

In [2]: #Importing the medical data file
df = pd.read_csv(r"C:\Users\arjun\OneDrive\Desktop\WGU\D206\medical_raw_data.csv")

In [3]: #data profiling to see if data was imported correctly
df.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 53 columns):

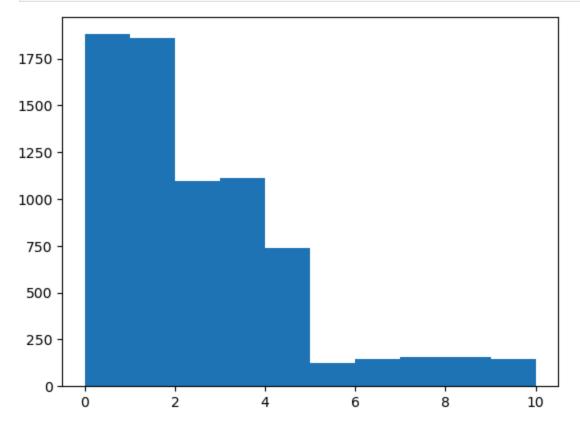
Data	columns (total 53 c	olumns):	
#	Column	Non-Null Count	Dtype
0	Unnamed: 0	10000 non-null	int64
1	CaseOrder	10000 non-null	int64
2	Customer_id	10000 non-null	object
3	Interaction	10000 non-null	object
4	UID	10000 non-null	object
5			_
	City	10000 non-null	object
6	State	10000 non-null	object
7	County	10000 non-null	object
8	Zip	10000 non-null	int64
9	Lat	10000 non-null	
10	Lng	10000 non-null	
11	Population	10000 non-null	int64
12	Area	10000 non-null	object
13	Timezone	10000 non-null	object
14	Job	10000 non-null	object
15	Children	7412 non-null	float64
16	Age	7586 non-null	float64
17	Education	10000 non-null	object
18	Employment	10000 non-null	object
19	Income	7536 non-null	float64
20	Marital	10000 non-null	object
21	Gender	10000 non-null	object
22	ReAdmis	10000 non-null	object
23	VitD_levels	10000 non-null	float64
24	Doc_visits	10000 non-null	int64
25	Full_meals_eaten	10000 non-null	int64
26	VitD_supp	10000 non-null	int64
27	Soft_drink	7533 non-null	object
28	Initial_admin	10000 non-null	object
29	HighBlood	10000 non-null	object
30	Stroke	10000 non-null	object
31	Complication_risk	10000 non-null	object
32	Overweight	9018 non-null	float64
33	Arthritis	10000 non-null	object
34	Diabetes	10000 non-null	object
35	Hyperlipidemia	10000 non-null	object
36	BackPain	10000 non-null	•
37	Anxiety	9016 non-null	object
	•		float64
38	Allergic_rhinitis	10000 non-null	object
39	Reflux_esophagitis	10000 non-null	object
40	Asthma	10000 non-null	object
41	Services	10000 non-null	object
42	Initial_days	8944 non-null	float64
43	TotalCharge	10000 non-null	float64
44	Additional_charges	10000 non-null	float64
45	Item1	10000 non-null	int64
46	Item2	10000 non-null	int64
47	Item3	10000 non-null	int64
48	Item4	10000 non-null	int64
49	Item5	10000 non-null	int64
50	Item6	10000 non-null	int64

```
51 Item7
                                10000 non-null int64
        52 Item8
                                10000 non-null int64
       dtypes: float64(11), int64(15), object(27)
       memory usage: 4.0+ MB
In [4]: #steps to removing duplicates
        # 1 Check to see how many rows are duplicated. I will check using the Customer_id,
        # 2 Check to see how many are duplicated for each
        # 3 Removal of the duplicates from the dataset
In [5]: # 1 Check to see how many rows are duplicated. I will check using the CaseOrder, Cu
        #Duplicates in CaseOrder
        df.CaseOrder.duplicated()
Out[5]: 0
                False
                False
        1
        2
                False
        3
                False
                False
                . . .
        9995
                False
              False
        9996
        9997
               False
        9998
                False
        9999
                False
        Name: CaseOrder, Length: 10000, dtype: bool
In [6]: #Duplicates in Customer_id
        df.Customer_id.duplicated()
Out[6]: 0
                False
                False
        1
        2
                False
        3
                False
                False
                . . .
        9995
              False
        9996
                False
        9997
               False
                False
        9998
        9999
                False
        Name: Customer_id, Length: 10000, dtype: bool
In [7]: #Duplicates in Interaction
        df.Interaction.duplicated()
```

```
Out[7]: 0
                 False
         1
                 False
         2
                 False
         3
                 False
                 False
                 . . .
         9995
                 False
         9996
               False
         9997
                False
         9998
                 False
         9999
                 False
         Name: Interaction, Length: 10000, dtype: bool
In [8]: #Duplicates in UID
         df.UID.duplicated()
Out[8]: 0
                 False
         1
                 False
         2
                 False
         3
                 False
                 False
                 . . .
         9995
                 False
         9996
               False
         9997
               False
         9998
                 False
         9999
                 False
         Name: UID, Length: 10000, dtype: bool
In [9]: # Steps to clean for missing data
         # 1 Detect which columns have missing data
         # 2 impute missing data
         # 3 Verify if the missing data has been corrected.
In [10]: #1 Detect which columns have missing data
         df.isnull().sum()
```

Out[10]:	Unnamed: 0	0
	CaseOrder	0
	Customer_id	0
	Interaction	0
	UID	0
	City	0
	State	0
	County	0
	Zip	0
	Lat	0
	Lng	0
	Population	0
	Area	0
	Timezone	0
	Job	0
	Children	2588
	Age	2414
	Education	0
	Employment	0
	Income	2464
	Marital	0
	Gender	0
	ReAdmis	0
	VitD_levels	0
	Doc_visits	0
	Full_meals_eaten	0
	VitD_supp	0
	Soft_drink	2467
	Initial_admin	0
	HighBlood	0
	Stroke	0
	Complication_risk	0
	Overweight	982
	Arthritis	0
	Diabetes	0
	Hyperlipidemia	0
	BackPain	0
	Anxiety	984
	Allergic_rhinitis	0
	Reflux_esophagitis	0
	Asthma	0
	Services	0
	Initial_days	1056
	TotalCharge	0
	Additional_charges	0
	Item1	0
	Item2	0
	Item3	0
	Item4	0
	Item5	0
	Item6	0
	Item7	0
	Item8	0
	dtype: int64	

```
In [11]: #impute data into Children variable
  plt.hist(df['Children'])
  plt.show()
```

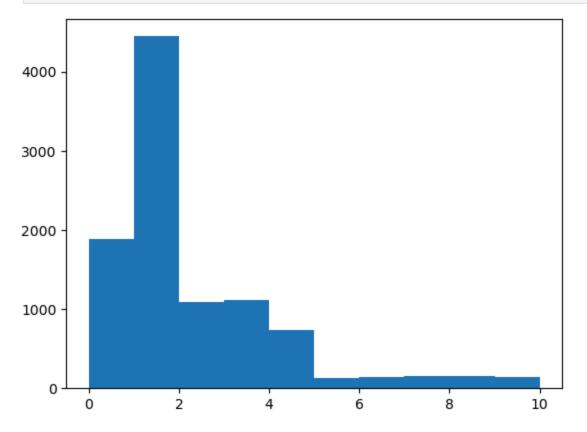


```
In [12]: #impute using median
df['Children'].fillna(df['Children'].median(), inplace= True)
```

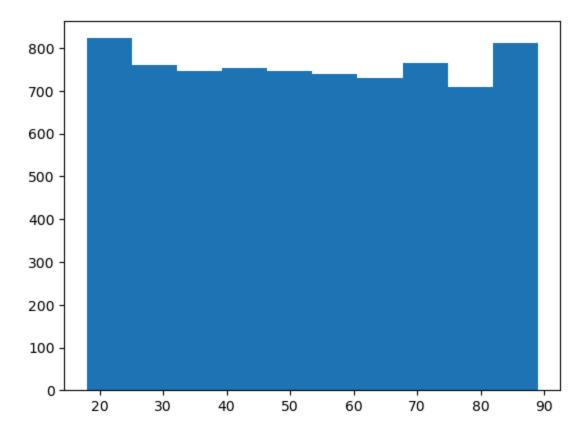
In [13]: df.isnull().sum()

