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In [1]: import pandas as pd

import numpy as np

import missingno as msno

import matplotlib.pyplot as plt

from pandas import DataFrame

import scipy.stats as stats

import seaborn

from sklearn.decomposition import PCA

#Loading the CSV of the default dataset
df = pd.read_csv(r'C:\Users\mmorg\WGU\D206\D206 Assessment Data\medical_raw_data.csv')

#Object or string is qualitative, int64 or float64 is quantitative
df.info()

#Detect duplicate rows
print(df.duplicated())

#Detect null values
print(df.isnull().sum())

#Missing data matrix showing where null values exist
msno.matrix(df, labels=True)
plt.title('Missing Data Matrix')
plt.show()

#Detection of outliers for quantitative values
boxplot=seaborn.boxplot(x='Children',data=df)
plt.show()
boxplot=seaborn.boxplot(x='Age',data=df)
plt.show()
boxplot=seaborn.boxplot(x='Income',data=df)
plt.show()
boxplot=seaborn.boxplot(x='VitD_levels',data=df)
plt.show()
boxplot=seaborn.boxplot(x='Doc_visits',data=df)
plt.show()
boxplot=seaborn.boxplot(x='Full_meals_eaten',data=df)
plt.show()
boxplot=seaborn.boxplot(x='VitD_supp',data=df)
plt.show()
boxplot=seaborn.boxplot(x='Overweight',data=df)
plt.show()
boxplot=seaborn.boxplot(x='Anxiety',data=df)
plt.show()
boxplot=seaborn.boxplot(x='Initial_days',data=df)
plt.show()
boxplot=seaborn.boxplot(x='TotalCharge',data=df)
plt.show()
boxplot=seaborn.boxplot(x='Additional_charges',data=df)
plt.show()
boxplot=seaborn.boxplot(x='Item1',data=df)
plt.show()
boxplot=seaborn.boxplot(x='Item2',data=df)
plt.show()
boxplot=seaborn.boxplot(x='Item3',data=df)
plt.show()
boxplot=seaborn.boxplot(x='Item4',data=df)
plt.show()
boxplot=seaborn.boxplot(x='Item5',data=df)
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plt.show()
boxplot=seaborn.boxplot(x='Item6',data=df)
plt.show()
boxplot=seaborn.boxplot(x='Item7',data=df)
plt.show()
boxplot=seaborn.boxplot(x='Item8',data=df)
plt.show()

#Visualizing distribution shapes
print('Children Original: ')
plt.hist(df['Children'])
plt.show()
print('Age Original: ')
plt.hist(df['Age'])
plt.show()
print('Income Original: ')
plt.hist(df['Income'])
plt.show()
print('Overweight Original: ')
plt.hist(df['Overweight'])
plt.show()
print('Anxiety Original: ')
plt.hist(df['Anxiety'])
plt.show()
print('Initial_days Original: ')
plt.hist(df['Initial_days'])
plt.show()

#Fixing missing data based on distribution shapes
df['Children'].fillna(method='ffill', inplace=True)
df['Age'].fillna(method='ffill', inplace=True)
df['Income'].fillna(df['Income'].median(), inplace=True)
df['Soft_drink'].fillna("No", inplace=True)
df['Overweight'].fillna(0, inplace=True)
df['Anxiety'].fillna(0, inplace=True)
df['Initial_days'].fillna(method='ffill', inplace=True)

#Checking to make sure all missing data has been replaced with selected values
print(df.isnull().sum())

#Visualizing distribution shapes after changes
print('Children Modified: ')
plt.hist(df['Children'])
plt.show()
print('Age Modified: ')
plt.hist(df['Age'])
plt.show()
print('Income Modified: ')
plt.hist(df['Income'])
plt.show()
print('Overweight Modified: ')
plt.hist(df['Overweight'])
plt.show()
print('Anxiety Modified: ')
plt.hist(df['Anxiety'])
plt.show()
print('Initial_days Modified: ')
plt.hist(df['Initial_days'])
plt.show()

#Checking Z-scores because why not
##df['ChildrenZScore']=stats.zscore(df['Children'])
##df[['Children', 'ChildrenZScore']].head
##plt.hist(df['ChildrenZScore'])
##plt.show()
##
##df['AgeZScore']=stats.zscore(df['Age'])
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##df[['Age', 'AgeZScore']].head
##plt.hist(df['AgeZScore'])
##plt.show()

##df['IncomeZScore']=stats.zscore(df['Income'])
##df[['Income', 'IncomeZScore']].head
##plt.hist(df['IncomeZScore'])
##plt.show()

##df['IncomeZScore']=stats.zscore(df['Income'])
##df[['Income', 'IncomeZScore']].head
##plt.hist(df['IncomeZScore'])
##plt.show()

##df['VitD_LevelsZScore']=stats.zscore(df['VitD_Levels'])
##df[['VitD_Levels', 'VitD_LevelsZScore']].head
##plt.hist(df['VitD_LevelsZScore'])
##plt.show()

##df['Doc_visitsZScore']=stats.zscore(df['Doc_visits'])
##df[['Doc_visits', 'Doc_visitsZScore']].head
##plt.hist(df['Doc_visitsZScore'])
##plt.show()

#Check for unique values to re-express categorical variables
print('Education: ', df.Education.unique(), "\n")
print('Marital: ', df.Marital.unique())
print('Gender: ', df.Gender.unique())
print('ReAdmis: ', df.ReAdmis.unique())
print('Soft_drink: ', df.Soft_drink.unique())
print('Initial_admin: ', df.Initial_admin.unique())
print('HighBlood: ', df.HighBlood.unique())
print('Stroke: ', df.Stroke.unique())
print('Complication_risk: ', df.Complication_risk.unique())
print('Arthritis: ', df.Arthritis.unique())
print('Diabetes: ', df.Diabetes.unique())
print('Hyperlipidemia: ', df.Hyperlipidemia.unique())
print('BackPain: ', df.BackPain.unique())
print('Allergic_rhinitis: ', df.Allergic_rhinitis.unique())
print('Reflux_esophagitis: ', df.Reflux_esophagitis.unique())
print('Ashtma: ', df.Asthma.unique())
print('Services: ', df.Services.unique())

#Turn categorical values into quantitative data
df['Education_numeric'] = df['Education']
dict_edu = {"Education_numeric": {"No Schooling Completed": 0, "Nursery School to 8th Grade": 1,
df.replace(dict_edu, inplace=True)

df['Marital_numeric'] = df['Marital']
dict_marital = {"Marital_numeric": {"Never Married": 0, "Separated": 1, "Widowed": 2, "Divorced": 3}
df.replace(dict_marital, inplace=True)

df['Gender_numeric'] = df['Gender']
dict_gender = {"Gender_numeric": {"Prefer not to answer": 0, "Male": 1, "Female": 2}}
df.replace(dict_gender, inplace=True)

df['ReAdmis_numeric'] = df['ReAdmis']
dict_ReAdmis = {"ReAdmis_numeric": {"No": 0, "Yes": 1}}
df.replace(dict_ReAdmis, inplace=True)

df['Soft_drink_numeric'] = df['Soft_drink']
dict_Soft_drink = {"Soft_drink_numeric": {"No": 0, "Yes": 1}}
df.replace(dict_Soft_drink, inplace=True)

df['Initial_admin_numeric'] = df['Initial_admin']
dict_Initial_admin = {"Initial_admin_numeric": {"Emergency Admission": 0, "Elective Admission": 1}
df.replace(dict_Initial_admin, inplace=True)

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df['HighBlood_numeric'] = df['HighBlood']
dict_HighBlood = {"HighBlood_numeric": {"No": 0, "Yes": 1}}
df.replace(dict_HighBlood, inplace=True)

df['Stroke_numeric'] = df['Stroke']
dict_stroke = {"Stroke_numeric": {"No": 0, "Yes": 1}}
df.replace(dict_stroke, inplace=True)

df['Complication_risk_numeric'] = df['Complication_risk']
dict_complication = {"Complication_risk_numeric": {"Low": 0, "Medium": 1, "High": 2}}
df.replace(dict_complication, inplace=True)

df['Arthritis_numeric'] = df['Arthritis']
dict_arthritis = {"Arthritis_numeric": {"No": 0, "Yes": 1}}
df.replace(dict_arthritis, inplace=True)

df['Diabetes_numeric'] = df['Diabetes']
dict_diabetes = {"Diabetes_numeric": {"No": 0, "Yes": 1}}
df.replace(dict_diabetes, inplace=True)

df['Hyperlipidemia_numeric'] = df['Hyperlipidemia']
dict_hyperlipidemia = {"Hyperlipidemia_numeric": {"No": 0, "Yes": 1}}
df.replace(dict_hyperlipidemia, inplace=True)

df['BackPain_numeric'] = df['BackPain']
dict_backpain = {"BackPain_numeric": {"No": 0, "Yes": 1}}
df.replace(dict_backpain, inplace=True)

df['Allergic_rhinitis_numeric'] = df['Allergic_rhinitis']
dict_allergies = {"Allergic_rhinitis_numeric": {"No": 0, "Yes": 1}}
df.replace(dict_allergies, inplace=True)

df['Reflux_esophagitis_numeric'] = df['Reflux_esophagitis']
dict_reflux = {"Reflux_esophagitis_numeric": {"No": 0, "Yes": 1}}
df.replace(dict_reflux, inplace=True)

df['Asthma_numeric'] = df['Asthma']
dict_asthma = {"Asthma_numeric": {"No": 0, "Yes": 1}}
df.replace(dict_asthma, inplace=True)

df['Services_numeric'] = df['Services']
dict_services = {"Services_numeric": {"Blood Work": 0, "Intravenous": 1, "CT Scan": 2, "MRI": 3}}
df.replace(dict_services, inplace=True)

#Check for new columns
df.info()

#Checking data in new columns to make sure replacements worked
print('Education_numeric: ', df.Education_numeric.unique())
print('Marital_numeric: ', df.Marital_numeric.unique())
print('Gender_numeric: ', df.Gender_numeric.unique())
print('ReAdmis_numeric: ', df.ReAdmis_numeric.unique())
print('Soft_drink_numeric: ', df.Soft_drink_numeric.unique())
print('Initial_admin_numeric: ', df.Initial_admin_numeric.unique())
print('HighBlood_numeric: ', df.HighBlood_numeric.unique())
print('Stroke_numeric: ', df.Stroke_numeric.unique())
print('Complication_risk_numeric: ', df.Complication_risk_numeric.unique())
print('Arthritis_numeric: ', df.Arthritis_numeric.unique())
print('Diabetes_numeric: ', df.Diabetes_numeric.unique())
print('Hyperlipidemia_numeric: ', df.Hyperlipidemia_numeric.unique())
print('BackPain_numeric: ', df.BackPain_numeric.unique())
print('Allergic_rhinitis_numeric: ', df.Allergic_rhinitis_numeric.unique())
print('Reflux_esophagitis_numeric: ', df.Reflux_esophagitis_numeric.unique())
print('Ashtma_numeric: ', df.Asthma_numeric.unique())
print('Services_numeric: ', df.Services_numeric.unique(), '\n')
```

```
##Running PCA
test_pca = X_stand_df
test_pca_normalized=(test_pca-test_pca.mean())/test_pca.std()
pca = PCA(n_components=test_pca.shape[1])
pca.fit(test_pca_normalized)
print(pca)

test_pca2 = pd.DataFrame(pca.transform(test_pca_normalized),columns=['PC1','PC2','PC3', 'PC4',
                                                                    'PC10', 'PC11', 'PC12'])

loadings = pd.DataFrame(pca.components_.T,
columns = ['PC1','PC2','PC3', 'PC4', 'PC5', 'PC6', 'PC7', 'PC8', 'PC9','PC10', 'PC11', 'PC12',
index=test_pca_normalized.columns)
loadings
print(loadings)

cov_matrix = np.dot(test_pca_normalized.T, test_pca_normalized) / test_pca.shape[0]
eigenvalues = [np.dot(eigenvector.T, np.dot(cov_matrix, eigenvector)) for eigenvector in pca.components_]

plt.plot(eigenvalues)
plt.xlabel('number of components')
plt.ylabel('eigenvalues')
plt.axhline(y=1, color="red")
plt.title('Scree Plot')
plt.show()