Out[13]:	Unnamed: 0	0
	CaseOrder	0
	Customer_id	0
	Interaction	0
	UID	0
	City	0
	State	0
	County	0
	Zip	0
	Lat	0
	Lng	0
	Population	0
	Area	0
	Timezone	0
	Job	0
	Children	0
	Age	2414
	Education	0
	Employment	0
	Income	2464
	Marital	0
	Gender	0
	ReAdmis	0
	VitD_levels	0
	Doc_visits	0
	Full_meals_eaten	0
	VitD_supp	0
	Soft_drink	2467
	Initial_admin	0
	HighBlood	0
	Stroke	0
	Complication_risk	982
	Overweight Arthritis	962
	Diabetes	0
	Hyperlipidemia	0
	BackPain	0
	Anxiety	984
	Allergic_rhinitis	0
	Reflux_esophagitis	0
	Asthma	0
	Services	0
	Initial_days	1056
	TotalCharge	0
	Additional_charges	0
	Item1	0
	Item2	0
	Item3	0
	Item4	0
	Item5	0
	Item6	0
	Item7	0
	Item8	0
	dtype: int64	

```
In [14]: plt.hist(df['Children'])
   plt.show()
```



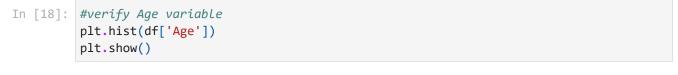
```
In [15]: #impute data into Age variable
  plt.hist(df['Age'])
  plt.show()
```

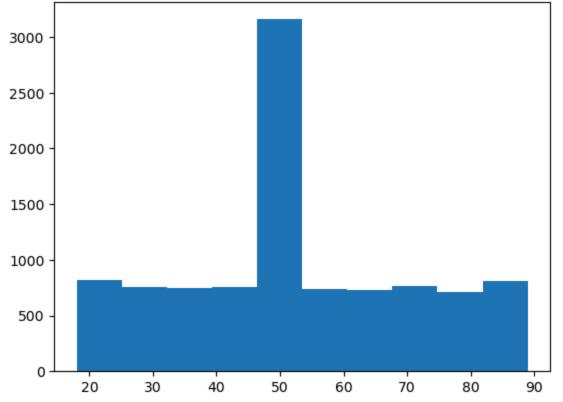


In [16]: #impute age using mean due to uniform distribution
 df['Age'].fillna(df['Age'].mean(), inplace= True)

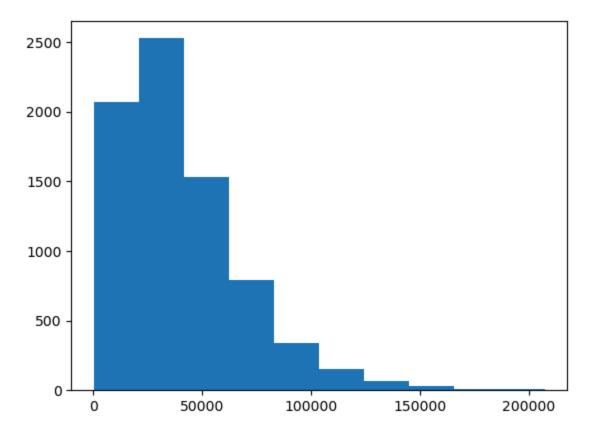
In [17]: df.isnull().sum()

Out[17]:	Unnamed: 0	0
	CaseOrder	0
	Customer_id	0
	Interaction	0
	UID	0
	City	0
	State	0
	County	0
	Zip	0
	Lat	0
	Lng	0
	Population	0
	Area	0
	Timezone	0
	Job	0
	Children	0
	Age	0
	Education	0
	Employment	0
	Income	2464
	Marital	2404
	Gender	0
	ReAdmis	0
	VitD_levels	0
	Doc_visits	0
	Full_meals_eaten	0
	VitD_supp	0
	Soft_drink	2467
	Initial_admin	0
	HighBlood	0
	Stroke	0
	Complication_risk	0
	Overweight	982
	Arthritis	0
	Diabetes	0
	Hyperlipidemia	0
	BackPain	0
	Anxiety	984
	Allergic_rhinitis	0
	Reflux_esophagitis	0
	Asthma	0
	Services	0
	<pre>Initial_days</pre>	1056
	TotalCharge	0
	Additional_charges	0
	Item1	0
	Item2	0
	Item3	0
	Item4	0
	Item5	0
	Item6	0
	Item7	0
	Item8	0
	dtype: int64	





```
In [19]: #impute data into the Income variable
    plt.hist(df['Income'])
    plt.show()
```

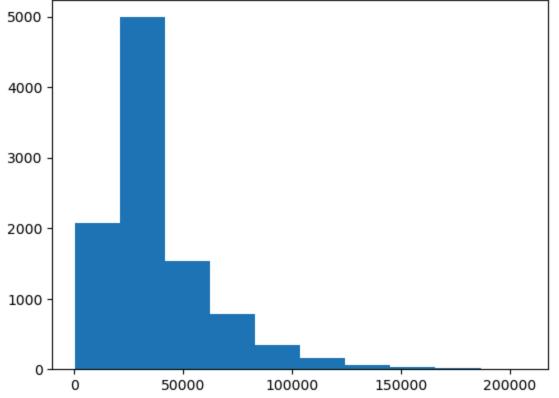


In [20]: #impute Income using median due to skewed right distribution
 df['Income'].fillna(df['Income'].median(),inplace= True)

In [21]: df.isnull().sum()

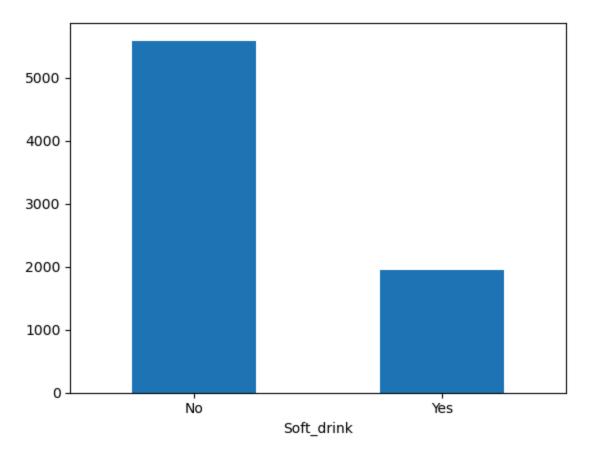
Out[21]:	Unnamed: 0	0
	CaseOrder	0
	Customer_id	0
	Interaction	0
	UID	0
	City	0
	State	0
	County	0
	•	
	Zip	0
	Lat	0
	Lng	0
	Population	0
	Area	0
	Timezone	0
	Job	0
	Children	0
	Age	0
	Education	0
	Employment	0
	Income	0
	Marital	0
	Gender	0
	ReAdmis	0
	VitD_levels	0
	Doc_visits	0
	Full_meals_eaten	0
	VitD_supp	0
		2467
	Soft_drink	
	Initial_admin	0
	HighBlood	0
	Stroke	0
	Complication_risk	0
	Overweight	982
	Arthritis	0
	Diabetes	0
	Hyperlipidemia	0
	BackPain	0
	Anxiety	984
	Allergic_rhinitis	0
	Reflux_esophagitis	0
	Asthma	0
	Services	0
	Initial_days	1056
	TotalCharge	0
	Additional_charges	0
	Item1	0
	Item2	0
	Item3	0
	Item4	0
	Item5	0
	Item6	0
	Item7	0
	Item8	0
		О
	dtype: int64	

```
In [22]: #verify Income variable
plt.hist(df['Income'])
plt.show()
```



```
In [23]: #impute data into the Soft_drink variable
df['Soft_drink'].value_counts().plot.bar(rot=0)
```

Out[23]: <Axes: xlabel='Soft_drink'>

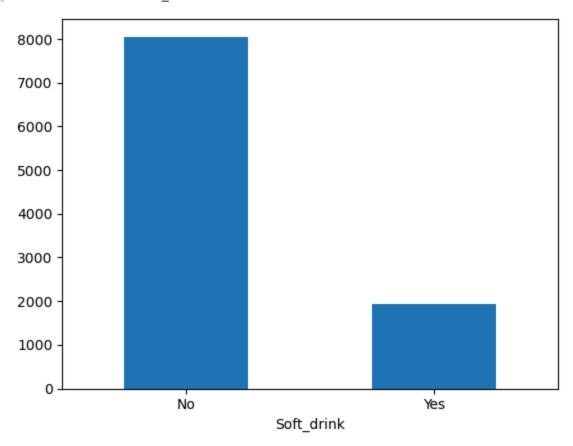


```
In [24]: #impute Soft_drink using mode due to categorical data
    df['Soft_drink'] = df['Soft_drink'].fillna(df['Soft_drink'].mode()[0])
In [25]: #verify Soft_drink variable
    df.isnull().sum()
```

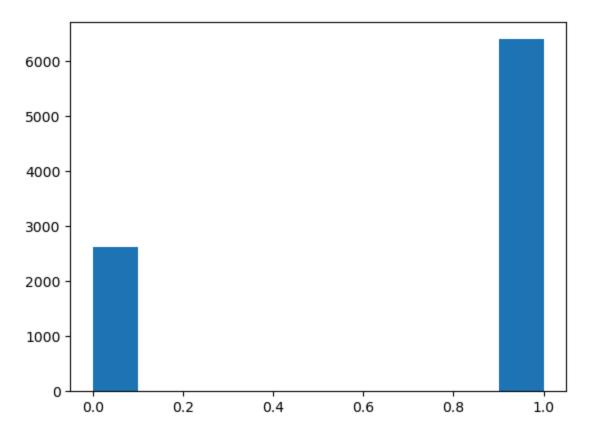
Out[25]:	Unnamed: 0	0
00.0[_0].	CaseOrder	0
	Customer_id	0
	Interaction	0
	UID	0
	City	0
	State	0
	County	0
	Zip	0
	Lat	0
	Lng	0
	Population	0
	•	
	Area	0
	Timezone	0
	Job	0
	Children	0
	Age	0
	Education	0
	Employment	0
	Income	0
	Marital	0
	Gender	0
	ReAdmis	0
	VitD_levels	0
	_	
	Doc_visits	0
	Full_meals_eaten	0
	VitD_supp	0
	Soft_drink	0
	Initial_admin	0
	HighBlood	0
	Stroke	0
	Complication_risk	0
	Overweight	982
	Arthritis	0
	Diabetes	0
	Hyperlipidemia	0
	BackPain	0
	Anxiety	984
	Allergic_rhinitis	0
	Reflux_esophagitis	0
	Asthma	
		0
	Services	0
	Initial_days	1056
	TotalCharge	0
	Additional_charges	0
	Item1	0
	Item2	0
	Item3	0
	Item4	0
	Item5	0
	Item6	0
	Item7	0
	Item8	0
	dtype: int64	J
	25,pc. 111007	

```
In [26]: df['Soft_drink'].value_counts().plot.bar(rot=0)
```

```
Out[26]: <Axes: xlabel='Soft_drink'>
```



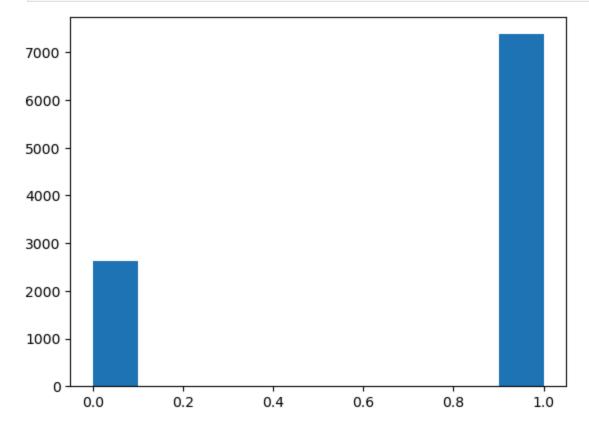
```
In [27]: #impute data into the Overweight variable
   plt.hist(df['Overweight'])
   plt.show()
```



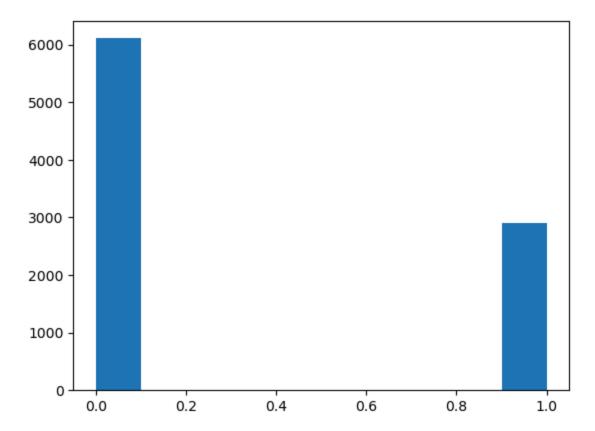
```
In [28]: #impute Overweight using mode due to categorical data
    df['Overweight'] = df['Overweight'].fillna(df['Overweight'].mode()[0])
In [29]: #verify Overweight variable
    df.isnull().sum()
```

Out[29]:	Unnamed: 0	0
	CaseOrder	0
	Customer_id	0
	Interaction	0
	UID	0
	City	0
	State	0
		0
	County	
	Zip	0
	Lat	0
	Lng	0
	Population	0
	Area	0
	Timezone	0
	Job	0
	Children	0
	Age	0
	Education	0
	Employment	0
	Income	0
	Marital	0
	Gender	0
	ReAdmis	0
	VitD_levels	0
	Doc_visits	0
	Full_meals_eaten	0
	VitD_supp	0
	Soft_drink	0
	Initial_admin	0
	HighBlood	0
	Stroke	0
	Complication_risk	0
	Overweight	0
	Arthritis	0
	Diabetes	0
	Hyperlipidemia	0
	BackPain	0
	Anxiety	984
	Allergic_rhinitis	0
	Reflux_esophagitis	0
	Asthma	0
	Services	0
	Initial_days	1056
	TotalCharge	0
	Additional_charges	0
	Item1	0
	Item2	0
	Item3	0
	Item4	0
	Item5	0
	Item6	0
	Item7	0
	Item8	0
	dtype: int64	

```
In [30]: #Verify imputation of data into the Overweight variable
   plt.hist(df['Overweight'])
   plt.show()
```



```
In [31]: #impute data into the Anxiety variable
   plt.hist(df['Anxiety'])
   plt.show()
```

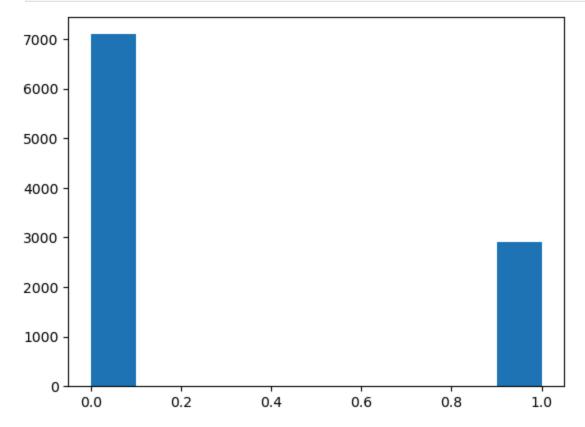


```
In [32]: #impute Anxiety using mode due to categorical data
df['Anxiety'] = df['Anxiety'].fillna(df['Anxiety'].mode()[0])
```

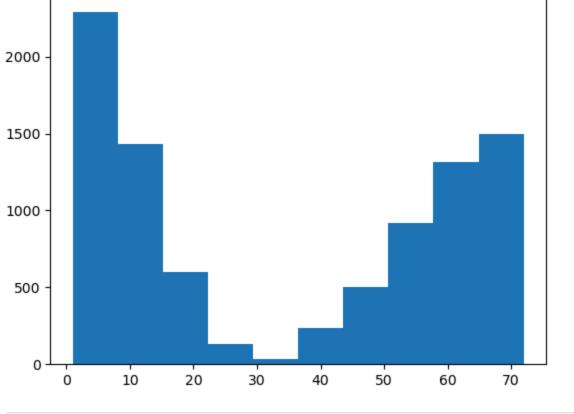
In [33]: #verify Anxiety variable
 df.isnull().sum()

Out[33]:	Unnamed: 0	0
	CaseOrder	0
	Customer_id	0
	Interaction	0
	UID	0
	City	0
	State	0
	County	0
	Zip	0
	Lat	0
		0
	Lng	0
	Population Area	0
	Timezone	0
	Job Children	0
		0
	Age	0
	Education	0
	Employment	0
	Income	0
	Marital	0
	Gender	0
	ReAdmis	0
	VitD_levels	0
	Doc_visits	0
	Full_meals_eaten	0
	VitD_supp	0
	Soft_drink	0
	Initial_admin	0
	HighBlood	0
	Stroke	0
	Complication_risk	0
	Overweight	0
	Arthritis	0
	Diabetes	0
	Hyperlipidemia	0
	BackPain	0
	Anxiety	0
	Allergic_rhinitis	0
	Reflux_esophagitis	0
	Asthma	0
	Services	0
	Initial_days	1056
	TotalCharge	0
	Additional_charges	0
	Item1	0
	Item2	0
	Item3	0
	Item4	0
	Item5	0
	Item6	0
	Item7	0
	Item8	0
	dtype: int64	J
	23,per 211co	

```
In [34]: #Verify imputation of data into the Anxiety variable
   plt.hist(df['Anxiety'])
   plt.show()
```



```
In [35]: #impute data into the Initial_days variable
   plt.hist(df['Initial_days'])
   plt.show()
```

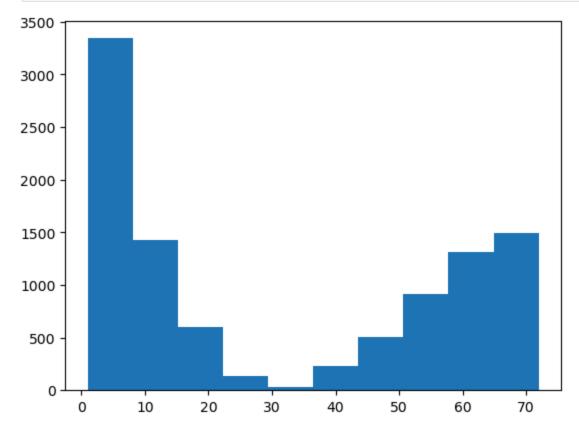


```
In [36]: #impute Initial_days using mode due to bimodal distribution
    df['Initial_days'] = df['Initial_days'].fillna(df['Initial_days'].mode()[0])
```