#Extracting cleaned data
df.to_csv(r'C:\Users\mmorg\Desktop\Cleaned206data.csv')
```

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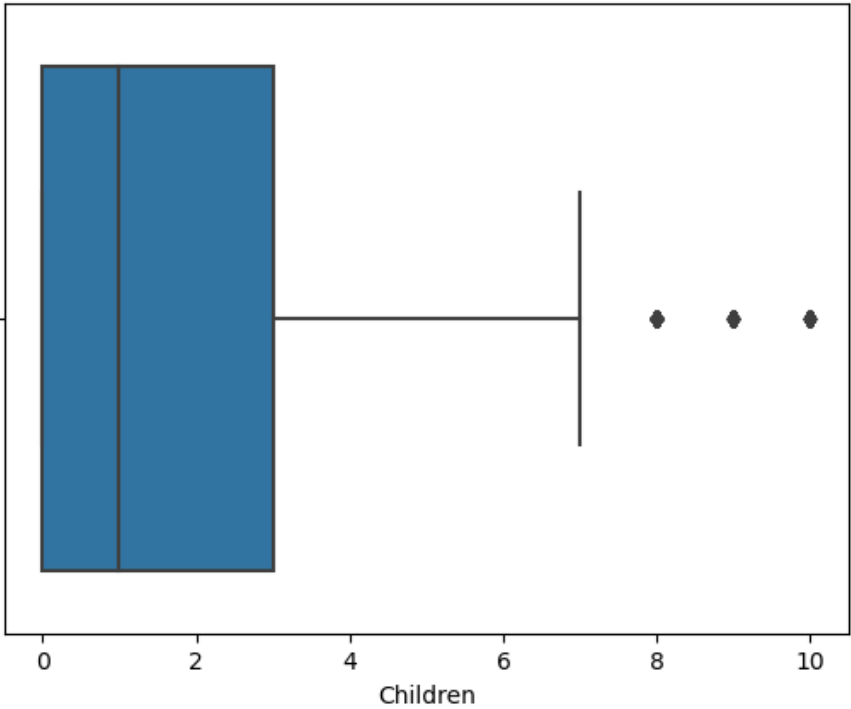
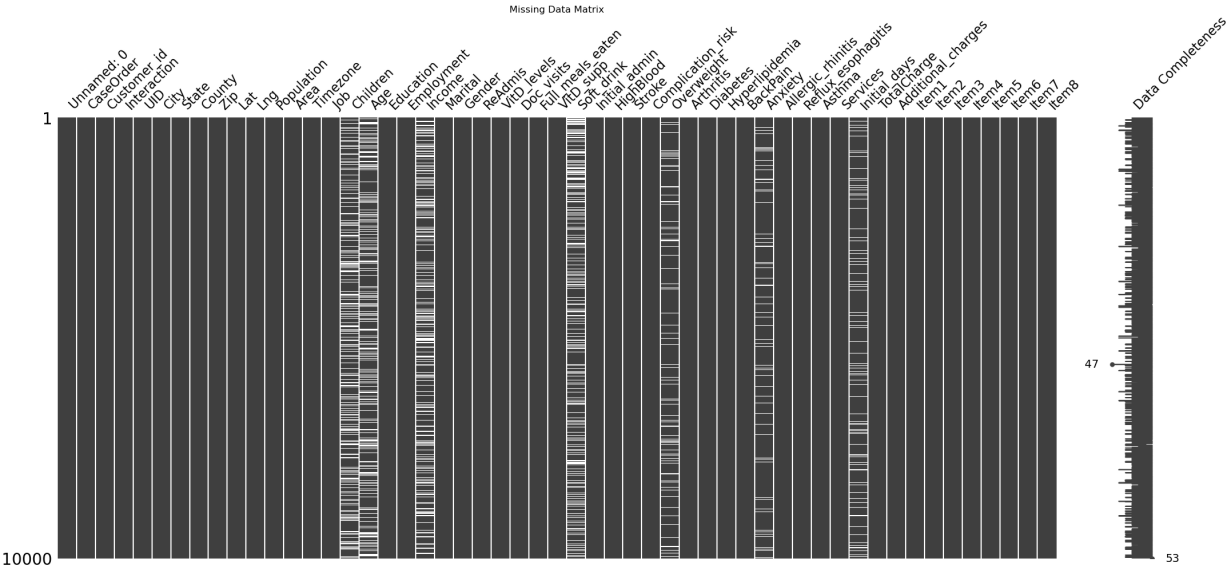
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RangeIndex: 10000 entries, 0 to 9999
Data columns (total 53 columns):
#   Column                                Non-Null Count  Dtype
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0   Unnamed: 0                            10000 non-null  int64
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                               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  object