In [37]: #verify Initial_days variable
 df.isnull().sum()

Out[37]:	Unnamed: 0	0
	CaseOrder	0
	Customer_id	0
	Interaction	0
	UID	0
	City	0
	State	0
	County	0
	Zip	0
	Lat	0
	Lng	0
	Population	0
	Area	0
	Timezone	0
	Job	0
	Children	0
	Age	0
	Education	0
	Employment	0
	Income	0
	Marital	0
	Gender	0
	ReAdmis	0
	VitD_levels	0
	Doc_visits	0
	Full_meals_eaten	0
	VitD_supp	0
	Soft_drink	0
	Initial_admin	0
	HighBlood	0
	Stroke	0
	Complication risk	0
	Overweight	0
	Arthritis	0
	Diabetes	0
	Hyperlipidemia	0
	BackPain	0
	Anxiety	0
	Allergic_rhinitis	0
	Reflux_esophagitis	0
	Asthma	0
	Services	0
	Initial_days	0
	TotalCharge	0
	•	0
	Additional_charges Item1	0
	Item2 Item3	0 0
	Item3	0
	Item5	0
	Item6	0
	Item7	0
	<pre>Item8 dtype: int64</pre>	0
	ucype. IIIC04	

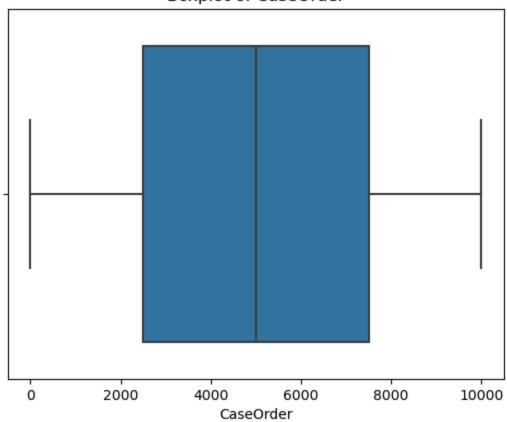
```
In [38]: #Verify imputation of data into the Initial_days variable
   plt.hist(df['Initial_days'])
   plt.show()
```



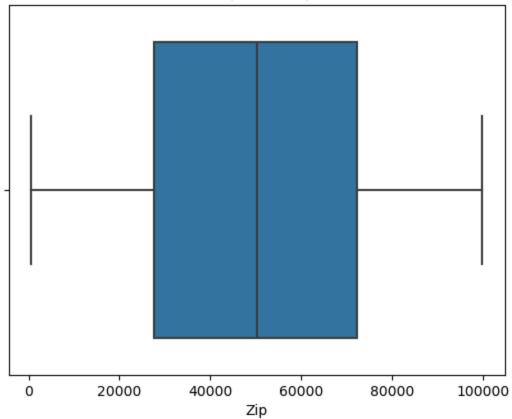
```
In [42]: #placing numerical columns in an array
outlier_col = ['CaseOrder', 'Zip', 'Lat', 'Lng', 'Population', 'Children', 'Age', '
```

```
In [43]: # creating boxplot graphs for each numeric variable
for i in outlier_col:
    plt.figure()
    sns.boxplot(x=i, data=df)
    plt.title(f'Boxplot of {i}')
    plt.show()
```

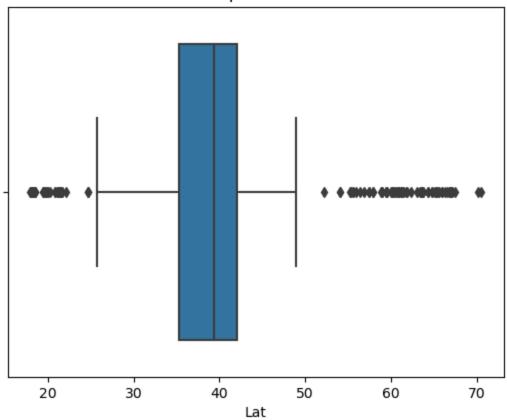
Boxplot of CaseOrder



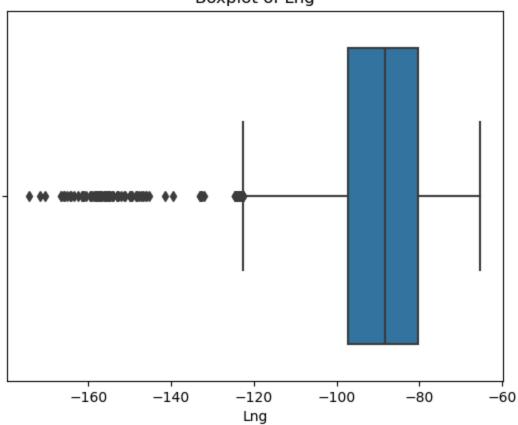
Boxplot of Zip



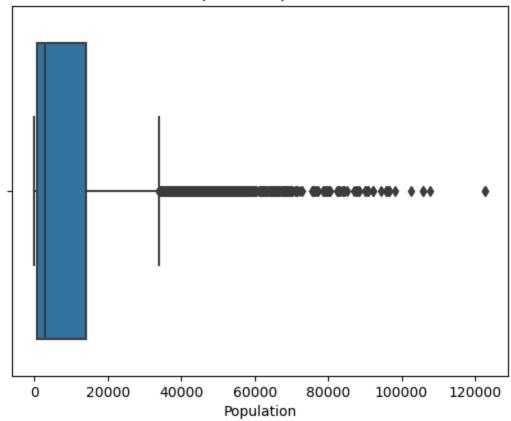
Boxplot of Lat



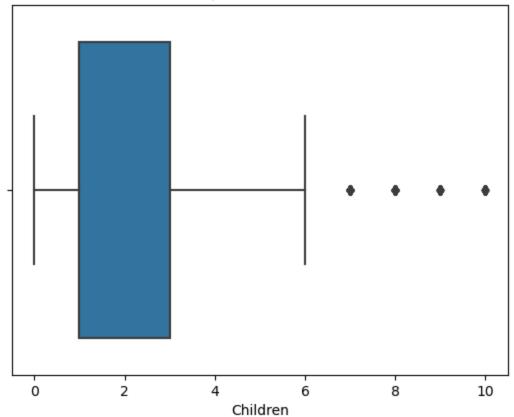
Boxplot of Lng



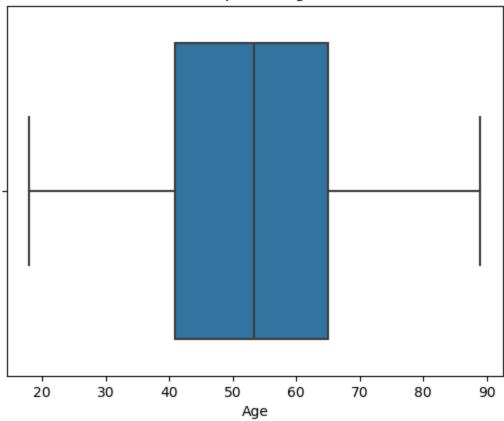
Boxplot of Population



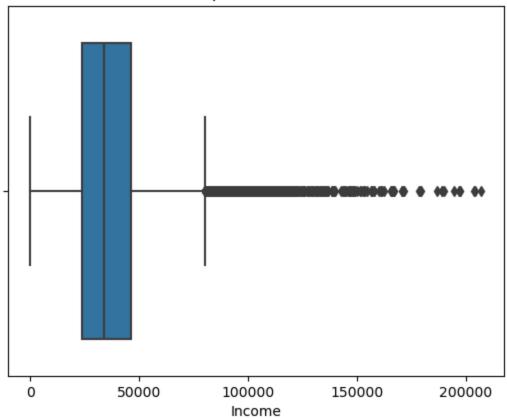
Boxplot of Children



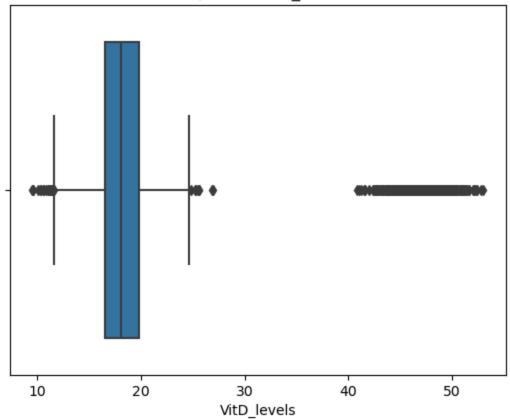
Boxplot of Age



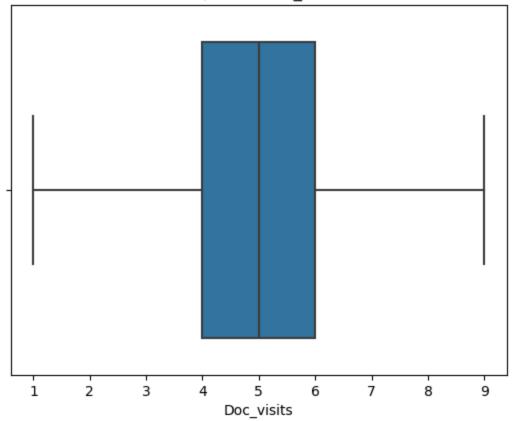
Boxplot of Income



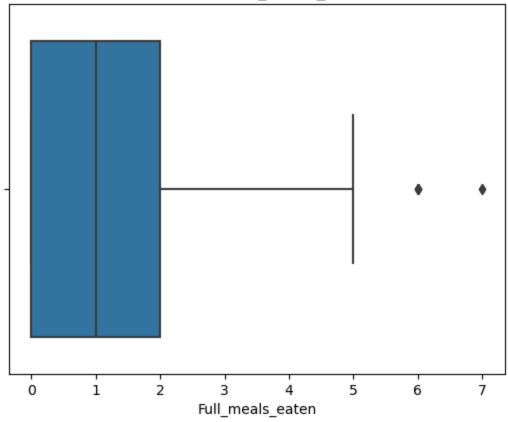
Boxplot of VitD_levels



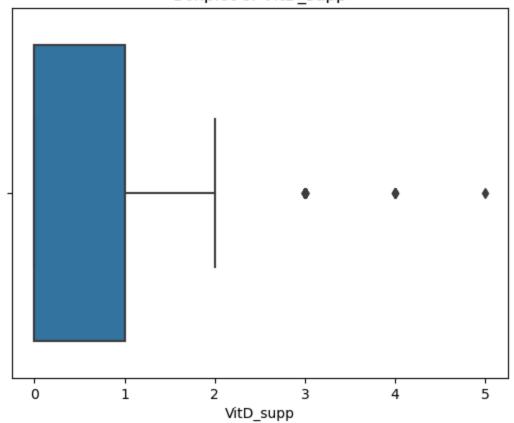
Boxplot of Doc_visits



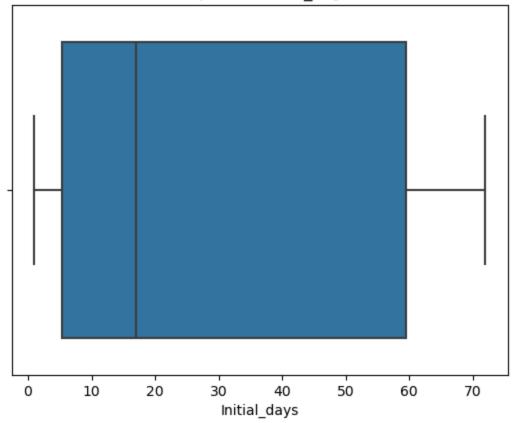
Boxplot of Full_meals_eaten



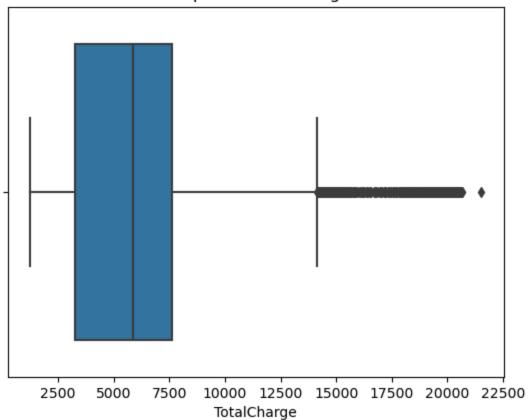
Boxplot of VitD_supp



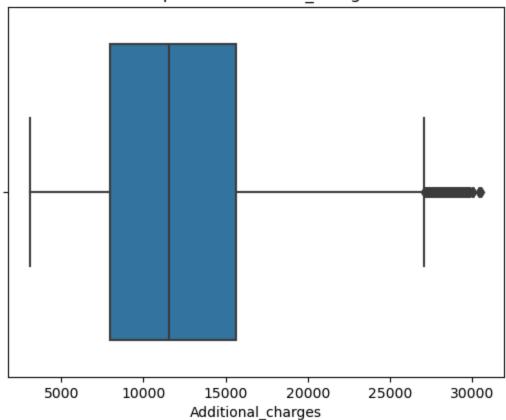
Boxplot of Initial_days



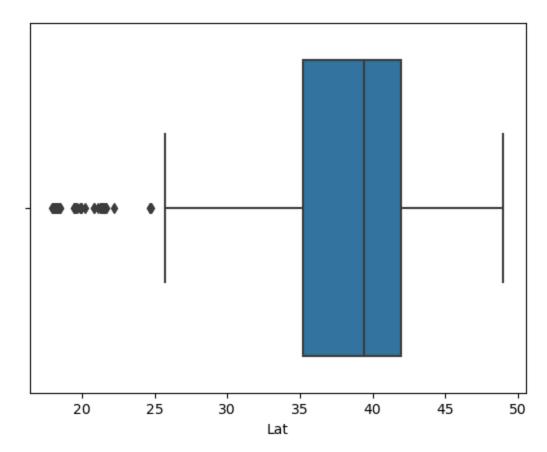
Boxplot of TotalCharge



Boxplot of Additional_charges



```
In [39]: #impute outliers with the median for Lat
df['Lat'] = np.where(df['Lat'] >50, np.nan , df['Lat'])
In [40]: boxplot=sns.boxplot(x='Lat',data=df)
```



```
In [41]: df['Lat'] = np.where(df['Lat'] <25, np.nan , df['Lat'])
In [42]: df['Lat'].fillna(df['Lat'].median(), inplace = True)
In [43]: df.info()</pre>
```

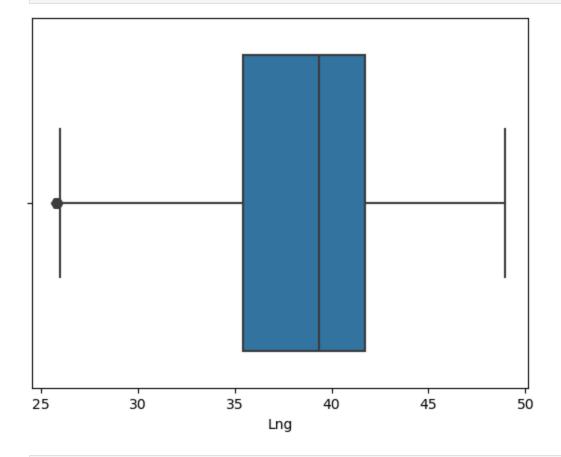
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 53 columns):

Data	columns (total 53 c	olumns):	
#	Column	Non-Null Count	Dtype
0	Unnamed: 0	10000 non-null	int64
1	CaseOrder	10000 non-null	int64
2	Customer_id	10000 non-null	object
3	Interaction	10000 non-null	object
4	UID	10000 non-null	object
			•
5	City	10000 non-null	object
6	State	10000 non-null	object
7	County	10000 non-null	object
8	Zip	10000 non-null	int64
9	Lat	10000 non-null	float64
10	Lng	10000 non-null	float64
11	Population	10000 non-null	int64
12	Area	10000 non-null	object
13	Timezone	10000 non-null	object
14	Job	10000 non-null	object
15	Children	10000 non-null	float64
16	Age	10000 non-null	float64
17	Education	10000 non-null	object
18	Employment	10000 non-null	object
19	Income	10000 non-null	float64
20	Marital	10000 non-null	object
21	Gender	10000 non-null	object
22	ReAdmis	10000 non-null	object
23	VitD_levels	10000 non-null	float64
24	Doc_visits	10000 non-null	int64
25	Full_meals_eaten	10000 non-null	int64
26	VitD_supp	10000 non-null	int64
27	Soft_drink	10000 non-null	object
28	Initial_admin	10000 non-null	object
29	HighBlood	10000 non-null	object
30	Stroke	10000 non-null	object
31	Complication_risk	10000 non-null	_
	· —		object float64
32	Overweight	10000 non-null	
33	Arthritis	10000 non-null	object
34	Diabetes	10000 non-null	object
35	Hyperlipidemia	10000 non-null	object
36	BackPain	10000 non-null	object
37	Anxiety	10000 non-null	float64
38	Allergic_rhinitis	10000 non-null	object
39	Reflux_esophagitis	10000 non-null	object
40	Asthma	10000 non-null	object
41	Services	10000 non-null	object
42	<pre>Initial_days</pre>	10000 non-null	float64
43	TotalCharge	10000 non-null	float64
44	Additional_charges	10000 non-null	float64
45	Item1	10000 non-null	int64
46	Item2	10000 non-null	int64
47	Item3	10000 non-null	int64
48	Item4	10000 non-null	int64
49	Item5	10000 non-null	int64
50	Item6	10000 non-null	int64