37  Anxiety                                9016 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                           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
0      False
1      False
2      False
3      False
4      False
...
9995   False

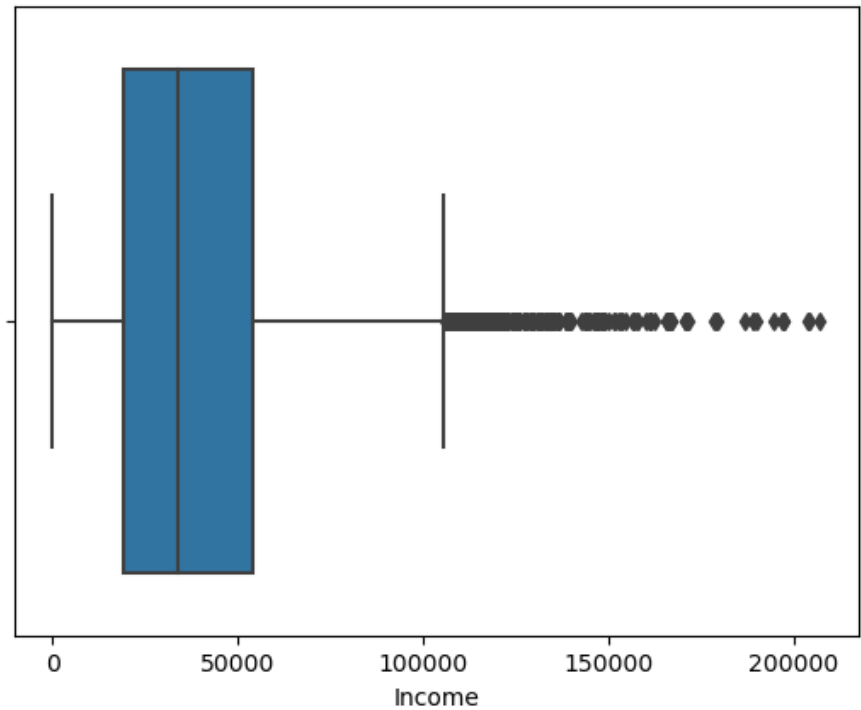
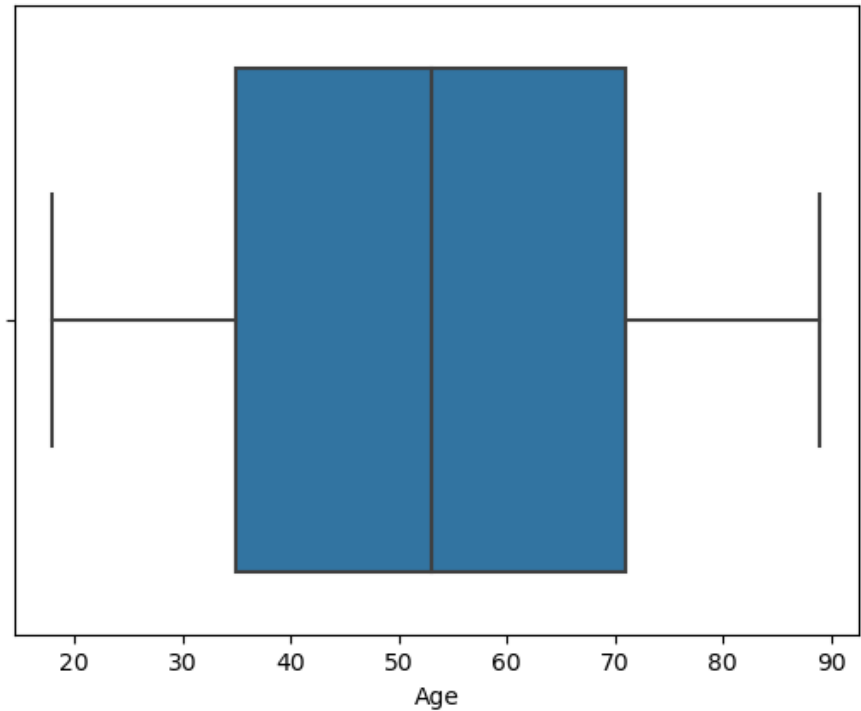
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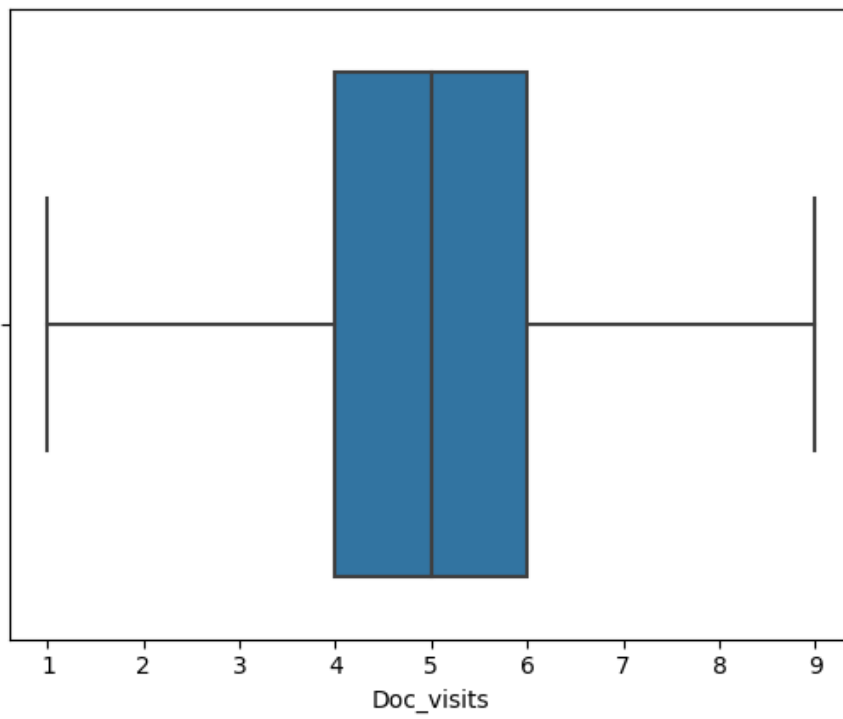
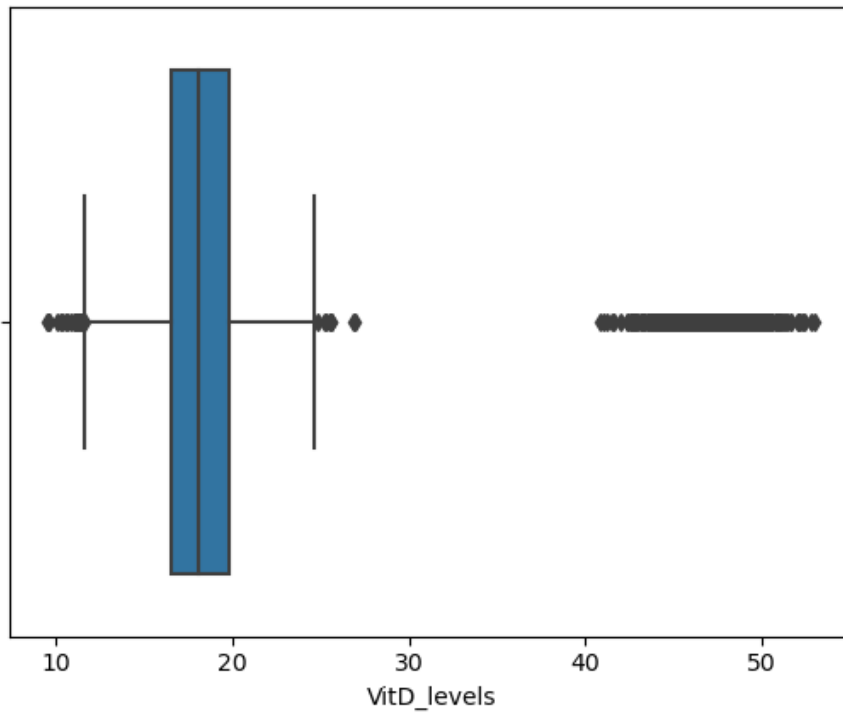
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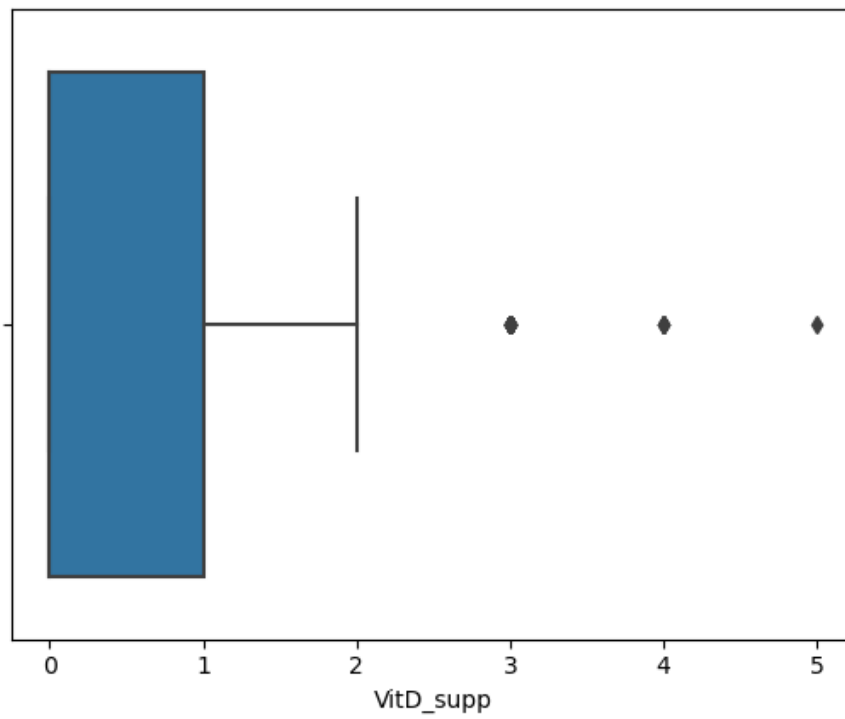
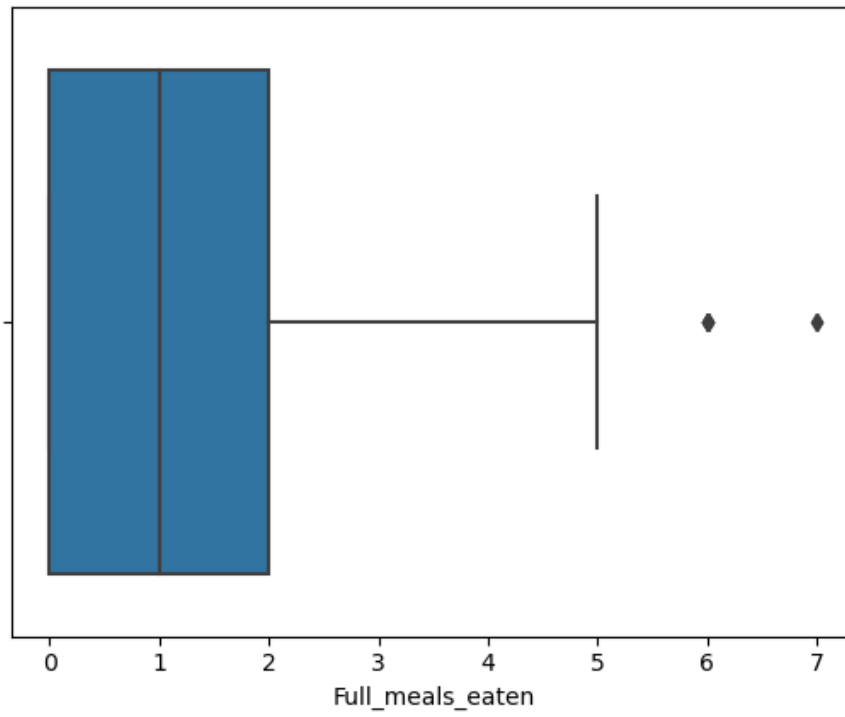
9996    False
9997    False
9998    False
9999    False
Length: 10000, dtype: bool
Unnamed: 0                0
CaseOrder                 0
Customer_id               0