Data	columns (total 53 c	olumns):	
#	Column	Non-Null Count	Dtype
0	Unnamed: 0	10000 non-null	int64
1	CaseOrder	10000 non-null	int64
2	Customer_id	10000 non-null	object
3	Interaction	10000 non-null	object
4	UID	10000 non-null	object
			•
5	City	10000 non-null	object
6	State	10000 non-null	object
7	County	10000 non-null	object
8	Zip	10000 non-null	int64
9	Lat	10000 non-null	float64
10	Lng	10000 non-null	float64
11	Population	10000 non-null	int64
12	Area	10000 non-null	object
13	Timezone	10000 non-null	object
14	Job	10000 non-null	object
15	Children	10000 non-null	float64
16	Age	10000 non-null	float64
17	Education	10000 non-null	object
18	Employment	10000 non-null	object
19	Income	10000 non-null	float64
20	Marital	10000 non-null	object
21	Gender	10000 non-null	object
22	ReAdmis	10000 non-null	object
23	VitD_levels	10000 non-null	float64
24	Doc_visits	10000 non-null	int64
25	Full_meals_eaten	10000 non-null	int64
26	VitD_supp	10000 non-null	int64
27	Soft_drink	10000 non-null	object
28	Initial_admin	10000 non-null	object
29	HighBlood	10000 non-null	object
30	Stroke	10000 non-null	object
31		10000 non-null	_
	Complication_risk		object float64
32	Overweight	10000 non-null	
33	Arthritis	10000 non-null	object
34	Diabetes	10000 non-null	object
35	Hyperlipidemia	10000 non-null	object
36	BackPain	10000 non-null	object
37	Anxiety	10000 non-null	float64
38	Allergic_rhinitis	10000 non-null	object
39	Reflux_esophagitis	10000 non-null	object
40	Asthma	10000 non-null	object
41	Services	10000 non-null	object
42	<pre>Initial_days</pre>	10000 non-null	float64
43	TotalCharge	10000 non-null	float64
44	Additional_charges	10000 non-null	float64
45	Item1	10000 non-null	int64
46	Item2	10000 non-null	int64
47	Item3	10000 non-null	int64
48	Item4	10000 non-null	int64
49	Item5	10000 non-null	int64
50	Item6	10000 non-null	int64

51 Item7 10000 non-null int64 52 Item8 10000 non-null int64 dtypes: float64(11), int64(15), object(27) memory usage: 4.0+ MB

```
In [45]: boxplot=sns.boxplot(x='Lng',data=df)
```

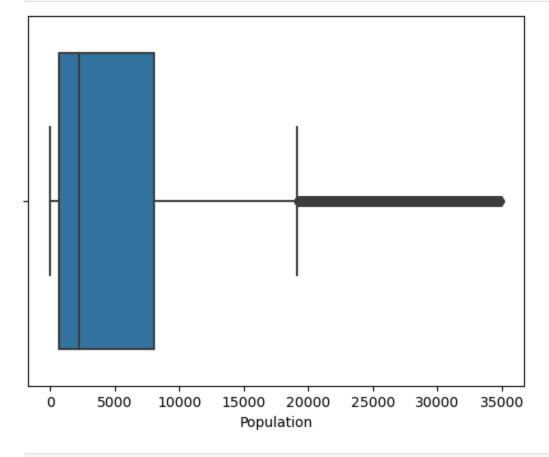


```
In [46]: #impute outliers with the median for Population
    df['Population'] = np.where(df['Population'] > 35000, np.nan , df['Population'])
    df['Population'].fillna(df['Population'].median(), inplace = True)
    df.info()
```

Data	columns (total 53 c	olumns):	
#	Column	Non-Null Count	Dtype
0	Unnamed: 0	10000 non-null	int64
1	CaseOrder	10000 non-null	int64
2	Customer_id	10000 non-null	object
3	Interaction	10000 non-null	object
4	UID	10000 non-null	object
5	City	10000 non-null	object
6	•		-
7	State	10000 non-null	object
	County	10000 non-null	object
8	Zip	10000 non-null	int64
9	Lat	10000 non-null	float64
10	Lng	10000 non-null	float64
11	Population	10000 non-null	float64
12	Area	10000 non-null	object
13	Timezone	10000 non-null	object
14	Job	10000 non-null	object
15	Children	10000 non-null	float64
16	Age	10000 non-null	float64
17	Education	10000 non-null	object
18	Employment	10000 non-null	object
19	Income	10000 non-null	float64
20	Marital	10000 non-null	object
21	Gender	10000 non-null	object
22	ReAdmis	10000 non-null	object
23	VitD_levels	10000 non-null	float64
24	_ Doc_visits	10000 non-null	int64
25	Full_meals_eaten	10000 non-null	
26	VitD_supp	10000 non-null	int64
27	Soft_drink	10000 non-null	object
28	Initial admin	10000 non-null	object
29	HighBlood	10000 non-null	object
30	Stroke	10000 non-null	•
31	Complication_risk	10000 non-null	object
32	Overweight	10000 non-null	float64
33	Arthritis	10000 non-null	object
34	Diabetes	10000 non-null	object
	Hyperlipidemia	10000 non-null	object
35	BackPain		-
36		10000 non-null	object
37	Anxiety	10000 non-null	float64
38	Allergic_rhinitis	10000 non-null	object
39	Reflux_esophagitis	10000 non-null	object
40	Asthma	10000 non-null	object
41	Services	10000 non-null	object
42	Initial_days	10000 non-null	float64
43	TotalCharge	10000 non-null	float64
44	Additional_charges	10000 non-null	float64
45	Item1	10000 non-null	int64
46	Item2	10000 non-null	int64
47	Item3	10000 non-null	int64
48	Item4	10000 non-null	int64
49	Item5	10000 non-null	int64
50	Item6	10000 non-null	int64

```
51 Item7 10000 non-null int64
52 Item8 10000 non-null int64
dtypes: float64(12), int64(14), object(27)
memory usage: 4.0+ MB
```

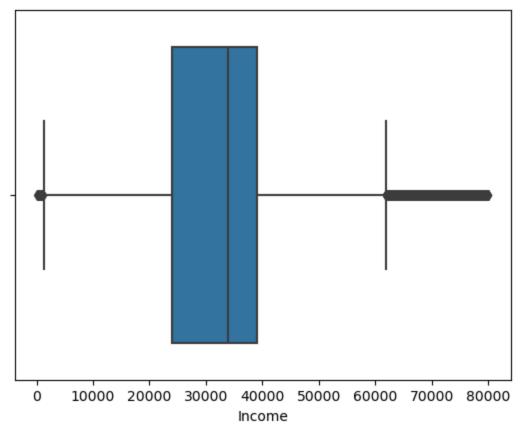
```
In [47]: boxplot=sns.boxplot(x='Population',data=df)
```



```
In [48]: #impute outliers with the median for Income
    df['Income'] = np.where(df['Income'] > 80000, np.nan , df['Income'])
    df['Income'].fillna(df['Income'].median(), inplace = True)
    df.info()
    boxplot=sns.boxplot(x='Income',data=df)
```

Data	columns (total 53 c	olumns):	
#	Column	Non-Null Count	Dtype
0	Unnamed: 0	10000 non-null	int64
1	CaseOrder	10000 non-null	int64
2	Customer_id	10000 non-null	object
3	Interaction	10000 non-null	object
4	UID	10000 non-null	-
			object
5	City	10000 non-null	object
6	State	10000 non-null	object
7	County	10000 non-null	object
8	Zip	10000 non-null	int64
9	Lat	10000 non-null	float64
10	Lng	10000 non-null	float64
11	Population	10000 non-null	float64
12	Area	10000 non-null	object
13	Timezone	10000 non-null	object
14	Job	10000 non-null	object
15	Children	10000 non-null	float64
16	Age	10000 non-null	float64
17	Education	10000 non-null	object
18	Employment	10000 non-null	object
19	Income	10000 non-null	float64
20	Marital	10000 non-null	object
21	Gender	10000 non-null	object
22	ReAdmis	10000 non-null	object
23	VitD_levels	10000 non-null	float64
24	Doc_visits	10000 non-null	int64
25	Full_meals_eaten	10000 non-null	int64
26	VitD_supp	10000 non-null	int64
27	Soft_drink	10000 non-null	object
28	Initial_admin	10000 non-null	object
29	HighBlood	10000 non-null	object
30	Stroke	10000 non-null	object
31	Complication_risk	10000 non-null	object
32	Overweight	10000 non-null	float64
33	Arthritis	10000 non-null	object
34	Diabetes	10000 non-null	object
35	Hyperlipidemia	10000 non-null	object
36	BackPain	10000 non-null	-
			object float64
37	Anxiety		
38	Allergic_rhinitis	10000 non-null	object
39	Reflux_esophagitis	10000 non-null	object
40	Asthma	10000 non-null	object
41	Services	10000 non-null	object
42	Initial_days	10000 non-null	float64
43	TotalCharge	10000 non-null	float64
44	Additional_charges	10000 non-null	float64
45	Item1	10000 non-null	int64
46	Item2	10000 non-null	int64
47	Item3	10000 non-null	int64
48	Item4	10000 non-null	int64
49	Item5	10000 non-null	int64
50	Item6	10000 non-null	int64

51 Item7 10000 non-null int64 52 Item8 10000 non-null int64 dtypes: float64(12), int64(14), object(27) memory usage: 4.0+ MB



```
In [49]: #impute outliers with the median for Vitamin D levels

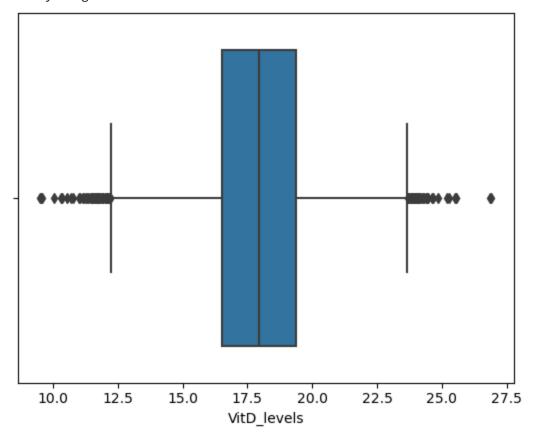
df['VitD_levels'] = np.where(df['VitD_levels'] > 30, np.nan , df['VitD_levels'])

df['VitD_levels'].fillna(df['VitD_levels'].median(), inplace = True)

df.info()
boxplot=sns.boxplot(x='VitD_levels',data=df)
```

Data	columns (total 53 c	olumns):	
#	Column	Non-Null Count	Dtype
0	Unnamed: 0	10000 non-null	int64
1	CaseOrder	10000 non-null	int64
2	Customer_id	10000 non-null	object
3	Interaction	10000 non-null	object
4	UID	10000 non-null	object
5			•
	City	10000 non-null	object
6	State	10000 non-null	object
7	County	10000 non-null	object
8	Zip	10000 non-null	int64
9	Lat	10000 non-null	float64
10	Lng	10000 non-null	float64
11	Population	10000 non-null	float64
12	Area	10000 non-null	object
13	Timezone	10000 non-null	object
14	Job	10000 non-null	object
15	Children	10000 non-null	float64
16	Age	10000 non-null	float64
17	Education	10000 non-null	object
18	Employment	10000 non-null	object
19	Income	10000 non-null	float64
20	Marital	10000 non-null	object
21	Gender	10000 non-null	object
22	ReAdmis	10000 non-null	object
23	VitD_levels	10000 non-null	float64
24	_ Doc_visits	10000 non-null	int64
25	Full_meals_eaten	10000 non-null	int64
26	VitD_supp	10000 non-null	
27	Soft_drink	10000 non-null	object
28	Initial_admin	10000 non-null	object
29	HighBlood	10000 non-null	object
30	Stroke	10000 non-null	object
31	Complication_risk	10000 non-null	object
32	Overweight	10000 non-null	float64
	Arthritis		
33 34		10000 non-null	object
	Diabetes	10000 non-null	object
35	Hyperlipidemia	10000 non-null	object
36	BackPain	10000 non-null	object
37	Anxiety	10000 non-null	float64
38	Allergic_rhinitis	10000 non-null	object
39	Reflux_esophagitis	10000 non-null	object
40	Asthma	10000 non-null	object
41	Services	10000 non-null	object
42	Initial_days	10000 non-null	float64
43	TotalCharge	10000 non-null	float64
44	Additional_charges	10000 non-null	float64
45	Item1	10000 non-null	int64
46	Item2	10000 non-null	int64
47	Item3	10000 non-null	int64
48	Item4	10000 non-null	int64
49	Item5	10000 non-null	int64
50	Item6	10000 non-null	int64

51 Item7 10000 non-null int64 52 Item8 10000 non-null int64 dtypes: float64(12), int64(14), object(27) memory usage: 4.0+ MB



```
In [50]: #impute outliers with the median for TotalCharge

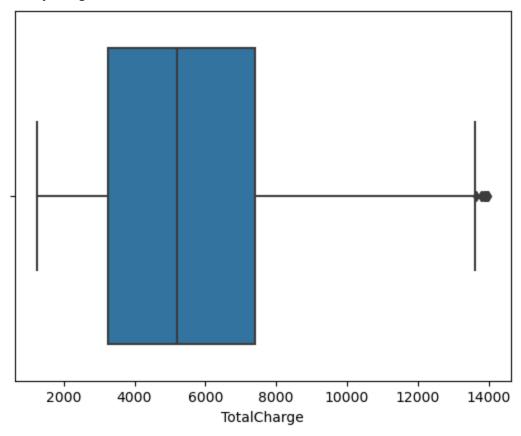
df['TotalCharge'] = np.where(df['TotalCharge'] > 14000, np.nan , df['TotalCharge'])

df['TotalCharge'].fillna(df['TotalCharge'].median(), inplace = True)

df.info()
boxplot=sns.boxplot(x='TotalCharge',data=df)
```

Data	columns (total 53 c	olumns):	
#	Column	Non-Null Count	Dtype
0	Unnamed: 0	10000 non-null	int64
1	CaseOrder	10000 non-null	int64
2	Customer_id	10000 non-null	object
3	Interaction	10000 non-null	object
4	UID	10000 non-null	object
5	City	10000 non-null	object
6	State	10000 non-null	object
7	County	10000 non-null	object
8	Zip	10000 non-null	int64
9	Lat	10000 non-null	float64
10	Lng	10000 non-null	
11	Population	10000 non-null	
12	Area	10000 non-null	object
13	Timezone	10000 non-null	object
14	Job	10000 non-null	object
15	Children	10000 non-null	float64
16	Age	10000 non-null	float64
17	Education	10000 non-null	object
18	Employment	10000 non-null	object
19	Income	10000 non-null	float64
20	Marital		
		10000 non-null	object
21	Gender	10000 non-null	object
22	ReAdmis	10000 non-null	object
23	VitD_levels	10000 non-null	float64
24	Doc_visits	10000 non-null	int64
25	Full_meals_eaten	10000 non-null	int64
26	VitD_supp	10000 non-null	int64
27	Soft_drink	10000 non-null	object
28	Initial_admin	10000 non-null	object
29	HighBlood	10000 non-null	object
30	Stroke	10000 non-null	object
31	Complication_risk	10000 non-null	object
32	Overweight	10000 non-null	float64
33	Arthritis	10000 non-null	object
34	Diabetes	10000 non-null	object
35	Hyperlipidemia	10000 non-null	object
36	BackPain	10000 non-null	object
37	Anxiety	10000 non-null	float64
38	Allergic_rhinitis	10000 non-null	object
39	Reflux_esophagitis	10000 non-null	object
40	Asthma	10000 non-null	object
41	Services	10000 non-null	object
42	<pre>Initial_days</pre>	10000 non-null	float64
43	TotalCharge	10000 non-null	float64
44	Additional_charges	10000 non-null	float64
45	Item1	10000 non-null	int64
46	Item2	10000 non-null	int64
47	Item3	10000 non-null	int64
48	Item4	10000 non-null	int64
49	Item5	10000 non-null	int64
50	Item6	10000 non-null	int64

51 Item7 10000 non-null int64 52 Item8 10000 non-null int64 dtypes: float64(12), int64(14), object(27) memory usage: 4.0+ MB



```
In [51]: #change boolean categorical to numeric
    # create an array of all the variables that are needed to be converted
    #create a dictionary for converting the values
    #create a for loop changing all of them to numeric
    var_cat = ['HighBlood', 'Stroke', 'Arthritis', 'Diabetes', 'Hyperlipidemia', 'BackP dict_var = {'numeric':{'No':0, 'Yes':1}}
    for i in var_cat:
        df['numeric'] = df[i]
        df.replace(dict_var, inplace = True)
```

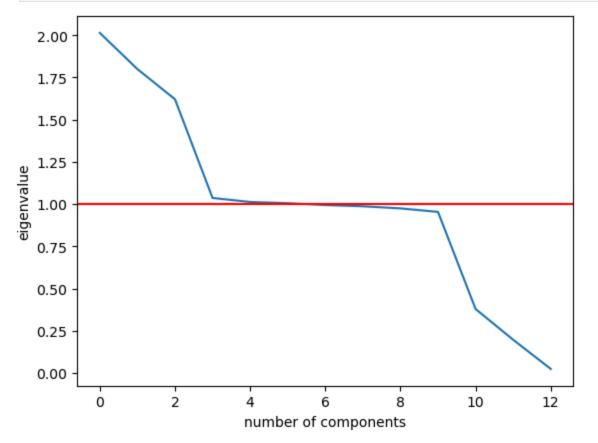
In [52]: df.info()