Interaction               0
UID                       0
City                     0
State                     0
County                   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
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Item5                     0
Item6                     0
Item7                     0
Item8                     0
dtype: int64

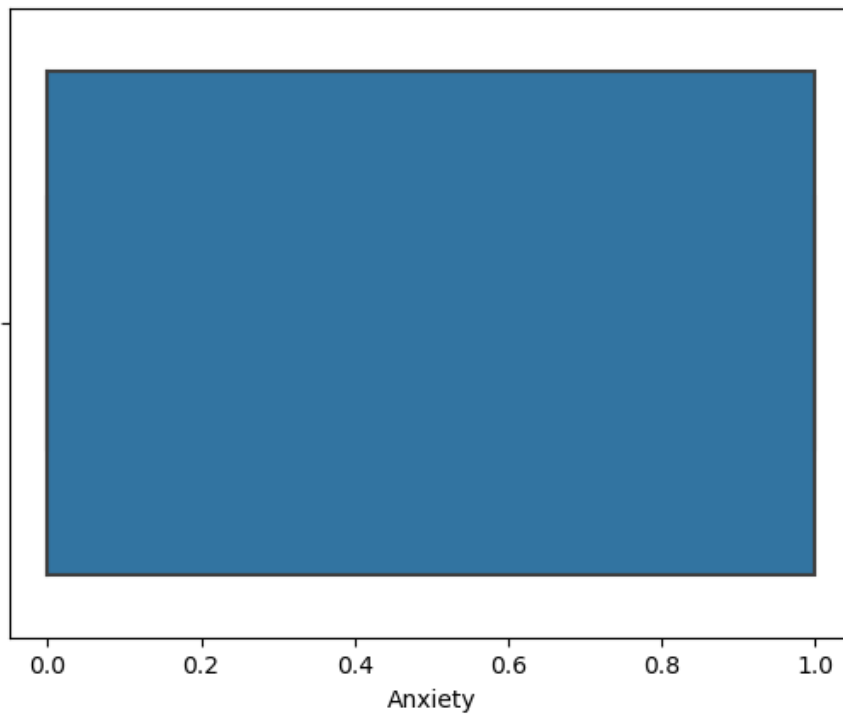
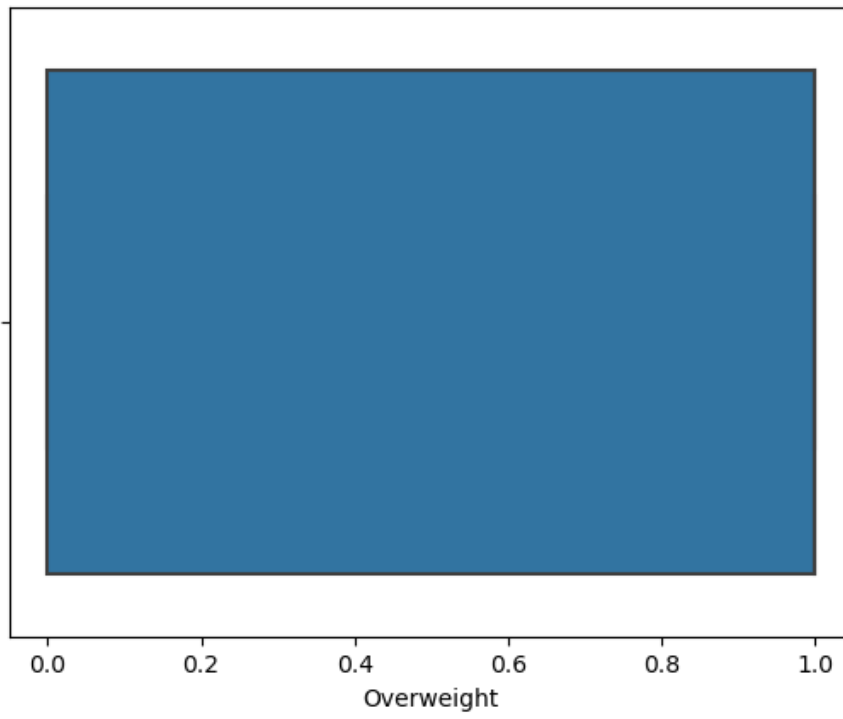
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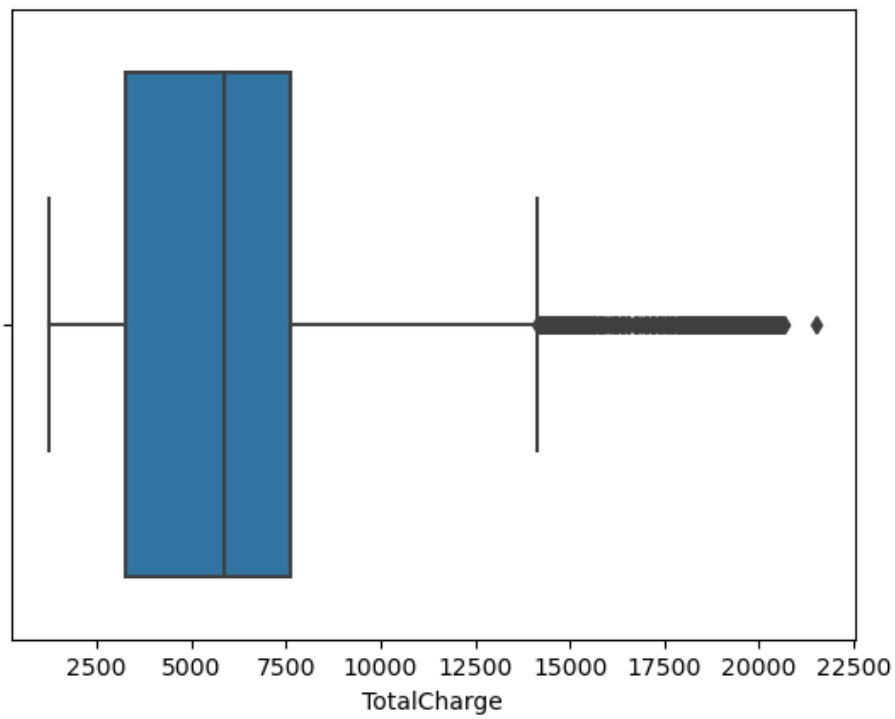
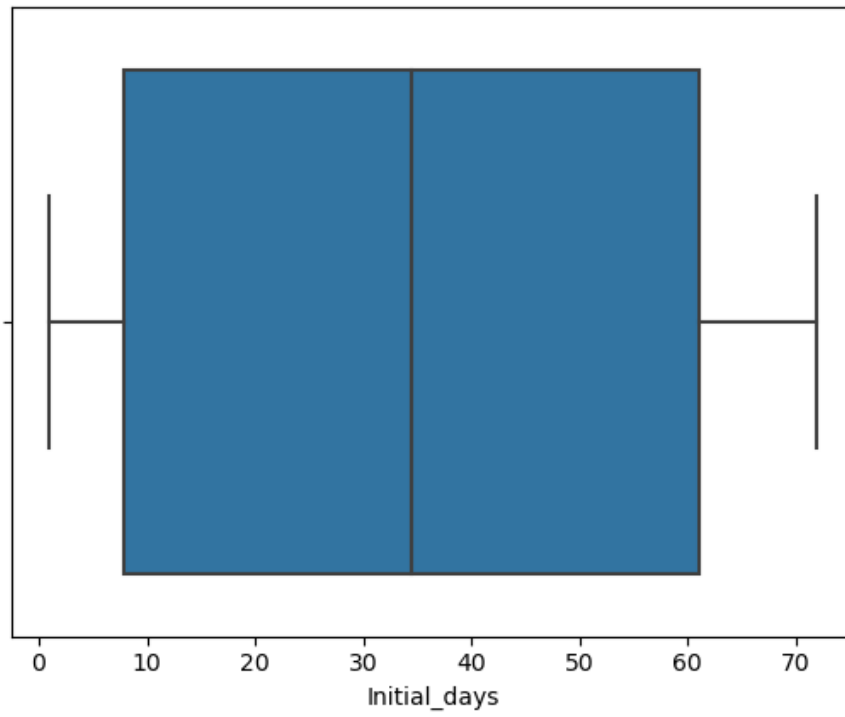


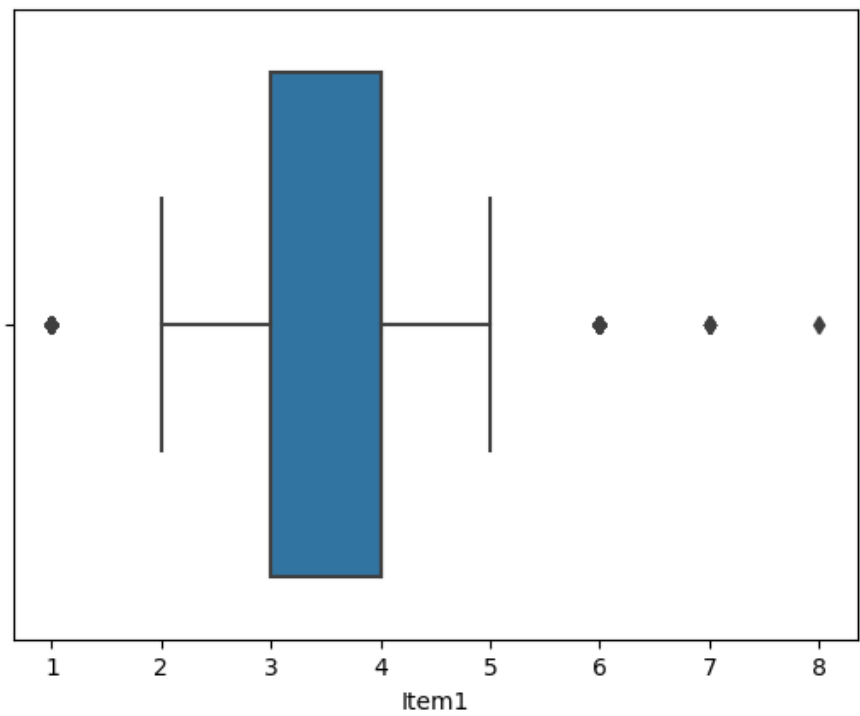
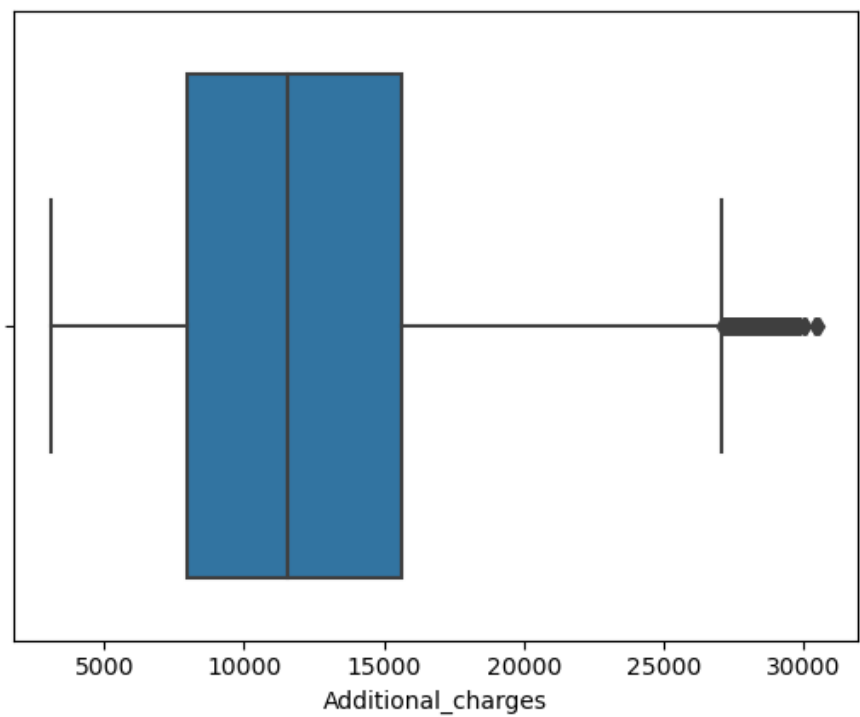


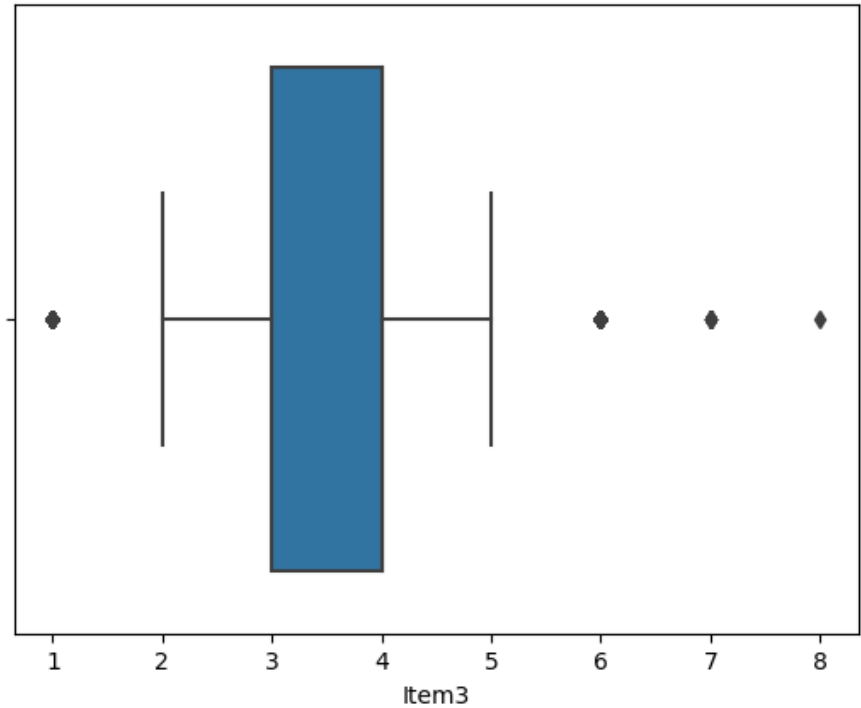
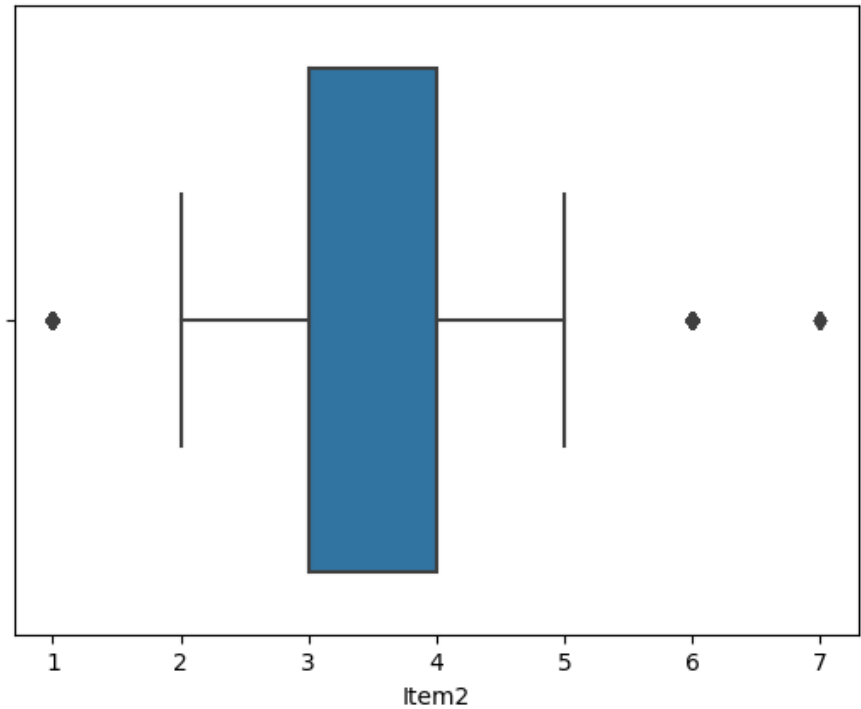


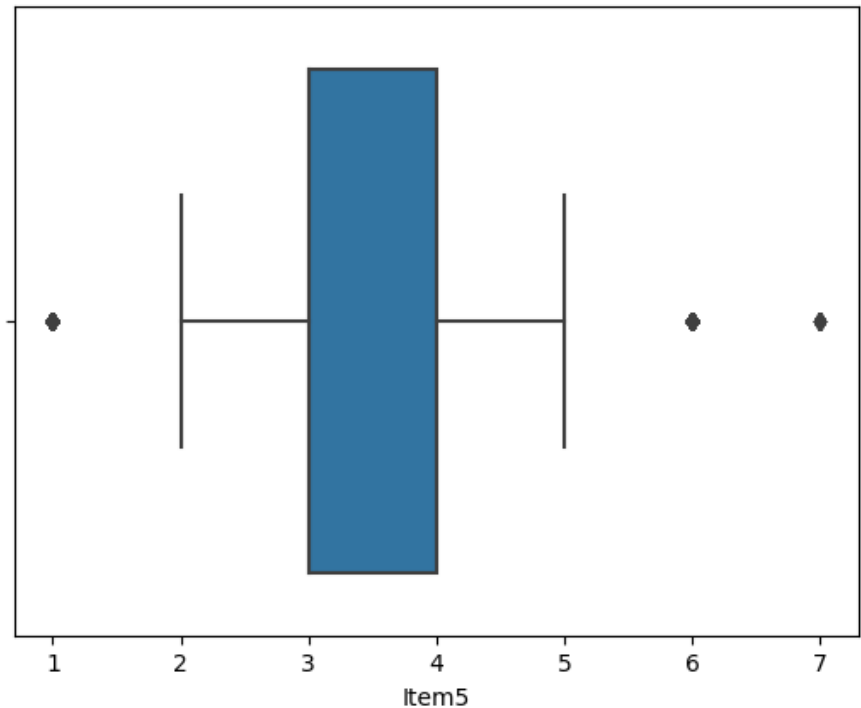
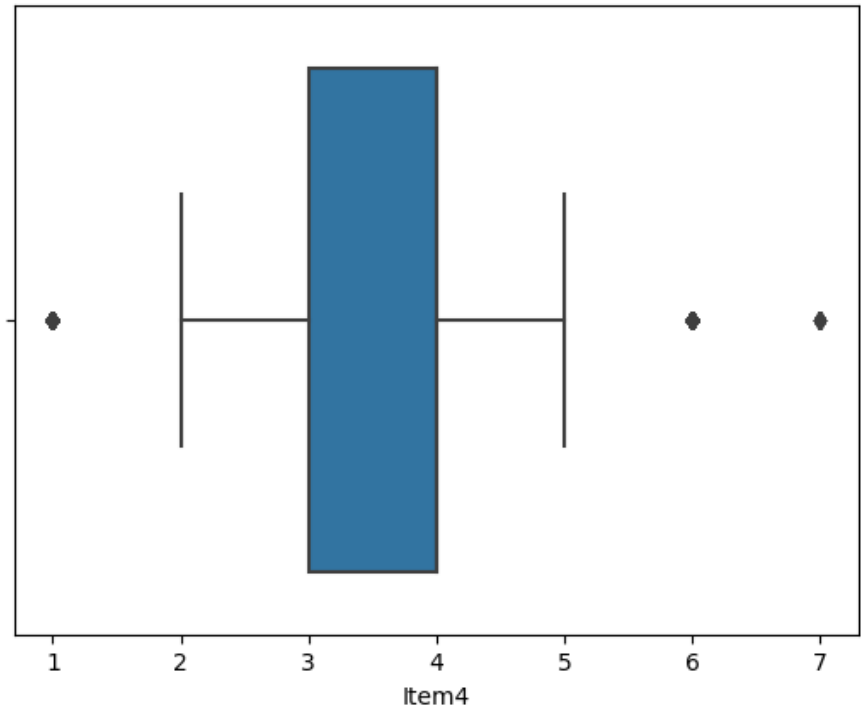


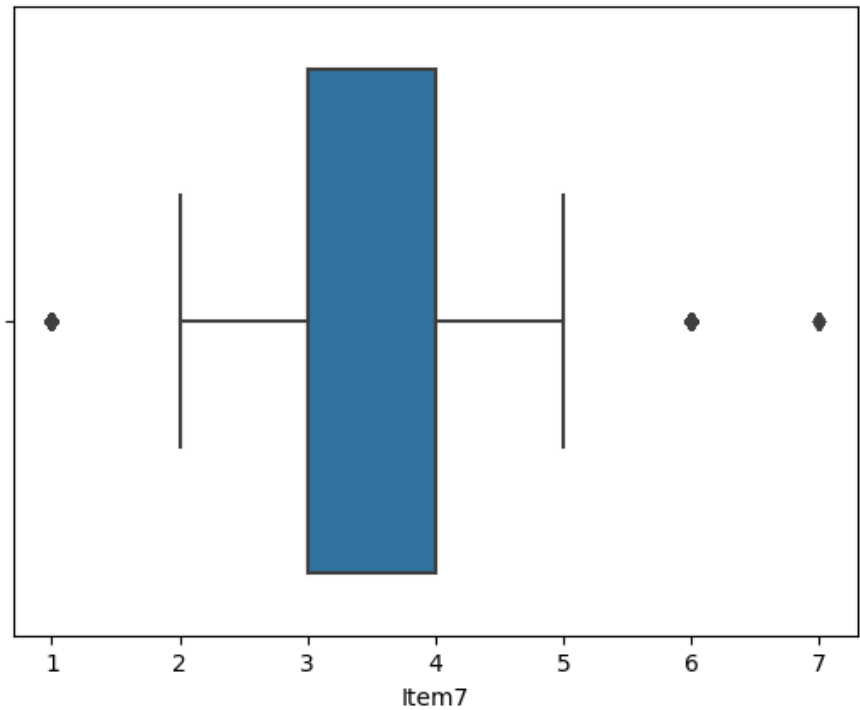
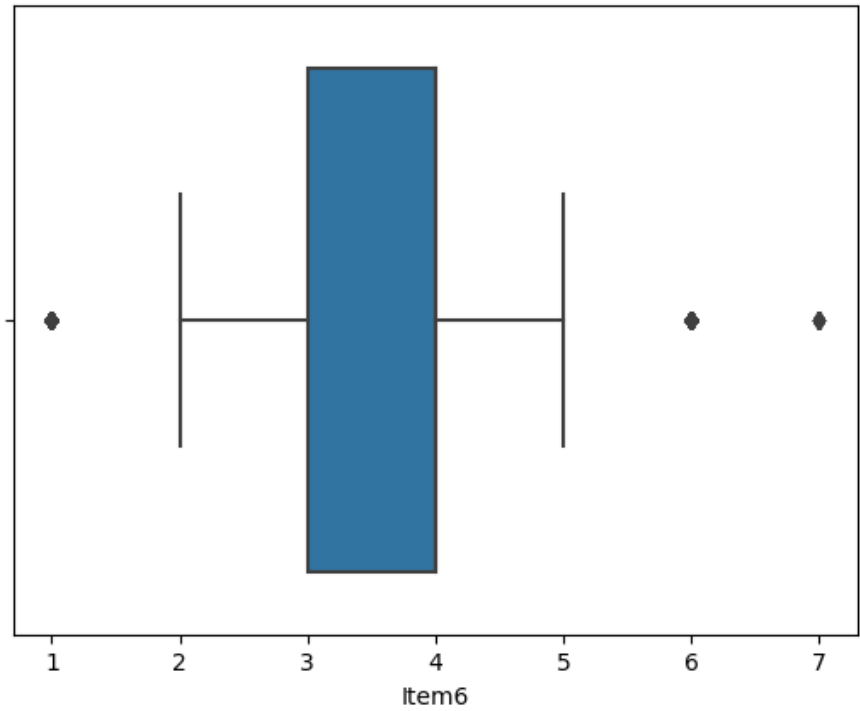


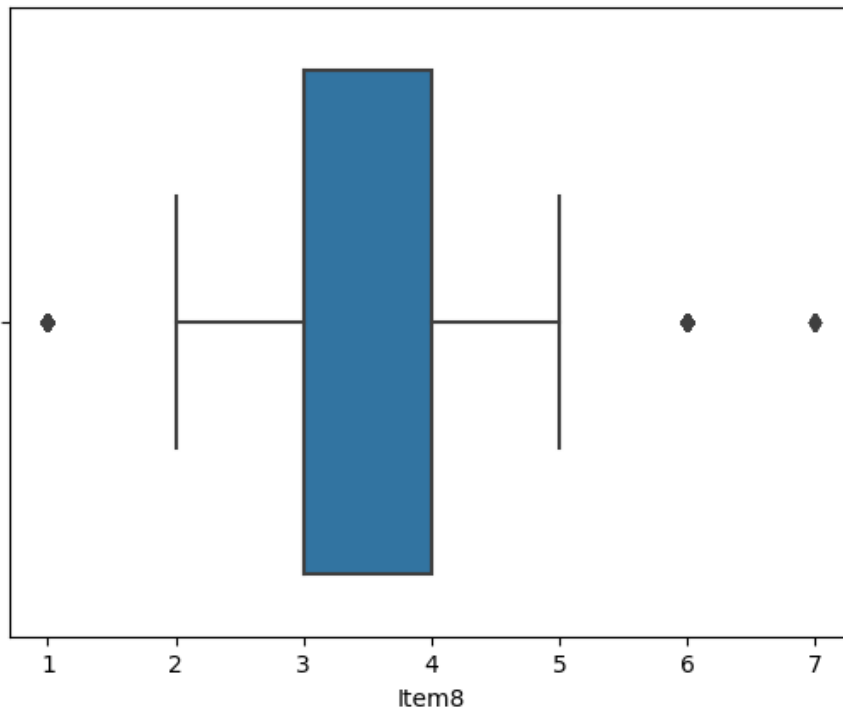




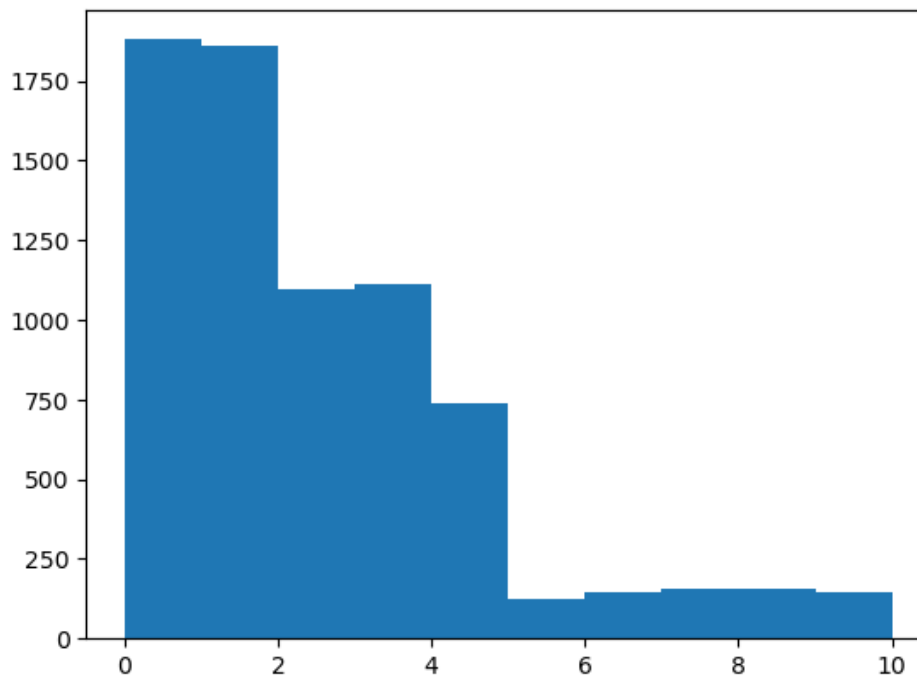




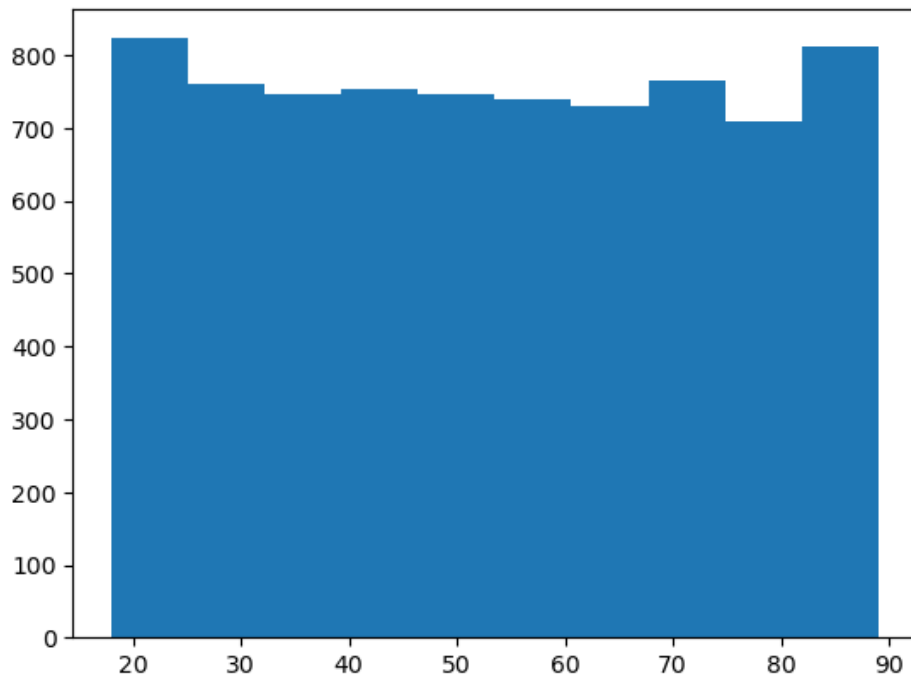




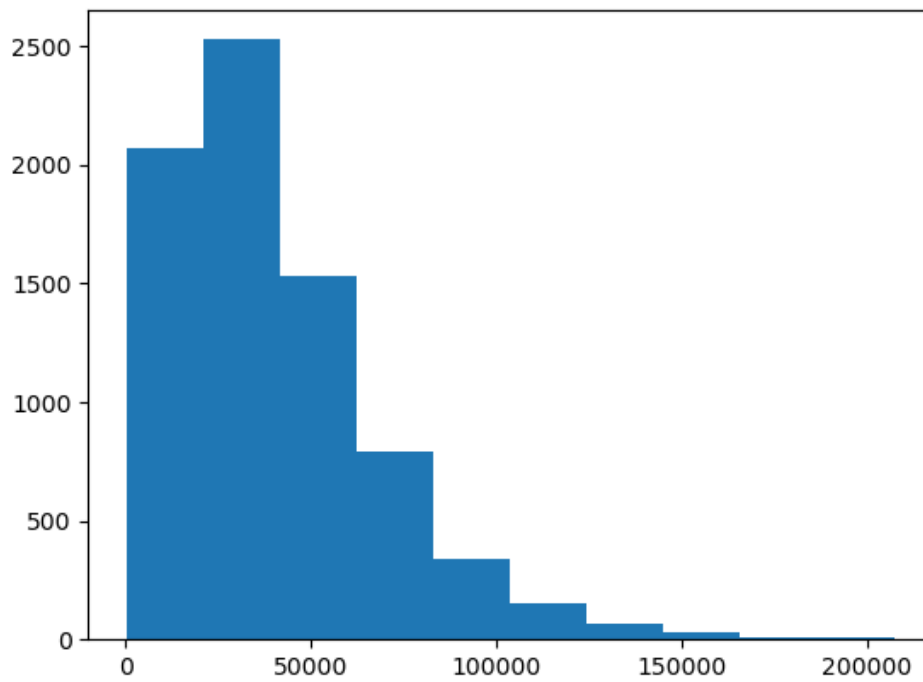
Children Original:



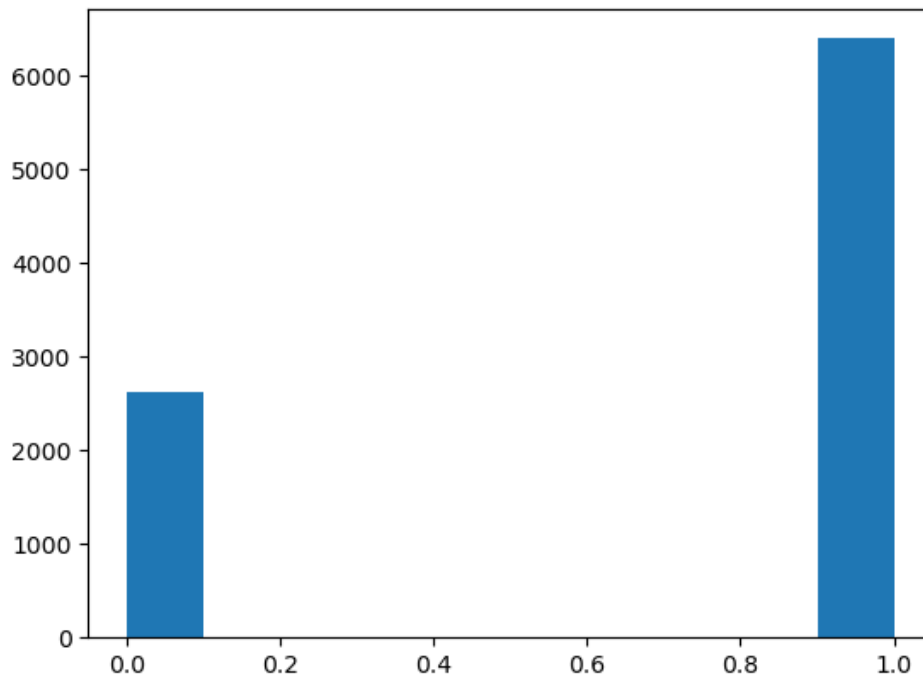
Age Original:



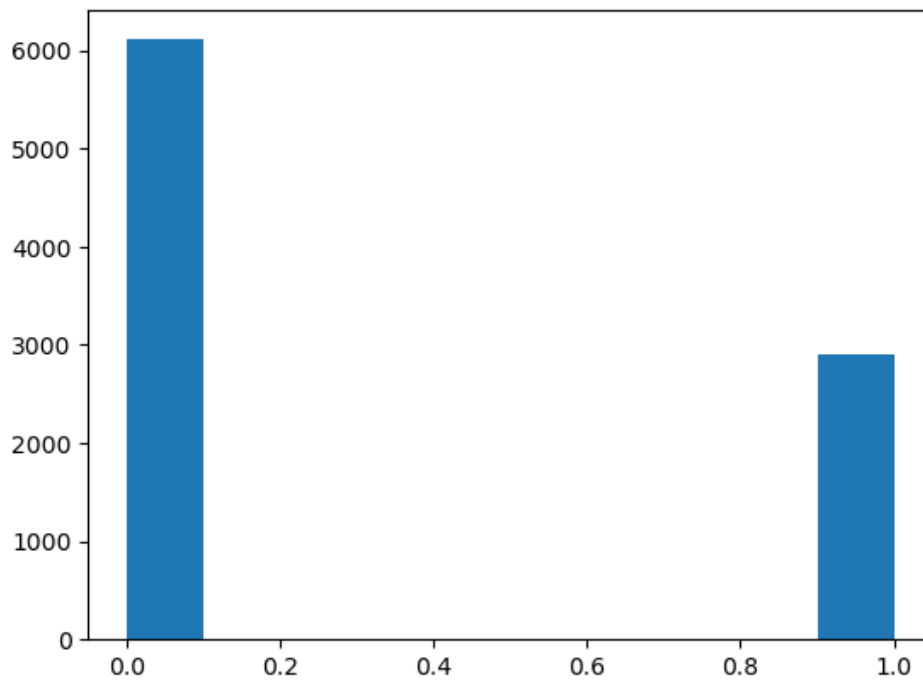
Income Original:



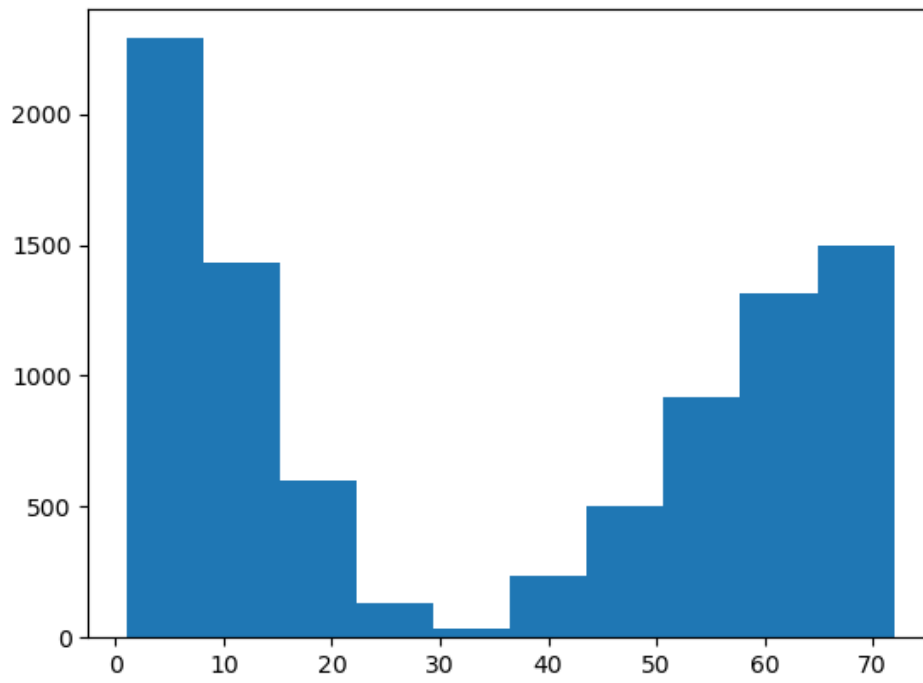
Overweight Original:



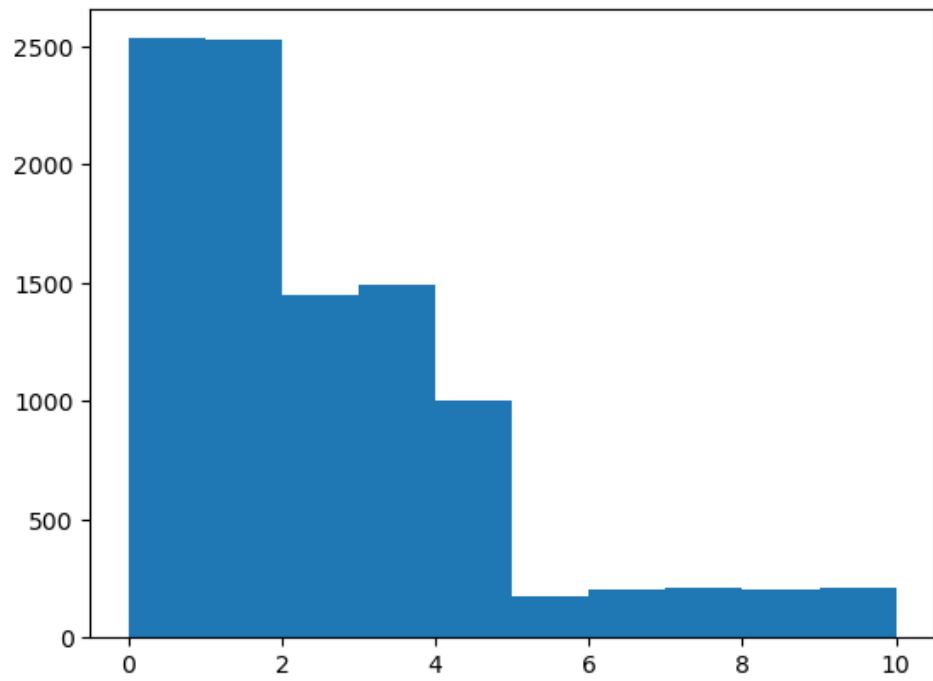
Anxiety Original:



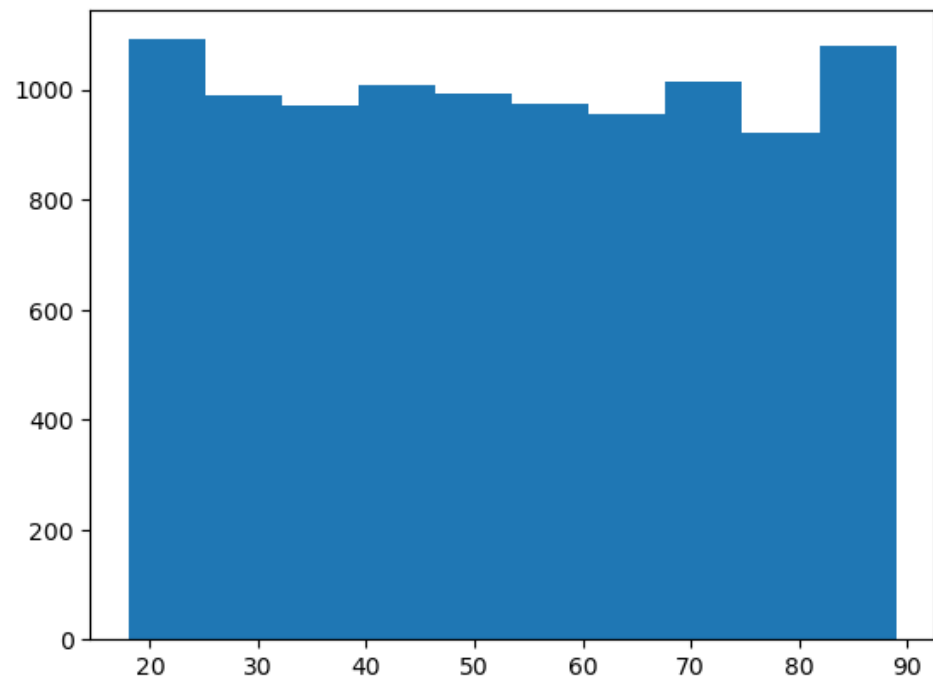
Initial_days Original:



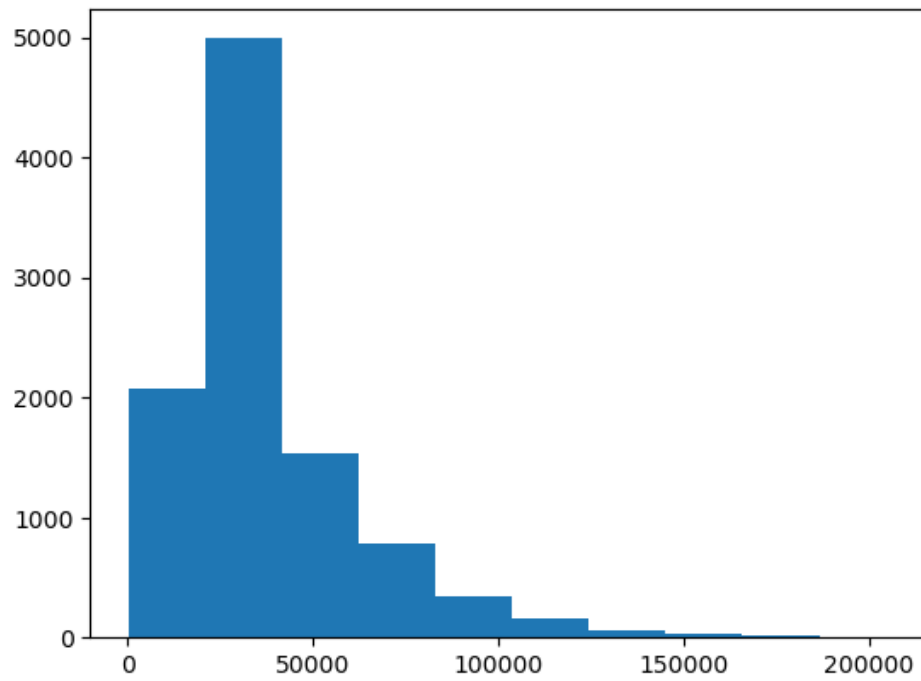