Data	columns (total 54 co	olumns):	
#	Column	Non-Null Count	Dtype
0	Unnamed: 0	10000 non-null	int64
1	CaseOrder	10000 non-null	int64
2	Customer_id	10000 non-null	object
3	Interaction	10000 non-null	object
4	UID	10000 non-null	object
			•
5	City	10000 non-null	object
6	State	10000 non-null	object
7	County	10000 non-null	object
8	Zip	10000 non-null	int64
9	Lat	10000 non-null	float64
10	Lng	10000 non-null	float64
11	Population	10000 non-null	float64
12	Area	10000 non-null	object
13	Timezone	10000 non-null	object
14	Job	10000 non-null	object
15	Children	10000 non-null	float64
16	Age	10000 non-null	float64
17	Education	10000 non-null	object
18	Employment	10000 non-null	object
19	Income	10000 non-null	float64
20	Marital	10000 non-null	object
21	Gender	10000 non-null	object
22	ReAdmis	10000 non-null	object
23	VitD_levels	10000 non-null	float64
24	Doc_visits	10000 non-null	int64
25	Full_meals_eaten	10000 non-null	
26	VitD_supp	10000 non-null	int64
27	Soft_drink	10000 non-null	object
28	Initial admin	10000 non-null	object
29	HighBlood	10000 non-null	object
30	Stroke	10000 non-null	object
31		10000 non-null	-
	Complication_risk		object
32	Overweight Arthritis	10000 non-null	float64
33		10000 non-null	object
34	Diabetes	10000 non-null	object
35	Hyperlipidemia	10000 non-null	object
36	BackPain	10000 non-null	object
37	Anxiety	10000 non-null	float64
38	Allergic_rhinitis	10000 non-null	object
39	Reflux_esophagitis	10000 non-null	object
40	Asthma	10000 non-null	object
41	Services	10000 non-null	object
42	Initial_days	10000 non-null	float64
43	TotalCharge	10000 non-null	float64
44	Additional_charges	10000 non-null	float64
45	Item1	10000 non-null	int64
46	Item2	10000 non-null	int64
47	Item3	10000 non-null	int64
48	Item4	10000 non-null	int64
49	Item5	10000 non-null	int64
50	Item6	10000 non-null	int64

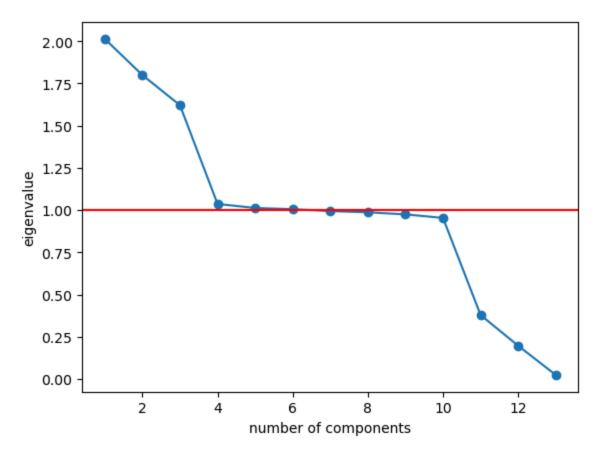
```
51 Item7
                                 10000 non-null int64
         52 Item8
                                10000 non-null int64
         53 numeric
                                 10000 non-null int64
        dtypes: float64(12), int64(15), object(27)
        memory usage: 4.1+ MB
In [53]: #Check if it worked
         df['Stroke']
Out[53]: 0
                   Nο
         1
                   No
          2
                  No
          3
                 Yes
          4
                  No
         9995
                  Nο
          9996
                  No
          9997
                  No
         9998
                  No
          9999
                   No
         Name: Stroke, Length: 10000, dtype: object
In [54]: #round the VitD_levels, Initial_days, TotalCharge, and Additional_charge to 2 place
         df = df.round({'VitD_levels': 2, 'Initial_days': 2, 'TotalCharge': 2, 'Additional_cha
In [55]: df['TotalCharge']
Out[55]: 0
                 3191.05
                 4214.91
         1
          2
                 2177.59
          3
                 2465.12
                 1885.66
          4
                  . . .
         9995
                 6651.24
               7851.52
          9996
          9997
                 7725.95
          9998
                 8462.83
          9999
                 8700.86
         Name: TotalCharge, Length: 10000, dtype: float64
 In [ ]: #steps taken for PCA
         #define features/variables for PCA
         #Normalize data and apply PCA
         #PCA Loadings
         #selecting PCs
In [56]: #selecting continuous variables
         pca_col = df[['Lat', 'Lng', 'Population', 'Children', 'Age', 'Income', 'VitD_levels'
In [57]: #Normalize data
         pca_normalized = (pca_col - pca_col.mean())/pca_col.std()
In [58]: #Applying PCA
         pca = PCA(n_components=pca_col.shape[1])
```

```
In [59]:
          print(pca)
        PCA(n_components=13)
In [60]: pca.fit(pca_normalized)
Out[60]:
                     PCA
          PCA(n components=13)
In [61]:
          df_pca = pd.DataFrame(pca.transform(pca_normalized),
          columns = ['PC1','PC2','PC3','PC4','PC5','PC6','PC7','PC8','PC9','PC10','PC11','PC1
In [62]:
          #PCA Loadings
          loadings = pd.DataFrame(pca.components_.T,
                                    columns= ['PC1', 'PC2', 'PC3', 'PC4', 'PC5', 'PC6', 'PC7', 'PC8', '
                                    index = pca_col.columns)
          loadings
                                   PC1
                                              PC2
                                                        PC3
                                                                   PC4
                                                                              PC5
                                                                                        PC6
                                                                                                   Ρ
Out[62]:
                             -0.688881
                                         0.090087
                                                    0.005902 -0.020168
                                                                         0.011733
                                                                                             -0.0528
                         Lat
                                                                                    0.028989
                                                    0.007406 -0.017607
                              -0.689853
                                         0.091700
                                                                         0.008570
                                                                                    0.029425
                                                                                             -0.0470
                  Population
                               0.181661
                                         -0.003183
                                                   -0.025145 -0.169625
                                                                         0.043939
                                                                                    0.193604
                                                                                             -0.3712
                    Children
                               0.007908
                                         0.005243
                                                    0.011403
                                                               0.123882
                                                                         0.032642
                                                                                    0.890027
                                                                                             -0.2674
                               0.018017
                                         0.076730
                                                    0.701364
                                                              -0.009192
                                                                         0.038900
                                                                                   -0.008535
                                                                                             -0.0059
                        Age
                     Income
                               0.002998
                                         -0.001692
                                                    0.001013
                                                              -0.486053
                                                                         -0.347674
                                                                                   -0.117513
                                                                                             -0.2362
                               0.005916
                  VitD levels
                                         0.048511
                                                    0.015771
                                                               0.299371
                                                                        -0.695813
                                                                                    0.230925
                                                                                              0.3652
                   Doc visits
                              -0.011559
                                        -0.009666
                                                    0.014245
                                                              -0.279159
                                                                        -0.541270
                                                                                   -0.084112 -0.4302
                                                                        -0.286246
                                                                                  -0.213568 -0.1224
             Full meals eaten
                              -0.004441
                                        -0.024515
                                                    0.037558
                                                               0.556597
                  VitD supp
                              -0.001700
                                         0.038639
                                                    0.012404
                                                              -0.490496
                                                                        -0.106798
                                                                                    0.217701
                                                                                              0.6275
                                                                                   -0.033248 -0.0479
                  Initial_days
                               0.085627
                                         0.693399
                                                   -0.088100
                                                               0.002814
                                                                         0.055428
                 TotalCharge
                               0.090274
                                         0.697039
                                                   -0.071925
                                                               0.024273
                                                                         -0.027377
                                                                                  -0.016387 -0.0078
          Additional_charges
                               0.021820
                                         0.079633
                                                    0.701630
                                                              -0.017189
                                                                         0.010465 -0.001133 -0.0125
In [63]:
          #Selecting PCs
          cov_matrix = np.dot(pca_normalized.T, pca_normalized) / pca_col.shape[0]
          eigenvalues = [np.dot(eigenvector.T,np.dot(cov_matrix,eigenvector)) for eigenvector
          plt.plot(eigenvalues)
          plt.xlabel('number of components')
          plt.ylabel('eigenvalue')
```

```
plt.axhline(y=1, color = 'red')
plt.show()
```



```
In [64]: #relabelling the axia
  plt.plot(np.arange(1,len(eigenvalues)+1),eigenvalues, marker='o')
  plt.xlabel('number of components')
  plt.ylabel('eigenvalue')
  plt.axhline(y=1, color = 'red')
  plt.show()
```



```
In [65]: #exporting as csv file
    df.to_csv(r'C:\Users\arjun\OneDrive\Desktop\WGU\D206\D206csv.csv')
In []:
```