Unnamed: 0	0
CaseOrder	0
Customer_id	0
Interaction	0
UID	0
City	0
State	0
County	0
Zip	0
Lat	0
Lng	0
Population	0
Area	0
Timezone	0
Job	0
Children	0
Age	0
Education	0
Employment	0
Income	0
Marital	0
Gender	0
ReAdmis	0
VitD_levels	0
Doc_visits	0
Full_meals_eaten	0
VitD_supp	0
Soft_drink	0
Initial_admin	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
Additional_charges	0
Item1	0
Item2	0
Item3	0
Item4	0
Item5	0
Item6	0
Item7	0
Item8	0
dtype: int64	
Children Modified:	



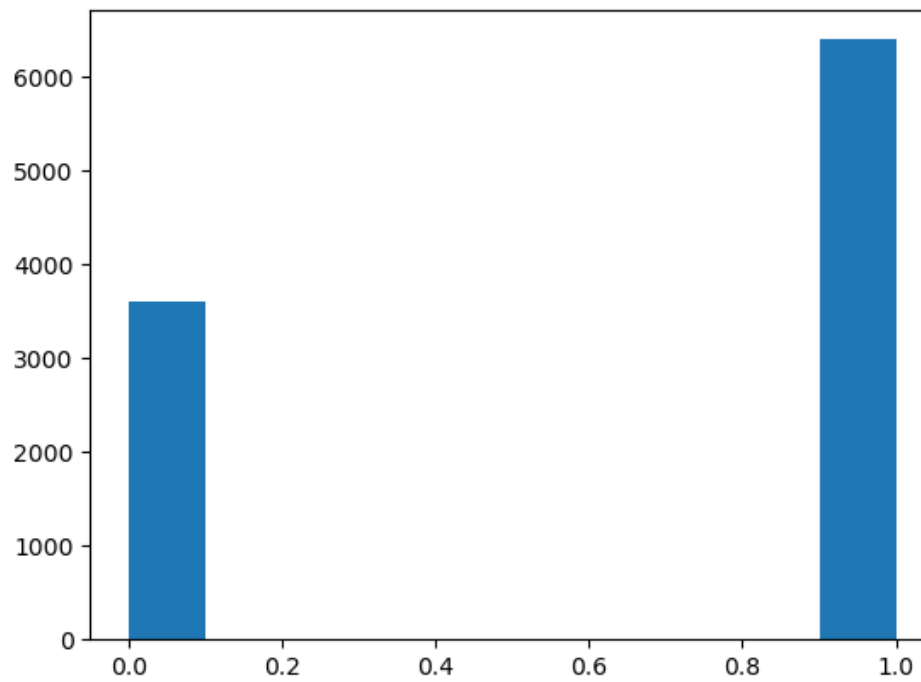
Age Modified:



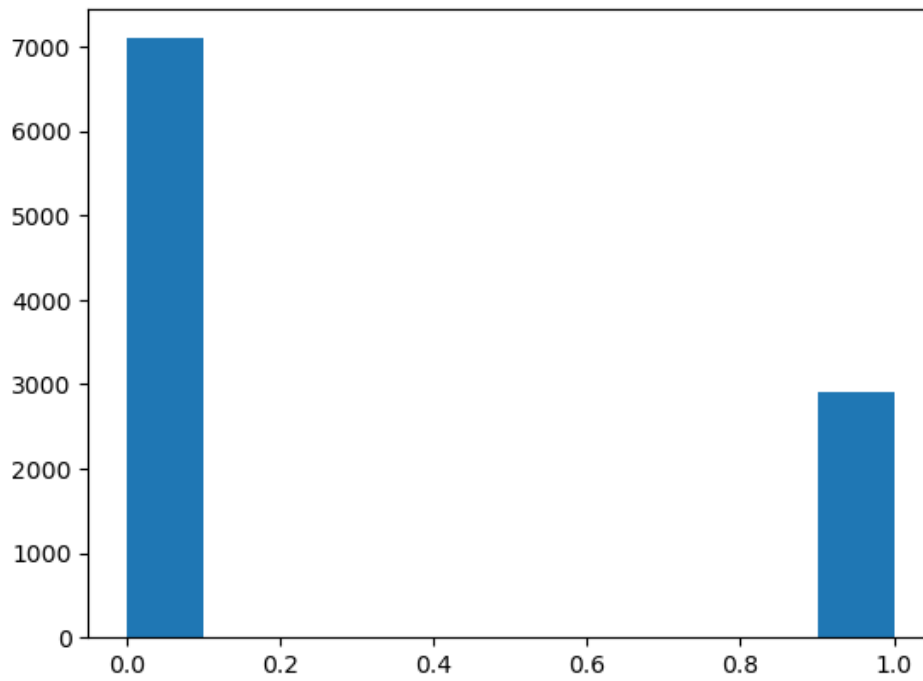
Income Modified:



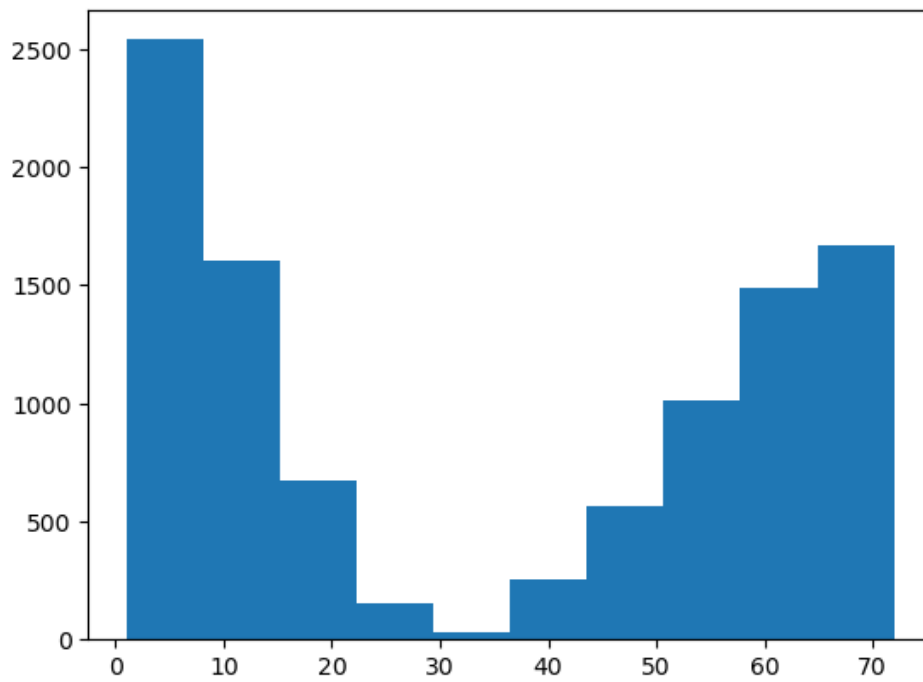
Overweight Modified:



Anxiety Modified:



Initial_days Modified:



```

Education: ['Some College, Less than 1 Year'
'Some College, 1 or More Years, No Degree'
'GED or Alternative Credential' 'Regular High School Diploma'
'Bachelor's Degree' 'Master's Degree' 'Nursery School to 8th Grade'
'9th Grade to 12th Grade, No Diploma' 'Doctorate Degree'
'Associate's Degree' 'Professional School Degree'
'No Schooling Completed']

Marital: ['Divorced' 'Married' 'Widowed' 'Never Married' 'Separated']
Gender: ['Male' 'Female' 'Prefer not to answer']
ReAdmis: ['No' 'Yes']
Soft_drink: ['No' 'Yes']
Initial_admin: ['Emergency Admission' 'Elective Admission' 'Observation Admission']
HighBlood: ['Yes' 'No']
Stroke: ['No' 'Yes']
Complication_risk: ['Medium' 'High' 'Low']
Arthritis: ['Yes' 'No']
Diabetes: ['Yes' 'No']
Hyperlipidemia: ['No' 'Yes']
BackPain: ['Yes' 'No']
Allergic_rhinitis: ['Yes' 'No']
Reflux_esophagitis: ['No' 'Yes']
Ashtma: ['Yes' 'No']
Services: ['Blood Work' 'Intravenous' 'CT Scan' 'MRI']
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10000 entries, 0 to 9999
Data columns (total 70 columns):

```

#	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
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 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 Education_numeric       10000 non-null int64
54 Marital_numeric         10000 non-null int64
55 Gender_numeric          10000 non-null int64
56 ReAdmis_numeric         10000 non-null int64
57 Soft_drink_numeric      10000 non-null int64
58 Initial_admin_numeric   10000 non-null int64
59 HighBlood_numeric       10000 non-null int64
60 Stroke_numeric          10000 non-null int64
61 Complication_risk_numeric 10000 non-null int64
62 Arthritis_numeric       10000 non-null int64
63 Diabetes_numeric        10000 non-null int64
64 Hyperlipidemia_numeric  10000 non-null int64
65 BackPain_numeric        10000 non-null int64
66 Allergic_rhinitis_numeric 10000 non-null int64
67 Reflux_esophagitis_numeric 10000 non-null int64
68 Asthma_numeric          10000 non-null int64
69 Services_numeric        10000 non-null int64

```

dtypes: float64(11), int64(32), object(27)

memory usage: 5.3+ MB

Education_numeric: [5 6 3 4 9 10 1 2 11 7 8 0]

Marital_numeric: [3 4 2 0 1]

Gender_numeric: [1 2 0]

ReAdmis_numeric: [0 1]

Soft_drink_numeric: [0 1]

Initial_admin_numeric: [0 1 2]

HighBlood_numeric: [1 0]

Stroke_numeric: [0 1]

Complication_risk_numeric: [1 2 0]

Arthritis_numeric: [1 0]

Diabetes_numeric: [1 0]

Hyperlipidemia_numeric: [0 1]

BackPain_numeric: [1 0]

Allergic_rhinitis_numeric: [1 0]

Reflux_esophagitis_numeric: [0 1]

Ashtma_numeric: [1 0]

Services_numeric: [0 1 2 3]

PCA(n_components=20)

	PC1	PC2	PC3	PC4	PC5	\
ReAdmis_numeric	0.019851	0.127795	0.175456	-0.353865	-0.032540	
Services_numeric	0.705512	0.009579	-0.013558	-0.004255	-0.010995	
Age	0.013203	-0.229001	-0.143450	0.074654	0.354792	
Gender_numeric	0.036482	-0.036906	0.062988	0.395736	-0.274561	
Doc_visits	-0.001959	-0.261453	0.133387	0.008292	-0.242440	
Soft_drink_numeric	0.011052	0.114656	0.543563	0.084041	0.164635	
Initial_admin_numeric	-0.021101	-0.112728	-0.255054	0.400224	-0.167758	
HighBlood_numeric	0.007924	-0.394309	-0.055899	-0.298067	-0.065738	
Stroke_numeric	-0.003465	-0.196086	0.016540	0.251773	-0.072744	
Complication_risk_numeric	0.004593	-0.140210	-0.025062	-0.192025	-0.258244	
Arthritis_numeric	-0.004227	0.047176	-0.436380	-0.255660	0.114494	
Diabetes_numeric	0.034947	0.094009	-0.086680	0.241931	0.356749	
Hyperlipidemia_numeric	-0.004628	0.427117	0.150956	-0.177230	0.298589	

BackPain_numeric	0.008511	-0.236902	0.538287	0.062751	0.062242
Allergic_rhinitis_numeric	0.006058	-0.242265	-0.058299	-0.164063	0.247245
Reflux_esophagitis_numeric	0.004447	0.157243	0.136674	0.022118	-0.331555
Asthma_numeric	0.004309	-0.159516	0.043155	0.303615	0.411165
Services_numeric	0.705512	0.009579	-0.013558	-0.004255	-0.010995
Education_numeric	-0.019142	0.238284	0.026542	-0.072749	-0.164102
Marital_numeric	0.012822	-0.454559	0.163839	-0.268373	0.081588

	PC6	PC7	PC8	PC9	PC10	\
ReAdmis_numeric	-0.143015	-0.119035	0.181851	0.522127	0.108240	
Services_numeric	0.004155	0.017485	0.019507	-0.009729	-0.010693	
Age	-0.062761	-0.149631	-0.112612	0.494622	0.017862	
Gender_numeric	-0.219364	-0.189680	-0.461167	0.033749	-0.077140	
Doc_visits	0.246096	-0.582819	0.107615	-0.035123	-0.031306	
Soft_drink_numeric	0.196500	-0.273601	-0.004180	-0.054392	0.139432	
Initial_admin_numeric	0.020542	0.095840	0.392278	0.006889	0.073086	
HighBlood_numeric	0.095530	0.093574	0.124019	-0.150690	0.437013	
Stroke_numeric	0.110094	0.149226	0.225268	0.426483	0.119244	
Complication_risk_numeric	0.494101	0.105350	-0.316720	0.033632	0.347314	
Arthritis_numeric	-0.369964	-0.313207	0.087425	0.013103	0.228266	
Diabetes_numeric	0.222501	-0.425421	0.146032	-0.072290	0.276184	
Hyperlipidemia_numeric	0.159638	0.191373	-0.063646	0.029092	0.110516	
BackPain_numeric	-0.338007	0.069437	0.161794	0.074169	0.033480	
Allergic_rhinitis_numeric	-0.144564	-0.223779	-0.321020	-0.215393	-0.121809	
Reflux_esophagitis_numeric	-0.441239	-0.036442	-0.120240	-0.084374	0.545925	
Asthma_numeric	-0.102526	0.238795	0.013412	-0.289638	0.330642	
Services_numeric	0.004155	0.017485	0.019507	-0.009729	-0.010693	
Education_numeric	-0.035461	-0.156908	0.460780	-0.308875	-0.114648	
Marital_numeric	-0.094869	0.086213	0.140313	-0.163975	-0.230519	

	PC11	PC12	PC13	PC14	PC15	\
ReAdmis_numeric	0.014549	0.449140	0.051658	0.140656	-0.415099	
Services_numeric	-0.007358	0.000096	0.003527	-0.021920	0.016814	
Age	-0.024631	-0.458214	-0.116199	-0.376736	-0.229437	
Gender_numeric	0.136788	-0.143189	0.042206	0.288660	-0.347084	
Doc_visits	-0.230820	0.045052	-0.021763	0.047394	0.185108	
Soft_drink_numeric	-0.071909	-0.002802	-0.102073	-0.294438	0.185735	
Initial_admin_numeric	-0.368419	-0.069619	0.460711	-0.110073	-0.115627	
HighBlood_numeric	0.086318	-0.038997	0.218765	0.120616	-0.131129	
Stroke_numeric	0.598561	0.017625	0.010997	0.040444	0.435130	
Complication_risk_numeric	-0.026769	-0.101371	-0.070682	-0.220261	-0.220000	
Arthritis_numeric	-0.169559	-0.078617	-0.158997	-0.032542	0.260781	
Diabetes_numeric	0.164254	-0.046727	0.067213	0.448329	-0.164821	
Hyperlipidemia_numeric	-0.058372	-0.400797	0.374411	0.211075	0.080255	
BackPain_numeric	-0.202252	-0.095895	0.236883	-0.071264	-0.040638	
Allergic_rhinitis_numeric	0.331937	0.225528	0.566124	-0.252433	0.069993	
Reflux_esophagitis_numeric	0.098464	-0.189384	-0.021739	-0.020853	0.170788	
Asthma_numeric	-0.016352	0.322078	-0.316182	-0.086194	-0.191273	
Services_numeric	-0.007358	0.000096	0.003527	-0.021920	0.016814	
Education_numeric	0.444033	-0.238318	-0.060882	-0.354637	-0.377126	
Marital_numeric	0.048836	-0.351229	-0.234861	0.373097	-0.028496	

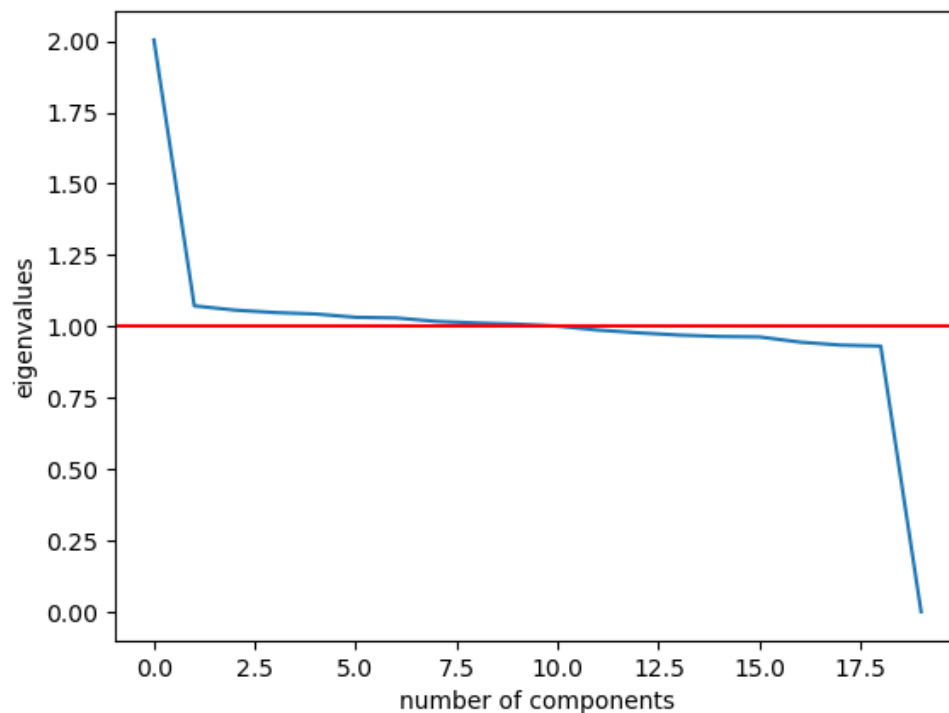
	PC16	PC17	PC18	PC19	\
ReAdmis_numeric	-0.050733	-0.058256	-0.108383	0.244485	
Services_numeric	-0.002500	0.008656	-0.000445	0.003786	
Age	0.062460	0.164435	-0.238058	-0.071583	
Gender_numeric	0.275468	-0.028774	0.269179	0.252051	
Doc_visits	-0.116821	0.541442	-0.035362	0.201534	
Soft_drink_numeric	0.376105	-0.450375	-0.065626	0.189284	
Initial_admin_numeric	-0.047727	-0.267675	-0.149689	0.299899	
HighBlood_numeric	0.604621	0.125542	-0.007621	-0.176800	
Stroke_numeric	-0.028099	0.008687	0.177146	0.171509	
Complication_risk_numeric	-0.401106	-0.207756	0.276240	0.074581	
Arthritis_numeric	0.045635	-0.178694	0.480117	0.200406	
Diabetes_numeric	-0.238413	-0.191279	-0.025169	-0.327723	
Hyperlipidemia_numeric	-0.008969	0.305429	0.105164	0.367868	
BackPain_numeric	-0.225011	0.043870	0.458607	-0.344312	

Allergic_rhinitis_numeric	-0.177501	-0.073691	-0.092135	0.137754
Reflux_esophagitis_numeric	-0.224839	0.037990	-0.449507	-0.028934
Asthma_numeric	-0.085380	0.278721	0.054463	0.338494
Services_numeric	-0.002500	0.008656	-0.000445	0.003786
Education_numeric	-0.021570	0.108999	0.140777	0.073220
Marital_numeric	-0.183192	-0.285961	-0.193797	0.312618

PC20

ReAdmis_numeric	-3.569954e-17
Services_numeric	7.071068e-01
Age	1.543904e-16
Gender_numeric	1.110223e-16
Doc_visits	-2.081668e-17
Soft_drink_numeric	1.110223e-16
Initial_admin_numeric	8.326673e-17
HighBlood_numeric	-1.040834e-17
Stroke_numeric	5.724587e-17
Complication_risk_numeric	-9.714451e-17
Arthritis_numeric	-2.775558e-17
Diabetes_numeric	-3.469447e-17
Hyperlipidemia_numeric	-3.469447e-17
BackPain_numeric	2.775558e-17
Allergic_rhinitis_numeric	4.857226e-17
Reflux_esophagitis_numeric	-8.326673e-17
Asthma_numeric	2.220446e-16
Services_numeric	-7.071068e-01
Education_numeric	4.510281e-17
Marital_numeric	-2.775558e-17

Scree Plot



In []: